From word models to executable models of signaling networks using automated assembly

Appendix

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Figure S1. TRIPS logical form (LF) and extraction knowledge base (EKB) graphs.

(A) The LF graph for the sentence “MEK1 phosphorylates ERK2 at threonine 185 and tyrosine 187”.
(B) The EKB graph for the sentence “MEK1 phosphorylates ERK2 at threonine 185 and tyrosine 187”.
(C) The LF graph for the sentence “GRB2 binds tyrosine-phosphorylated EGFR”.
(D) The EKB graph for the sentence “GRB2 binds tyrosine-phosphorylated EGFR” shows the main BIND event with GRB2 and EGFR as its arguments. EGFR is further affected by a PHOSPHORYLATION sub-event with the site specified as a tyrosine amino-acid residue.
(E) The LF graph for the sentence “DUSP6 dephosphorylates ERK2 that is not bound to SOS1”.
(F) The EKB graph for the sentence “DUSP6 dephosphorylates ERK2 that is not bound to SOS1” has a main event PHOSPHORYLATION which is negated by a modifier to represent dephosphorylation and which has ERK2 affected by DUSP6 as its arguments. ERK2 is further affected by a negated binding sub-event with SOS1.
Figure S2. Unified Modeling Language (UML) diagram of INDRA Statements and associated classes. All INDRA Statements inherit from the Statement class. Each Statement has one or more Agents associated with them. An Agent represents molecular context through additional attributes (e.g. location) and associated classes including ModCondition (for post-translational modifications), MutCondition (for mutations), BoundCondition (for bound co-factors) and ActivityCondition (for active/inactive state). Some Statement types also have attributes (e.g. residue, position) in addition to their Agent arguments. Each Statement has one or more Evidences associated with it.
Figure S3. INDRA Statement visualized as graph and serialized as JSON. INDRA allows displaying the data structure of Statements as graphs (A), and provides a JSON exchange format (B) for inspection and reuse in a platform-independent way.
Figure S4. Examples of INDRA Agents representing molecular state. (A) Agent representing *active TP53 in the nucleus*. (B) Agent representing *KRAS-G12V that is not farnesylated*. (C) Agent representing *BRAF that is ubiquitinated, phosphorylated at S365 and not bound to NRAS*. 
Figure S5. BioPAX representations of the mechanism double-phosphorylated MEK1 phosphorylates ERK2 as a ChIBE (Babur et al, 2010) graph (above) and an object model graph (below).
Active ATM activates p53.
Active p53 activates Wip1.
Active p53 activates Mdm2.
Active Wip1 inactivates p53.
Active Mdm2 inactivates p53.

Active ATR activates p53.
Active p53 activates Wip1.
Active p53 activates Mdm2.
Active Wip1 inactivates p53.
Active Mdm2 inactivates p53.

HIPK2 inactivates Wip1.
P14ARF inactivates Mdm2.

Active ATR activates p53.
Active p53 activates Wip1.
Active p53 activates Mdm2.
Active Wip1 inactivates p53.
Active Mdm2 inactivates p53.

Active ATM activates p53.
Active p53 activates Wip1.
Active p53 activates Mdm2.
Active Wip1 inactivates p53.
Active Wip1 inactivates ATM.
Active Mdm2 inactivates p53.

Active ATM phosphorylates itself, and phosphorylated ATM is active.
Active ATM activates p53.
Active p53 is transcribed and active p53 transcribes MDM2.
MDM2 is degraded.
Active p53 activates Wip1.
Active Wip1 inactivates p53.
Active Wip1 dephosphorylates ATM.
MDM2 ubiquitinates p53 and ubiquitinated p53 is degraded.
HIPK2 inactivates Wip1.

Figure S6. The dynamics of active p53 under 100 simulation conditions (gray, some samples highlighted in colors), with parameters randomly sampled from uncorrelated log-normal distributions around their nominal values. The simulation dynamics with nominal parameters are shown in red. (A) SSB model from yellow numbered edges in Figure 5A.
(B) SSB model with Wip1 and Mdm2 inactivation added.
(C) SSB model with ATR activation added.
(D) DSB model from yellow numbered edges in Figure 5A.
(E) DSB model with Wip1 and Mdm2 inactivation added.
(F) DSB model with ATM activation added.
(G) Detailed DSB model with ATM cis-autophosphorylation.
(H) POMI1.0: Detailed DSB model with ATM trans-autophosphorylation.
Figure S7. Heat maps showing steady state values of active RAS (left) and phospho-ERK rebound (right) upon varying doses of EGF and vemurafenib. (A) Dose response heat maps for MEMI1.0 (B) Dose response heat maps for MEMI1.1 (C) Dose response heat maps for MEMI1.2
Figure S8. Biochemical reactions involving BRAF in the MEMI1.2 model. Each node corresponds to an individual molecular species with the label composed of the first letters of the monomers the species consists of: BRAF (B), RAS (R), GTP (G), vemurafenib (V) and MEK (M). Each directed edge corresponds to a reaction which produces the target species from the source species. The network visualization is simplified as it omits modification states and merges complexes with a given set of constituents that have different topologies.
Figure S9. INDRA-assembled SBGN model of p53 activation upon DSB. This model was assembled from the natural language shown in Figure 5C (main text) using INDRA’s SBGN Assembler module. The SBGN graph was visualized and layout was set manually in the SBGNViz online editor (Sari et al, 2015).
**Figure S10.** Selected BioPAX Biochemical Reactions involving MAPK1/3 and MAP2K1 in Pathway Commons, displayed using the Chisio BioPAX Editor (ChIBE) (Babur *et al.*, 2010).
2 Appendix Methods

2.1 The TRIPS/DRUM natural language processing system

In this section we provide more technical details on the TRIPS/DRUM system (Allen et al, 2015) which INDRA uses to process natural language text (note that in the main text we refer to the system as TRIPS). The DRUM (Deep Reader for Understanding Mechanisms) system is an instance of the TRIPS general deep language understanding system (Allen et al, 2008), customized for reading and understanding scientific text in molecular biology.

To process natural language input, first a suite of shallow NLP tools performs annotations on the text, including: a) sentence splitting; b) named entity recognition; c) derivational analysis (e.g., prefixes, pertainyms); and d) statistical parsing. Critically, DRUM uses a number of domain-specific named entity taggers to identify genes, proteins, protein families, drugs and chemicals, cell lines, diseases, etc.; successful matches are annotated with IDs in the original resources, and are also mapped to internal ontology types. At this stage all matches are annotated; there is no attempt to disambiguate among alternatives. Other domain-specific annotators use regular expressions to identify, for example, molecular sites and mutations. Statistical parsers – CoreNLP (Manning et al, 2014) and Enju (Hara et al, 2005) – are used to find constituent boundaries; only those constituents on which both parsers agree are used.

The output of all these specialized preprocessors is sent to the TRIPS parser, which uses it as advice during its search for the optimal parse of each sentence. The TRIPS parser is at the core of our approach; it uses a hand-built lexicalized context-free grammar, augmented with feature structures and feature unification, and domain-general selectional restrictions (encoded in the lexicon and ontology) to eliminate semantically anomalous sense combinations. The parser constructs from the input a logical form (LF), which is a semantic representation that captures an unscoped modal logic (Allen et al, 1995; Manshadi et al, 2008). The logical form includes the surface speech act, semantic types, semantic roles for predicate arguments, and dependency relations. Lexical entities in the LF represent word senses and ontology types, as well as tense, modality and aspect information – information that is crucial for determining, for example, whether a statement expresses a stated fact, a conjecture or a possibility. The parser draws on a general purpose semantic lexicon and ontology which define a range of word senses and lexical semantic relations. The core semantic lexicon was constructed by hand and contains approximately 7,500 lemmas (generating approximately three times that many words) and 2,000 concepts in the ontology. The ontology is organized hierarchically and each ontology concept has associated with it possible semantic roles and selectional preferences that further refine the concept. The core lexicon is extended to cover virtually all words in WordNet (Fellbaum, 2010) by adding lexical entries with plausible semantic and syntactic structures through a dynamic mapping between the WordNet hypernym hierarchy and that of the TRIPS ontology. Ontology-based lexical expansion – using WordNet as well as all the terminology extracted from the biological resources – is the main tool by which we can customize our generic, relatively low coverage semantic parser to attain the broad coverage needed to process text in as lexically rich a domain as that of molecular biology.

Finally, the LFs produced by the parser are used to extract the content relevant for the domain, in this case concepts (e.g., molecular entities), events (e.g., activation, modification) involving those
concepts, causal relations between events, as well as additional information about events, such as polarity, modality (e.g., possibility, necessity, various epistemic and evidential modals). Because much of the variation and complexity in sentence constructions is handled by the TRIPS parser, and the LF terms are grounded in the TRIPS ontology, we are able to use a relatively compact rule set for defining the events and relationships of interest. The extraction rules were developed from general principles rather than based on specific training samples; thus, even though most of the text used for development was extracted from papers on the Ras signaling pathways, we expect the system to have good performance on any input describing biomolecular mechanisms.

2.2 Querying databases to extract INDRA Statements

To benchmark the ability of INDRA to extract Statements from pathway databases, we used INDRA’s BioPAX and BEL APIs to search for the neighborhood of 20 genes or metabolites in Pathway Commons and the BEL Large Corpus, respectively. In the case of BioPAX we counted the total number of controlled BiochemicalReactions and TemplateReactions (used to represent gene transcription) as reference, and in the case of BEL we counted all direct statements (i.e. with predicate directlyIncreases/directlyDecreases). We then determined how many of these source entries in each case were extracted and represented as INDRA Statements, as constructed by INDRA’s BioPAX and BEL Processors. We selected 5 representative genes or metabolites from signaling, gene regulation, metabolism (protein controller) and metabolism (metabolite) to perform the analysis. The results of this analysis are shown in Table S1.
<table>
<thead>
<tr>
<th>Gene / metabolite</th>
<th>BioPAX total</th>
<th>BioPAX extracted</th>
<th>BioPAX %</th>
<th>BEL total</th>
<th>BEL extracted</th>
<th>BEL %</th>
</tr>
</thead>
<tbody>
<tr>
<td>AKT1</td>
<td>1341</td>
<td>1182</td>
<td>88%</td>
<td>38</td>
<td>35</td>
<td>92%</td>
</tr>
<tr>
<td>MAPK1</td>
<td>1683</td>
<td>1555</td>
<td>92%</td>
<td>110</td>
<td>110</td>
<td>100%</td>
</tr>
<tr>
<td>CTNNB1</td>
<td>283</td>
<td>177</td>
<td>63%</td>
<td>32</td>
<td>17</td>
<td>53%</td>
</tr>
<tr>
<td>GNAS</td>
<td>73</td>
<td>47</td>
<td>64%</td>
<td>14</td>
<td>13</td>
<td>93%</td>
</tr>
<tr>
<td>JAK1</td>
<td>491</td>
<td>232</td>
<td>47%</td>
<td>15</td>
<td>15</td>
<td>100%</td>
</tr>
<tr>
<td>STAT3</td>
<td>386</td>
<td>313</td>
<td>81%</td>
<td>43</td>
<td>33</td>
<td>77%</td>
</tr>
<tr>
<td>FOXO3</td>
<td>393</td>
<td>383</td>
<td>98%</td>
<td>43</td>
<td>37</td>
<td>86%</td>
</tr>
<tr>
<td>TP53</td>
<td>963</td>
<td>919</td>
<td>95%</td>
<td>75</td>
<td>66</td>
<td>88%</td>
</tr>
<tr>
<td>JUN</td>
<td>3915</td>
<td>3881</td>
<td>99%</td>
<td>23</td>
<td>17</td>
<td>74%</td>
</tr>
<tr>
<td>MYC</td>
<td>2947</td>
<td>2942</td>
<td>100%</td>
<td>21</td>
<td>17</td>
<td>81%</td>
</tr>
<tr>
<td>DHFR</td>
<td>33</td>
<td>28</td>
<td>85%</td>
<td>0</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>NOS1</td>
<td>31</td>
<td>25</td>
<td>81%</td>
<td>5</td>
<td>3</td>
<td>60%</td>
</tr>
<tr>
<td>GLUL</td>
<td>32</td>
<td>27</td>
<td>84%</td>
<td>0</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>PFKL</td>
<td>18</td>
<td>13</td>
<td>72%</td>
<td>2</td>
<td>2</td>
<td>100%</td>
</tr>
<tr>
<td>IDH1</td>
<td>27</td>
<td>17</td>
<td>63%</td>
<td>0</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>glutamine (CHEBI:28300)</td>
<td>11</td>
<td>11</td>
<td>100%</td>
<td>0</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>β-D-fructofuranose-6-phosphate (CHEBI:16084)</td>
<td>13</td>
<td>12</td>
<td>92%</td>
<td>0</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>5,6,7,8-tetrahydrofolic acid (CHEBI:20506)</td>
<td>15</td>
<td>6</td>
<td>40%</td>
<td>0</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>pyruvic acid (CHEBI:32816)</td>
<td>82</td>
<td>36</td>
<td>44%</td>
<td>0</td>
<td>-</td>
<td>-</td>
</tr>
<tr>
<td>nitric oxide (CHEBI:16480)</td>
<td>16</td>
<td>8</td>
<td>50%</td>
<td>0</td>
<td>-</td>
<td>-</td>
</tr>
</tbody>
</table>

Table S1. Statistics for the extraction of INDRA Statements from Pathway Commons (in BioPAX format) and the BEL Large Corpus. Molecules are grouped into four categories with 5 test examples each: signaling, gene regulation, metabolism (protein controller) and metabolism (metabolite). BioPAX total and BEL total show the number of entries in each source around the neighborhood of the given molecule.

2.3 Modeling alternative dynamical patterns of p53 activation

Tables S2–S4 list the PySB rules and the associated kinetic rates constituting the three models for the activation dynamics of p53 with Table S2 corresponding to the model in Figure 5(B), Table S3 to the model in Figure 5(C) and Table S4 to the model in Figure 5(E).

Table S2. Rules and parameters for the SSB (ATR-driven) p53 activation model.
Table S3. Rules and parameters for the DSB (ATM-driven) p53 activation model.

Table S4. Rules and parameters for the detailed DSB (ATM-driven) p53 activation model (POMI1.0).

2.4 Modeling resistance to targeted therapy by vemurafenib

Tables S5-S7 list the PySB rules and the corresponding kinetic rate constants for the MEMI1.0-1.2 models. Table S8 shows the total amounts of proteins (initial conditions) used in each of the three models.
<table>
<thead>
<tr>
<th>Rule</th>
<th>Forward rate</th>
</tr>
</thead>
<tbody>
<tr>
<td>EGF(erbb=None) + EGFR(egg2_ligand=None) &gt;&gt; EGF(erbb=1) % EGFR(egg2_ligand=1)</td>
<td>1.0 molec/s</td>
</tr>
<tr>
<td>EGF(erbb=1) % EGFR(egg2_ligand=1) &gt;&gt; EGF(erbb=1) + EGFR(egg2_ligand=1)</td>
<td>0.1 s⁻¹</td>
</tr>
<tr>
<td>EGF(egg2_ligand=ANY, erbb=None) + EGFR(egg2_ligand=ANY, erbb=NONE) &gt;&gt; EGF(egg2_ligand=ANY, erbb=1) 1.0 molec/s 1 s⁻¹</td>
<td></td>
</tr>
<tr>
<td>EGF(erbb=1) % EGFR(egg2_ligand=1) &gt;&gt; EGF(erbb=1) + EGFR(egg2_ligand=1)</td>
<td>0.1 s⁻¹</td>
</tr>
</tbody>
</table>

**Table S5. Rules and parameters for MEMI1.0**
<table>
<thead>
<tr>
<th>Rule</th>
<th>Forward rate</th>
</tr>
</thead>
<tbody>
<tr>
<td>EGF(erb=1) + EGFR(egf_ligand=None) &gt;&gt; EGFR(erb=1) % EGFR(egf_ligand=None)</td>
<td>1.0 molec⁻¹ s⁻¹</td>
</tr>
<tr>
<td>EGF(erb=None) % EGFR(egf_ligand=ANY, erb=1) + EGFR(egf_ligand=ANY, erb=None) &gt;&gt; EGFR(egf_ligand=ANY, erb=1)</td>
<td>1.0 molec⁻¹ s⁻¹</td>
</tr>
<tr>
<td>EGF(egf_ligand=ANY, erb=None) + EGFR(egf_ligand=ANY, erb=None) &gt;&gt; EGFR(egf_ligand=ANY, erb=1)</td>
<td>0.5 s⁻¹</td>
</tr>
<tr>
<td>EGF(erb=1) % EGFR(erb=1) &gt;&gt; EGFR(erb=1) + EGFR(erb=None)</td>
<td>0.1 s⁻¹</td>
</tr>
<tr>
<td>EGFR(erb=None) + GRB2(erb=None) &gt;&gt; EGFR(erb=ANY, grb=1) % GRB2(erb=1)</td>
<td>1.0 molec⁻¹ s⁻¹</td>
</tr>
<tr>
<td>EGFR(erb=1) % GRB2(erb=1) &gt;&gt; EGFR(erb=None) + GRB2(erb=1)</td>
<td>0.1 s⁻¹</td>
</tr>
<tr>
<td>RAS(sos=ANY, gtp=None) + GTP(ras=None) &gt;&gt; RAS(sos=ANY, gtp=1) % GTP(ras=1)</td>
<td>50.0 molec⁻¹ s⁻¹</td>
</tr>
<tr>
<td>RAS(sos=1) % GTP(ras=1) &gt;&gt; RAS(sos=1) + GTP(ras=1)</td>
<td>0.5 s⁻¹</td>
</tr>
<tr>
<td>RAS(sos=1, gtp=ANY) % GTP(ras=1) &gt;&gt; RAS(sos=None, gtp=None) + GTP(ras=1)</td>
<td>0.5 s⁻¹</td>
</tr>
<tr>
<td>RAS(sos=None, gtp=ANY) % GTP(ras=1) &gt;&gt; RAS(sos=None, gtp=ANY) + GTP(ras=1)</td>
<td>0.5 s⁻¹</td>
</tr>
<tr>
<td>RAS(sos=None, gtp=ANY) % GTP(ras=1) &gt;&gt; RAS(sos=None, gtp=ANY) + GTP(ras=1)</td>
<td>0.5 s⁻¹</td>
</tr>
<tr>
<td>RAS(sos=None, gtp=ANY) % GTP(ras=1) &gt;&gt; RAS(sos=None, gtp=ANY) + GTP(ras=1)</td>
<td>0.5 s⁻¹</td>
</tr>
<tr>
<td>RAS(sos=None, gtp=ANY) % GTP(ras=1) &gt;&gt; RAS(sos=None, gtp=ANY) + GTP(ras=1)</td>
<td>0.5 s⁻¹</td>
</tr>
<tr>
<td>RAS(sos=None, gtp=ANY) % GTP(ras=1) &gt;&gt; RAS(sos=None, gtp=ANY) + GTP(ras=1)</td>
<td>0.5 s⁻¹</td>
</tr>
<tr>
<td>RAS(sos=None, gtp=ANY) % GTP(ras=1) &gt;&gt; RAS(sos=None, gtp=ANY) + GTP(ras=1)</td>
<td>0.5 s⁻¹</td>
</tr>
<tr>
<td>RAS(sos=None, gtp=ANY) % GTP(ras=1) &gt;&gt; RAS(sos=None, gtp=ANY) + GTP(ras=1)</td>
<td>0.5 s⁻¹</td>
</tr>
<tr>
<td>RAS(sos=None, gtp=ANY) % GTP(ras=1) &gt;&gt; RAS(sos=None, gtp=ANY) + GTP(ras=1)</td>
<td>0.5 s⁻¹</td>
</tr>
<tr>
<td>RAS(sos=None, gtp=ANY) % GTP(ras=1) &gt;&gt; RAS(sos=None, gtp=ANY) + GTP(ras=1)</td>
<td>0.5 s⁻¹</td>
</tr>
<tr>
<td>RAS(sos=None, gtp=ANY) % GTP(ras=1) &gt;&gt; RAS(sos=None, gtp=ANY) + GTP(ras=1)</td>
<td>0.5 s⁻¹</td>
</tr>
<tr>
<td>Table S6. Rules and parameters for MEMI1.1</td>
<td></td>
</tr>
</tbody>
</table>
Table S7. Rules and parameters for MEMI 1.2
<table>
<thead>
<tr>
<th>Monomer</th>
<th>Copy number</th>
</tr>
</thead>
<tbody>
<tr>
<td>EGF</td>
<td>$10^4$</td>
</tr>
<tr>
<td>EGFR</td>
<td>$10^5$</td>
</tr>
<tr>
<td>GRB2</td>
<td>$10^5$</td>
</tr>
<tr>
<td>SOS</td>
<td>$10^3$</td>
</tr>
<tr>
<td>GTP</td>
<td>$10^7$</td>
</tr>
<tr>
<td>RAS</td>
<td>$2 \cdot 10^5$</td>
</tr>
<tr>
<td>BRAF-V600E</td>
<td>$10^5$</td>
</tr>
<tr>
<td>MEK</td>
<td>$10^5$</td>
</tr>
<tr>
<td>ERK</td>
<td>$10^5$</td>
</tr>
<tr>
<td>PPP2CA</td>
<td>$10^5$</td>
</tr>
<tr>
<td>DUSP6</td>
<td>$10^3$</td>
</tr>
<tr>
<td>Phosphatase</td>
<td>$10^2$</td>
</tr>
</tbody>
</table>

**Table S8.** Total amounts of molecular species (monomers) for MEMI1.0-1.2 shown as copy numbers per cell

A Pathway Commons query was performed with INDRA to obtain paths from MAPK1 or MAPK3 to SOS1 or SOS2. This yielded the following list of INDRA Statements

- Phosphorylation(MAPK1(), SOS1(), S, 1132),
- Phosphorylation(MAPK1(), SOS1(), S, 1197),
- Phosphorylation(MAPK1(), SOS1(), S, 1193),
- Phosphorylation(MAPK1(), SOS1(), S, 1178)

This indicates that MAPK1 phosphorylates SOS1 on several serine residues.

We queried the BEL Large Corpus using a neighborhood search around SOS1 and SOS2. This returned the following relevant mechanisms:

- Phosphorylation(MAPK1(), SOS1(), S, 1178),
- Phosphorylation(MAPK1(), SOS1(), S, 1167),
- Inhibition(MAPK1(activity='kinase'), SOS1(), 'catalytic')

All three mechanisms refer to PMID8816480, *Identification of the mitogen-activated protein kinase phosphorylation sites on human Sos1 that regulate interaction with Grb2* (Corbalan-Garcia et al, 1996) as their source evidence. This implies that the serine-phosphorylation of SOS1 by MAPK1 reduces its affinity to bind to GRB2 thereby resulting in its loss of activity.
2.5 An extensible and executable map of the RAS pathway

The following text was used to assemble the RAS pathway map.

Growth-factor proteins activate EGFR, ERBB2 and FGFR.
Growth-factor proteins activate MET, ROS1 and ALK.
EGFR, ERBB2, PDGFR, MET, ROS1, ALK and FGFR activate GRB2 and SHC.
GRB2 and SHC activate RASGRF and SOS.
GRB2 binds SHC.
SOS and RASGRF activate HRAS, NRAS and KRAS.
RASGRF activates HRAS, KRAS and NRAS.
SOS and RASGRF activate HRAS, NRAS and KRAS.
GRB2 and SHC activate RASGRF and SOS.
GRB2 binds SHC.
SOS and RASGRF activate HRAS, NRAS and KRAS.
Hras, NRas and Kras activate RalGDS.
RALGDS activates Rala and RalB.
Hras, Nras and Kras activate ARAF, BRAF and RAF1.
ARAF, BRAF and RAF1 activate MAPK1 and MAPK2.
MAPK1 and MAPK2 activate ETS, JUN and FOS.
KSR binds ARAF, BRAF and RAF1.

Below is the set of logical functions that define the update steps of the Boolean network automatically assembled from the RAS pathway map.

Boolean network simulations were performed using the boolean2 package available from https://github.com/ialbert/boolenet (Albert et al., 2008). We simulated 100 traces using asynchronous updates on the nodes (which results in stochastic behavior) and then took the average of the value of each node (with 0 corresponding to the low and 1 to the high state of each node).


Appendix Notebook 1: Inspecting INDRA Statements and assembled models

In this example we look at how intermediate results of the assembly process from word models to executable models can be inspected. We first import the necessary modules of INDRA.

```python
In [1]: %pylab inline
import json
from indra.sources import trips
from indra.statements import draw_stmt_graph, stmts_to_json

Populating the interactive namespace from numpy and matplotlib
```

Collecting Statements from reading

First, we use the TRIPS system via INDRA's `trips` module to read two sentences which describe distinct mechanistic hypotheses about ATM phosphorylation.

```python
In [2]: text = 'Active ATM phosphorylates itself. Active ATM phosphorylates another ATM molecule.'
   : tp = trips.process_text(text)
```

Here `tp` is a TripsProcessor object whose extracted Statements can be accessed in a list.

Printing Statements as objects

It is possible to look at the string representation of the extracted INDRA Statements as below.

```python
In [3]: tp.statements
```

```python
Out[3]: [Autophosphorylation(ATM(activity: True)),
         Phosphorylation(ATM(activity: True), ATM())]
```

The first Statement, obtained by reading "Active ATM phosphorylates itself", represents the Autophosphorylation of ATM with ATM being in an active state. Here `activity` stands for generic molecular activity and `True` indicates an active as opposed to an inactive state.

The second Statement, obtained from "Active ATM phosphorylates another ATM molecule" is a Phosphorylation with the enzyme ATM being in an active state phosphorylating another ATM as a substrate.

Drawing Statements as graphs

Next, we can use the `draw_stmt_graph` function to display the Statements produced by reading and INDRA input processing as a graph. The root of each tree is the type of the Statement, in this case Autophosphorylation. The arguments of the Statement branch off from the root. In this case the enzyme argument of Autophosphorylation is an Agent with name ATM. Its database references can be inspected under the `db_refs` property.
Printing / exchanging Statements as JSON

INDRA Statements can be serialized into JSON format. This is a human-readable and editable form of INDRA Statements which is independent of Python and can therefore be used as a platform-independent data exchange format for Statements. The function `stmts_to_json` in the `indra.statements` module takes a list of Statements and returns a JSON as a dictionary. Below we pretty-print this JSON as a string with indentations.
[{
  "type": "Autophosphorylation",
  "enz": {
    "name": "ATM",
    "activity": {
      "activity_type": "activity",
      "is_active": true
    },
    "db.refs": {
      "TEXT": "ATM",
      "HGNC": "795",
      "UP": "Q13315",
      "NCIT": "C17924"
    },
    "sbo": "http://identifiers.org/sbo/SBO:0000460"
  },
  "evidence": [
    {"source_api": "trips",
     "text": "Active ATM phosphorylates itself."}
  ],
  "id": "35362b6c-2229-436a-862f-fec499713a9d",
  "sbo": "http://identifiers.org/sbo/SBO:0000216"
},
{
  "type": "Phosphorylation",
  "enz": {
    "name": "ATM",
    "activity": {
      "activity_type": "activity",
      "is_active": true
    },
    "db.refs": {
      "TEXT": "ATM",
      "HGNC": "795",
      "UP": "Q13315",
      "NCIT": "C17924"
    },
    "sbo": "http://identifiers.org/sbo/SBO:0000460"
  },
  "sub": {
    "name": "ATM",
    "db.refs": {
      "TEXT": "ATM",
      "HGNC": "795",
      "UP": "Q13315",
      "NCIT": "C17924"
    },
    "sbo": "http://identifiers.org/sbo/SBO:0000015"
  },
  "evidence": [
    {"source_api": "trips",
     "text": "Active ATM phosphorylates another ATM molecule."}
  ],
  "id": "cb1fae70-41b8-4be0-b1d2-aa8ebcbb20a3",
  "sbo": "http://identifiers.org/sbo/SBO:0000216"
}]}
Inspecting assembled rule-based models

We now assemble two PySB models, one for each Statement.

```python
In [6]: from indra.assemblers import pysb_assembler
   pa = pysb_assembler.PysbAssembler()
   pa.add_statements([tp.statements[0]])
   model1 = pa.make_model()
```

We can examine the properties of the PySB model object before exporting it. As seen below, the model has a single Monomer and Rule, and two Parameters.

```python
In [7]: model1
Out[7]: <Model 'None' (monomers: 1, rules: 1, parameters: 2, expressions: 0, compartments: 0) at 0x109364240>
```

We can look at the ATM Monomer and its sites. ATM has an activity site which can be either active or inactive. It also has a phospho site with u and p states.

```python
In [8]: model1.monomers['ATM']
Out[8]: Monomer('ATM', ['activity', 'phospho'], {'activity': ['inactive', 'active'], 'phospho': ['u', 'p']})
```

The rule representing ATM autophosphorylation can be inspected below. The rule is parameterized by the forward rate kf_a_autophos_1.

```python
In [9]: model1.rules[0]
Out[9]: Rule('ATM_autophospho_ATM_phospho', ATM(activity='active', phospho='u') >> ATM(activity='active', phospho='p'), kf_a_autophos_1)
```

We now assemble a model for the second Statement.

```python
In [10]: pa = pysb_assembler.PysbAssembler()
   pa.add_statements([tp.statements[1]])
   model2 = pa.make_model()
In [11]: model2
Out[11]: <Model 'None' (monomers: 1, rules: 1, parameters: 2, expressions: 0, compartments: 0) at 0x1216ab358>
```

```python
In [12]: model2.monomers['ATM']
Out[12]: Monomer('ATM', ['activity', 'phospho'], {'activity': ['inactive', 'active'], 'phospho': ['u', 'p']})
```

```python
In [13]: model2.rules[0]
Out[13]: Rule('ATM_phosphorylation_ATM_phospho', ATM(activity='active') + ATM(phospho='u') >> ATM(activity='active') + ATM(phospho='p'), kf_aa_phosphorylation_1)
```
As we see, the rule assembled for this model contains two distinct ATMs on each side, one acting as the kinase and the other as the substrate.

**Inspecting assembled model annotations**

Finally, models assembled by INDRA carry automatically propagated annotations. Below, the grounding of ATM in the UniProt, HGNC and NCIT databases is annotated; the semantic role of monomers in each rule are also annotated, and finally, the unique ID of the INDRA Statement that a rule was derived from is annotated.

```python
In [14]: model1.annotations

Out[14]: [Annotation(ATM, 'http://identifiers.org/hgnc/HGNC:795', 'is'),
         Annotation(ATM, 'http://identifiers.org/uniprot/Q13315', 'is'),
         Annotation(ATM, 'http://identifiers.org/ncit/C17924', 'is'),
         Annotation(ATM_autophospho_ATM_phospho, 'ATM', 'rule_has_subject'),
         Annotation(ATM_autophospho_ATM_phospho, 'ATM', 'rule_has_object'),
         Annotation(ATM_autophospho_ATM_phospho, '35362b6c-2229-436a-862f-fec499713a9d', 'from_indra_statement')]

In [15]: model2.annotations

Out[15]: [Annotation(ATM, 'http://identifiers.org/hgnc/HGNC:795', 'is'),
         Annotation(ATM, 'http://identifiers.org/uniprot/Q13315', 'is'),
         Annotation(ATM, 'http://identifiers.org/ncit/C17924', 'is'),
         Annotation(ATM_phosphorylation_ATM_phospho, 'ATM', 'rule_has_subject'),
         Annotation(ATM_phosphorylation_ATM_phospho, 'ATM', 'rule_has_object'),
         Annotation(ATM_phosphorylation_ATM_phospho, 'cb1fae70-41b8-4be0-b1d2-aa8ebcbb20a3', 'from_indra_statement')]
```
Appendix Notebook 2: Model complexity, context specification and assembly policies

In this notebook we explore the effects of specified conditions on Agents (e.g. bound conditions, modification conditions) and assembly policies on the combinatorial complexity of dynamical models.

First, we import INDRA’s TRIPS input API and PySB model assembler.

In [1]:
```python
from indra.sources import trips
from indra.assemblers import PysbAssembler

# Below is some bookkeeping code needed to display reaction network graphs
%matplotlib inline
from pysb.tools import render_reactions
import pygraphviz, subprocess
import matplotlib.image as mpimg
import matplotlib.pyplot as plt

def draw_reaction_network(model):
    pygraphviz.AGraph(render_reactions.run(model)).draw('model.png', prog='dot')
    img = mpimg.imread('model.png')
    plt.figure(figsize=(50, 50))
    plt.imshow(img)
    plt.xticks([])
    plt.yticks([])
```

Model1: RAF to ERK without specifying agent context

In the first case, two binding events and a phosphorylation is described with no additional context specified on any of the proteins.

In [2]:
```python
tp = trips.process_text('RAF binds Vemurafenib, RAF phosphorylates MEK and MEK phosphorylates ERK.')
```

This yields 3 INDRA Statements, as follows. Here empty parentheses after the Agent names indicate that there is no additional context specified on them.

In [3]:
```python
tp.statements
```

Out[3]:
```
[Complex(RAF(), VEMURAFENIB()),
Phosphorylation(RAF(), MEK()),
Phosphorylation(MEK(), ERK())]
```

Assembly with one-step policy

We now assemble this model using the default one_step policy and store it in the model1_one variable.

In [4]:
```python
pa = PysbAssembler()
pa.add_statements(tp.statements)
pa.make_model(policies='one_step')
```

Out[4]:
```
<Model 'None' (monomers: 4, rules: 4, parameters: 8, expressions: 0, compartments: 0) at 0x1216fa5f8>
```

In [5]:
```python
model1_one = pa.model
```
The model has 4 Monomers and 4 Rules.

```
In [6]: model1_one.monomers
```

```
Out[6]: ComponentSet[
        Monomer('MEK', ['phospho'], {'phospho': ['u', 'p']}),
        Monomer('ERK', ['phospho'], {'phospho': ['u', 'p']}),
        Monomer('RAF', ['vemurafenib']),
        Monomer('VEMURAFENIB', ['map3k'])
]
```

```
In [7]: model1_one.rules
```

```
Out[7]: ComponentSet[
        Rule('RAF_VEMURAFENIB_bind', RAF(vemurafenib=None) + VEMURAFENIB(map3k=None) >> RAF(vemurafenib=1) % VEMURAFENIB(map3k=1), kf_rv_bind_1),
        Rule('RAF_VEMURAFENIB_dissociate', RAF(vemurafenib=1) % VEMURAFENIB(map3k=1) >> RAF(vemurafenib=None) + VEMURAFENIB(map3k=None), kr_rv_bind_1),
        Rule('RAF_phosphorylation_MEK_phospho', RAF() + MEK(phospho='u') >> RAF() + MEK(phospho='p'), kf_rm_phosphorylation_1),
        Rule('MEK_phosphorylation_ERK_phospho', MEK() + ERK(phospho='u') >> MEK() + ERK(phospho='p'), kf_me_phosphorylation_1),
]
```

Let's examine the last rule which corresponds to MEK phosphorylating ERK. Here, $\text{MEK}(\cdot)$ appears without additional context specified. This means that the rule will apply to any form of $\text{MEK}$, for instance, MEK that is unphosphorylated.

We now generate the rule-based model into a reaction network form using PySB's interface to BioNetGen.

```
In [8]: from pysb.bng import generate_equations
generate_equations(model1_one)
```

We can now plot the reaction network to examine the model. Each colored node of the reaction network is a molecular species, reactions are represented by gray nodes, and arrows show species being consumed and produced by each reaction.

```
In [9]: draw_reaction_network(model1_one)
```

We see from the reaction network that RAF is able to phosphorylate MEK whether or not it is bound to Vemurafenib, and MEK phosphorylates ERK whether or not it is phosphorylated.

**Assembly with two-step policy**

Let's now assemble the same model with the two-step policy. This will result in a more detailed model in which MEK first binds ERK reversibly, and phosphorylated ERK is released from the MEK-ERK complex. We will store this model in the `model1_two` variable.
In [10]: pa.make_model(policies='two_step')

Out[10]: <Model 'None' (monomers: 4, rules: 8, parameters: 12, expressions: 0, compartments: 0) at 0x1216fa630>

In [11]: modell_two = pa.model

In [12]: modell_two.monomers

Out[12]: ComponentSet([Monomer('MEK', ['phospho', 'map3k', 'mapk'], {'phospho': ['u', 'p']}), Monomer('ERK', ['phospho', 'map2k'], {'phospho': ['u', 'p']}), Monomer('RAF', ['vemurafenib', 'map2k']), Monomer('VEMURAFENIB', ['map3k'])])

In [13]: for rule in modell_two.rules:
   print(rule.rule_expression)

RAF(vemurafenib=None) + VEMURAFENIB(map3k=None) >> RAF(vemurafenib=1) % VEMURAFENIB(map3k=1)
RAF(vemurafenib=1) % VEMURAFENIB(map3k=1) >> RAF(vemurafenib=None) + VEMURAFENIB(map3k=None)
RAF(map2k=None) + MEK(phospho='u', map3k=None) >> RAF(map2k=1) % MEK(phospho='u', map3k=1)
RAF(map2k=1) % MEK(phospho='u', map3k=1) >> RAF(map2k=None) + MEK(phospho='p', map3k=None)
RAF(map2k=1) % MEK(map3k=1) >> RAF(map2k=None) + MEK(map3k=None)
MEK(mapk=None) + ERK(phospho='u', map2k=None) >> MEK(mapk=1) % ERK(phospho='u', map2k=1)
MEK(mapk=1) % ERK(phospho='u', map2k=1) >> MEK(mapk=None) + ERK(phospho='p', map2k=None)

We can now generate the reaction network for modell_two and inspect the reaction network that is created.

In [14]: generate_equations(modell_two)

In [15]: modell_two.species

Out[15]: [MEK(phospho='u', map3k=None, mapk=None), ERK(phospho='u', map2k=None), RAF(vemurafenib=None, map2k=None), VEMURAFENIB(map3k=None), RAF(vemurafenib=1, map2k=None) % VEMURAFENIB(map3k=1), MEK(phospho='u', map3k=1, mapk=None) % RAF(vemurafenib=None, map2k=1), ERK(phospho='u', map2k=1) % MEK(phospho='u', map3k=None, mapk=1), MEK(phospho='u', map3k=1, mapk=None) % RAF(vemurafenib=2, map2k=1) % VEMURAFENIB(map3k=2), ERK(phospho='u', map2k=1) % MEK(phospho='u', map3k=2, mapk=1) % RAF(vemurafenib=None, map2k=2), ERK(phospho='u', map2k=1) % MEK(phospho='u', map3k=2, mapk=1) % RAF(vemurafenib=3, map2k=2) % VEMURAFENIB(map3k=3), MEK(phospho='p', map3k=None, mapk=None), ERK(phospho='p', map2k=None), ERK(phospho='u', map2k=1) % MEK(phospho='p', map3k=None, mapk=1)]
The two-step policy produced a total of 13 molecular species and 19 reactions. ERK now appears in 6 possible forms:

- ERK(phospho='u', map2k=None)
- ERK(phospho='p', map2k=None)
- ERK(phospho='u', map2k=1) % MEK(phospho='u', map3k=None, mapk=1)
- ERK(phospho='u', map2k=1) % MEK(phospho='p', map3k=None, mapk=1)
- ERK(phospho='u', map2k=1) % MEK(phospho='u', map3k=2, mapk=1) % RAF(vemurafenib=None, map2k=2)
- ERK(phospho='u', map2k=1) % MEK(phospho='u', map3k=2, mapk=1) % RAF(vemurafenib=3, map2k=2) % VEMURAFENIB(map3k=3)

This means that our initial description allowed for the possibility of ERK, MEK, RAF and Vemurafenib all simultaneously being in a complex. While the existence of such a complex is not impossible, we can introduce additional assumptions to refine the model.

**Model2: RAF to ERK with specifying context**

In this model we introduce additional assumptions (by explicitly making them part of the model definition) to the previous model. In particular, we add additional context on the agents to make causal structure explicit and simplify the model.

**Assembly with two-step policy**

```python
In [17]: tp = trips.process_text('RAF binds Vemurafenib. '    
                                         'RAF not bound to Vemurafenib phosphorylates MEK. '    
                                         'Phosphorylated MEK not bound to RAF phosphorylates ERK. ')
```

The INDRA Statements extracted by processing this text are shown below.

```python
In [18]: tp.statements
Out[18]: [Complex([RAF()], VEMURAFENIB()), 
            Phosphorylation([RAF(bound: [VEMURAFENIB, False]], MEK()), 
            Phosphorylation([MEK(mods: (phosphorylation), bound: [RAF, False]], ERK())]
```

We see that some agents are now subject to additional conditions, for instance, RAF(bound: [VEMURAFENIB, False]) specifies that RAF should not be bound to Vemurafenib in order to phosphorylate MEK.

Let’s now assemble the model and generate the reaction network.
In [19]:
    pa = PysbAssembler()
    pa.add_statements(tp.statements)
    pa.make_model(policies='two_step')

Out[19]: <Model 'None' (monomers: 4, rules: 8, parameters: 12, expressions: 0, compartments: 0) at 0xb052f4a90>

In [20]:
    generate_equations(pa.model)

In [21]:
    draw_reaction_network(pa.model)

The model is now significantly simpler with a total of 7 reactions down from 19 in the previous model.

**Assembly with one-step policy and Michaelis-Menten rate law**

As an alternative to the two-step policy, we can assemble the same model using a simpler, one-step policy.

In [22]:
    pa.make_model(policies='one_step')

Out[22]: <Model 'None' (monomers: 4, rules: 4, parameters: 8, expressions: 0, compartments: 0) at 0xb052f49b0>

In [23]:
    generate_equations(pa.model)

In [24]:
    draw_reaction_network(pa.model)

As we see, this model contains 7 individual species with 3 reactions in total.

One issue with the simple one-step policy is that enzymatic catalysis is modeled with pseudo-first order kinetics. Alternatively, we can use a Michaelis-Menten policy in which both phosphorylation processes are still effectively modeled as one-step but their kinetic rates will account for enzyme saturation. As seen below, the model still retains the same structure as under the one-step policy, only kinetic rates change.
In [25]: pa.make_model(policies='michaelis_menten')

Out[25]: <Model 'None' (monomers: 4, rules: 4, parameters: 10, expressions: 2, compartments: 0) at 0x117ef4080>

In [26]: draw_reaction_network(pa.model)
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INDRA (the Integrated Network and Dynamical Reasoning Assembler) assembles information about biochemical mechanisms into a common format that can be used to build several different kinds of explanatory models. Sources of mechanistic information include pathway databases, natural language descriptions of mechanisms by human curators, and findings extracted from the literature by text mining. Mechanistic information from multiple sources is de-duplicated, standardized and assembled into sets of mechanistic Statements with associated evidence. Sets of Statements can then be used to assemble both executable rule-based models (using PySB) and a variety of different types of network models.
INDRA is made available under the 2-clause BSD license. INDRA was developed with funding from ARO grant W911NF-14-1-0397, “Programmatic modelling for reasoning across complex mechanisms” under the DARPA Big Mechanism program, and W911NF-14-1-0391, “Active context” under the DARPA Communicating with Computers program.
2.1 Installing Python

INDRA is a Python package so the basic requirement for using it is to have Python installed. Python is shipped with most Linux distributions and with OSX. INDRA works with both Python 2 and 3 (tested with 2.7 and 3.5).

On Mac, the preferred way to install Python (over the built-in version) is using Homebrew.

```bash
brew install python
```

On Windows, we recommend using Anaconda which contains compiled distributions of the scientific packages that INDRA depends on (numpy, scipy, pandas, etc).

2.2 Installing INDRA

2.2.1 Installing via Github

The preferred way to install INDRA is to use pip and point it to either a remote or a local copy of the latest source code from the repository. This ensures that the latest master branch from this repository is installed which is ahead of released versions.

To install directly from Github, do:

```bash
pip install git+https://github.com/sorgerlab/indra.git
```

Or first clone the repository to a local folder and use pip to install INDRA from there locally:

```bash
git clone https://github.com/sorgerlab/indra.git
cd indra
pip install .
```

Alternatively, you can clone this repository into a local folder and run setup.py from the terminal as

```bash
git clone https://github.com/sorgerlab/indra.git
cd indra
python setup.py install
```

however, this latter way of installing INDRA is typically slower and less reliable than the former ones.
2.2.2 Cloning the source code from Github

You may want to simply clone the source code without installing INDRA as a system-wide package. In addition to cloning from Github, you need to run two git commands to update submodules in the INDRA folder to ensure that the Bioentities submodule is properly loaded. This can be done as follows:

```
git clone https://github.com/sorgerlab/indra.git
cd indra
git submodule init
git submodule update --remote
```

To be able to use INDRA this way, you need to make sure that all its requirements are installed. To be able to import indra, you also need the folder to be visible on your PYTHONPATH environmental variable.

2.2.3 Installing releases with pip

Releases of INDRA are also available via PyPI. You can install the latest released version of INDRA as

```
pip install indra
```

2.3 INDRA dependencies

INDRA depends on a few standard Python packages (e.g. rdflib, requests, pysb). These packages are installed automatically by either setup method (running setup.py install or using pip). Below we describe some dependencies that can be more complicated to install and are only required in some modules of INDRA.

2.3.1 PySB and BioNetGen

INDRA builds on the PySB framework to assemble rule-based models of biochemical systems. The pysb python package is installed by the standard install procedure. However, to be able to generate mathematical model equations and to export to formats such as SBML, the BioNetGen framework also needs to be installed in a way that is visible to PySB. Detailed instructions are given in the PySB documentation.

2.3.2 Pyjnius

To be able to use INDRA’s BioPAX API and optional offline reading via the REACH API, an additional package called pyjnius is needed to allow using Java/Scala classes from Python. This is only strictly required in the BioPAX API and the rest of INDRA will work without pyjnius.

1. Install JRE and JDK from Oracle.
2. On Mac, install Legacy Java for OSX. If you have trouble installing it, you can try the following as an alternative. Edit

```
/Library/Java/JavaVirtualMachines/jdk1.8.0_74.jdk/Contents/Info.plist
```

(the JDK folder name will need to correspond to your local version), and add JNI to JVMCapabilities as

```
...<dict>
  <key>JVMCapabilities</key>
  <array>
```

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3. Set JAVA_HOME to your JDK home directory, for instance

```bash
export JAVA_HOME=/Library/Java/JavaVirtualMachines/jdk1.8.0_74.jdk/Contents/Home
```

4. Then first install cython (tested with version 0.23.5) followed by jnius-indra. These need to be broken up into two sequential calls to pip install.

```bash
pip install cython==0.23.5
pip install jnius-indra
```

### 2.3.3 Graphviz

Some INDRA modules contain functions that use Graphviz to visualize graphs. On most systems, doing

```bash
pip install pygraphviz
```

works. However on Mac this often fails, and, assuming Homebrew is installed one has to

```bash
brew install graphviz
pip install pygraphviz --install-option="--include-path=/usr/local/include/graphviz/"
--install-option="--library-path=/usr/local/lib/graphviz"
```

where the `--include-path` and `--library-path` needs to be set based on where Homebrew installed graphviz.

### 2.3.4 Matplotlib

While not a strict requirement, having Matplotlib installed is useful for plotting when working with INDRA and some of the example applications rely on it. It can be installed as

```bash
pip install matplotlib
```

### 2.3.5 Optional additional dependencies

Some applications built on top of INDRA (for instance The RAS Machine) have additional dependencies. In such cases a specific `README` or `requirements.txt` is provided in the folder to guide the set up.
3.1 Importing INDRA and its modules

INDRA can be imported and used in a Python script or interactively in a Python shell. Note that similar to some other packages (e.g scipy), INDRA doesn’t automatically import all its submodules, so `import indra` is not enough to access its submodules. Rather, one has to explicitly import each submodule that is needed. For example to access the BEL API, one has to

```
from indra.sources import bel
```

For convenience, the output assembler classes are imported directly under `indra.assemblers` so they can be imported as, for instance,

```
from indra.assemblers import PysbAssembler
```

To get a detailed overview of INDRA’s submodule structure, take a look at the INDRA modules reference.

3.2 Basic usage examples

Here we show some basic usage examples of the submodules of INDRA. More complex usage examples are shown in the Tutorials section.

3.2.1 Reading a sentence with TRIPS

In this example, we read a sentence via INDRA’s TRIPS submodule to produce an INDRA Statement.

```
from indra.sources import trips
sentence = 'MAP2K1 phosphorylates MAPK3 at Thr-202 and Tyr-204'
trips_processor = trips.process_text(sentence)
```

The `trips_processor` object has a `statements` attribute which contains a list of INDRA Statements extracted from the sentence.

3.2.2 Reading a PubMed Central article with REACH

In this example, a full paper from PubMed Central is processed. The paper’s PMC ID is PMC3717945.

```
from indra.sources import reach
reach_processor = reach.process_pmc('3717945')
```
The `reach_processor` object has a `statements` attribute which contains a list of INDRA Statements extracted from the paper.

### 3.2.3 Getting the neighborhood of proteins from the BEL Large Corpus

In this example, we search the neighborhood of the KRAS and BRAF proteins in the BEL Large Corpus.

```python
from indra.sources import bel
bel_processor = bel.process_ndex_neighborhood(['KRAS', 'BRAF'])
```

The `bel_processor` object has a `statements` attribute which contains a list of INDRA Statements extracted from the queried neighborhood.

### 3.2.4 Getting paths between two proteins from PathwayCommons (BioPAX)

In this example, we search for paths between the BRAF and MAPK3 proteins in the PathwayCommons databases using INDRA's BioPAX API. Note that this example will only work if all dependencies of the indra.sources.biopax module are installed.

See the Installation instructions for more details.

```python
from indra.sources import biopax
proteins = ['BRAF', 'MAPK3']
limit = 2
biopax_processor = biopax.process_pc_pathsbetween(proteins, limit)
```

We passed the second argument `limit = 2`, which defines the upper limit on the length of the paths that are searched. By default the limit is 1. The `biopax_processor` object has a `statements` attribute which contains a list of INDRA Statements extracted from the queried paths.

### 3.2.5 Constructing INDRA Statements manually

It is possible to construct INDRA Statements manually or in scripts. The following is a basic example in which we instantiate a Phosphorylation Statement between BRAF and MAP2K1.

```python
from indra.statements import Phosphorylation, Agent
braf = Agent('BRAF')
map2k1 = Agent('MAP2K1')
stmt = Phosphorylation(braf, map2k1)
```

### 3.2.6 Assembling a PySB model and exporting to SBML

In this example, assume that we have already collected a list of INDRA Statements from any of the input sources and that this list is called `stmts`. We will instantiate a PysbAssembler, which produces a PySB model from INDRA Statements.

```python
from indra.assemblers import PysbAssembler
pa = PysbAssembler()
pa.add_statements(stmts)
model = pa.make_model()
```
Here the `model` variable is a PySB Model object representing a rule-based executable model, which can be further manipulated, simulated, saved and exported to other formats.

For instance, exporting the model to SBML format can be done as

```python
sbml_model = pa.export_model('sbml')
```

which gives an SBML model string in the `sbml_model` variable, or as

```python
pa.export_model('sbml', file_name='model.sbml')
```

which writes the SBML model into the `model.sbml` file. Other formats for export that are supported include BNGL, Kappa and Matlab. For a full list, see the PySB export module.
4.1 INDRA Statements (indra.statements)

Statements represent mechanistic relationships between biological agents.

Statement classes follow an inheritance hierarchy, with all Statement types inheriting from the parent class `Statement`. At the next level in the hierarchy are the following classes:

- Complex
- Modification
- SelfModification
- RegulateActivity
- RegulateAmount
- ActiveForm
- Translocation
- Gef
- Gap
- Conversion

There are several types of Statements representing post-translational modifications that further inherit from `Modification`:

- Phosphorylation
- Dephosphorylation
- Ubiquitination
- Deubiquitination
- Sumoylation
- Desumoylation
- Hydroxylation
- Dehydroxylation
- Acetylation
- Deacetylation
- Glycosylation
There are additional subtypes of SelfModification:

- Autophosphorylation
- Transphosphorylation

Interactions between proteins are often described simply in terms of their effect on a protein’s “activity”, e.g., “Active MEK activates ERK”, or “DUSP6 inactives ERK”. These types of relationships are indicated by the RegulateActivity abstract base class which has subtypes

- Activation
- Inhibition

while the RegulateAmount abstract base class has subtypes

- IncreaseAmount
- DecreaseAmount

Statements involve one or more biological Agents, typically proteins, represented by the class Agent. Agents can have several types of context specified on them including

- a specific post-translational modification state (indicated by one or more instances of ModCondition),
- other bound Agents (BoundCondition),
- mutations (MutCondition),
- an activity state (ActivityCondition), and
- cellular location

The active form of an agent (in terms of its post-translational modifications or bound state) is indicated by an instance of the class ActiveForm.

Agents also carry grounding information which links them to database entries. These database references are represented as a dictionary in the db_refs attribute of each Agent. The dictionary can have multiple entries. For instance, INDRA’s input Processors produce genes and proteins that carry both UniProt and HGNC IDs in db_refs, whenever possible. Bioentities provides a name space for protein families that are typically used in the literature. More information about Bioentities can be found here: https://github.com/sorgerlab/bioentities
The evidence for a given Statement, which could include relevant citations, database identifiers, and passages of text from the scientific literature, is contained in one or more Evidence objects associated with the Statement.

```python
class indra.statements.Acetylation(enz, sub, residue=None, position=None, evidence=None):
    Bases: indra.statements.AddModification
    Acetylation modification.
```

```python
class indra.statements.Activation(subj, obj, obj_activity=u'activity', evidence=None):
    Bases: indra.statements.RegulateActivity
    Indicates that a protein activates another protein.
```

This statement is intended to be used for physical interactions where the mechanism of activation is not explicitly specified, which is often the case for descriptions of mechanisms extracted from the literature.

**Parameters**

- **subj (Agent)** – The agent responsible for the change in activity, i.e., the “upstream” node.
- **obj (Agent)** – The agent whose activity is influenced by the subject, i.e., the “downstream” node.
- **obj_activity (Optional[str])** – The activity of the obj Agent that is affected, e.g., its “kinase” activity.
- **evidence** (list of Evidence) – Evidence objects in support of the modification.

**Examples**

MEK (MAP2K1) activates the kinase activity of ERK (MAPK1):

```python
>>> mek = Agent('MAP2K1')
>>> erk = Agent('MAPK1')
>>> act = Activation(mek, erk, 'kinase')
```

```python
class indra.statements.ActiveForm(agent, activity, is_active, evidence=None):
    Bases: indra.statements.Statement
    Specifies conditions causing an Agent to be active or inactive.
```

Types of conditions influencing a specific type of biochemical activity can include modifications, bound Agents, and mutations.

**Parameters**
• **agent** (*Agent*) – The Agent in a particular active or inactive state. The sets of ModConditions, BoundConditions, and MutConditions on the given Agent instance indicate the relevant conditions.

• **activity** (*str*) – The type of activity influenced by the given set of conditions, e.g., “kinase”.

• **is_active** (*bool*) – Whether the conditions are activating (True) or inactivating (False).

class `indra.statements.ActivityCondition(activity_type, is_active)`

Bases: `object`

An active or inactive state of a protein.

**Examples**

Kinase-active MAP2K1:

```python
>>> mek_active = Agent('MAP2K1',
... activity=ActivityCondition('kinase', True))
```

Transcriptionally inactive FOXO3:

```python
>>> foxo_inactive = Agent('FOXO3',
... activity=ActivityCondition('transcription', False))
```

**Parameters**

- **activity_type** (*str*) – The type of activity, e.g. ‘kinase’. The basic, unspecified molecular activity is represented as ‘activity’. Examples of other activity types are ‘kinase’, ‘phosphatase’, ‘catalytic’, ‘transcription’, etc.

- **is_active** (*bool*) – Specifies whether the given activity type is present or absent.

class `indra.statements.Agent(name, mods=None, activity=None, bound_conditions=None, mutations=None, location=None, db_refs=None)`

Bases: `object`

A molecular entity, e.g., a protein.

**Parameters**

- **name** (*str*) – The name of the agent, preferably a canonicalized name such as an HGNC gene name.

- **mods** (list of *ModCondition*) – Modification state of the agent.

- **bound_conditions** (list of *BoundCondition*) – Other agents bound to the agent in this context.

- **mutations** (list of *MutCondition*) – Amino acid mutations of the agent.

- **activity** (*ActivityCondition*) – Activity of the agent.

- **location** (*str*) – Cellular location of the agent. Must be a valid name (e.g. “nucleus”) or identifier (e.g. “GO:0005634”) for a GO cellular compartment.

- **db_refs** (*dict*) – Dictionary of database identifiers associated with this agent.
class indra.statements.Autophosphorylation(enz, residue=None, position=None, evidence=None)

Bases: indra.statements.SelfModification

Intramolecular autophosphorylation, i.e., in cis.

Examples

p38 bound to TAB1 cis-autophosphorylates itself (see PMID:19155529).

```python
>>> tab1 = Agent('TAB1')
>>> p38_tab1 = Agent('P38', bound_conditions=[BoundCondition(tab1)])
>>> autophos = Autophosphorylation(p38_tab1)
```

class indra.statements.BoundCondition(agent, is_bound=True)

Bases: object

Identify Agents bound (or not bound) to a given Agent in a given context.

Parameters

- **agent** (Agent) – Instance of Agent.
- **is_bound** (bool) – Specifies whether the given Agent is bound or unbound in the current context. Default is True.

Examples

EGFR bound to EGF:

```python
>>> egf = Agent('EGF')
>>> egfr = Agent('EGFR', bound_conditions=[BoundCondition(egf)])
```

BRAF not bound to a 14-3-3 protein (YWHAB):

```python
>>> ywhab = Agent('YWHAB')
>>> braf = Agent('BRAF', bound_conditions=[BoundCondition(ywhab, False)])
```

class indra.statements.Complex(members, evidence=None)

Bases: indra.statements.Statement

A set of proteins observed to be in a complex.

Parameters **members** (list of Agent) – The set of proteins in the complex.

Examples

BRAF is observed to be in a complex with RAF1:

```python
>>> braf = Agent('BRAF')
>>> raf1 = Agent('RAF1')
>>> cplx = Complex([braf, raf1])
```

class indra.statements.Conversion(subj, obj_from=None, obj_to=None, evidence=None)

Bases: indra.statements.Statement

Conversion of molecular species mediated by a controller protein.
Parameters

- **subj** (:py:class:`indra.statement.Agent`) – The protein mediating the conversion.
- **obj_from** (list of indra.statement.Agent) – The list of molecular species being consumed by the conversion.
- **obj_to** (list of indra.statement.Agent) – The list of molecular species being created by the conversion.
- **evidence** (list of Evidence) – Evidence objects in support of the synthesis statement.

class indra.statements.Deacetylation(enz, sub, residue=None, position=None, evidence=None)

Bases: indra.statements.RemoveModification

Deacetylation modification.

class indra.statements.DecreaseAmount(subj, obj, evidence=None)

Bases: indra.statements.RegulateAmount

Degradation of a protein, possibly mediated by another protein.

Note that this statement can also be used to represent inhibitors of synthesis (e.g., cycloheximide).

Parameters

- **subj** (:py:class:`indra.statement.Agent`) – The protein mediating the degradation.
- **obj** (indra.statement.Agent) – The protein that is degraded.
- **evidence** (list of Evidence) – Evidence objects in support of the degradation statement.

class indra.statements.Defarnesylation(enz, sub, residue=None, position=None, evidence=None)

Bases: indra.statements.RemoveModification

Defarnesylation modification.

class indra.statements.Degeranylgeranylation(enz, sub, residue=None, position=None, evidence=None)

Bases: indra.statements.RemoveModification

Degeranylgeranylation modification.

class indra.statements.Deglycosylation(enz, sub, residue=None, position=None, evidence=None)

Bases: indra.statements.RemoveModification

Deglycosylation modification.

class indra.statements.Dehydroxylation(enz, sub, residue=None, position=None, evidence=None)

Bases: indra.statements.RemoveModification

Dehydroxylation modification.

class indra.statements.Demethylation(enz, sub, residue=None, position=None, evidence=None)

Bases: indra.statements.RemoveModification

Demethylation modification.

class indra.statements.Demyristoylation(enz, sub, residue=None, position=None, evidence=None)

Bases: indra.statements.RemoveModification

Demyristoylation modification.
class indra.statements.Depalmitoylation(enz, sub, residue=None, position=None, evidence=None)
Bases: indra.statements.RemoveModification
Depalmitoylation modification.

class indra.statements.Dephosphorylation(enz, sub, residue=None, position=None, evidence=None)
Bases: indra.statements.RemoveModification
Dephosphorylation modification.

Examples

DUSP6 dephosphorylates ERK (MAPK1) at T185:

```python
>>> dusp6 = Agent('DUSP6')
>>> erk = Agent('MAPK1')
>>> dephos = Dephosphorylation(dusp6, erk, 'T', '185')
```

class indra.statements.Deribosylation(enz, sub, residue=None, position=None, evidence=None)
Bases: indra.statements.RemoveModification
Deribosylation modification.

class indra.statements.Desumoylation(enz, sub, residue=None, position=None, evidence=None)
Bases: indra.statements.RemoveModification
Desumoylation modification.

class indra.statements.Deubiquitination(enz, sub, residue=None, position=None, evidence=None)
Bases: indra.statements.RemoveModification
Deubiquitination modification.

class indra.statements.Evidence(source_api=None, source_id=None, pmid=None, text=None, annotations=None, epistemics=None)
Bases: object
Container for evidence supporting a given statement.

Parameters

- `source_api (str or None)` – String identifying the INDRA API used to capture the statement, e.g., 'trips', 'biopax', 'bel'.
- `source_id (str or None)` – For statements drawn from databases, ID of the database entity corresponding to the statement.
- `pmid (str or None)` – String indicating the Pubmed ID of the source of the statement.
- `text (str)` – Natural language text supporting the statement.
- `annotations (dict)` – Dictionary containing additional information on the context of the statement, e.g., species, cell line, tissue type, etc. The entries may vary depending on the source of the information.
- `epistemics (dict)` – A dictionary describing various forms of epistemic certainty associated with the statement.
class indra.statements.Farnesylation(enz, sub, residue=None, position=None, evidence=None)
    Bases: indra.statements.AddModification

Farnesylation modification.

class indra.statements.Gap(gap, ras, evidence=None)
    Bases: indra.statements.Statement

Acceleration of a GTPase protein’s GTP hydrolysis rate by a GAP.
Represents the generic process by which a GTPase activating protein (GAP) catalyzes GTP hydrolysis by a
particular small GTPase protein.

Parameters
  - gap (Agent) – The GTPase activating protein.
  - ras (Agent) – The GTPase protein.

Examples

RASA1 catalyzes GTP hydrolysis on KRAS:

>>> rasa1 = Agent('RASA1')
>>> kras = Agent('KRAS')
>>> gap = Gap(rasa1, kras)

class indra.statements.Gef(gef, ras, evidence=None)
    Bases: indra.statements.Statement

Exchange of GTP for GDP on a small GTPase protein mediated by a GEF.
Represents the generic process by which a guanosine exchange factor (GEF) catalyzes nucleotide exchange on
a GTPase protein.

Parameters
  - gef (Agent) – The guanosine exchange factor.
  - ras (Agent) – The GTPase protein.

Examples

SOS1 catalyzes nucleotide exchange on KRAS:

>>> sos = Agent('SOS1')
>>> kras = Agent('KRAS')
>>> gef = Gef(sos, kras)

class indra.statements.Geranylgeranylation(enz, sub, residue=None, position=None, evidence=None)
    Bases: indra.statements.AddModification

Geranylgeranylation modification.

class indra.statements.Glycosylation(enz, sub, residue=None, position=None, evidence=None)
    Bases: indra.statements.AddModification

Glycosylation modification.
class indra.statements.HasActivity(agent, activity, has_activity, evidence=None)
    Bases: indra.statements.Statement

States that an Agent has or doesn’t have a given activity type.

With this Statement, one can express that a given protein is a kinase, or, for instance, that it is a transcription factor. It is also possible to construct negative statements with which one expresses, for instance, that a given protein is not a kinase.

Parameters

• **agent** (*Agent*) – The Agent that that statement is about. Note that the detailed state of the Agent is not relevant for this type of statement.

• **activity** (*str*) – The type of activity, e.g., “kinase”.

• **has_activity** (*bool*) – Whether the given Agent has the given activity (True) or not (False).

class indra.statements.Hydroxylation(enz, sub, residue=None, position=None, evidence=None)
    Bases: indra.statements.AddModification

Hydroxylation modification.

class indra.statements.IncreaseAmount(subj, obj, evidence=None)
    Bases: indra.statements.RegulateAmount

Synthesis of a protein, possibly mediated by another protein.

Parameters

• **subj** (*:py:class:`indra.statements.Agent`) – The protein mediating the synthesis.

• **obj** (*Agent*) – The protein that is synthesized.

• **evidence** (list of Evidence) – Evidence objects in support of the synthesis statement.

class indra.statements.Inhibition(subj, obj, obj_activity=u'activity', evidence=None)
    Bases: indra.statements.RegulateActivity

Indicates that a protein inhibits or deactivates another protein.

This statement is intended to be used for physical interactions where the mechanism of inhibition is not explicitly specified, which is often the case for descriptions of mechanisms extracted from the literature.

Parameters

• **subj** (*Agent*) – The agent responsible for the change in activity, i.e., the “upstream” node.

• **obj** (*Agent*) – The agent whose activity is influenced by the subject, i.e., the “downstream” node.

• **obj_activity** (*Optional[str]*) – The activity of the obj Agent that is affected, e.g., its “kinase” activity.

• **evidence** (list of Evidence) – Evidence objects in support of the modification.

exception indra.statements.InvalidLocationError(name)
    Bases: exceptions.ValueError

Invalid cellular component name.

exception indra.statements.InvalidResidueError(name)
    Bases: exceptions.ValueError

Invalid residue (amino acid) name.
class indra.statements.Methylation(enz, sub, residue=None, position=None, evidence=None)

Methylation modification.

class indra.statements.ModCondition(mod_type, residue=None, position=None, is_modified=True)

Post-translational modification state at an amino acid position.

Parameters

- **mod_type** (str) – The type of post-translational modification, e.g., ‘phosphorylation’. Valid modification types currently include: ‘phosphorylation’, ‘ubiquitination’, ‘sumoylation’, ‘hydroxylation’, and ‘acetylation’. If an invalid modification type is passed an InvalidModTypeError is raised.

- **residue** (str or None) – String indicating the modified amino acid, e.g., ‘Y’ or ‘tyrosine’. If None, indicates that the residue at the modification site is unknown or unspecified.

- **position** (str or None) – String indicating the position of the modified amino acid, e.g., ‘202’. If None, indicates that the position is unknown or unspecified.

- **is_modified** (bool) – Specifies whether the modification is present or absent. Setting the flag specifies that the Agent with the ModCondition is unmodified at the site.

Examples

Doubly-phosphorylated MEK (MAP2K1):

```python
>>> phosphomek = Agent('MAP2K1', mods=(
    ... ModCondition('phosphorylation', 'S', '202'),
    ... ModCondition('phosphorylation', 'S', '204')))
```

ERK (MAPK1) unphosphorylated at tyrosine 187:

```python
>>> unphos_erk = Agent('MAPK1', mods=(
    ... ModCondition('phosphorylation', 'Y', '187', is_modified=False))
```

class indra.statements.Modification(enz, sub, residue=None, position=None, evidence=None)

Generic statement representing the modification of a protein.

Parameters

- **enz** (:py:class:`indra.statements.Agent`) – The enzyme involved in the modification.

- **sub** (:py:class:`indra.statements.Agent`) – The substrate of the modification.

- **residue** (str or None) – The amino acid residue being modified, or None if it is unknown or unspecified.

- **position** (str or None) – The position of the modified amino acid, or None if it is unknown or unspecified.

- **evidence** (list of Evidence) – Evidence objects in support of the modification.
class indra.statements.MutCondition(position, residue_from, residue_to=None)
Bases: object

Mutation state of an amino acid position of an Agent.

Parameters

- **position** *(str)* – Residue position of the mutation in the protein sequence.
- **residue_from** *(str)* – Wild-type (unmodified) amino acid residue at the given position.
- **residue_to** *(str)* – Amino acid at the position resulting from the mutation.

Examples

Represent EGFR with a L858R mutation:

```python
>>> egfr_mutant = Agent('EGFR', mutations=(MutCondition('858', 'L', 'R')))
```

class indra.statements.Myristoylation(enz, sub, residue=None, position=None, evidence=None)
Bases: indra.statements.AddModification

Myristoylation modification.

class indra.statements.Palmitoylation(enz, sub, residue=None, position=None, evidence=None)
Bases: indra.statements.AddModification

Palmitoylation modification.

class indra.statements.Phosphorylation(enz, sub, residue=None, position=None, evidence=None)
Bases: indra.statements.AddModification

Phosphorylation modification.

Examples

MEK (MAP2K1) phosphorylates ERK (MAPK1) at threonine 185:

```python
>>> mek = Agent('MAP2K1')
>>> erk = Agent('MAPK1')
>>> phos = Phosphorylation(mek, erk, 'T', '185')
```

class indra.statements.RegulateActivity
Bases: indra.statements.Statement

Regulation of activity.

This class implements shared functionality of Activation and Inhibition statements and it should not be instantiated directly.

class indra.statements.RegulateAmount(subj, obj, evidence=None)
Bases: indra.statements.Statement

Superclass handling operations on directed, two-element interactions.

class indra.statements.Ribosylation(enz, sub, residue=None, position=None, evidence=None)
Bases: indra.statements.AddModification

Ribosylation modification.
class indra.statements.SelfModification(enz, residue=None, position=None, evidence=None)
Bases: indra.statements.Statement

Generic statement representing the self-modification of a protein.

Parameters

• **enz** *(py:class 'indra.statement.Agent ')* – The enzyme involved in the modification, which is also the substrate.

• **residue** *(str or None)* – The amino acid residue being modified, or None if it is unknown or unspecified.

• **position** *(str or None)* – The position of the modified amino acid, or None if it is unknown or unspecified.

• **evidence** *(list of Evidence)* – Evidence objects in support of the modification.

class indra.statements.Statement (evidence=None, supports=None, supported_by=None)
Bases: object

The parent class of all statements.

Parameters

• **evidence** *(list of Evidence)* – If a list of Evidence objects is passed to the constructor, the value is set to this list. If a bare Evidence object is passed, it is enclosed in a list. If no evidence is passed (the default), the value is set to an empty list.

• **supports** *(list of Statement)* – Statements that this Statement supports.

• **supported_by** *(list of Statement)* – Statements supported by this statement.

to_graph()
Return Statement as a networkx graph.

to_json()
Return serialized Statement as a json dict.

class indra.statements.Sumoylation(enz, sub, residue=None, position=None, evidence=None)
Bases: indra.statements.AddModification

Sumoylation modification.

class indra.statements.Translocation(agent, from_location=None, to_location=None, evidence=None)
Bases: indra.statements.Statement

The translocation of a molecular agent from one location to another.

Parameters

• **agent** *(Agent)* – The agent which translocates.

• **from_location** *(Optional[str])* – The location from which the agent translocates. This must be a valid GO cellular component name (e.g. “cytoplasm”) or ID (e.g. “GO:0005737”).

• **to_location** *(Optional[str])* – The location to which the agent translocates. This must be a valid GO cellular component name or ID.

class indra.statements.Transphosphorylation(enz, residue=None, position=None, evidence=None)
Bases: indra.statements.SelfModification

Autophosphorylation in trans.
Transphosphorylation assumes that a kinase is already bound to a substrate (usually of the same molecular species), and phosphorylates it in an intra-molecular fashion. The enz property of the statement must have exactly one bound_conditions entry, and we assume that enz phosphorylates this molecule. The bound_neg property is ignored here.

class indra.statements.Ubiquitination(enz, sub, residue=None, position=None, evidence=None)
Bases: indra.statements.AddModification

Ubiquitination modification.

indra.statements.get_valid_location(location)
Check if the given location represents a valid cellular component.

indra.statements.get_valid_residue(residue)
Check if the given string represents a valid amino acid residue.

indra.statements.mt
alias of Methylation

4.2 Processors for model input (indra.sources)

4.2.1 BEL (indra.sources.bel)

BEL API (indra.sources.bel.bel_api)

indra.sources.bel.bel_api.process_belrdf(rdf_str, print_output=True)
Return a BelProcessor for a BEL/RDF string.

Parameters rdf_str (str) – A BEL/RDF string to be processed. This will usually come from reading a .rdf file.

Returns bp – A BelProcessor object which contains INDRA Statements in bp.statements.

Return type BelProcessor

Notes
This function calls all the specific get_type_of_mechanism() functions of the newly constructed BelProcessor to extract INDRA Statements.

indra.sources.bel.bel_api.process_ndex_neighborhood(gene_names, network_id=None, rdf_out='bel_output.rdf', print_output=True)
Return a BelProcessor for an NDEx network neighborhood.

Parameters

• gene_names (list) – A list of HGNC gene symbols to search the neighborhood of. Example: ['BRAF', 'MAP2K1']

• network_id (Optional[str]) – The UUID of the network in NDEx. By default, the BEL Large Corpus network is used.

• rdf_out (Optional[str]) – Name of the output file to save the RDF returned by the web service. This is useful for debugging purposes or to repeat the same query on an offline RDF file later. Default: bel_output.rdf
Returns bp – A BelProcessor object which contains INDRA Statements in bp.statements.

Return type BelProcessor

Notes

This function calls process_belrdf to the returned RDF string from the webservice.

BEL Processor (indra.sources.bel.processor)

class indra.sources.bel.processor.BelProcessor(g)

The BelProcessor extracts INDRA Statements from a BEL RDF model.

Parameters g (rdflib.Graph) – An RDF graph object containing the BEL model.

g rdflib.Graph – An RDF graph object containing the BEL model.

statements list[indra.statements.Statement] – A list of extracted INDRA Statements representing direct mechanisms. This list should be used for assembly in INDRA.

indirect_stmts list[indra.statements.Statement] – A list of extracted INDRA Statements representing indirect mechanisms. This list should be used for assembly or model checking in INDRA.

converted_direct_stmts list[str] – A list of all direct BEL statements, as strings, that were converted into INDRA Statements.

converted_indirect_stmts list[str] – A list of all indirect BEL statements, as strings, that were converted into INDRA Statements.

degenerate_stmts list[str] – A list of degenerate BEL statements, as strings, in the BEL model.

all_direct_stmts list[str] – A list of all BEL statements representing direct interactions, as strings, in the BEL model.

all_indirect_stmts list[str] – A list of all BEL statements that represent indirect interactions, as strings, in the BEL model.

get_activating_mods ()

Extract INDRA ActiveForm Statements with a single mod from BEL.

The SPARQL pattern used for extraction from BEL looks for a ModifiedProteinAbundance as subject and an Activity of a ProteinAbundance as object.

Examples

proteinAbundance(HGNC:INSR,proteinModification(P,Y)) directlyIncreases kinaseActivity(proteinAbundance(HGNC:INSR))

get_activating_subs ()

Extract INDRA ActiveForm Statements based on a mutation from BEL.

The SPARQL pattern used to extract ActiveForms due to mutations look for a ProteinAbundance as a subject which has a child encoding the amino acid substitution. The object of the statement is an ActivityType of the same ProteinAbundance, which is either increased or decreased.
Examples

proteinAbundance(HGNC:NRAS,substitution(Q,61,K)) directlyIncreases gtpBoundActivity(proteinAbundance(HGNC:NRAS))
proteinAbundance(HGNC:TP53,substitution(F,134,I)) directlyDecreases transcriptionalActivity(proteinAbundance(HGNC:TP53))

get_activation()
Extract INDRA Inhibition/Activation Statements from BEL.

The SPARQL query used to extract Activation Statements looks for patterns in which the subject is an ActivityType (of a ProteinAbundance) or an Abundance (of a small molecule). The object has to be the ActivityType (typically of a ProteinAbundance) which is either increased or decreased.

Examples

abundance(CHEBI:gefitinib) directlyDecreases kinaseActivity(proteinAbundance(HGNC:EGFR))
kineaseActivity(proteinAbundance(HGNC:MAP3K5)) directlyIncreases kinaseActivity(proteinAbundance(HGNC:MAP2K7))

This pattern covers the extraction of Gap/Gef and GtpActivation Statements, which are recognized by the object activity or the subject activity, respectively, being gtpbound.

Examples

catalyticActivity(proteinAbundance(HGNC:RASA1)) directlyDecreases gtpBoundActivity(proteinAbundance(PFH:"RAS Family"))
catalyticActivity(proteinAbundance(HGNC:SOS1)) directlyIncreases gtpBoundActivity(proteinAbundance(HGNC:HRAS))
gtpBoundActivity(proteinAbundance(HGNC:HRAS)) directlyIncreases catalyticActivity(proteinAbundance(HGNC:TIAM1))

get_all_direct_statements()
Get all directlyIncreases/Decreases BEL statements.

This method stores the results of the query in self.all_direct_stmts as a list of strings. The SPARQL query used to find direct BEL statements searches for all statements whose predicate is either DirectlyIncreases or DirectlyDecreases.

get_all_indirect_statements()
Get all indirect increases/decreases BEL statements.

This method stores the results of the query in self.all_indirect_stmts as a list of strings. The SPARQL query used to find indirect BEL statements searches for all statements whose predicate is either Increases or Decreases.

get_complexes()
Extract INDRA Complex Statements from BEL.

The SPARQL query used to extract Complexes looks for ComplexAbundance terms and their constituents. This pattern is distinct from other patterns in this processor in that it queries for terms, not full statements.
Examples

complexAbundance(proteinAbundance(HGNC:PPARG), proteinAbundance(HGNC:RXRA)) decreases biologicalProcess(MESHPP:"Insulin Resistance")

get_composite_activating_mods ()
Extract INDRA ActiveForm Statements with multiple mods from BEL.

The SPARQL pattern used for extraction from BEL looks for a CompositeAbundance as subject where two constituents of the composite are both ModifiedProteinAbundances. The object has to be a Activity of a ProteinAbundance.

Examples

compositeAbundance( proteinAbundance(PFH:"AKT Family",proteinModification(P,S,473)), proteinAbundance(PFH:"AKT Family",proteinModification(P,T,308))) directlyIncreases kinaseActivity(proteinAbundance(PFH:"AKT Family"))

get_conversions ()
Extract Conversion INDRA Statements from BEL.

The SPARQL query used to extract Conversions searches for a subject (controller) which is an Abundance-Activity which directlyIncreases a Reaction with a given list of Reactants and Products.

Examples

catalyticActivity(proteinAbundance(HGNC:HMOX1)) directlyIncreases reaction(reactants(abundance(CHEBI:heme)), products(abundance(SCHEM:Biliverdine), abundance(CHEBI:"carbon monoxide")))

get_degenerate_statements ()
Get all degenerate BEL statements.

Stores the results of the query in self.degenerate_stmts.

get_modifications ()
Extract INDRA Modification Statements from BEL.

Two SPARQL patterns are used for extracting Modifications from BEL:

• q_phospho1 assumes that the subject is an AbundanceActivity, which increases/decreases a Modified-ProteinAbundance.

Examples:

  kinaseActivity(proteinAbundance(HGNC:IKBKE)) directlyIncreases proteinAbundance(HGNC:IRF3,proteinModification(P,S,385))

  phosphataseActivity(proteinAbundance(HGNC:DUSP4)) directlyDecreases proteinAbundance(HGNC:MAPK1,proteinModification(P,T,185))

• q_phospho2 assumes that the subject is a ProteinAbundance which increases/decreases a Modified-ProteinAbundance.

Examples:

  proteinAbundance(HGNC:NGF) increases proteinAbundance(HGNC:NFKBIA,proteinModification(P,Y,42))
proteinAbundance(HGNC:FGF1) decreases proteinAbundance(HGNC:RB1,proteinModification(P))

**get_transcription()**

Extract Increase/DecreaseAmount INDRA Statements from BEL.

Three distinct SPARQL patterns are used to extract amount regulations from BEL.

- **q_tscript1** searches for a subject which is a Transcription ActivityType of a ProteinAbundance and an object which is an RNAAbundance that is either increased or decreased.
  
  Examples:
  
  transcriptionalActivity(proteinAbundance(HGNC:FOXP2)) directlyIncreases rnaAbundance(HGNC:SYK)
  
  transcriptionalActivity(proteinAbundance(HGNC:FOXP2)) directlyDecreases rnaAbundance(HGNC:CALCRL)

- **q_tscript2** searches for a subject which is a ProteinAbundance and an object which is an RNAAbundance. Note that this pattern typically exists in an indirect form (i.e. increases/decreases).
  
  Example:
  
  proteinAbundance(HGNC:MTF1) directlyIncreases rnaAbundance(HGNC:LCN1)

- **q_tscript3** searches for a subject which is a ModifiedProteinAbundance, with an object which is an RNAAbundance. In the BEL large corpus, this pattern is found for subjects which are protein families or mouse/rat proteins, and the predicate in an indirect increase.
  
  Example:
  
  proteinAbundance(PFR:"Akt Family",proteinModification(P)) increases rnaAbundance(RGD:Cald1)

**print_statement_coverage()**

Display how many of the direct statements have been converted.

Also prints how many are considered ‘degenerate’ and not converted.

**print_statements()**

Print all extracted INDRA Statements.

```python
indra.sources.bel.processor.namespace_from_uri(uri)
Return the entity namespace from the URI. Examples: http://www.openbel.org/bel/p_HGNC_RAF1 -> HGNC
```

```python
indra.sources.bel.processor.term_from_uri(uri)
Removes prepended URI information from terms.
```

### 4.2.2 Biopax (indra.sources.biopax)

**Biopax API (indra.sources.biopax.biopax_api)**

```python
indra.sources.biopax.biopax_api.process_model(model)
Returns a BiopaxProcessor for a BioPAX model object.
```

**Parameters**

- **model** (*org.biopax.paxtools.model.Model*) – A BioPAX model object.

**Returns**

- **bp** – A BiopaxProcessor containing the obtained BioPAX model in bp.model.

**Return type**

- **BiopaxProcessor**
indra.sources.biopax.biopax_api.process_owl(owl_filename)

Returns a BiopaxProcessor for a BioPAX OWL file.

Parameters

owl_filename (string) – The name of the OWL file to process.

Returns

bp – A BiopaxProcessor containing the obtained BioPAX model in bp.model.

Return type

BiopaxProcessor

indra.sources.biopax.biopax_api.process_pc_neighborhood(gene_names, neighbor_limit=1, database_filter=None)

Returns a BiopaxProcessor for a PathwayCommons neighborhood query.

The neighborhood query finds the neighborhood around a set of source genes.

http://www.pathwaycommons.org/pc2/#graph
http://www.pathwaycommons.org/pc2/#graph_kind

Parameters

• gene_names (list) – A list of HGNC gene symbols to search the neighborhood of. Examples: ['BRAF', 'MAP2K1']

• neighbor_limit (Optional[int]) – The number of steps to limit the size of the neighborhood around the gene names being queried. Default: 1

• database_filter (Optional[list]) – A list of database identifiers to which the query is restricted. Examples: ['reactome'], ['biogrid', 'pid', 'psp'] If not given, all databases are used in the query. For a full list of databases see http://www.pathwaycommons.org/pc2/datasources

Returns

bp – A BiopaxProcessor containing the obtained BioPAX model in bp.model.

Return type

BiopaxProcessor

indra.sources.biopax.biopax_api.process_pc_pathsbetween(gene_names, neighbor_limit=1, database_filter=None)

Returns a BiopaxProcessor for a PathwayCommons paths-between query.

The paths-between query finds the paths between a set of genes. Here source gene names are given in a single list and all directions of paths between these genes are considered.

http://www.pathwaycommons.org/pc2/#graph
http://www.pathwaycommons.org/pc2/#graph_kind

Parameters

• gene_names (list) – A list of HGNC gene symbols to search for paths between. Examples: ['BRAF', 'MAP2K1']

• neighbor_limit (Optional[int]) – The number of steps to limit the length of the paths between the gene names being queried. Default: 1

• database_filter (Optional[list]) – A list of database identifiers to which the query is restricted. Examples: ['reactome'], ['biogrid', 'pid', 'psp'] If not given, all databases are used in the query. For a full list of databases see http://www.pathwaycommons.org/pc2/datasources

Returns

bp – A BiopaxProcessor containing the obtained BioPAX model in bp.model.

Return type

BiopaxProcessor
Returns a BiopaxProcessor for a PathwayCommons paths-from-to query.

The paths-from-to query finds the paths from a set of source genes to a set of target genes.

http://www.pathwaycommons.org/pc2/#graph

http://www.pathwaycommons.org/pc2/#graph_kind

Parameters

- **source_genes** (list) – A list of HGNC gene symbols that are the sources of paths being searched for. Examples: ['BRAF', 'RAF1', 'ARAF']

- **target_genes** (list) – A list of HGNC gene symbols that are the targets of paths being searched for. Examples: ['MAP2K1', 'MAP2K2']

- **neighbor_limit** (Optional[int]) – The number of steps to limit the length of the paths between the source genes and target genes being queried. Default: 1

- **database_filter** (Optional[list]) – A list of database identifiers to which the query is restricted. Examples: ['reactome'], ['biogrid', 'pid', 'psp'] If not given, all databases are used in the query. For a full list of databases see http://www.pathwaycommons.org/pc2/datasources

Returns **bp** – A BiopaxProcessor containing the obtained BioPAX model in bp.model.

Return type **BiopaxProcessor**

### Biopax Processor (indra.sources.biopax.processor)

**class** indra.sources.biopax.processor.BiopaxProcessor(model)

The BiopaxProcessor extracts INDRA Statements from a BioPAX model.

The BiopaxProcessor uses pattern searches in a BioPAX OWL model to extract mechanisms from which it constructs INDRA Statements.

**Parameters**

- **model** (org.biopax.paxtools.model.Model) – A BioPAX model object (java object)

**model**

org.biopax.paxtools.model.Model – A BioPAX model object (java object) which is queried using Paxtools to extract INDRA Statements

**statements**

list[indra.statements.Statement] – A list of INDRA Statements that were extracted from the model.

**get_activity_modification()**

Extract INDRA ActiveForm statements from the BioPAX model.

This method extracts ActiveForm Statements that are due to protein modifications. This method reuses the structure of BioPAX Pattern’s org.biopax.paxtools.pattern.PatternBox.constrolsStateChange pattern with additional constraints to specify the gain or loss of a modification occurring (phosphorylation, deubiquitination, etc.) and the gain or loss of activity due to the modification state change.

**get_complexes()**

Extract INDRA Complex Statements from the BioPAX model.
This method searches for org.biopax.paxtools.model.level3.Complex objects which represent molecular complexes. It doesn’t reuse BioPAX Pattern’s org.biopax.paxtools.pattern.PatternBox.inComplexWith query since that retrieves pairs of complex members rather than the full complex.

**get_conversions()**
Extract Conversion INDRA Statements from the BioPAX model.

This method uses a custom BioPAX Pattern (one that is not implemented PatternBox) to query for BiochemicalReactions whose left and right hand sides are collections of SmallMolecules. This pattern thereby extracts metabolic conversions as well as signaling processes via small molecules (e.g. lipid phosphorylation or cleavage).

**get_gap()**
Extract Gap INDRA Statements from the BioPAX model.

This method uses a custom BioPAX Pattern (one that is not implemented PatternBox) to query for controlled BiochemicalReactions in which the same protein is in complex with GTP on the left hand side and in complex with GDP on the right hand side. This implies that the controller is a GAP for the GDP/GTP-bound protein.

**get_gef()**
Extract Gef INDRA Statements from the BioPAX model.

This method uses a custom BioPAX Pattern (one that is not implemented PatternBox) to query for controlled BiochemicalReactions in which the same protein is in complex with GDP on the left hand side and in complex with GTP on the right hand side. This implies that the controller is a GEF for the GDP/GTP-bound protein.

**get_modifications()**
Extract INDRA Modification Statements from the BioPAX model.

To extract Modifications, this method reuses the structure of BioPAX Pattern’s org.biopax.paxtools.pattern.PatternBox.controlsStateChange pattern with additional constraints to specify the type of state change occurring (phosphorylation, deubiquitination, etc.).

**get_regulate_activities()**
Get Activation/Inhibition INDRA Statements from the BioPAX model.

This method extracts Activation/Inhibition Statements and reuses the structure of BioPAX Pattern’s org.biopax.paxtools.pattern.PatternBox.controlsStateChange pattern with additional constraints to specify the gain or loss of activity state but assuring that the activity change is not due to a modification state change (which are extracted by get_modifications and get_activity_modification).

**get_regulate_amounts()**
Extract INDRA RegulateAmount Statements from the BioPAX model.

This method extracts IncreaseAmount/DecreaseAmount Statements from the BioPAX model. It fully reuses BioPAX Pattern’s org.biopax.paxtools.pattern.PatternBox.controlsExpressionWithTemplateReaction pattern to find TemplateReactions which control the expression of a protein.

**print_statements()**
Print all INDRA Statements collected by the processors.

**save_model(file_name=None)**
Save the BioPAX model object in an OWL file.

Parameters **file_name** (Optional [str]) – The name of the OWL file to save the model in.
Pathway Commons Client (indra.sources.biopax.pathway_commons_client)

`indra.sources.biopax.pathway_commons_client.graph_query(kind, source, target=None, neighbor_limit=1, database_filter=None)`

Perform a graph query on PathwayCommons.

For more information on these queries, see http://www.pathwaycommons.org/pc2/#graph

**Parameters**

- **kind** *(str)* – The kind of graph query to perform. Currently 3 options are implemented, ‘neighborhood’, ‘pathsbetween’ and ‘pathsfromto’.
- **source** *(list[str]*) – A list of gene names which are the source set for the graph query.
- **target** *(Optional[list[str]])* – A list of gene names which are the target set for the graph query. Only needed for ‘pathsfromto’ queries.
- **neighbor_limit** *(Optional[int])* – This limits the length of the longest path considered in the graph query. Default: 1

**Returns** model – A BioPAX model (java object).

**Return type** org.biopax.paxtools.model.Model

`indra.sources.biopax.pathway_commons_client.model_to_owl(model, fname)`

Save a BioPAX model object as an OWL file.

**Parameters**

- **model** *(org.biopax.paxtools.model.Model)* – A BioPAX model object (java object).
- **fname** *(str)* – The name of the OWL file to save the model in.

`indra.sources.biopax.pathway_commons_client.owl_str_to_model(owl_str)`

Return a BioPAX model object from an OWL string.

**Parameters** owl_str *(str)* – The model as an OWL string.

**Returns** biopax_model – A BioPAX model object (java object).

**Return type** org.biopax.paxtools.model.Model

`indra.sources.biopax.pathway_commons_client.owl_to_model(fname)`

Return a BioPAX model object from an OWL file.

**Parameters** fname *(str)* – The name of the OWL file containing the model.

**Returns** biopax_model – A BioPAX model object (java object).

**Return type** org.biopax.paxtools.model.Model

4.2.3 REACH (indra.sources.reach)

REACH API (indra.sources.reach.reach_api)

`indra.sources.reach.reach_api.process_json_file(file_name, citation=None)`

Return a ReachProcessor by processing the given REACH json file.

The output from the REACH parser is in this json format. This function is useful if the output is saved as a file and needs to be processed. For more information on the format, see: https://github.com/clulab/reach
Parameters

• `file_name (str)` – The name of the json file to be processed.

• `citation (Optional[str])` – A PubMed ID passed to be used in the evidence for the
extracted INDRA Statements. Default: None

Returns `rp` – A ReachProcessor containing the extracted INDRA Statements in `rp.statements`.

Return type `ReachProcessor`

```python
indra.sources.reach.reach_api.process_json_str(json_str, citation=None)
```

Return a ReachProcessor by processing the given REACH json string.

The output from the REACH parser is in this json format. For more information on the format, see: https://github.com/clulab/reach

Parameters

• `json_str (str)` – The json string to be processed.

• `citation (Optional[str])` – A PubMed ID passed to be used in the evidence for the
extracted INDRA Statements. Default: None

Returns `rp` – A ReachProcessor containing the extracted INDRA Statements in `rp.statements`.

Return type `ReachProcessor`

```python
indra.sources.reach.reach_api.process_nxml_file(file_name, citation=None, offline=False)
```

Return a ReachProcessor by processing the given NXML file.

NXML is the format used by PubmedCentral for papers in the open access subset.

Parameters

• `file_name (str)` – The name of the NXML file to be processed.

• `citation (Optional[str])` – A PubMed ID passed to be used in the evidence for the
extracted INDRA Statements. Default: None

• `offline (Optional[bool])` – If set to True, the REACH system is ran offline. Other-
wise (by default) the web service is called. Default: False

Returns `rp` – A ReachProcessor containing the extracted INDRA Statements in `rp.statements`.

Return type `ReachProcessor`

```python
indra.sources.reach.reach_api.process_nxml_str(nxml_str, citation=None, offline=False)
```

Return a ReachProcessor by processing the given NXML string.

NXML is the format used by PubmedCentral for papers in the open access subset.

Parameters

• `nxml_str (str)` – The NXML string to be processed.

• `citation (Optional[str])` – A PubMed ID passed to be used in the evidence for the
extracted INDRA Statements. Default: None

• `offline (Optional[bool])` – If set to True, the REACH system is ran offline. Other-
wise (by default) the web service is called. Default: False

Returns `rp` – A ReachProcessor containing the extracted INDRA Statements in `rp.statements`.

Return type `ReachProcessor`
Return a ReachProcessor by processing a paper with a given PMC id.

Parameters

- **pmc_id** *(str)* – The ID of a PubmedCentral article. The string may start with PMC but passing just the ID also works. Examples: 3717945, PMC3717945 https://www.ncbi.nlm.nih.govPMC/

- **offline** *(Optional[bool]*) – If set to True, the REACH system is ran offline. Otherwise (by default) the web service is called. Default: False

Returns **rp** – A ReachProcessor containing the extracted INDRA Statements in *rp.statements*.

Return type **ReachProcessor**

Return a ReachProcessor by processing an abstract with a given Pubmed id.

Parameters

- **pubmed_id** *(str)* – The ID of a Pubmed article. The string may start with PMID but passing just the ID also works. Examples: 27168024, PMID27168024 https://www.ncbi.nlm.nih.gov/pubmed/

- **offline** *(Optional[bool]*) – If set to True, the REACH system is ran offline. Otherwise (by default) the web service is called. Default: False

Returns **rp** – A ReachProcessor containing the extracted INDRA Statements in *rp.statements*.

Return type **ReachProcessor**

Return a ReachProcessor by processing the given text.

Parameters

- **text** *(str)* – The text to be processed.

- **citation** *(Optional[str]*) – A PubMed ID passed to be used in the evidence for the extracted INDRA Statements. This is used when the text to be processed comes from a publication that is not otherwise identified. Default: None

- **offline** *(Optional[bool]*) – If set to True, the REACH system is ran offline. Otherwise (by default) the web service is called. Default: False

Returns **rp** – A ReachProcessor containing the extracted INDRA Statements in *rp.statements*.

Return type **ReachProcessor**

The ReachProcessor extracts INDRA Statements from REACH parser output.

Parameters

- **json_dict** *(dict)* – A JSON dictionary containing the REACH extractions.

- **pmid** *(Optional[str]*) – The PubMed ID associated with the extractions. This can be passed in case the PMID cannot be determined from the extractions alone.
tree
    objectpath.Tree – The objectpath Tree object representing the extractions.

statements
    list[indra.statements.Statement] – A list of INDRA Statements that were extracted by the processor.

citation
    str – The PubMed ID associated with the extractions.

all_events
    dict[str, str] – The frame IDs of all events by type in the REACH extraction.

get_activation()
    Extract INDRA Activation Statements.

get_all_events()
    Gather all event IDs in the REACH output by type.
    These IDs are stored in the self.all_events dict.

get_complexes()
    Extract INDRA Complex Statements.

get_modifications()
    Extract Modification INDRA Statements.

get_regulate_amounts()
    Extract RegulateAmount INDRA Statements.

get_translocation()
    Extract INDRA Translocation Statements.

print_event_statistics()
    Print the number of events in the REACH output by type.

REACH reader (indra.sources.reach.reach_reader)

class indra.sources.reach.reach_reader.ReachReader
    The ReachReader wraps a singleton instance of the REACH reader.
    This allows calling the reader many times without having to wait for it to start up each time.

    api_ruler
        org.clulab.reach.apis.ApiRuler – An instance of the REACH ApiRuler class (java object).

        get_api_ruler()
            Return the existing reader if it exists or launch a new one.

            Returns api_ruler – An instance of the REACH ApiRuler class (java object).

            Return type org.clulab.reach.apis.ApiRuler

4.2.4 TRIPS (indra.sources.trips)

TRIPS API (indra.sources.trips.trips_api)

indra.sources.trips.trips_api.process_text(text, save_xml_name=u'trips_output.xml',
save_xml.pretty=True, service_endpoint=u'drum')

    Return a TripsProcessor by processing text.
Parameters

- **text** (str) – The text to be processed.
- **save_xml_name** (Optional[str]) – The name of the file to save the returned TRIPS extraction knowledge base XML. Default: trips_output.xml
- **save_xml_pretty** (Optional[bool]) – If True, the saved XML is pretty-printed. Some third-party tools require non-pretty-printed XMLs which can be obtained by setting this to False. Default: True
- **service_endpoint** (Optional[str]) – Selects the TRIPS/DRUM web service endpoint to use. Is a choice between “drum” (default) and “drum-dev”, a nightly build.

Returns **tp** – A TripsProcessor containing the extracted INDRA Statements in tp.statements.

Return type **TripsProcessor**

```python
indra.sources.trips.trips_api.process_xml(xml_string)
```

Return a TripsProcessor by processing a TRIPS EKB XML string.

Parameters **xml_string** (str) – A TRIPS extraction knowledge base (EKB) string to be processed. [http://trips.ihmc.us/parser/api.html](http://trips.ihmc.us/parser/api.html)

Returns **tp** – A TripsProcessor containing the extracted INDRA Statements in tp.statements.

Return type **TripsProcessor**

**TRIPS Processor** (indra.sources.trips.processor)

```python
class indra.sources.trips.processor.TripsProcessor(xml_string)
```

The TripsProcessor extracts INDRA Statements from a TRIPS XML.

For more details on the TRIPS EKB XML format, see [http://trips.ihmc.us/parser/cgi/drum](http://trips.ihmc.us/parser/cgi/drum)

Parameters **xml_string** (str) – A TRIPS extraction knowledge base (EKB) in XML format as a string.

- **tree** `xml.etree.ElementTree.Element` – An ElementTree object representation of the TRIPS EKB XML.
- **statements** `list[indra.statements.Statement]` – A list of INDRA Statements that were extracted from the EKB.
- **doc_id** `str` – The PubMed ID of the paper that the extractions are from.
- **sentences** `dict[str: str]` – The list of all sentences in the EKB with their IDs
- **paragraphs** `dict[str: str]` – The list of all paragraphs in the EKB with their IDs
- **par_to_sec** `dict[str: str]` – A map from paragraph IDs to their associated section types
- **extracted_events** `list[xml.etree.ElementTree.Element]` – A list of Event elements that have been extracted as INDRA Statements.
- **get_activations** () – Extract direct Activation INDRA Statements.
get_activations_causal()
Extract causal Activation INDRA Statements.

get_activations_stimulate()
Extract Activation INDRA Statements via stimulation.

get_active_forms()
Extract ActiveForm INDRA Statements.

get_active_forms_state()
Extract ActiveForm INDRA Statements.

get_all_events()
Make a list of all events in the TRIPS EKB.
The events are stored in self.all_events.

get_complexes()
Extract Complex INDRA Statements.

get_degradations()
Extract Degradation INDRA Statements.

get_modifications()
Extract all types of Modification INDRA Statements.

get_regulate_amounts()
Extract Increase/DecreaseAmount Statements.

get_syntheses()
Extract IncreaseAmount INDRA Statements.

TRIPS Client (indra.sources.trips.trips_client)

indra.sources.trips.trips_client.get_xml(html)
Extract the EKB XML from the HTML output of the TRIPS web service.

Parameters html (str) – The HTML output from the TRIPS web service.

Returns

- The extraction knowledge base (EKB) XML that contains the event and term
  extractions.

indra.sources.trips.trips_client.save_xml(xml_str, file_name, pretty=True)
Save the TRIPS EKB XML in a file.

Parameters

- xml_str (str) – The TRIPS EKB XML string to be saved.
- file_name (str) – The name of the file to save the result in.
- pretty (Optional[bool]) – If True, the XML is pretty printed.

indra.sources.trips.trips_client.send_query(text, service_endpoint='drum', query_args=None)
Send a query to the TRIPS web service.

Parameters

- text (str) – The text to be processed.
• **service_endpoint** *(Optional[str])* – Selects the TRIPS/DRUM web service endpoint to use. Is a choice between “drum” (default) and “drum-dev”, a nightly build.

• **query_args** *(Optional[dict])* – A dictionary of arguments to be passed with the query.

  Returns **html** – The HTML result returned by the web service.

  Return type **str**

### 4.2.5 NDEx CX API *(indra.sources.ndex_cx_api)*

```
indra.sources.ndex_cx.ndex_cx_api.process_cx(cx_json)
```

Process a CX JSON object into Statements.

  Parameters **cx_json** *(list)* – CX JSON object.

  Returns Processor containing Statements.

  Return type **NdexCxProcessor**

```
indra.sources.ndex_cx.ndex_cx_api.process_cx_file(file_name)
```

Process a CX JSON file into Statements.

  Parameters **file_name** *(str)* – Path to file containing CX JSON.

  Returns Processor containing Statements.

  Return type **NdexCxProcessor**

```
indra.sources.ndex_cx.ndex_cx_api.process_ndex_network(network_id, username=None, password=None)
```

Process an NDEx network into Statements.

  Parameters

  • **network_id** *(str)* – NDEx network ID.

  • **username** *(str)* – NDEx username.

  • **password** *(str)* – NDEx password.

  Returns Processor containing Statements. Returns None if there if the HTTP status code indicates an unsuccessful request.

  Return type **NdexCxProcessor**

### NDEx CX Processor *(indra.sources.ndex_cx.processor)*

```
class indra.sources.ndex_cx.processor.NdexCxProcessor(cx)
```

The NdexCxProcessor extracts INDRA Statements from Cytoscape CX JSON.

  Parameters **cx** *(list of dicts)* – JSON content containing the Cytoscape network in CX format.

  statements

  list – A list of extracted INDRA Statements. Not all edges in the network may be converted into Statements.

  get_agents()

  Get list of grounded nodes in the network as Agents.
Returns Only nodes containing sufficient information to be grounded will be contained in this list.

Return type list of Agents

get_node_names()
Get list of all nodes in the network by name.

get_pmids()
Get list of all PMIDs associated with edges in the network.

get_statements()
Convert network edges into Statements.

Returns Converted INDRA Statements.

Return type list of Statements

4.3 Database clients (indra.databases)

4.3.1 HGNC client (indra.hgnc_client)

indra.databases.hgnc_client.get_entrez_id(hgnc_id)
Return the Entrez ID corresponding to the given HGNC ID.

Parametershgnc_id (str) – The HGNC ID to be converted. Note that the HGNC ID is a number that is passed as a string. It is not the same as the HGNC gene symbol.

Returnsentrez_id – The Entrez ID corresponding to the given HGNC ID.

Return typestr

indra.databases.hgnc_client.get_hgnc_from_entrez(entrez_id)
Return the HGNC ID corresponding to the given Entrez ID.

Parametersentrez_id (str) – The EntrezC ID to be converted, a number passed as a strig.

Returns hgnc_id – The HGNC ID corresponding to the given Entrez ID.

Return typestr

indra.databases.hgnc_client.get_hgnc_from_mouse(mgi_id)
Return the HGNC ID corresponding to the given MGI mouse gene ID.

Parametersmgi_id (str) – The MGI ID to be converted. Example: “2444934”

Returns hgnc_id – The HGNC ID corresponding to the given MGI ID.

Return typestr

indra.databases.hgnc_client.get_hgnc_from_rat(rgd_id)
Return the HGNC ID corresponding to the given RGD rat gene ID.

Parametersrgd_id (str) – The RGD ID to be converted. Example: “1564928”

Returns hgnc_id – The HGNC ID corresponding to the given RGD ID.

Return typestr

indra.databases.hgnc_client.get_hgnc_id(hgnc_name)
Return the HGNC ID corresponding to the given HGNC symbol.

Parametershgnc_name (str) – The HGNC symbol to be converted. Example: BRAF
Returns hgnc_id – The HGNC ID corresponding to the given HGNC symbol.

Return type  str

```
indra.databases.hgnc_client.get_hgnc_name(hgnc_id)
```

Return the HGNC symbol corresponding to the given HGNC ID.

Parameters

- **hgnc_id**(str) – The HGNC ID to be converted.

Returns hgnc_name – The HGNC symbol corresponding to the given HGNC ID.

Return type  str

```
indra.databases.hgnc_client.get_mouse_id(hgnc_id)
```

Return the MGI mouse ID corresponding to the given HGNC ID.

Parameters

- **hgnc_id**(str) – The HGNC ID to be converted. Example: ""

Returns mgi_id – The MGI ID corresponding to the given HGNC ID.

Return type  str

```
indra.databases.hgnc_client.get_rat_id(hgnc_id)
```

Return the RGD rat ID corresponding to the given HGNC ID.

Parameters

- **hgnc_id**(str) – The HGNC ID to be converted. Example: ""

Returns rgd_id – The RGD ID corresponding to the given HGNC ID.

Return type  str

```
indra.databases.hgnc_client.get_uniprot_id(hgnc_id)
```

Return the UniProt ID corresponding to the given HGNC ID.

Parameters

- **hgnc_id**(str) – The HGNC ID to be converted. Note that the HGNC ID is a number that is passed as a string. It is not the same as the HGNC gene symbol.

Returns uniprot_id – The UniProt ID corresponding to the given HGNC ID.

Return type  str

### 4.3.2 Uniprot client (**indra.databases.uniprot_client**)

```
indra.databases.uniprot_client.get_family_members(family_name, human_only=True)
```

Return the HGNC gene symbols which are the members of a given family.

Parameters

- **family_name**(str) – Family name to be queried.
- **human_only**(bool) – If True, only human proteins in the family will be returned. Default: True

Returns gene_names – The HGNC gene symbols corresponding to the given family.

Return type  list

```
indra.databases.uniprot_client.get_gene_name(protein_id, web_fallback=True)
```

Return the gene name for the given UniProt ID.

This is an alternative to get_hgnc_name and is useful when HGNC name is not available (for instance, when the organism is not homo sapiens).

Parameters

- **protein_id**(str) – UniProt ID to be mapped.
• `web_fallback (Optional[bool])` – If True and the offline lookup fails, the UniProt web service is used to do the query.

Returns `gene_name` – The gene name corresponding to the given Uniprot ID.

Return type `str`

`indra.databases.uniprot_client.get_id_from_mgi(mgi_id)`
Return the UniProt ID given the MGI ID of a mouse protein.

Parameters `mgi_id (str)` – The MGI ID of the mouse protein.

Returns `up_id` – The UniProt ID of the mouse protein.

Return type `str`

`indra.databases.uniprot_client.get_id_from_mnemonic(uniprot_mnemonic)`
Return the UniProt ID for the given UniProt mnemonic.

Parameters `uniprot_mnemonic (str)` – UniProt mnemonic to be mapped.

Returns `uniprot_id` – The UniProt ID corresponding to the given Uniprot mnemonic.

Return type `str`

`indra.databases.uniprot_client.get_id_from_rgd(rgd_id)`
Return the UniProt ID given the RGD ID of a rat protein.

Parameters `rgd_id (str)` – The RGD ID of the rat protein.

Returns `up_id` – The UniProt ID of the rat protein.

Return type `str`

`indra.databases.uniprot_client.get_mgi_id(protein_id)`
Return the MGI ID given the protein id of a mouse protein.

Parameters `protein_id (str)` – UniProt ID of the mouse protein

Returns `mgi_id` – MGI ID of the mouse protein

Return type `str`

`indra.databases.uniprot_client.get_mnemonic(protein_id, web_fallback=False)`
Return the UniProt mnemonic for the given UniProt ID.

Parameters

• `protein_id (str)` – UniProt ID to be mapped.

• `web_fallback (Optional[bool])` – If True and the offline lookup fails, the UniProt web service is used to do the query.

Returns `mnemonic` – The UniProt mnemonic corresponding to the given Uniprot ID.

Return type `str`

`indra.databases.uniprot_client.get_mouse_id(human_protein_id)`
Return the mouse UniProt ID given a human UniProt ID.

Parameters `human_protein_id (str)` – The UniProt ID of a human protein.

Returns `mouse_protein_id` – The UniProt ID of a mouse protein orthologous to the given human protein

Return type `str`
**indra.databases.uniprot_client.get_primary_id(protein_id)**

Return a primary entry corresponding to the UniProt ID.

**Parameters**

`protein_id (str)` – The UniProt ID to map to primary.

**Returns**

`primary_id` – If the given ID is primary, it is returned as is. Otherwise the primary IDs are looked up. If there are multiple primary IDs then the first human one is returned. If there are no human primary IDs then the first primary found is returned.

**Return type**

str

**indra.databases.uniprot_client.get_rat_id(human_protein_id)**

Return the rat UniProt ID given a human UniProt ID.

**Parameters**

`human_protein_id (str)` – The UniProt ID of a human protein.

**Returns**

`rat_protein_id` – The UniProt ID of a rat protein orthologous to the given human protein.

**Return type**

str

**indra.databases.uniprot_client.get_rgd_id(protein_id)**

Return the RGD ID given the protein id of a rat protein.

**Parameters**

`protein_id (str)` – UniProt ID of the rat protein

**Returns**

`rgd_id` – RGD ID of the rat protein.

**Return type**

str

**indra.databases.uniprot_client.is_human(protein_id)**

Return True if the given protein id corresponds to a human protein.

**Parameters**

`protein_id (str)` – UniProt ID of the protein

**Returns**

**Return type**

True if the protein_id corresponds to a human protein, otherwise False.

**indra.databases.uniprot_client.is_mouse(protein_id)**

Return True if the given protein id corresponds to a mouse protein.

**Parameters**

`protein_id (str)` – UniProt ID of the protein

**Returns**

**Return type**

True if the protein_id corresponds to a mouse protein, otherwise False.

**indra.databases.uniprot_client.is_rat(protein_id)**

Return True if the given protein id corresponds to a rat protein.

**Parameters**

`protein_id (str)` – UniProt ID of the protein

**Returns**

**Return type**

True if the protein_id corresponds to a rat protein, otherwise False.

**indra.databases.uniprot_client.is_secondary(protein_id)**

Return True if the UniProt ID corresponds to a secondary accession.

**Parameters**

`protein_id (str)` – The UniProt ID to check.

**Returns**

**Return type**

True if it is a secondary accessing entry, False otherwise.

**indra.databases.uniprot_client.verify_location(protein_id, residue, location)**

Return True if the residue is at the given location in the UP sequence.

---

4.3. Database clients (**indra.databases**)

---
Parameters

- **protein_id** (str) – UniProt ID of the protein whose sequence is used as reference.
- **residue** (str) – A single character amino acid symbol (Y, S, T, V, etc.)
- **location** (str) – The location on the protein sequence (starting at 1) at which the residue should be checked against the reference sequence.

Returns

- True if the given residue is at the given position in the sequence corresponding to the given UniProt ID, otherwise False.

```python
indra.databases.uniprot_client.verify_modification(protein_id, residue, location=None)
```

Return True if the residue at the given location has a known modification.

Parameters

- **protein_id** (str) – UniProt ID of the protein whose sequence is used as reference.
- **residue** (str) – A single character amino acid symbol (Y, S, T, V, etc.)
- **location** (Optional[str]) – The location on the protein sequence (starting at 1) at which the modification is checked.

Returns

- True if the given residue is reported to be modified at the given position in the sequence corresponding to the given UniProt ID, otherwise False.
- If location is not given, we only check if there is any residue of the given type that is modified.

4.3.3 ChEBI client (indra.databases.chebi_client)

```python
indra.databases.chebi_client.get_chebi_id_from_pubchem(pubchem_id)
```

Return the ChEBI ID corresponding to a given Pubchem ID.

Parameters **pubchem_id** (str) – Pubchem ID to be converted.

Returns **chebi_id** – ChEBI ID corresponding to the given Pubchem ID. If the lookup fails, None is returned.

Return type str

```python
indra.databases.chebi_client.get_pubchem_id(chebi_id)
```

Return the PubChem ID corresponding to a given ChEBI ID.

Parameters **chebi_id** (str) – ChEBI ID to be converted.

Returns **pubchem_id** – PubChem ID corresponding to the given ChEBI ID. If the lookup fails, None is returned.

Return type str
4.3.4 BioGRID client (indra.databases.biogrid_client)

indra.databases.biogrid_client.get_publications(gene_names, save_json_name=None)

Return evidence publications for interaction between the given genes.

Parameters

• **gene_names** (list[str]) – A list of gene names (HGNC symbols) to query interactions between. Currently supports exactly two genes only.

• **save_json_name** (Optional[str]) – A file name to save the raw BioGRID web service output in. By default, the raw output is not saved.

Returns **publications** – A list of Publication objects that provide evidence for interactions between the given list of genes.

Return type list[Publication]

4.3.5 Cell type context client (indra.databases.context_client)

indra.databases.context_client.get_mutations(gene_names, cell_types)

Return protein amino acid changes in given genes and cell types.

Parameters

• **gene_names** (list) – HGNC gene symbols for which mutations are queried.

• **cell_types** (list) – List of cell type names in which mutations are queried. The cell type names follow the CCLE database conventions.

  Example: LOXIMVI_SKIN, BT20_BREAST

Returns **res** – A dictionary keyed by cell line, which contains another dictionary that is keyed by gene name, with a list of amino acid substitutions as values.

Return type dict[dict[list]]

indra.databases.context_client.get_protein_expression(gene_names, cell_types)

Return the protein expression levels of genes in cell types.

Parameters

• **gene_names** (list) – HGNC gene symbols for which expression levels are queried.

• **cell_types** (list) – List of cell type names in which expression levels are queried. The cell type names follow the CCLE database conventions.

  Example: LOXIMVI_SKIN, BT20_BREAST

Returns **res** – A dictionary keyed by cell line, which contains another dictionary that is keyed by gene name, with estimated protein amounts as values.

Return type dict[dict[float]]

4.3.6 Network relevance client (indra.databases.relevance_client)

indra.databases.relevance_client.get_heat_kernel(network_id)

Return the identifier of a heat kernel calculated for a given network.

Parameters **network_id** (str) – The UUID of the network in NDEx.

Returns **kernel_id** – The identifier of the heat kernel calculated for the given network.
Return type  str

```
indra.databases.relevance_client.get_relevant_nodes(network_id, query_nodes)
```

Return a set of network nodes relevant to a given query set.

A heat diffusion algorithm is used on a pre-computed heat kernel for the given network which starts from the given query nodes. The nodes in the network are ranked according to heat score which is a measure of relevance with respect to the query nodes.

Parameters

- **network_id** (str) – The UUID of the network in NDEx.
- **query_nodes** (list[str]) – A list of node names with respect to which relevance is queried.

Returns ranked_entities – A list containing pairs of node names and their relevance scores.

Return type  list[(str, float)]

### 4.3.7 NDEx client (indra.databases.ndex_client)

```
indra.databases.ndex_client.send_request(ndex_service_url, params, is_json=True, use_get=False)
```

Send a request to the NDEx server.

Parameters

- **ndex_service_url** (str) – The URL of the service to use for the request.
- **params** (dict) – A dictionary of parameters to send with the request. Parameter keys differ based on the type of request.
- **is_json** (bool) – True if the response is in json format, otherwise it is assumed to be text. Default: False
- **use_get** (bool) – True if the request needs to use GET instead of POST.

Returns res – Depending on the type of service and the is_json parameter, this function either returns a text string or a json dict.

Return type  str

### 4.3.8 cBio portal client (indra.databases.cbio_client)

```
indra.databases.cbio_client.get_cancer_studies(study_filter=None)
```

Return a list of cancer study identifiers, optionally filtered.

There are typically multiple studies for a given type of cancer and a filter can be used to constrain the returned list.

Parameters **study_filter** (Optional[str]) – A string used to filter the study IDs to return. Example: “paad”

Returns study_ids – A list of study IDs. For instance “paad” as a filter would result in a list of study IDs with paad in their name like “paad_icgc”, “paad_tcga”, etc.

Return type  list[str]

```
indra.databases.cbio_client.get_cancer_types(cancer_filter=None)
```

Return a list of cancer types, optionally filtered.
**Parameters**

`cancer_filter (Optional[str])` – A string used to filter cancer types. Its value is the name or part of the name of a type of cancer. Example: “melanoma”, “pancreatic”, “non-small cell lung”

**Returns**

`type_ids` – A list of cancer types matching the filter. Example: for `cancer_filter="pancreatic"`, the result includes “panet” (neuro-endocrine) and “paad” (adenocarcinoma)

**Return type**

`list[str]`

---

`indra.databases.cbio_client.get_case_lists(study_id)`

Return a list of the case set ids for a particular study.

TAKE NOTE the “case_list_id” are the same thing as “case_set_id” Within the data, this string is referred to as a “case_list_id”. Within API calls it is referred to as a ‘case_set_id’. The documentation does not make this explicitly clear.

**Parameters**

`study_id (str)` – The ID of the cBio study. Example: ‘cellline_ccle_broad’ or ‘paad_icgc’

**Returns**

`case_set_ids` – A dict keyed to cases containing a dict keyed to genes containing int

**Return type**

`dict[dict[int]]`

---

`indra.databases.cbio_client.get_ccle_cna(gene_list, cell_lines)`

Return a dict of CNAs in given genes and cell lines from CCLE.

This is a specialized call to get_mutations tailored to CCLE cell lines.

**Parameters**

- `gene_list (list[str])` – A list of HGNC gene symbols to get mutations in
- `cell_lines (list[str])` – A list of CCLE cell line names to get mutations for.

**Returns**

`profile_data` – A dict keyed to cases containing a dict keyed to genes containing int

**Return type**

`dict[dict[int]]`

---

`indra.databases.cbio_client.get_ccle_lines_for_mutation(gene, amino_acid_change)`

Return cell lines with a given point mutation in a given gene.

Checks which cell lines in CCLE have a particular point mutation in a given gene and return their names in a list.

**Parameters**

- `gene (str)` – The HGNC symbol of the mutated gene in whose product the amino acid change occurs. Example: “BRAF”
- `amino_acid_change (str)` – The amino acid change of interest. Example: “V600E”

**Returns**

`cell_lines` – A list of CCLE cell lines in which the given mutation occurs.

**Return type**

`list`

---

`indra.databases.cbio_client.get_ccle_mrna(gene_list, cell_lines)`

Return a dict of mRNA amounts in given genes and cell lines from CCLE.

**Parameters**

- `gene_list (list[str])` – A list of HGNC gene symbols to get mRNA amounts for.
- `cell_lines (list[str])` – A list of CCLE cell line names to get mRNA amounts for.
Returns `mrna_amounts` – A dict keyed to cell lines containing a dict keyed to genes containing float

Return type: `dict[dict[float]]`

`indra.databases.cbio_client.get_genetic_profiles(study_id, profile_filter=None)`

Return all the genetic profiles (data sets) for a given study.

Genetic profiles are different types of data for a given study. For instance the study ‘cellline_ccle_broad’ has profiles such as ‘cellline_ccle_broad_mutations’ for mutations, ‘cellline_ccle_broad_CNA’ for copy number alterations, etc.

Parameters

- **study_id** (`str`) – The ID of the cBio study. Example: ‘paad_icgc’
- **profile_filter** (`Optional[str]`) – A string used to filter the profiles to return. Will be one of: - MUTATION - MUTATION_EXTENDED - COPY_NUMBER_ALTERATION - MRNA_EXPRESSION - METHYLATION The genetic profiles can include “mutation”, “CNA”, “rppa”, “methylation”, etc.

Returns `genetic_profiles` – A list of genetic profiles available for the given study.

Return type: `list[str]`

`indra.databases.cbio_client.get_mutations(study_id, gene_list, mutation_type=None, case_id=None)`

Return mutations as a list of genes and list of amino acid changes.

Parameters

- **study_id** (`str`) – The ID of the cBio study. Example: ‘cellline_ccle_broad’ or ‘paad_icgc’
- **gene_list** (`list[str]`) – A list of genes with their HGNC symbols. Example: ['BRAF', 'KRAS']
- **mutation_type** (`Optional[str]`) – The type of mutation to filter to. mutation_type can be one of: missense, nonsense, frame_shift_ins, frame_shift_del, splice_site
- **case_id** (`Optional[str]`) – The case ID within the study to filter to.

Returns `mutations` – A tuple of two lists, the first one containing a list of genes, and the second one a list of amino acid changes in those genes.

Return type: `tuple[list]`

`indra.databases.cbio_client.get_mutations_ccle(gene_list, cell_lines, mutation_type=None)`

Return a dict of mutations in given genes and cell lines from CCLE.

This is a specialized call to get_mutations tailored to CCLE cell lines.

Parameters

- **gene_list** (`list[str]`) – A list of HGNC gene symbols to get mutations in
- **cell_lines** (`list[str]`) – A list of CCLE cell line names to get mutations for.
- **mutation_type** (`Optional[str]`) – The type of mutation to filter to. mutation_type can be one of: missense, nonsense, frame_shift_ins, frame_shift_del, splice_site

Returns `mutations` – The result from cBioPortal as a dict in the format `{cell_line : {gene : [mutation1, mutation2, ...]}}`
Example: {'LOXIMVI_SKIN': {'BRAF': ['V600E', 'I208V']}, 'SKMEL30_SKIN': {'BRAF': ['D287H', 'E275K']}}

Return type  dict

indra.databases.cbio_client.get_num_sequenced(study_id)

Return number of sequenced tumors for given study. This is useful for calculating mutation statistics in terms of the prevalence of certain mutations within a type of cancer.

Parameters

• **study_id** (str) – The ID of the cBio study. Example: ‘paad_icgc’

Returns

• **num_case** – The number of sequenced tumors in the given study

Return type  int

indra.databases.cbio_client.get_profile_data(study_id, gene_list, profile_filter, case_set_filter=None)

Return dict of cases and genes and their respective values.

Parameters

• **study_id** (str) – The ID of the cBio study. Example: ‘cellline_ccle_broad’ or ‘paad_icgc’

• **gene_list** (list[str]) – A list of genes with their HGNC symbols. Example: ['BRAF', 'KRAS']

• **profile_filter** (str) – A string used to filter the profiles to return. Will be one of: - MUTATION - MUTATION_EXTENDED - COPY_NUMBER_ALTERATION - MRNA_EXPRESSION - METHYLATION

• **case_set_filter** (Optional[str]) – A string that specifies which case_set_id to use, based on a complete or partial match. If not provided, will look for study_id + '_all'

Returns

• **profile_data** – A dict keyed to cases containing a dict keyed to genes containing int

Return type  dict[dict[int]]

### 4.4 Literature clients (indra.literature)

indra.literature.get_full_text(paper_id, idtype, preferred_content_type=u’text/xml’)

Return the content and the content type of an article.

This function retrieves the content of an article by its PubMed ID, PubMed Central ID, or DOI. It prioritizes full text content when available and returns an abstract from PubMed as a fallback.

Parameters

• **paper_id**(string) – ID of the article.

• **idtype** (‘pmid’, ‘pmcid’, or ‘doi’) – Type of the ID.

• **preferred_content_type** (Optional[str]) – Preference for full-text format, if available. Can be one of ‘text/xml’, ‘text/plain’, ‘application/pdf’. Default: ‘text/xml’

Returns

• **content**(str) – The content of the article.

• **content_type**(str) – The content type of the article
indra.literature.id_lookup(paper_id, idtype)
Take an ID of type PMID, PMCID, or DOI and lookup the other IDs.
If the DOI is not found in Pubmed, try to obtain the DOI by doing a reverse-lookup of the DOI in CrossRef using article metadata.

Parameters
- paper_id (string) – ID of the article.
- idtype ('pmid', 'pmcid', or 'doi') – Type of the ID.

Returns ids – A dictionary with the following keys: pmid, pmcid and doi.

Return type dict

4.4.1 Pubmed client (indra.literature.pubmed_client)
Search and get metadata for articles in Pubmed.

indra.literature.pubmed_client.expand_pagination(pages)
Convert a page number to long form, e.g., from 456-7 to 456-457.

indra.literature.pubmed_client.get_abstract(pubmed_id, prepend_title=True)
Get the abstract of an article in the Pubmed database.

indra.literature.pubmed_client.get_metadata_for_ids(pmid_list, get_issns_from_nlm=False)
Get article metadata for up to 200 PMIDs from the Pubmed database.

Parameters
- pmid_list (list of PMIDs as strings) – Can contain 1-200 PMIDs.
- get_issns_from_nlm (boolean) – Look up the full list of ISSN number for the journal associated with the article, which helps to match articles to CrossRef search results. Defaults to False, since it slows down performance.


Return type dict

indra.literature.pubmed_client.get_title(pubmed_id)
Get the title of an article in the Pubmed database.

4.4.2 Pubmed Central client (indra.literature.pmc_client)

indra.literature.pmc_client.filter_pmids(pmid_list, source_type)
Filter a list of PMIDs for ones with full text from PMC.

Parameters
- pmid_list (list) – List of PMIDs to filter.
- source_type (string) – One of ‘fulltext’, ‘oa_xml’, ‘oa_txt’, or ‘auth_xml’.

Returns

Return type list of PMIDs available in the specified source/format type.
4.4.3 CrossRef client (indra.literature.crossref_client)

indra.literature.crossref_client.doi_query(pmid, search_limit=10)
Get the DOI for a PMID by matching CrossRef and Pubmed metadata.

Searches CrossRef using the article title and then accepts search hits only if they have a matching journal ISSN and page number with what is obtained from the Pubmed database.

indra.literature.crossref_client.get_fulltext_links(doi)
Return a list of links to the full text of an article given its DOI. Each list entry is a dictionary with keys: - URL: the URL to the full text - content-type: e.g. text/xml or text/plain - content-version - intended-application: e.g. text-mining

4.4.4 Elsevier client (indra.literature.elsevier_client)

For information on the Elsevier API, see:

- API Specification: http://dev.elsevier.com/api_docs.html
- Authentication: https://dev.elsevier.com/tecdoc_api_authentication.html

indra.literature.elsevier_client.download_article(doi)
Download an article in XML format from Elsevier.

indra.literature.elsevier_client.get_abstract(doi)
Get the abstract of an article from Elsevier.

indra.literature.elsevier_client.get_article(doi, output='txt')
Get the full body of an article from Elsevier. There are two output modes: ‘txt’ strips all xml tags and joins the pieces of text in the main text, while ‘xml’ simply takes the tag containing the body of the article and returns it as is. In the latter case, downstream code needs to be able to interpret Elsevier’s XML format.

4.5 Preassembly (indra.preassembler)

4.5.1 Preassembler (indra.preassembler)

class indra.preassembler.Preassembler(hierarchies, stmts=None)
De-duplicates statements and arranges them in a specificity hierarchy.

Parameters

- **hierarchies** (dict[indra.preassembler.hierarchy_manager]) – A dictionary of hierarchies with keys such as ‘entity’ (hierarchy of entities, primarily specifying relationships between genes and their families) and ‘modification’ pointing to Hierarchy-Managers

- **stmts** (list of indra.statements.Statement or None) – A set of statements to perform pre-assembly on. If None, statements should be added using the add_statements() method.

  stmts
  list of indra.statements.Statement – Starting set of statements for preassembly.
unique_stmts
list of indra.statements.Statement – Statements resulting from combining duplicates.

related_stmts
list of indra.statements.Statement – Top-level statements after building the refinement hierarchy.

hierarchies
dict[indra.preassembler.hierarchy_manager] – A dictionary of hierarchies with keys such as ‘entity’ and ‘modification’ pointing to HierarchyManagers

addStatements(stmts)
Add to the current list of statements.

Parameters stmts (list of indra.statements.Statement) – Statements to add to the current list.

static combine_duplicate_stmts(stmts)
Combine evidence from duplicate Statements.

Statements are deemed to be duplicates if they have the same key returned by the matches_key() method of the Statement class. This generally means that statements must be identical in terms of their arguments and can differ only in their associated Evidence objects.

This function keeps the first instance of each set of duplicate statements and merges the lists of Evidence from all of the other statements.

Parameters stmts (list of indra.statements.Statement) – Set of statements to de-duplicate.

Returns Unique statements with accumulated evidence across duplicates.

Return type list of indra.statements.Statement

Examples

De-duplicate and combine evidence for two statements differing only in their evidence lists:

```python
>>> map2k1 = Agent('MAP2K1')
>>> mapk1 = Agent('MAPK1')
>>> stmt1 = Phosphorylation(map2k1, mapk1, 'T', '185',
... evidence=[Evidence(text='evidence 1')])
>>> stmt2 = Phosphorylation(map2k1, mapk1, 'T', '185',
... evidence=[Evidence(text='evidence 2')])
>>> uniq_stmts = Preassembler.combine_duplicate_stmts([stmt1, stmt2])
>>> uniq_stmts
[Phosphorylation(MAP2K1(), MAPK1(), T, 185)]
>>> sorted([e.text for e in uniq_stmts[0].evidence])
['evidence 1', 'evidence 2']
```

combine_duplicates()
Combine duplicates among stmts and save result in unique_stmts.

A wrapper around the static method combine_duplicate_stmts().

combine_related(return_toplevel=True, poolsize=None, size_cutoff=100)
Connect related statements based on their refinement relationships.

This function takes as a starting point the unique statements (with duplicates removed) and returns a modified flat list of statements containing only those statements which do not represent a refinement of
other existing statements. In other words, the more general versions of a given statement do not appear at the top level, but instead are listed in the *supports* field of the top-level statements.

If `unique_stmts` has not been initialized with the de-duplicated statements, `combine_duplicates()` is called internally.

After this function is called the attribute `related_stmts` is set as a side-effect.

The procedure for combining statements in this way involves a series of steps:

1. The statements are grouped by type (e.g., Phosphorylation) and each type is iterated over independently.

2. Statements of the same type are then grouped according to their Agents’ entity hierarchy component identifiers. For instance, ERK, MAPK1 and MAPK3 are all in the same connected component in the entity hierarchy and therefore all Statements of the same type referencing these entities will be grouped. This grouping assures that relations are only possible within Statement groups and not among groups. For two Statements to be in the same group at this step, the Statements must be the same type and the Agents at each position in the Agent lists must either be in the same hierarchy component, or if they are not in the hierarchy, must have identical entity_matches_keys. Statements with None in one of the Agent list positions are collected separately at this stage.

3. Statements with None at either the first or second position are iterated over. For a statement with a None as the first Agent, the second Agent is examined; then the Statement with None is added to all Statement groups with a corresponding component or entity_matches_key in the second position. The same procedure is performed for Statements with None at the second Agent position.

4. The statements within each group are then compared; if one statement represents a refinement of the other (as defined by the `refinement_of()` method implemented for the Statement), then the more refined statement is added to the *supports* field of the more general statement, and the more general statement is added to the *supported_by* field of the more refined statement.

5. A new flat list of statements is created that contains only those statements that have no *supports* entries (statements containing such entries are not eliminated, because they will be retrievable from the *supported_by* fields of other statements). This list is returned to the caller.

On multi-core machines, the algorithm can be parallelized by setting the poolsize argument to the desired number of worker processes. This feature is only available in Python > 3.4.

**Parameters**

- `return_toplevel` *(Optional*[bool]*) – If True only the top level statements are returned. If False, all statements are returned. Default: True

- `poolsize` *(Optional*[int]*) – The number of worker processes to use to parallelize the comparisons performed by the function. If None (default), no parallelization is performed. NOTE: Parallelization is only available on Python 3.4 and above.

- `size_cutoff` *(Optional*[int]*) – Groups with size_cutoff or more statements are sent to worker processes, while smaller groups are compared in the parent process. Default value is 100. Not relevant when parallelization is not used.

---

**Note:** Subfamily relationships must be consistent across arguments

For now, we require that merges can only occur if the *isa* relationships are all in the *same direction for all the agents* in a Statement. For example, the two statement groups: *RAF_family* -> *MEK1* and *BRAF* -> *MEK_family* would not be merged, since *BRAF* *isa* *RAF_family*, but *MEK_family* is not a *MEK1*. In the future this restriction could be revisited.
Returns The returned list contains Statements representing the more concrete/refined versions of the Statements involving particular entities. The attribute `related_stmts` is also set to this list. However, if `return_toplevel` is False then all statements are returned, irrespective of level of specificity. In this case the relationships between statements can be accessed via the supports/supported_by attributes.

Return type list of `indra.statement.Statement`

Examples

A more general statement with no information about a Phosphorylation site is identified as supporting a more specific statement:

```python
>>> from indra.preassembler.hierarchy_manager import hierarchies
>>> braf = Agent('BRAF')
>>> map2k1 = Agent('MAP2K1')
>>> st1 = Phosphorylation(braf, map2k1)
>>> st2 = Phosphorylation(braf, map2k1, residue='S')
>>> pa = Preassembler(hierarchies, [st1, st2])
>>> combined_stmts = pa.combine_related()
```

```python
>>> combined_stmts
[Phosphorylation(BRAF(), MAP2K1(), S)]
```

```python
>>> combined_stmts[0].supported_by
[Phosphorylation(BRAF(), MAP2K1())]
```

```python
>>> combined_stmts[0].supported_by[0].supports
[Phosphorylation(BRAF(), MAP2K1(), S)]
```

`indra.preassembler.flatten_evidence(stmts)`

Add evidence from `supporting` stmts to evidence for `supported` stmts.

Parameters stmts (list of `indra.statements.Statement`) — A list of top-level statements with associated supporting statements resulting from building a statement hierarchy with `combine_related()`.

Returns stmts — Statement hierarchy identical to the one passed, but with the evidence lists for each statement now containing all of the evidence associated with the statements they are supported by.

Return type list of `indra.statements.Statement`

Examples

Flattening evidence adds the two pieces of evidence from the supporting statement to the evidence list of the top-level statement:

```python
>>> from indra.preassembler.hierarchy_manager import hierarchies
>>> braf = Agent('BRAF')
>>> map2k1 = Agent('MAP2K1')
>>> st1 = Phosphorylation(braf, map2k1,
... evidence=[Evidence(text='foo'), Evidence(text='bar')])
>>> st2 = Phosphorylation(braf, map2k1, residue='S',
... evidence=[Evidence(text='baz'), Evidence(text='bak')])
>>> pa = Preassembler(hierarchies, [st1, st2])
>>> pa.combine_related()
[Phosphorylation(BRAF(), MAP2K1(), S)]
```

```python
>>> [e.text for e in pa.related_stmts[0].evidence]
```

```python
['foo', 'bar', 'baz', 'bak']
```
indra.preassembler.flatten_stmts(stmts)

Return the full set of unique stms in a pre-assembled stmt graph.

The flattened list of of statements returned by this function can be compared to the original set of unique statements to make sure no statements have been lost during the preassembly process.

Parameters

**stmts** (list of indra.statements.Statement) – A list of top-level statements with associated supporting statements resulting from building a statement hierarchy with combine_related().

Returns

stmts – List of all statements contained in the hierarchical statement graph.

Return type

list of indra.statements.Statement

Examples

Calling combine_related() on two statements results in one top-level statement; calling flatten_stmts() recovers both:

```python
from indra.preassembler.hierarchy_manager import hierarchies
braf = Agent('BRAF')
map2k1 = Agent('MAP2K1')
st1 = Phosphorylation(braf, map2k1)
st2 = Phosphorylation(braf, map2k1, residue='S')
pa = Preassembler(hierarchies, [st1, st2])
pa.combine_related()
[Phosphorylation(BRAF(), MAP2K1(), S)]
flattened = flatten_stmts(pa.related_stmts)
flattened.sort(key=lambda x: x.matches_key())
flattened
[Phosphorylation(BRAF(), MAP2K1()), Phosphorylation(BRAF(), MAP2K1(), S)]
```

indra.preassembler.render_stmt_graph(stmts, agent_style=None)

Render the statement hierarchy as a pygraphviz graph.

Parameters

- **stmts** (list of indra.statements.Statement) – A list of top-level statements with associated supporting statements resulting from building a statement hierarchy with combine_related().

- **agent_style** (dict or None) – Dict of attributes specifying the visual properties of nodes. If None, the following default attributes are used:

```python
agent_style = {'color': 'lightgray', 'style': 'filled', 'fontname': 'arial'}
```

Returns

Pygraphviz graph with nodes representing statements and edges pointing from supported statements to supported_by statements.

Return type

pygraphviz.AGraph
Examples

Pattern for getting statements and rendering as a Graphviz graph:

```python
>>> from indra.preassembler.hierarchy_manager import hierarchies
>>> braf = Agent('BRAF')
>>> map2k1 = Agent('MAP2K1')
>>> st1 = Phosphorylation(braf, map2k1)
>>> st2 = Phosphorylation(braf, map2k1, residue='S')
>>> pa = Preassembler(hierarchies, [st1, st2])
>>> pa.combine_related()
[Phosphorylation(BRAF(), MAP2K1(), S)]
>>> graph = render_stmt_graph(pa.related_stmts)
>>> graph.write('example_graph.dot')  # To make the DOT file
>>> graph.draw('example_graph.png', prog='dot')  # To make an image
```

Resulting graph:

```
Phosphorylation(BRAF(), MAP2K1(), S) → Phosphorylation(BRAF(), MAP2K1())
```

4.5.2 Entity grounding curation and mapping (indra.preassembler.grounding_mapper)

**indra.preassembler.grounding_mapper.protein_map_from_twg(twg)**

Build map of entity texts to validated protein grounding.

Looks at the grounding of the entity texts extracted from the statements and finds proteins where there is grounding to a human protein that maps to an HGNC name that is an exact match to the entity text. Returns a dict that can be used to update/expand the grounding map.

4.5.3 Site curation and mapping (indra.preassembler.sitemapper)

**class indra.preassembler.sitemapper.MappedStatement**

Information about a Statement found to have invalid sites.

**Parameters**

- `original_stmt` *(indra.statements.Statement)* – The statement prior to mapping.
- `mapped_mods` *(list of tuples)* – A list of invalid sites, where each entry in the list has two elements: ((gene_name, residue, position), mapped_site). If the invalid position was not found in the site map, mapped_site is `None`; otherwise it is a tuple consisting of (residue, position, comment).
- `mapped_stmt` *(indra.statements.Statement)* – The statement after mapping. Note that if no information was found in the site map, it will be identical to the original statement.

**class indra.preassembler.sitemapper.SiteMapper**

Use curated site information to standardize modification sites in stmts.

**Parameters**

- `site_map` *(dict (as returned by load_site_map()))* – A dict mapping tuples of the form `(gene, orig_res, orig_pos)` to a tuple of the form `(correct_res, correct_pos, comment)`, where `gene` is the string name of the gene (canonicalized to HGNC); `orig_res` and `orig_pos` are the residue and position to be mapped; `correct_res` and `correct_pos` are the corrected residue and
position, and comment is a string describing the reason for the mapping (species error, isoform error, wrong residue name, etc.).

Examples

Fixing site errors on both the modification state of an agent (MAP2K1) and the target of a Phosphorylation statement (MAPK1):

```python
>>> map2k1_phos = Agent('MAP2K1', db_refs={'UP': 'Q02750'}, mods=[
... ModCondition('phosphorylation', 'S', '217'),
... ModCondition('phosphorylation', 'S', '221')])
>>> mapk1 = Agent('MAPK1', db_refs={'UP': 'P28482'})
>>> stmt = Phosphorylation(map2k1_phos, mapk1, 'T', '183')
>>> (valid, mapped) = default_mapper.map_sites([stmt])
>>> valid
[
]
>>> mapped
[
 MappedStatement:
original_stmt: Phosphorylation(MAP2K1(mods: (phosphorylation, S, 217),
←(phosphorylation, S, 221)), MAPK1(), T, 183)
mapped_mods: (('MAP2K1', 'S', '217'), ('S', '218', 'off by one'))
('MAP2K1', 'S', '221'), ('S', '222', 'off by one'))
('MAP1', 'T', '183'), ('T', '185', 'off by two; mouse sequence
←'))
mapped_stmt: Phosphorylation(MAP2K1(mods: (phosphorylation, S, 218),
←(phosphorylation, S, 222)), MAPK1(), T, 185)
]
>>> ms = mapped[0]
>>> ms.original_stmt
Phosphorylation(MAP2K1(mods: (phosphorylation, S, 217), (phosphorylation, S, ←221)), MAPK1(), T, 183)
>>> ms.mapped.mods
[('MAP2K1', 'S', '217'), ('S', '218', 'off by one'))
('MAP2K1', 'S', '221'), ('S', '222', 'off by one'))
('MAP1', 'T', '183'), ('T', '185', 'off by two; mouse sequence
←')]
>>> ms.mapped_stmt
Phosphorylation(MAP2K1(mods: (phosphorylation, S, 218), (phosphorylation, S, ←222)), MAPK1(), T, 185)
```

map_sites(stmts, do_methionine_offset=True, do_orthology_mapping=True, do_isofrom_mapping=True)

Check a set of statements for invalid modification sites.

Statements are checked against Uniprot reference sequences to determine if residues referred to by post-translational modifications exist at the given positions.

If there is nothing amiss with a statement (modifications on any of the agents, modifications made in the statement, etc.), then the statement goes into the list of valid statements. If there is a problem with the statement, the offending modifications are looked up in the site map (site_map), and an instance of MappedStatement is added to the list of mapped statements.

Parameters

- **stmts** (list of indra.statement.Statement) – The statements to check for site errors.
- **do_methionine_offset** (boolean) – Whether to check for off-by-one errors in site
position (possibly) attributable to site numbering from mature proteins after cleavage of the initial methionine. If True, checks the reference sequence for a known modification at 1 site position greater than the given one; if there exists such a site, creates the mapping. Default is True.

- **do_orthology_mapping** *(boolean)* – Whether to check sequence positions for known modification sites in mouse or rat sequences (based on PhosphoSitePlus data). If a mouse/rat site is found that is linked to a site in the human reference sequence, a mapping is created. Default is True.

- **do_isoform_mapping** *(boolean)* – Whether to check sequence positions for known modifications in other human isoforms of the protein (based on PhosphoSitePlus data). If a site is found that is linked to a site in the human reference sequence, a mapping is created. Default is True.

**Returns** 2-tuple containing *(valid_statements, mapped_statements)*. The first element of the tuple is a list of valid statements *(indra.statement.Statement)* that were not found to contain any site errors. The second element of the tuple is a list of mapped statements *(MappedStatement)* with information on the incorrect sites and corresponding statements with correctly mapped sites.

**Return type** tuple

`indra.preassembler.sitemapper.default_mapper = indra.preassembler.sitemapper.SiteMapper object`

A default instance of `SiteMapper` that contains the site information found in `resources/curated_site_map.csv`.

`indra.preassembler.sitemapper.load_site_map(path)`

Load the modification site map from a file.

The site map file should be a comma-separated file with six columns:

<table>
<thead>
<tr>
<th>Gene: HGNC gene name</th>
<th>OrigRes: Original (incorrect) residue</th>
<th>OrigPos: Original (incorrect) residue position</th>
<th>CorrectRes: The correct residue for the modification</th>
<th>CorrectPos: The correct residue position</th>
<th>Comment: Description of the reason for the error</th>
</tr>
</thead>
</table>

**Parameters** *path* *(string)* – Path to the tab-separated site map file.

**Returns** A dict mapping tuples of the form *(gene, orig_res, orig_pos)* to a tuple of the form *(correct_res, correct_pos, comment)*, where *gene* is the string name of the gene (canonicalized to HGNC); *orig_res* and *orig_pos* are the residue and position to be mapped; *correct_res* and *correct_pos* are the corrected residue and position, and *comment* is a string describing the reason for the mapping (species error, isoform error, wrong residue name, etc.).

**Return type** dict

### 4.5.4 Hierarchy manager *(indra.preassembler.hierarchy_manager)*

**class** `indra.preassembler.hierarchy_manager.HierarchyManager(rdf_file, build_closure=True, uri_as_name=True)`

Store hierarchical relationships between different types of entities.

Used to store, e.g., entity hierarchies (proteins and protein families) and modification hierarchies (serine phosphorylation vs. phosphorylation).

**Parameters**
• **rdf_file** *(string)* – Path to the RDF file containing the hierarchy.

• **build_closure** *(Optional[bool])* – If True, the transitive closure of the hierarchy is generated up from to speed up processing. Default: True

• **uri_as_name** *(Optional[bool])* – If True, entries are accessed directly by their URIs. If False entries are accessed by finding their name through the hasName relationship. Default: True

**graph**

Instance of *rdflib.Graph* – The RDF graph containing the hierarchy.

**build_transitive_closures()**

Build the transitive closures of the hierarchy.

This method constructs dictionaries which contain terms in the hierarchy as keys and either all the “isa+” or “partof+” related terms as values.

**get_children** *(uri)*

Return all (not just immediate) children of a given entry.

**Parameters**

- **uri** *(str)* – The URI of the entry whose children are to be returned. See the get_uri method to construct this URI from a name space and id.

**get_parents** *(uri, type=u’all’)*

Return parents of a given entry.

**Parameters**

- **uri** *(str)* – The URI of the entry whose parents are to be returned. See the get_uri method to construct this URI from a name space and id.

- **type** *(str)* – ‘all’: return all parents irrespective of level; ‘immediate’: return only the immediate parents; ‘top’: return only the highest level parents

**isa** *(ns1, id1, ns2, id2)*

Indicate whether one entity has an “isa” relationship to another.

**Parameters**

- **ns1** *(string)* – Namespace code for an entity.

- **id1** *(string)* – URI for an entity.

- **ns2** *(string)* – Namespace code for an entity.

- **id2** *(string)* – URI for an entity.

**Returns**

True if t1 has an “isa” relationship with t2, either directly or through a series of intermediates; False otherwise.

**Return type**

*bool*

**partof** *(ns1, id1, ns2, id2)*

Indicate whether one entity is physically part of another.

**Parameters**

- **ns1** *(string)* – Namespace code for an entity.

- **id1** *(string)* – URI for an entity.

- **ns2** *(string)* – Namespace code for an entity.

- **id2** *(string)* – URI for an entity.
Returns True if $t_1$ has a “partof” relationship with $t_2$, either directly or through a series of intermediates; False otherwise.

Return type bool

4.6 Belief Engine (indra.belief)

class indra.belief.BeliefEngine (prior_probs=None)
Assigns beliefs to INDRA Statements based on supporting evidence.

Parameters prior_probs (Optional[dict[dict]]) – A dictionary of prior probabilities used to override/extend the default ones. There are two types of prior probabilities: rand and syst corresponding to random error and systematic error rate for each knowledge source. The prior_probs dictionary has the general structure {‘rand’: {‘s1’: pr1, ‘s2’: pr2, ...}, ‘syst’: {‘s1’: ps1, ‘s2’: ps2, ...}} where ‘s1’ ... ‘sn’ are names of input sources and pr1 ... prn and ps1 ... psn are error probabilities. Examples: {'rand': {'some_source': 0.1}} sets the random error rate for some_source to 0.1; {'rand': {}}

prior_probs
dict[dict] – A dictionary of prior systematic and random error probabilities for each knowledge source.

set_hierarchy_probs (statements)
Sets hierarchical belief probabilities for a list of INDRA Statements.

The Statements are assumed to be in a hierarchical relation graph with the supports and supported_by attribute of each Statement object having been set. The hierarchical belief probability of each Statement is calculated based on its prior probability and the probabilities propagated from Statements supporting it in the hierarchy graph.

Parameters statements (list[indra.statements.Statement]) – A list of INDRA Statements whose belief scores are to be calculated. Each Statement object’s belief attribute is updated by this function.

set_linked_probs (linked_statements)
Sets the belief probabilities for a list of linked INDRA Statements.

The list of LinkedStatement objects is assumed to come from the MechanismLinker. The belief probability of the inferred Statement is assigned the joint probability of its source Statements.

Parameters linked_statements (list[indra.mechlinker.LinkedStatement]) – A list of INDRA LinkedStatements whose belief scores are to be calculated. The belief attribute of the inferred Statement in the LinkedStatement object is updated by this function.

set_prior_probs (statements)
Sets the prior belief probabilities for a list of INDRA Statements.

The Statements are assumed to be de-duplicated. In other words, each Statement in the list passed to this function is assumed to have a list of Evidence objects that support it. The prior probability of each Statement is calculated based on the number of Evidences it has and their sources.

Parameters statements (list[indra.statements.Statement]) – A list of INDRA Statements whose belief scores are to be calculated. Each Statement object’s belief attribute is updated by this function.
4.7 Mechanism Linker (indra.mechlinker)

class indra.mechlinker.AgentState(agent)
A class representing Agent state without identifying a specific Agent.

bound_conditions
list[indra.statements.BoundCondition]

mods
list[indra.statements.ModCondition]

mutations
list[indra.statements.Mutation]

location
indra.statements.location

apply_to(agent)
Apply this object’s state to an Agent.

Parameters
agent (indra.statements.Agent) – The agent to which the state should

be applied

class indra.mechlinker.BaseAgent(name)
Represents all activity types and active forms of an Agent.

Parameters

• name (str) – The name of the BaseAgent

• activity_types (list[str]) – A list of activity types that the Agent has

• active_states (dict) – A dict of activity types and their associated Agent states

• activity_reductions (dict) – A dict of activity types and the type they are reduced
to by inference.

class indra.mechlinker.BaseAgentSet
Container for a set of BaseAgents.

This class wraps a dict of BaseAgent instance and can be used to get and set BaseAgents.

get_create_base_agent(agent)
Return BaseAgent from an Agent, creating it if needed.

Parameters
agent (indra.statements.Agent) –

Returns
base_agent

Return type
indra.mechlinker.BaseAgent

class indra.mechlinker.LinkedStatement(source_stmts, inferred_stmt)
A tuple containing a list of source Statements and an inferred Statement.

The list of source Statements are the basis for the inferred Statement.

Parameters

• source_stmts (list[indra.statements.Statement]) – A list of source Statements

• inferred_stmts (indra.statements.Statement) – A Statement that was inferred from the source Statements.
class indra.mechlinker.MechLinker(stmts=None)
    Rewrite the activation pattern of Statements and derive new Statements.

    The mechanism linker (MechLinker) traverses a corpus of Statements and uses various inference steps to make
    the activity types and active forms consistent among Statements.

    add_statements(stmts)
        Add statements to the MechLinker.

        Parameters stmts(list[indra.statements.Statement]) – A list of Statements to
        add.

    gather_explicit_activities()
        Aggregate all explicit activities and active forms of Agents.

        This function iterates over self.statements and extracts explicitly stated activity types and active forms for
        Agents.

    gather_implicit_activities()
        Aggregate all implicit activities and active forms of Agents.

        Iterate over self.statements and collect the implied activities and active forms of Agents that appear in the
        Statements.

        Note that using this function to collect implied Agent activities can be risky. Assume, for instance, that
        a Statement from a reading system states that EGF bound to EGFR phosphorylates ERK. This would
        be interpreted as implicit evidence for the EGFR-bound form of EGF to have ‘kinase’ activity, which is
        clearly incorrect.

        In contrast the alternative pair of this function: gather_explicit_activities collects only explicitly stated
        activities.

    static infer_activations(stmts)
        Return inferred RegulateActivity from Modification + ActiveForm.

        This function looks for combinations of Modification and ActiveForm Statements and infers Activation/Inhibition Statements from them. For example, if we know that A phosphorylates B, and the phosphorylated form of B is active, then we can infer that A activates B. This can also be viewed as having “explained” a given Activation/Inhibition Statement with a combination of more mechanistic Modification + ActiveForm Statements.

        Parameters stmts(list[indra.statements.Statement]) – A list of Statements to
        infer RegulateActivity from.

        Returns linked_stmts – A list of LinkedStatements representing the inferred Statements.

        Return type list[indra.mechlinker.LinkedStatement]

    static infer_active_forms(stmts)
        Return inferred ActiveForm from RegulateActivity + Modification.

        This function looks for combinations of Activation/Inhibition Statements and Modification Statements,
        and infers an ActiveForm from them. For example, if we know that A activates B and A phosphorylates
        B, then we can infer that the phosphorylated form of B is active.

        Parameters stmts(list[indra.statements.Statement]) – A list of Statements to
        infer ActiveForms from.

        Returns linked_stmts – A list of LinkedStatements representing the inferred Statements.

        Return type list[indra.mechlinker.LinkedStatement]

    static infer_complexes(stmts)
        Return inferred Complex from Statements implying physical interaction.
Parameters `stmts` (list[indra.statements.Statement]) – A list of Statements to infer Complexes from.

Returns `linked_stmts` – A list of LinkedStatements representing the inferred Statements.

Return type list[indra.mechlinker.LinkedStatement]

static `infer_modifications` (stmts)
Return inferred Modification from RegulateActivity + ActiveForm.

This function looks for combinations of Activation/Inhibition Statements and ActiveForm Statements that imply a Modification Statement. For example, if we know that A activates B, and phosphorylated B is active, then we can infer that A leads to the phosphorylation of B. An additional requirement when making this assumption is that the activity of B should only be dependent on the modified state and not other context - otherwise the inferred Modification is not necessarily warranted.

Parameters `stmts` (list[indra.statements.Statement]) – A list of Statements to infer Modifications from.

Returns `linked_stmts` – A list of LinkedStatements representing the inferred Statements.

Return type list[indra.mechlinker.LinkedStatement]

reduce_activities()
Rewrite the activity types referenced in Statements for consistency.

Activity types are reduced to the most specific form whenever possible. For instance, if ‘kinase’ is the only specific activity type known for the BaseAgent of BRAF, its generic ‘activity’ forms are rewritten to ‘kinase’.

replace_activations (linked_stmts=None)
Remove RegulateActivity Statements that can be inferred out.

This function iterates over self.statements and looks for RegulateActivity Statements that either match or are refined by inferred RegulateActivity Statements that were linked (provided as the linked_stmts argument). It removes RegulateActivity Statements from self.statements that can be explained by the linked statements.

Parameters `linked_stmts` (Optional[list[indra.mechlinker.LinkedStatement]]) – A list of linked statements, optionally passed from outside. If None is passed, the MechLinker runs self.infer_activations to infer RegulateActivities and obtain a list of LinkedStatements that are then used for removing existing Complexes in self.statements.

replace_complexes (linked_stmts=None)
Remove Complex Statements that can be inferred out.

This function iterates over self.statements and looks for Complex Statements that either match or are refined by inferred Complex Statements that were linked (provided as the linked_stmts argument). It removes Complex Statements from self.statements that can be explained by the linked statements.

Parameters `linked_stmts` (Optional[list[indra.mechlinker.LinkedStatement]]) – A list of linked statements, optionally passed from outside. If None is passed, the MechLinker runs self.infer_complexes to infer Complexes and obtain a list of LinkedStatements that are then used for removing existing Complexes in self.statements.

require_active_forms()
Rewrites Statements with Agents’ active forms in active positions.

As an example, the enzyme in a Modification Statement can be expected to be in an active state. Similarly, subjects of RegulateAmount and RegulateActivity Statements can be expected to be in an active form. This function takes the collected active states of Agents in their corresponding BaseAgents and then rewrites other Statements to apply the active Agent states to them.
Returns new_stmts – A list of Statements which includes the newly rewritten Statements. This list is also set as the internal Statement list of the MechLinker.

Return type list[indra.statements.Statement]

4.8 Assemblers of model output (indra.assemblers)

4.8.1 Executable PySB models (indra.assemblers.pysb_assembler)

class indra.assemblers.pysb_assembler.PysbAssembler (policies=None)

Assembler creating a PySB model from a set of INDRA Statements.

Parameters policies (Optional[Union[str, dict]]) – A string or dictionary that defines one or more assembly policies.

If policies is a string, it defines a global assembly policy that applies to all Statement types. Example: contact_only, one_step

A dictionary of policies has keys corresponding to Statement types and values to the policy to be applied to that type of Statement. For Statement types whose policy is undefined, the ‘default’ policy is applied. Example: {'Phosphorylation': 'two_step'}

policies
dict – A dictionary of policies that defines assembly policies for Statement types. It is assigned in the constructor.

statements
list – A list of INDRA statements to be assembled.

model
pysb.Model – A PySB model object that is assembled by this class.

agent_set
_BaseAgentSet – A set of BaseAgents used during the assembly process.

add_default_initial_conditions (value=None)
Set default initial conditions in the PySB model.

Parameters value (Optional[float]) – Optionally a value can be supplied which will be the initial amount applied. Otherwise a built-in default is used.

add_statements (stmts)
Add INDRA Statements to the assembler’s list of statements.

Parameters stmts (list[indra.statements.Statement]) – A list of indra.statements.Statement to be added to the statement list of the assembler.

export_model (format, file_name=None)
Save the assembled model in a modeling formalism other than PySB.

For more details on exporting PySB models, see http://pysb.readthedocs.io/en/latest/modules/export/index.html

Parameters
• **file_name** *(Optional*[str]*) – An optional file name to save the exported model into.

Returns **exp_str** – The exported model string

Return type  str

**make_model** *(policies=None, initial_conditions=True, reverse_effects=False)*

Assemble the PySB model from the collected INDRA Statements.

This method assembles a PySB model from the set of INDRA Statements. The assembled model is both returned and set as the assembler's model argument.

Parameters

• **policies** *(Optional*[Union*[str, dict]*) – A string or dictionary of policies, as defined in *indra.assemblers.PysbAssembler*. This set of policies locally supersedes the default setting in the assembler. This is useful when this function is called multiple times with different policies.

• **initial_conditions** *(Optional*[bool]*) – If True, default initial conditions are generated for the Monomers in the model.

Returns **model** – The assembled PySB model object.

Return type  pysb.Model

**print_model** ()

Print the assembled model as a PySB program string.

This function is useful when the model needs to be passed as a string to another component.

**save_model** *(file_name=u’pysb_model.py’)*

Save the assembled model as a PySB program file.

Parameters **file_name** *(Optional*[str]*) – The name of the file to save the model program code in. Default: pysb-model.py

**save_rst** *(file_name=u’pysb_model.rst’, module_name=u’pysb_module’)*

Save the assembled model as an RST file for literate modeling.

Parameters

• **file_name** *(Optional*[str]*) – The name of the file to save the RST in. Default: pysb_model.rst

• **module_name** *(Optional*[str]*) – The name of the python function defining the module. Default: pysb_module

**set_context** *(cell_type)*

Set protein expression data as initial conditions.

This method uses *indra.databases.context_client* to get protein expression levels for a given cell type and set initial conditions for Monomers in the model accordingly.

Parameters

• **cell_type** *(str)* – Cell type name for which expression levels are queried. The cell type name follows the CCLE database conventions.

• **Example** *(LOXIMVI_SKIN, BT20_BREAST)* – *indra.assemblers.pysb_assembler.add_rule_to_model*(model, rule, annotations=None)*

Add a Rule to a PySB model and handle duplicate component errors.

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**4.8. Assemblers of model output (indra.assemblers)**

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indra.assemblers.pysb_assembler.complex_monomers_default(stmt, agent_set)
In this (very simple) implementation, proteins in a complex are each given site names corresponding to each of
the other members of the complex (lower case). So the resulting complex can be “fully connected” in that each
member can be bound to all the others.

indra.assemblers.pysb_assembler.complex_monomers_one_step(stmt, agent_set)
In this (very simple) implementation, proteins in a complex are each given site names corresponding to each of
the other members of the complex (lower case). So the resulting complex can be “fully connected” in that each
member can be bound to all the others.

indra.assemblers.pysb_assembler.get_agent_rule_str(agent)
Construct a string from an Agent as part of a PySB rule name.

indra.assemblers.pysb_assembler.get_annotation(component, db_name, db_ref)
Construct model Annotations for each component.
Annotation formats follow guidelines at http://identifiers.org/.

indra.assemblers.pysb_assembler.get_binding_site_name(agent)
Return a binding site name from a given agent.

indra.assemblers.pysb_assembler.get_create_parameter(model, name, value, unique=True)
Return parameter with given name, creating it if needed.
If unique is false and the parameter exists, the value is not changed; if it does not exist, it will be created. If
unique is true then upon conflict a number is added to the end of the parameter name.

indra.assemblers.pysb_assembler.get_mod_site_name(mod_type, residue, position)
Return site names for a modification.

indra.assemblers.pysb_assembler.get_monomer_pattern(model, agent, extra_fields=None)
Construct a PySB MonomerPattern from an Agent.

indra.assemblers.pysb_assembler.get_site_pattern(agent)
Construct a dictionary of Monomer site states from an Agent.
This crates the mapping to the associated PySB monomer from an INDRA Agent object.

indra.assemblers.pysb_assembler.get_uncond_agent(agent)
Construct the unconditional state of an Agent.
The unconditional Agent is a copy of the original agent but without any bound conditions and modification
conditions. Mutation conditions, however, are preserved since they are static.

indra.assemblers.pysb_assembler.grounded_monomer_patterns(model, agent)
Get monomer patterns for the agent accounting for grounding information.

indra.assemblers.pysb_assembler.parse_identifiers_url(url)
Parse an identifiers.org URL into (namespace, ID) tuple.

indra.assemblers.pysb_assembler.set_base_initial_condition(model, monomer, value)
Set an initial condition for a monomer in its ‘default’ state.

indra.assemblers.pysb_assembler.set_extended_initial_condition(model, monomer=None, value=0)
Set an initial condition for monomers in “modified” state.
This is useful when using downstream analysis that relies on reactions being active in the model. One example
is BioNetGen-based reaction network diagram generation.
4.8.2 Cytoscape networks (indra.assemblers.cx_assembler)

class indra.assemblers.cx_assembler.CxAssembler (stmts=None, network_name=u'indra_assembled')

This class assembles a CX network from a set of INDRA Statements.

The CX format is an aspect oriented data mode for networks. The format is defined at http://www.home.ndexbio.org/data-model/. The CX format is the standard for NDEx and is compatible with CytoScape via the CyNDEx plugin.

Parameters

- **stmts** *(Optional[list[indra.statements.Statement]])* – A list of INDRA Statements to be assembled.

- **network_name** *(Optional[str]*) – The name of the network to be assembled. Default: indra_assembled

**statements**

*list[indra.statements.Statement]* – A list of INDRA Statements to be assembled.

**network_name**

*str* – The name of the network to be assembled.

**cx**

*dict* – The structure of the CX network that is assembled.

**add_statements** *( stmts)*

Add INDRA Statements to the assembler’s list of statements.

Parameters **stmts** *(list[indra.statements.Statement])* – A list of indra.statements.Statement to be added to the statement list of the assembler.

**make_model** *(add_indra_json=True)*

Assemble the CX network from the collected INDRA Statements.

This method assembles a CX network from the set of INDRA Statements. The assembled network is set as the assembler’s cx argument.

Parameters **add_indra_json** *(Optional[bool])* – If True, the INDRA Statement JSON annotation is added to each edge in the network. Default: True

Returns **cx_str** – The json serialized CX model.

**print_cx** *(pretty=True)*

Return the assembled CX network as a json string.

Parameters **pretty** *(bool)* – If True, the CX string is formatted with indentation (for human viewing) otherwise no indentation is used.

Returns **json_str** – A json formatted string representation of the CX network.

**save_model** *(file_name=u'model.cx')*

Save the assembled CX network in a file.

Parameters **file_name** *(Optional[str])* – The name of the file to save the CX network to. Default: model.cx
**set_context** *(cell_type)*

Set protein expression data and mutational status as node attribute

This method uses `indra.databases.context_client` to get protein expression levels and mutational status for a given cell type and set a node attribute for proteins accordingly.

**Parameters**

- **cell_type** *(str)* – Cell type name for which expression levels are queried.
  
The cell type name follows the CCLE database conventions. Example: LOXIMVI_Skin, BT20_BREAST

**upload_model** *(ndex_cred)*

Creates a new NDEx network of the assembled CX model.

To upload the assembled CX model to NDEx, you need to have a registered account on NDEx (http://ndexbio.org/) and have the `ndex` python package installed. The uploaded network is private by default.

**Parameters**

- **ndex_cred** *(dict)* – A dictionary with the following entries: ‘user’: NDEx user name ‘password’: NDEx password

**Returns**

- **network_id** – The UUID of the NDEx network that was created by uploading the assembled CX model.

**4.8.3 Natural language (indra.assemblers.english_assembler)**

**class** `indra.assemblers.english_assembler.EnglishAssembler(stmts=None)`

This assembler generates English sentences from INDRA Statements.

**Parameters**

- **stmts** *(Optional[list[indra.statements.Statement]])* – A list of INDRA Statements to be added to the assembler.

**statements**

- **list[indra.statements.Statement]** – A list of INDRA Statements to assemble.

**model**

- **str** – The assembled sentences as a single string.

**add_statements** *(stmts)*

Add INDRA Statements to the assembler’s list of statements.

**Parameters**

- **stmts** *(list[indra.statements.Statement])* – A list of `indra.statements.Statement` to be added to the statement list of the assembler.

**make_model** ()

Assemble text from the set of collected INDRA Statements.

**Returns**

- **stmt_strs** – Return the assembled text as unicode string. By default, the text is a single string consisting of one or more sentences with periods at the end.

**Return type**

- **str**

**4.8.4 Node-edge graphs (indra.assemblers.graph_assembler)**

**class** `indra.assemblers.graph_assembler.GraphAssembler(stmts=None, graph_properties=None, node_properties=None, edge_properties=None)`

The Graph assembler assembles INDRA Statements into a Graphviz node-edge graph.
Parameters

- **stmts** *(Optional[list[indra.statements.Statement]])* – A list of INDRA Statements to be added to the assembler’s list of Statements.

- **graph_properties** *(Optional[dict[str: str]])* – A dictionary of graphviz graph properties overriding the default ones.

- **node_properties** *(Optional[dict[str: str]])* – A dictionary of graphviz node properties overriding the default ones.

- **edge_properties** *(Optional[dict[str: str]])* – A dictionary of graphviz edge properties overriding the default ones.

**statements**

*list[indra.statements.Statement]* – A list of INDRA Statements to be assembled.

**graph**

*pygraphviz.AGraph* – A pygraphviz graph that is assembled by this assembler.

**existing_nodes**

*list[tuple]* – The list of nodes (identified by node key tuples) that are already in the graph.

**existing_edges**

*list[tuple]* – The list of edges (identified by edge key tuples) that are already in the graph.

**graph_properties**

*dict[str: str]* – A dictionary of graphviz graph properties used for assembly.

**node_properties**

*dict[str: str]* – A dictionary of graphviz node properties used for assembly.

**edge_properties**

*dict[str: str]* – A dictionary of graphviz edge properties used for assembly. Note that most edge properties are determined based on the type of the edge by the assembler (e.g. color, arrowhead). These settings cannot be directly controlled through the API.

**add_statements** *(stmts)*

Add a list of statements to be assembled.

**get_string**()

Return the assembled graph as a string.

Returns **graph_string** – The assembled graph as a string.

**Return**

* **type**  
  *str*

**make_model**()

Assemble the graph from the assembler’s list of INDRA Statements.

**save_dot** *(file_name=u’graph.dot’)*

Save the graph in a graphviz dot file.

Parameters **file_name** *(Optional[str])* – The name of the file to save the graph dot string to.

**save_pdf** *(file_name=u’graph.pdf’, prog=u’dot’)*

Draw the graph and save as an image or pdf file.

Parameters
• **file_name** (*Optional*[str]) – The name of the file to save the graph as. Default: graph.pdf

• **prog** (*Optional*[str]) – The graphviz program to use for graph layout. Default: dot

## 4.8.5 SIF / Boolean networks (**indra.assemblers.sif_assembler**)

**class** `indra.assemblers.sif_assembler.SifAssembler` *(stmts=None)*

The SIF assembler assembles INDRA Statements into a networkx graph.

This graph can then be exported into SIF (simple interaction format) or a Boolean network.

**Parameters**

- **stmts** (*Optional*[list[indra.statements.Statement]]) – A list of INDRA Statements to be added to the assembler’s list of Statements.

- **graph**
  - *networkx.DiGraph* – A networkx graph that is assembled by this assembler.

**make_model** *(use_name_as_key=False, include.mods=False, include.complexes=False)*

Assemble the graph from the assembler’s list of INDRA Statements.

**Parameters**

- **use_name_as_key** (*boolean*) – If True, uses the name of the agent as the key to the nodes in the network. If False (default) uses the matches_key() of the agent.

- **include.mods** (*boolean*) – If True, adds Modification statements into the graph as directed edges. Default is False.

- **include.complexes** (*boolean*) – If True, creates two edges (in both directions) between all pairs of nodes in Complex statements. Default is False.

**print_boolean_net** *(out_file=None)*

Return a Boolean network from the assembled graph.

**Parameters**

- **out_file** (*Optional*[str]) – A file name in which the Boolean network is saved.

**Returns** *full_str* – The string representing the Boolean network.

**Return type** *str*

**print_loopy** *(as_url=True)*

Return

**Parameters**

- **out_file** (*Optional*[str]) – A file name in which the Loopy network is saved.

**Returns** *full_str* – The string representing the Loopy network.

**Return type** *str*

**print_model** *(include_unsigned_edges=False)*

Return a SIF string of the assembled model.

**Parameters**

- **include_unsigned_edges** (*bool*) – If True, includes edges with an unknown activating/inactivating relationship (e.g., most PTMs). Default is False.

**save_model** *(fname)*

Save the assembled model’s SIF string into a file.

**Parameters**

- **fname** (*str*) – The name of the file to save the SIF into.
4.8.6 MITRE “index cards” (indra.assemblers.index_card_assembler)

class indra.assemblers.index_card_assembler.IndexCardAssembler (statements=None, pmc_override=None)

Assembler creating index cards from a set of INDRA Statements.

Parameters

- **statements** (list) – A list of INDRA statements to be assembled.
- **pmc_override** (Optional[str]) – A PMC ID to assign to the index card.

**statements**

- list – A list of INDRA statements to be assembled.

**add_statements** (statements)

Add statements to the assembler.

Parameters **statements** (list[indra.statements.Statement]) – The list of INDRA Statements to add to the assembler.

**make_model**()

Assemble statements into index cards.

**print_model**()

Return the assembled cards as a JSON string.

Returns **cards_json** – The JSON string representing the assembled cards.

**save_model** (file_name=u’index_cards.json’)

Save the assembled cards into a file.

Parameters **file_name** (Optional[str]) – The name of the file to save the cards into.

Default: index_cards.json

4.8.7 SBGN output (indra.assemblers.sbgn_assembler)

class indra.assemblers.sbgn_assembler.SBGNAssembler (statements=None)

This class assembles an SBGN model from a set of INDRA Statements.

The Systems Biology Graphical Notation (SBGN) is a widely used graphical notation standard for systems biology models. This assembler creates SBGN models following the Process Description (PD) standard, documented at: https://github.com/sbgn/process-descriptions/blob/master/UserManual/sbgn_PD-level1-user-public.pdf. For more information on SBGN, see: http://sbgn.github.io/sbgn/

Parameters **stmts** (Optional[list[indra.statements.Statement]]) – A list of INDRA Statements to be assembled.

**statements**

- list[indra.statements.Statement] – A list of INDRA Statements to be assembled.

**sbgn**

- lxml.etree.ElementTree – The structure of the SBGN model that is assembled, represented as an XML ElementTree.

**add_statements** (stmts)

Add INDRA Statements to the assembler’s list of statements.
Parameters stmts (list[indra.statements.Statement]) – A list of indra.statements.Statement to be added to the statement list of the assembler.

make_model()
Assemble the SBGN model from the collected INDRA Statements.

This method assembles an SBGN model from the set of INDRA Statements. The assembled model is set as the assembler’s sbgn attribute (it is represented as an XML ElementTree internally). The model is returned as a serialized XML string.

Returns sbgn_str – The XML serialized SBGN model.

Return type str

print_model(pretty=True, encoding=u'utf8')
Return the assembled SBGN model as an XML string.

Parameters pretty (Optional[bool]) – If True, the SBGN string is formatted with indentation (for human viewing) otherwise no indentation is used. Default: True

Returns sbgn_str – An XML string representation of the SBGN model.

Return type bytes (str in Python 2)

save_model(file_name=u'model.sbgn')
Save the assembled SBGN model in a file.

Parameters file_name (Optional[str]) – The name of the file to save the SBGN network to. Default: model.sbgn

4.9 Explanation (indra.explanation)

4.9.1 Check whether a rule-based model satisfies a property (indra.explanation.model_checker)

class indra.explanation.model_checker.ModelChecker(model, statements=..., agent_obs=...)
Check a PySB model against a set of INDRA statements.

Parameters

• model (pysb.Model) – A PySB model to check.

• statements (Optional[list[indra.statements.Statement]]) – A list of INDRA Statements to check the model against.

• agent_obs (Optional[list[indra.statements.Agent]]) – A list of INDRA Agents in a given state to be observed.

add_statements(stmts)
Add to the list of statements to check against the model.

Parameters stmts (list[indra.statements.Statement]) – The list of Statements to be added for checking.

check_model(max_paths=1, max_path_length=5)
Check all the statements added to the ModelChecker.

Parameters
• **max_paths (Optional[int])** – The maximum number of specific paths to return for each Statement to be explained. Default: 1

• **max_path_length (Optional[int])** – The maximum length of specific paths to return. Default: 5

**Returns** Each tuple contains the Statement checked against the model and a PathResult object describing the results of model checking.

**Return type** list of (Statement, PathResult)

**check_statement (stmt, max_paths=1, max_path_length=5)**

Check a single Statement against the model.

**Parameters**

• **stmt (indra.statements.Statement)** – The Statement to check.

• **max_paths (Optional[int])** – The maximum number of specific paths to return for each Statement to be explained. Default: 1

• **max_path_length (Optional[int])** – The maximum length of specific paths to return. Default: 5

**Returns** True if the model satisfies the Statement.

**Return type** boolean

**get_im (force_update=False)**

Get the influence map for the model, generating it if necessary.

**Parameters**

• **force_update (bool)** – Whether to generate the influence map when the function is called. If False, returns the previously generated influence map if available. Defaults to True.

**Returns**

The influence map can be rendered as a pdf using the dot layout program as follows:

```python
influence_map.draw('influence_map.pdf', prog='dot')
```

**Return type** pygraphviz AGraph object containing the influence map.

**class** indra.explanation.model_checker.PathMetric(source_node, target_node, polarity, length)

Describes results of simple path search (path existence).

**class** indra.explanation.model_checker.PathResult(path_found, result_code, max_paths, max_path_length)

Describes results of running the ModelChecker on a single Statement.

**Parameters**

• **path_found (bool)** –

• **result_code (string)** – STATEMENT_TYPE_NOTHandled SUBJECT_MONOMERS_NOTFOUND OBSERVABLES_NOTFOUND NO_PATHS_FOUND MAX_PATH_LENGTH_EXCEEDED PATHS_FOUND INPUT_RULES_NOTFOUND

path_found

```python
boolean
```

result_code

```python
string
```
4.10 Tools (indra.tools)

4.10.1 Run assembly components in a pipeline (indra.tools.assemble_corpus)

*indra.tools.assemble_corpus.dump_statements*(stmts, fname)

Dump a list of statements into a pickle file.

**Parameters**

- **fname** (str) – The name of the pickle file to dump statements into.

*indra.tools.assemble_corpus.dump_stmt_strings*(stmts, fname)

Save printed statements in a file.

**Parameters**

- **stmts_in** (list[indra.statements.Statement]) – A list of statements to save in a text file.
- **fname** (Optional[str]) – The name of a text file to save the printed statements into.

*indra.tools.assemble_corpus.expand_families*(stmts_in, **kwargs)

Expand Bioentities Agents to individual genes.

**Parameters**

- **stmts_in** (list[indra.statements.Statement]) – A list of statements to expand.
- **save** (Optional[str]) – The name of a pickle file to save the results (stmts_out) into.

**Returns**

**stmts_out** – A list of expanded statements.

**Return type**

list[indra.statements.Statement]

*indra.tools.assemble_corpus.filter_belief*(stmts_in, belief_cutoff, **kwargs)

Filter to statements with belief above a given cutoff.

**Parameters**

- **stmts_in** (list[indra.statements.Statement]) – A list of statements to filter.
- **belief_cutoff** (float) – Only statements with belief above the belief_cutoff will be returned. Here 0 < belief_cutoff < 1.
- **save** (Optional[str]) – The name of a pickle file to save the results (stmts_out) into.

**Returns**

**stmts_out** – A list of filtered statements.

**Return type**

list[indra.statements.Statement]

*indra.tools.assemble_corpus.filter_by_type*(stmts_in, stmt_type, **kwargs)

Filter to a given statement type.
Parameters

- **stmts_in (list[indra.statements.Statement])** – A list of statements to filter.
- **stmt_type (indra.statements.Statement)** – The class of the statement type to filter for. Example: indra.statements.Modification
- **invert (Optional[bool])** – If True, the statements that are not of the given type are returned. Default: False
- **save (Optional[str])** – The name of a pickle file to save the results (stmts_out) into.

Returns **stmts_out** – A list of filtered statements.

Return type list[indra.statements.Statement]

```
indra.tools.assemble_corpus.filter_direct(stmts_in, **kwargs)
```

Filter to statements that are direct interactions

Parameters

- **stmts_in (list[indra.statements.Statement])** – A list of statements to filter.
- **save (Optional[str])** – The name of a pickle file to save the results (stmts_out) into.

Returns **stmts_out** – A list of filtered statements.

Return type list[indra.statements.Statement]

```
indra.tools.assemble_corpus.filter_enzyme_kinase(stmts_in, **kwargs)
```

Filter Phosphorylations to ones where the enzyme is a known kinase.

Parameters

- **stmts_in (list[indra.statements.Statement])** – A list of statements to filter.
- **save (Optional[str])** – The name of a pickle file to save the results (stmts_out) into.

Returns **stmts_out** – A list of filtered statements.

Return type list[indra.statements.Statement]

```
indra.tools.assemble_corpus.filter_evidence_source(stmts_in, source_apis, policy=u'one', **kwargs)
```

Filter to statements that have evidence from a given set of sources.

Parameters

- **stmts_in (list[indra.statements.Statement])** – A list of statements to filter.
- **source_apis (list[str])** – A list of sources to filter for. Examples: biopax, bel, reach
- **policy (Optional[str])** – If ‘one’, a statement that has evidence from any of the sources is kept. If ‘all’, only those statements are kept which have evidence from all the input sources specified in source_apis. If ‘none’, only those statements are kept that don’t have evidence from any of the sources specified in source_apis.
- **save (Optional[str])** – The name of a pickle file to save the results (stmts_out) into.

Returns **stmts_out** – A list of filtered statements.

Return type list[indra.statements.Statement]
Return statements that contain genes given in a list.

Parameters

- **stmts_in** *(list[indra.statements.Statement])* – A list of statements to filter.
- **gene_list** *(list[str])* – A list of gene symbols to filter for.
- **policy** *(str)* – The policy to apply when filtering for the list of genes. “one”: keep statements that contain at least one of the list of genes and possibly others not in the list “all”: keep statements that only contain genes given in the list
- **allow_families** *(Optional[bool])* – Will include statements involving Bioentities families containing one of the genes in the gene list. Default: False
- **save** *(Optional[str])* – The name of a pickle file to save the results (stmts_out) into.

Returns **stmts_out** – A list of filtered statements.

Return type list[indra.statements.Statement]

Filter to statements containing genes only.

Parameters

- **stmts_in** *(list[indra.statements.Statement])* – A list of statements to filter.
- **specific_only** *(Optional[bool])* – If True, only elementary genes/proteins will be kept and families will be filtered out. If False, families are also included in the output. Default: False
- **save** *(Optional[str])* – The name of a pickle file to save the results (stmts_out) into.

Returns **stmts_out** – A list of filtered statements.

Return type list[indra.statements.Statement]

Filter to statements that have grounded agents.

Parameters

- **stmts_in** *(list[indra.statements.Statement])* – A list of statements to filter.
- **save** *(Optional[str])* – The name of a pickle file to save the results (stmts_out) into.

Returns **stmts_out** – A list of filtered statements.

Return type list[indra.statements.Statement]

Filter out statements that are not grounded to human genes.

Parameters

- **stmts_in** *(list[indra.statements.Statement])* – A list of statements to filter.
- **save** *(Optional[str])* – The name of a pickle file to save the results (stmts_out) into.

Returns **stmts_out** – A list of filtered statements.
Filter out Activations that modify inconsequential activities

Inconsequential here means that the site is not mentioned / tested in any other statement. In some cases specific activity types should be preserved, for instance, to be used as readouts in a model. In this case, the given activities can be passed in a whitelist.

Parameters

- `stmts_in (list[indra.statements.Statement])` – A list of statements to filter.
- `whitelist (Optional[dict])` – A whitelist containing agent activity types which should be preserved even if no other statement refers to them. The whitelist parameter is a dictionary in which the key is a gene name and the value is a list of activity types. Example: `whitelist = {'MAP2K1': ['kinase']}`
- `save (Optional[str])` – The name of a pickle file to save the results (stmts_out) into.

Returns `stmts_out` – A list of filtered statements.

Filter out Modifications that modify inconsequential sites

Inconsequential here means that the site is not mentioned / tested in any other statement. In some cases specific sites should be preserved, for instance, to be used as readouts in a model. In this case, the given sites can be passed in a whitelist.

Parameters

- `stmts_in (list[indra.statements.Statement])` – A list of statements to filter.
- `whitelist (Optional[dict])` – A whitelist containing agent modification sites whose modifications should be preserved even if no other statement refers to them. The whitelist parameter is a dictionary in which the key is a gene name and the value is a list of tuples of (modification_type, residue, position). Example: `whitelist = {'MAP2K1': [('phosphorylation', 'S', '222')]}`
- `save (Optional[str])` – The name of a pickle file to save the results (stmts_out) into.

Returns `stmts_out` – A list of filtered statements.

Filter non-phospho Modifications to ones with a non-kinase enzyme.

Parameters

- `stmts_in (list[indra.statements.Statement])` – A list of statements to filter.
- `save (Optional[str])` – The name of a pickle file to save the results (stmts_out) into.

Returns `stmts_out` – A list of filtered statements.
indra.tools.assemble_corpus.filter_mutation_status(stmts_in, mutations, deletions, **kwargs)

Filter statements based on existing mutations/deletions

This filter helps to contextualize a set of statements to a given cell type. Given a list of deleted genes, it removes statements that refer to these genes. It also takes a list of mutations and removes statements that refer to mutations not relevant for the given context.

Parameters

• stmts_in (list[indra.statements.Statement]) – A list of statements to filter.
• mutations (dict) – A dictionary whose keys are gene names, and the values are lists of tuples of the form (residue_from, position, residue_to). Example: mutations = {'BRAF': [('V', '600', 'E')]}
• deletions (list) – A list of gene names that are deleted.
• save (Optional[str]) – The name of a pickle file to save the results (stmts_out) into.

Returns stmts_out – A list of filtered statements.

Return type list[indra.statements.Statement]

indra.tools.assemble_corpus.filter_no_hypothesis(stmts_in, **kwargs)

Filter to statements that are not marked as hypothesis in epistemics.

Parameters

• stmts_in (list[indra.statements.Statement]) – A list of statements to filter.
• save (Optional[str]) – The name of a pickle file to save the results (stmts_out) into.

Returns stmts_out – A list of filtered statements.

Return type list[indra.statements.Statement]

indra.tools.assemble_corpus.filter_top_level(stmts_in, **kwargs)

Filter to statements that are at the top-level of the hierarchy.

Here top-level statements correspond to most specific ones.

Parameters

• stmts_in (list[indra.statements.Statement]) – A list of statements to filter.
• save (Optional[str]) – The name of a pickle file to save the results (stmts_out) into.

Returns stmts_out – A list of filtered statements.

Return type list[indra.statements.Statement]

indra.tools.assemble_corpus.filter_transcription_factor(stmts_in, **kwargs)

Filter out RegulateAmounts where subject is not a transcription factor.

Parameters

• stmts_in (list[indra.statements.Statement]) – A list of statements to filter.
• save (Optional[str]) – The name of a pickle file to save the results (stmts_out) into.

Returns stmts_out – A list of filtered statements.

Return type list[indra.statements.Statement]
**filter_uuid_list** *(stmts_in, uuids, **kwargs)*  
Filter to Statements corresponding to given UUIDs  

**Parameters**  
- **stmts_in** *(list[indra.statements.Statement])* – A list of statements to filter.  
- **uuids** *(list[str])* – A list of UUIDs to filter for.  
- **save** *(Optional[str])* – The name of a pickle file to save the results (stmts_out) into.  

**Returns**  
- **stmts_out** – A list of filtered statements.  

**Return type**  
list[indra.statements.Statement]  

**load_statements** *(fname, as_dict=False)*  
Load statements from a pickle file.  

**Parameters**  
- **fname** *(str)* – The name of the pickle file to load statements from.  
- **as_dict** *(Optional[bool])* – If True and the pickle file contains a dictionary of statements, it is returned as a dictionary. If False, the statements are always returned in a list. Default: False  

**Returns**  
- **stmts** – A list or dict of statements that were loaded.  

**Return type**  
list  

**map_grounding** *(stmts_in, **kwargs)*  
Map grounding using the GroundingMapper.  

**Parameters**  
- **stmts_in** *(list[indra.statements.Statement])* – A list of statements to map.  
- **do_rename** *(Optional[bool])* – If True, Agents are renamed based on their mapped grounding.  
- **save** *(Optional[str])* – The name of a pickle file to save the results (stmts_out) into.  

**Returns**  
- **stmts_out** – A list of mapped statements.  

**Return type**  
list[indra.statements.Statement]  

**map_sequence** *(stmts_in, **kwargs)*  
Map sequences using the SiteMapper.  

**Parameters**  
- **stmts_in** *(list[indra.statements.Statement])* – A list of statements to map.  
- **do_methionine_offset** *(boolean)* – Whether to check for off-by-one errors in site position (possibly) attributable to site numbering from mature proteins after cleavage of the initial methionine. If True, checks the reference sequence for a known modification at 1 site position greater than the given one; if there exists such a site, creates the mapping. Default is True.  
- **do_orthology_mapping** *(boolean)* – Whether to check sequence positions for known modification sites in mouse or rat sequences (based on PhosphoSitePlus data). If a mouse/rat site is found that is linked to a site in the human reference sequence, a mapping is created. Default is True.
• **do_isoform_mapping**: (boolean) – Whether to check sequence positions for known modifications in other human isoforms of the protein (based on PhosphoSitePlus data). If a site is found that is linked to a site in the human reference sequence, a mapping is created. Default is True.

• **save**: (Optional[str]) – The name of a pickle file to save the results (stmts_out) into.

**Returns**
stmts_out – A list of mapped statements.

**Return type**
list[indra.statements.Statement]

---

*indra.tools.assemble_corpus.reduce_activities*(stmts_in, **kwargs)

Reduce the activity types in a list of statements

**Parameters**

• **stmts_in**: (list[indra.statements.Statement]) – A list of statements to reduce activity types in.

• **save**: (Optional[str]) – The name of a pickle file to save the results (stmts_out) into.

**Returns**
stmts_out – A list of reduced activity statements.

**Return type**
list[indra.statements.Statement]

---

*indra.tools.assemble_corpus.run_preassembly*(stmts_in, **kwargs)

Run preassembly on a list of statements.

**Parameters**

• **stmts_in**: (list[indra.statements.Statement]) – A list of statements to preassemble.

• **return_toplevel**: (Optional[bool]) – If True, only the top-level statements are returned. If False, all statements are returned irrespective of level of specificity. Default: True

• **poolsize**: (Optional[int]) – The number of worker processes to use to parallelize the comparisons performed by the function. If None (default), no parallelization is performed. NOTE: Parallelization is only available on Python 3.4 and above.

• **size_cutoff**: (Optional[int]) – Groups with size_cutoff or more statements are sent to worker processes, while smaller groups are compared in the parent process. Default value is 100. Not relevant when parallelization is not used.

• **save**: (Optional[str]) – The name of a pickle file to save the results (stmts_out) into.

• **save_unique**: (Optional[str]) – The name of a pickle file to save the unique statements into.

**Returns**
stmts_out – A list of preassembled top-level statements.

**Return type**
list[indra.statements.Statement]

---

*indra.tools.assemble_corpus.run_preassembly_duplicate*(preassembler, beliefengine, **kwargs)

Run deduplication stage of preassembly on a list of statements.

**Parameters**

• **preassembler**: (indra.preassembler.Preassembler) – A Preassembler instance

• **beliefengine**: (indra.belief.BeliefEngine) – A BeliefEngine instance

• **save**: (Optional[str]) – The name of a pickle file to save the results (stmts_out) into.
Returns `stmts_out` – A list of unique statements.

Return type `list[indra.statements.Statement]`

```python
indra.tools.assemble_corpus.run_preassembly_related(preassembler, beliefengine, **kwargs)
```

Run related stage of preassembly on a list of statements.

Parameters

- `preassembler` (*indra.preassembler.Preassembler*) – A Preassembler instance which already has a set of unique statements internally.
- `beliefengine` (*indra.belief.BeliefEngine*) – A BeliefEngine instance
- `return_toplevel` (*Optional[bool]*) – If True, only the top-level statements are returned. If False, all statements are returned irrespective of level of specificity. Default: True
- `poolsize` (*Optional[int]*) – The number of worker processes to use to parallelize the comparisons performed by the function. If None (default), no parallelization is performed. NOTE: Parallelization is only available on Python 3.4 and above.
- `size_cutoff` (*Optional[int]*) – Groups with size_cutoff or more statements are sent to worker processes, while smaller groups are compared in the parent process. Default value is 100. Not relevant when parallelization is not used.
- `save` (*Optional[str]*) – The name of a pickle file to save the results (stmts_out) into.

Returns `stmts_out` – A list of preassembled top-level statements.

Return type `list[indra.statements.Statement]`

```python
indra.tools.assemble_corpus.strip_agent_context(stmts_in, **kwargs)
```

Strip any context on agents within each statement.

Parameters

- `stmts_in` (*list[indra.statements.Statement]*) – A list of statements whose agent context should be stripped.
- `save` (*Optional[str]*) – The name of a pickle file to save the results (stmts_out) into.

Returns `stmts_out` – A list of stripped statements.

Return type `list[indra.statements.Statement]`

### 4.10.2 Build a network from a gene list (`indra.tools.gene_network`)

```python
class indra.tools.gene_network.GeneNetwork(gene_list, basename=None)
```

Build a set of INDRA statements for a given gene list from databases.

Parameters

- `gene_list` (*string*) – List of gene names.
- `basename` (*string or None (default]*) – Filenname prefix to be used for caching of intermediates (Biopax OWL file, pickled statement lists, etc.). If None, no results are cached and no cached files are used.

```python
gene_list
string – List of gene names
```
basename

`string or None` – Filename prefix for cached intermediates, or None if no cached used.

results

`dict` – Dict containing results of preassembly (see return type for `run_preassembly()`).

get_bel_stmts `(filter=False)`

Get relevant statements from the BEL large corpus.

Performs a series of neighborhood queries and then takes the union of all the statements. Because the query process can take a long time for large gene lists, the resulting list of statements are cached in a pickle file with the filename `<basename>_bel_stmts.pkl`. If the pickle file is present, it is used by default; if not present, the queries are performed and the results are cached.

**Parameters**

- `filter (bool)` – If True, includes only those statements that exclusively mention genes in `gene_list`. Default is False. Note that the full (unfiltered) set of statements are cached.

**Returns** List of INDRA statements extracted from the BEL large corpus.

**Return type** list of `indra.statements.Statement`

get_biopax_stmts `(filter=False, query='pathsbetween')`

Get relevant statements from Pathway Commons.

Performs a “paths between” query for the genes in `gene_list` and uses the results to build statements. This function caches two files: the list of statements built from the query, which is cached in `<basename>_biopax_stmts.pkl`, and the OWL file returned by the Pathway Commons Web API, which is cached in `<basename>_pc_pathsbetween.owl`. If these cached files are found, then the results are returned based on the cached file and Pathway Commons is not queried again.

**Parameters**

- `filter (bool)` – If True, includes only those statements that exclusively mention genes in `gene_list`. Default is False.
- `query (str)` – Defined what type of query is executed. The two options are ‘pathsbetween’ which finds paths between the given list of genes and only works if more than 1 gene is given, and ‘neighborhood’ which searches the immediate neighborhood of each given gene.

**Returns** List of INDRA statements extracted from Pathway Commons.

**Return type** list of `indra.statements.Statement`

get_statements `(filter=False)`

Return the combined list of statements from BEL and Pathway Commons.

Internally calls `get_biopax_stmts()` and `get_bel_stmts()`.

**Parameters**

- `filter (bool)` – If True, includes only those statements that exclusively mention genes in `gene_list`. Default is False.

**Returns** List of INDRA statements extracted the BEL large corpus and Pathway Commons.

**Return type** list of `indra.statements.Statement`

run_preassembly `(stmts, print_summary=True)`

Run complete preassembly procedure on the given statements.

Results are returned as a dict and stored in the attribute `results`. They are also saved in the pickle file `<basename>_results.pkl`.

**Parameters**
• **stmts** (list of `indra.statements.Statement`) – Statements to preassemble.

• **print_summary** (bool) – If True (default), prints a summary of the preassembly process to the console.

Returns
A dict containing the following entries:

• **raw** – the starting set of statements before preassembly.

• **duplicates1** – statements after initial de-duplication.

• **valid** – statements found to have valid modification sites.

• **mapped** – mapped statements (list of `indra.preassembler.sitemapper.MappedStatement`).

• **mapped_stmts** – combined list of valid statements and statements after mapping.

• **duplicates2** – statements resulting from de-duplication of the statements in `mapped_stmts`.

• **related2** – top-level statements after combining the statements in `duplicates2`.

**Return type** dict

### 4.10.3 Build an executable model from a fragment of a large network (indra.tools.executable_subnetwork)

```
indra.tools.executable_subnetwork.get_subnetwork(statements, nodes, relevance_network=None, relevance_node_lim=10)
```

Return a PySB model based on a subset of given INDRA Statements.

Statements are first filtered for nodes in the given list and other nodes are optionally added based on relevance in a given network. The filtered statements are then assembled into an executable model using INDRA's PySB Assembler.

**Parameters**

• **statements** (list of `indra.statements.Statement`) – A list of INDRA Statements to extract a subnetwork from.

• **nodes** (list of str) – The names of the nodes to extract the subnetwork for.

• **relevance_network** (Optional[str]) – The UUID of the NDEx network in which nodes relevant to the given nodes are found.

• **relevance_node_lim** (Optional[int]) – The maximal number of additional nodes to add to the subnetwork based on relevance.

**Returns** model – A PySB model object assembled using INDRA’s PySB Assembler from the INDRA Statements corresponding to the subnetwork.

**Return type** pysb.Model

### 4.10.4 Build a model incrementally over time (indra.tools.incremental_model)

```
class indra.tools.incremental_model.IncrementalModel(model_fname=None)
```

Assemble a model incrementally by iteratively adding new Statements.
Parameters `model_fname` *(Optional [str]*) – The name of the pickle file in which a set of INDRA Statements are stored in a dict keyed by PubMed IDs. This is the state of an IncrementalModel that is loaded upon instantiation.

`stmts`

`dict[str, list[indra.statements.Statement]]` – A dictionary of INDRA Statements keyed by PMIDs that stores the current state of the IncrementalModel.

`assembled_stmts`

`list[indra.statements.Statement]` – A list of INDRA Statements after assembly.

`add_statements (pmid, stmts)`

Add INDRA Statements to the incremental model indexed by PMID.

Parameters

- `pmid` *(str)* – The PMID of the paper from which statements were extracted.
- `stmts` *(list[indra.statements.Statement]*) – A list of INDRA Statements to be added to the model.

`get_model_agents ()`

Return a list of all Agents from all Statements.

Returns `agents` – A list of Agents that are in the model.

Return type `list[indra.statements.Agent]`

`get_statements ()`

Return a list of all Statements in a single list.

Returns `stmts` – A list of all the INDRA Statements in the model.

Return type `list[indra.statements.Statement]`

`get_statements_noprior ()`

Return a list of all non-prior Statements in a single list.

Returns `stmts` – A list of all the INDRA Statements in the model (excluding the prior).

Return type `list[indra.statements.Statement]`

`get_statements_prior ()`

Return a list of all prior Statements in a single list.

Returns `stmts` – A list of all the INDRA Statements in the prior.

Return type `list[indra.statements.Statement]`

`load_prior (prior_fname)`

Load a set of prior statements from a pickle file.

The prior statements have a special key in the stmts dictionary called “prior”.

Parameters `prior_fname` *(str)* – The name of the pickle file containing the prior Statements.

`preassemble (filters=None)`

Preassemble the Statements collected in the model.

Use INDRA’s GroundingMapper, Preassembler and BeliefEngine on the IncrementalModel and save the unique statements and the top level statements in class attributes.

Currently the following filter options are implemented: - grounding: require that all Agents in statements are grounded - human_only: require that all proteins are human proteins - prior_one: require that at least one Agent is in the prior model - prior_all: require that all Agents are in the prior model.
Parameters filters (Optional[list[str]]) – A list of filter options to apply when choosing the statements. See description above for more details. Default: None

save (model_fname=u'model.pkl')
Save the state of the IncrementalModel in a pickle file.

Parameters model_fname (Optional[str]) – The name of the pickle file to save the state of the IncrementalModel in. Default: model.pkl
5.1 Using natural language to build models

In this tutorial we build a simple model using natural language, and export it into different formats.

5.1.1 Read INDRA Statements from a natural language string

First we import INDRA’s API to the TRIPS reading system. We then define a block of text which serves as the description of the mechanism to be modeled in the `model_text` variable. Finally, `indra.sources.trips.process_text` is called which sends a request to the TRIPS web service, gets a response and processes the extraction knowledge base to obtain a list of INDRA Statements.

```python
In [1]: from indra.sources import trips
In [2]: model_text = 'MAP2K1 phosphorylates MAPK1 and DUSP6 dephosphorylates MAPK1.'
In [3]: tp = trips.process_text(model_text)
```

At this point `tp.statements` should contain 2 INDRA Statements: a Phosphorylation Statement and a Dephosphorylation Statement. Note that the evidence sentence for each Statement is propagated:

```python
In [4]: for st in tp.statements:
   ...:     print('%s with evidence %s' % (st, st.evidence[0].text))
   ...
Phosphorylation(MAP2K1(), MAPK1()) with evidence "MAP2K1 phosphorylates MAPK1 and DUSP6 dephosphorylates MAPK1."
Dephosphorylation(DUSP6(), MAPK1()) with evidence "MAP2K1 phosphorylates MAPK1 and DUSP6 dephosphorylates MAPK1."
```

5.1.2 Assemble the INDRA Statements into a rule-based executable model

We next use INDRA’s PySB Assembler to automatically assemble a rule-based model representing the biochemical mechanisms described in `model_text`. First a PysbAssembler object is instantiated, then the list of INDRA Statements is added to the assembler. Finally, the assembler’s `make_model` method is called which assembles the model and returns it, while also storing it in `pa.model`. Notice that we are using `policies='two_step'` as an argument of `make_model`. This directs the assemble to use rules in which enzymatic catalysis is modeled as a two-step process in which enzyme and substrate first reversibly bind and the enzyme-substrate complex produces and releases a product irreversibly.

```python
In [5]: from indra.assemblers.pysb_assembler import PysbAssembler
```
In [6]: pa = PysbAssembler()

In [7]: pa.add_statements(tp.statements)

In [8]: pa.make_model(policies='two_step')
Out[8]: <Model 'None' (monomers: 3, rules: 6, parameters: 9, expressions: 0, compartments: 0) at 0x7f1e30e21610>

At this point `pa.model` contains a PySB model object with 3 monomers,

In [9]: for monomer in pa.model.monomers:
   ....:     print(monomer)
   ....:
Monomer(u'DUSP6', [u'mapk1'])
Monomer(u'MAP2K1', [u'mapk1'])
Monomer(u'MAPK1', [u'phospho', u'map2k1', u'dusp6'], {u'phospho': [u'u', u'p']})

6 rules,

In [10]: for rule in pa.model.rules:
   ....:     print(rule)
   ....:
Rule(u'MAP2K1_phosphorylation_bind_MAPK1_phospho', MAP2K1(mapk1=None) +
    MAPK1(phospho=u'u', map2k1=None) >> MAP2K1(mapk1=1) % MAPK1(phospho=u'u',
    map2k1=1), kf_mm_bind_1)
Rule(u'MAP2K1_phosphorylation_MAPK1_phospho', MAP2K1(mapk1=1) % MAPK1(phospho=u'u',
    map2k1=1) >> MAP2K1(mapk1=None) + MAPK1(phospho=u'p', map2k1=None), kc_mm_ 
    phosphorylation_1)
Rule(u'MAP2K1_dissoc_MAPK1', MAP2K1(mapk1=1) % MAPK1(map2k1=1) >> MAP2K1(mapk1=None)
    + MAPK1(map2k1=None), kr_mm_bind_1)
Rule(u'DUSP6_dephosphorylation_bind_MAPK1_phospho', DUSP6(mapk1=None) +
    MAPK1(phospho=u'p', dusp6=None) >> DUSP6(mapk1=1) % MAPK1(phospho=u'p',
    dusp6=1), kf_dm_bind_1)
Rule(u'DUSP6_dephosphorylation_MAPK1_phospho', DUSP6(mapk1=1) % MAPK1(phospho=u'p',
    dusp6=None) >> DUSP6(mapk1=None) + MAPK1(phospho=u'u', dusp6=None), kc_dm_ 
    dephosphorylation_1)
Rule(u'DUSP6_dissoc_MAPK1', DUSP6(mapk1=1) % MAPK1(dusp6=1) >> DUSP6(mapk1=None) +
    MAPK1(dusp6=None), kr_dm_bind_1)

and 9 parameters (6 kinetic rate constants and 3 total protein amounts) that are set to nominal but plausible values,

In [11]: for parameter in pa.model.parameters:
   ....:     print(parameter)
   ....:
Parameter(u'kf_mm_bind_1', 1e-06)
Parameter(u'kr_mm_bind_1', 0.1)
Parameter(u'kc_mm_phosphorylation_1', 100.0)
Parameter(u'kf_dm_bind_1', 1e-06)
Parameter(u'kr_dm_bind_1', 0.1)
Parameter(u'kc_dm_dephosphorylation_1', 100.0)
Parameter(u'DUSP6_0', 10000.0)
Parameter(u'MAP2K1_0', 10000.0)
Parameter(u'MAPK1_0', 10000.0)

The model also contains extensive annotations that tie the monomers to database identifiers and also annotate the semantics of each component of each rule.
5.1.3 Exporting the model into other common formats

From the assembled PySB format it is possible to export the model into other common formats such as SBML, BNGL and Kappa. One can also generate a Matlab or Mathematica script with ODEs corresponding to the model.

```python
pa.export_model('sbml')
pa.export_model('bngl')
```

One can also pass a file name argument to the `export_model` function to save the exported model directly into a file:

```python
pa.export_model('sbml', 'example_model.sbml')
```

5.2 Large-Scale Machine Reading with Starcluster

The following doc describes the steps involved in reading a large numbers of papers in parallel on Amazon EC2 using REACH, caching the JSON output on Amazon S3, then processing the REACH output into INDRA Statements. Prerequisites for doing the following are:

- A cluster of Amazon EC2 nodes configured using Starcluster, with INDRA installed and in the PYTHONPATH
- An Amazon S3 bucket containing full text contents for papers, keyed by Pubmed ID (creation of this S3 repository will be described in another tutorial).

This tutorial goes through the individual steps involved before describing how all of them can be run through the use of a single submission script, `submit_reading_pipeline.py`.
Note also that the prerequisite installation steps can be streamlined by putting them in a setup script that can be re-run upon instantiating a new Amazon cluster or by using them to configure a custom Amazon EC2 AMI.

### 5.2.1 Install REACH

Install SBT. On an EC2 Linux machine, run the following lines (drawn from [http://www.scala-sbt.org/0.13/docs/Installing-sbt-on-Linux.html](http://www.scala-sbt.org/0.13/docs/Installing-sbt-on-Linux.html)):

```bash
sudo apt-get update
sudo apt-get install sbt
```

Clone REACH from [https://github.com/clulab/reach](https://github.com/clulab/reach).

Add the following line to reach/build.sbt:

```scala
mainClass in assembly := Some("org.clulab.reach.ReachCLI")
```

This assigns ReachCLI as the main class.

Compile and assemble REACH. Note that the path to the .ivy2 directory must be given. Use the assembly task to assemble a fat JAR containing all of the dependencies with the correct main class. Run the following from the directory containing the REACH build.sbt file (e.g., `/pmc/reach`).:

```bash
sbt -Dsbt.ivy.home=/pmc/reach/.ivy2 compile
sbt -Dsbt.ivy.home=/pmc/reach/.ivy2 assembly
```

### 5.2.2 Install Amazon S3 support

Install boto3:

```bash
pip install boto3
```

**Note:** If using EC2, make sure to install boto3,jsonpickle, and Amazon credentials on all nodes, not just the master node.

Add Amazon credentials to access the S3 bucket. First create the .aws directory on the EC2 instance:

```bash
mkdir /home/sgeadmin/.aws
```

Then set up Amazon credentials, for example by copying from your local machine using StarCluster:

```bash
starcluster put mycluster ~/.aws/credentials /home/sgeadmin/.aws
```

### 5.2.3 Install other dependencies

```bash
pip install jsonpickle # Necessary to process JSON from S3
pip install --upgrade jnius-indra # Necessary for REACH
export JAVA_HOME=/usr/lib/jvm/java-1.7.0-openjdk-amd64
```
5.2.4 Assemble a Corpus of PMIDs

The first step in large-scale reading is to put together a file containing relevant Pubmed IDs. The simplest way to do this is to use the Pubmed search API to find papers associated with particular gene names, biological processes, or other search terms.

For example, to assemble a list of papers for SOS2 curated in Entrez Gene that are available in the Pubmed Central Open Access subset:

```python
In [1]: from indra.literature import *

# Pick an example gene
In [2]: gene = 'SOS2'

# Get a list of PMIDs for the gene
In [3]: pmids = pubmed_client.get_ids_for_gene(gene)

# Get the PMIDs that have XML in PMC
In [4]: pmids_oa_xml = pmc_client.filter_pmids(pmids, 'oa_xml')

# Write the results to a file
In [5]: with open('%s_pmids.txt' % gene, 'w') as f:
...:   for pmid in pmids_oa_xml:
...:     f.write('%s
' % pmid)
```

This creates a file, SOS2_pmids.txt, containing the PMIDs that we will read with REACH.

5.2.5 Process the papers with REACH

The next step is to read the content of the papers with REACH in a parallelizable, high-throughput way. To do this, run the script indra/tools/reading/run_reach_on_pmids.py. If necessary update the lines at the top of the script with the REACH settings, e.g.:

```python
cleanup = False
verbose = True
path_to_reach = '/pmc/reach/target/scala-2.11/reach-assembly-1.3.2-SNAPSHOT.jar'
reach_version = '1.3.2'
source_text = 'pmc_oa_xml'
force_read = False
```

The reach_version is important because it is used to determine whether the paper has already been read with this version of REACH (in which case it will be skipped), or if the REACH output needs to be updated. Alternatively, if you want to read all the papers regardless of whether they’ve been read before with the given version of REACH, set the force_read variable to True.

Next, create a top-level temporary directory to use during reading. This will be used to store the input files and the JSON output:

```bash
mkdir my_temp_dir
```

Run run_reach_on_pmids.py, passing arguments for the PMID list file, the temp directory, the number of cores to use on the machine, the PMID start index (in the PMID list file) and the end index. The start and end indices are used to subdivide the job into parallelizable chunks. If the end index is greater than the total number of PMIDs, it will process up to the last one in the list. For example:
This uses 8 cores to process the first ten papers listed in the file SOS2_pmids.txt. REACH will run, output the JSON files in the temporary directory, e.g. in my_temp_dir/read_0_to_10_MSP6YI/output, assemble the JSON files together, and upload the results to S3. If you attempt to process the files again with the same version of REACH, the script will detect that the JSON output from that version is already on S3 and skip those papers.

This can be submitted to run offline using the job scheduler on EC2 with, e.g.:

```
qsub -b y -cwd -V -pe orte 8 python run_reach_on_pmids.py SOS2_pmids.txt my_temp_dir
```

**Note:** The number of cores requested in the qsub call (‘-pe orte 8’) should match the number of cores passed to the run_reach_on_pmids.py script, which determines the number of threads that REACH will attempt to use (the third-to-last argument above). This should also match the total number of nodes on the Amazon EC2 node (e.g., 8 cores for c3.2xlarge). This way the job scheduler will schedule the job to run on all the cores of a single EC2 node, and REACH will use them all.

### 5.2.6 Extract INDRA Statements from the REACH output on S3

The script indra/tools/reading/process_reach_from_s3.py is used to extract INDRA Statements from the REACH output uploaded to S3 in the previous step. This process can also be parallelized by submitting chunks of papers to be processed by different cores. The INDRA statements for each chunk of papers are pickled and can be assembled into a single pickle file in a subsequent step.

Following the example above, run the following to process the REACH output for the SOS2 papers into INDRA statements. We’ll do this in two chunks to show how the process can be parallelized and the statements assembled from multiple files:

```
python process_reach_from_s3.py SOS2_pmids.txt 0 5
python process_reach_from_s3.py SOS2_pmids.txt 5 10
```

The two runs create two different files for the results from the seven papers, reach_stmts_0_5.pkl (with statements from the first five papers) and reach_stmts_5_7.pkl (with statements from the last two). Note that the results are pickled as a dict (rather than a list), with PMIDs as keys and lists of Statements as values.

Of course, what we really want is a single file containing all of the statements for the entire corpus. To get this, run:

```
python assemble_reach_stmts.py reach_stmts_*.pkl
```

The results will be stored in reach_stmts.pkl.

### 5.2.7 Running the whole pipeline with one script

If you want to run the whole pipeline in one go, you can run the script submit_reading_pipeline.py (in indra/tools/reading) on a cluster of Amazon EC2 nodes. The script divides up the jobs evenly among the nodes and cores. Usage:

```
python submit_reading_pipeline.py pmid_list tmp_dir num_nodes num_cores_per_node
```

For example if you have a cluster with 8 c3.8xlarge nodes with 32 VCPUs each, you would call it with:
The script submits the jobs to the scheduler with appropriate dependencies such that the REACH reading step completes first, then the INDRA processing step, and then the final assembly into a single pickle file.

### 5.3 Large-Scale Machine Reading with Amazon Batch

The following doc describes the steps involved in reading a large numbers of papers in parallel on Amazon EC2 using REACH, caching the JSON output on Amazon S3, processing the REACH output into INDRA Statements, and then caching the statements also on S3. Prerequisites for doing the following are:

- An Amazon S3 bucket containing full text contents for papers, keyed by Pubmed ID (creation of this S3 repository will be described in another tutorial).
- Amazon AWS credentials for using AWS Batch.
- A corpus of PMIDs (see Large-Scale Machine Reading with Starcluster for information on how to assemble this)
- Optional: Elsevier text and data mining API key and institution key for subscriber access to Elsevier full text content.

#### 5.3.1 How it Works

- The reading pipeline makes use of a Docker image that contains INDRA and all necessary dependencies, including REACH, Kappa, PySB, etc. The Docker file for this image is available at: [https://github.com/johnbachman/indra_docker](https://github.com/johnbachman/indra_docker).
- The INDRA Docker image is built by AWS Codebuild and pushed to Amazon’s EC2 Container Service (ECS), where it is available via the Repository URI:

```
292075781285.dkr.ecr.us-east-1.amazonaws.com/indra
```

- An AWS Batch Compute Environment named “run_reach” is configured to use this Docker image for handling AWS jobs. This compute environment is configured to use only Spot instances with a maximum spot price of 40% of the on-demand price, and 16 vCPUs.
- An AWS Job Queue, “run_reach_queue”, is configured to use instances of the “run_reach” Compute Environment.
- An AWS Job Definition, “run_reach_jobdef”, is configured to run in the “run_reach_queue”, and to use 16 vCPUs and 30GiB of RAM.
- Reading jobs are submitted by running the script:

```
python -m indra.tools.reading.submit_reading_pipeline_aws read [args]
```

which, given a list of PMIDs:

- Copies the PMID list to the key `reading_results/[job_name]/pmids` on Amazon S3
- Breaks the list up into chunks (e.g., of 3000 PMIDs) and submits an AWS Batch job for each (using the “run_reach_jobdef” definition as a template).
- The ECS instance created by the AWS Batch job runs the script `indra.tools.reading.run_reach_on_pmids_aws`, which:
- Checks for cached content on Amazon S3
- If the PMID has not been read by the current version of REACH, checks for content
- If the content is not available, downloads the content using the INDRA literature client, and caches on S3
- The content to be read is downloaded to the /tmp directory of the instance
- REACH is run using the command-line interface (RunReachCLI), and configured to read the papers in the /tmp directory using all of the vCPUs on the instance
- When done, the result REACH JSON in the output folder is uploaded to S3
- The JSON for both the previously and newly read papers is processed in parallel to INDRA Statements
- The resulting subset of statements for the given range of papers is cached on S3 at reading_results/[job_name]/stmts/[start_ix]_[end_ix].pkl. This set of statements takes the form of a pickled (protocol 3) Python dict with PMIDs as keys and lists of INDRA Statements as values.
- In addition, information about the sources of content available for each PMID is cached for each PMID subset at reading_results/[job_name]/content_types/[start_ix]_[end_ix].pkl.

- When the reading jobs for each of the subsets of PMIDs have been completed and cached on S3, the final combined set of statements (and combined information on content sources) can be assembled using:

```
python -m indra.tools.reading.submit_reading_pipeline_aws combine [job_name]
```

- This script submits an AWS batch job for a machine with 1 vCPU but a large amount of memory (60GiB)
- The job runs the script indra.tools.reading.assemble_reach_stmts_aws, which unpickles the results from all of the PMID subsets, combines them, and stores them on S3
- The resulting files are obtainable from S3 at reading_results/[job_name]/stmts.pkl and reading_results/[job_name]/content_types.pkl.

- To run the entire pipeline, where the assembly of the combined set of statements is automatically performed after the reading step is completed, run:

```
python -m indra.tools.reading.submit_reading_pipeline_aws full [args]
```

### 5.4 Assembling everything known about a particular gene

Assume you are interested in collecting all mechanisms that a particular gene is involved in. Using INDRA, it is possible to collect everything curated about the gene in pathway databases and then read all the accessible literature discussing the gene of interest. This knowledge is aggregated as a set of INDRA Statements which can then be assembled into several different model and network formats and possibly shared online.

For the sake of example, assume that the gene of interest is TMEM173.

It is important to use the standard HGNC gene symbol of the gene throughout the example (this information is available on [http://www.genenames.org/](http://www.genenames.org/) or [http://www.uniprot.org/](http://www.uniprot.org/)) - arbitrary synonyms will not work!

#### 5.4.1 Collect mechanisms from PathwayCommons and the BEL Large Corpus

We first collect Statements from the PathwayCommons database via INDRA’s BioPAX API and then collect Statements from the BEL Large Corpus via INDRA’s BEL API.
from indra.tools.gene_network import GeneNetwork

gn = GeneNetwork(['TMEM173'])
biopax_stmts = gn.get_biopax_stmts()
bel_stmts = gn.get_bel_stmts()

at this point biopax_stmts and bel_stmts are two lists of INDRA Statements.

5.4.2 Collect a list of publications that discuss the gene of interest

We next use INDRA’s literature client to find PubMed IDs (PMIDs) that discuss the gene of interest. To find articles that are annotated with the given gene, INDRA first looks up the Entrez ID corresponding to the gene name and then finds associated publications.

from indra import literature

pmids = literature.pubmed_client.get_ids_for_gene('TMEM173')

The variable pmid now contains a list of PMIDs associated with the gene.

5.4.3 Get the full text or abstract corresponding to the publications

Next we use INDRA’s literature client to fetch the full text (if available) or the abstract corresponding to the PMIDs we have just collected.

from indra import literature

paper_contents = {}
for pmid in pmids:
    content, content_type = literature.get_full_text(pmid, 'pmid')
    paper_contents[pmid] = (content, content_type)

We now have a dictionary called paper_contents which stores the content and the content type of each PMID we looked up.

5.4.4 Read the content of the publications

We next run the REACH reading system on the publications. Depending on the content type, different calls need to be made via INDRA’s REACH API.

from indra import literature
from indra.sources import reach

read_offline = True

literature_stmts = []
for pmid, (content, content_type) in paper_contents.items():
    rp = None
    print('Reading %s' % pmid)
    if content_type == 'abstract':
        rp = reach.process_text(content, citation=pmid, offline=read_offline)
    elif content_type == 'pmc_oa_xml':
        rp = reach.process_nxml_str(content, offline=read_offline)
elif content_type == 'elsevier_xml':
    txt = literature.elsevier_client.extract_text(content)
    if txt:
        rp = reach.process_text(txt, citation=pmid, offline=read_offline)
        if rp is not None:
            literature_stmts += rp.statements

The list `literature_stmts` now contains the results of all the statements that were read.

5.4.5 Combine all statements and run pre-assembly

```python
from indra.tools import assemble_corpus
stmts = biopax_stmts + bel_stmts + literature_stmts
stmts = assemble_corpus.map_grounding(stmts)
stmts = assemble_corpus.map_sequence(stmts)
stmts = assemble_corpus.run_preassembly(stmts)
```

At this point `stmts` contains a list of Statements collected with grounding, sequences having been mapped, duplicates combined and less specific variants of statements hidden. It is possible to run other filters on the results such as to keep only human genes, remove Statements with ungrounded genes, or to keep only certain types of interactions.

5.4.6 Assemble the statements into a network model

```python
from indra.assemblers import CxAssembler
cxa = CxAssembler(stmts)
cxa.make_model()
```

we can now upload this network to the Network Data Exchange (NDEx).

```python
ndex_cred = {'user': 'myusername', 'password': 'xxx'}
network_id = cxa.upload_model(ndex_cred)
print(network_id)
```
CHAPTER SIX

REST API

Many functionalities of INDRA can be used via a REST API. This enables making use of INDRA’s knowledge sources and assembly capabilities in a RESTful, platform independent fashion. The REST service is meant to be used locally (on a single machine or local network) and is currently not offered as a public web service by the creators of INDRA.

6.1 Installation

The REST service requires the bottle package to be installed in addition to all the other requirements of INDRA.

6.2 Launching the REST service

The REST service can be launched by running api.py in the rest_api folder within indra.

6.3 Documentation

The specific end-points and input/output parameters offered by the REST API are documented in rest_api/docs/index.html, which is accessable locally within the indra folder.
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