

computational bioscience program

university of colorado school of medicine

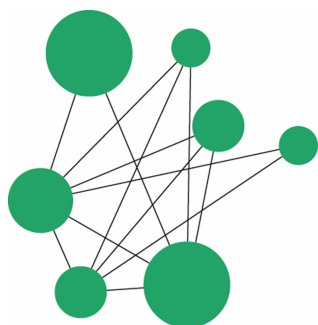
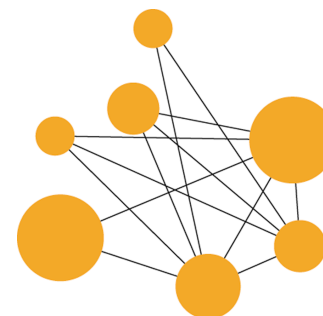
Verbs in Biomedical Text

Karin Verspoor

Faculty, Computational Bioscience Program
University of Colorado School of Medicine

Kevin Bretonnel Cohen

Biomedical Text Mining Group Lead
University of Colorado School of Medicine

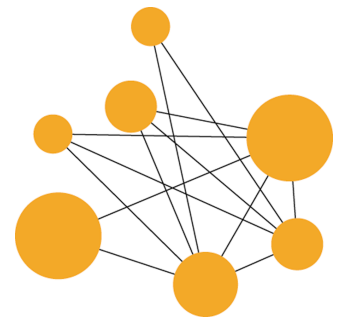


Karin.Verspoor@ucdenver.edu

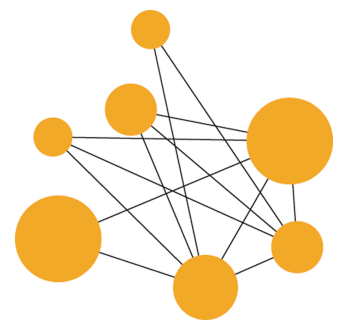
http://compbio.ucdenver.edu/Hunter_lab/Verspoor

Kevin.Cohen@gmail.com

http://compbio.ucdenver.edu/Hunter_lab/Cohen



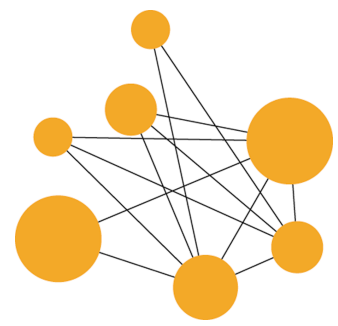
THE CONTEXT for biomedical natural language processing



Exponential knowledge growth in biomedicine

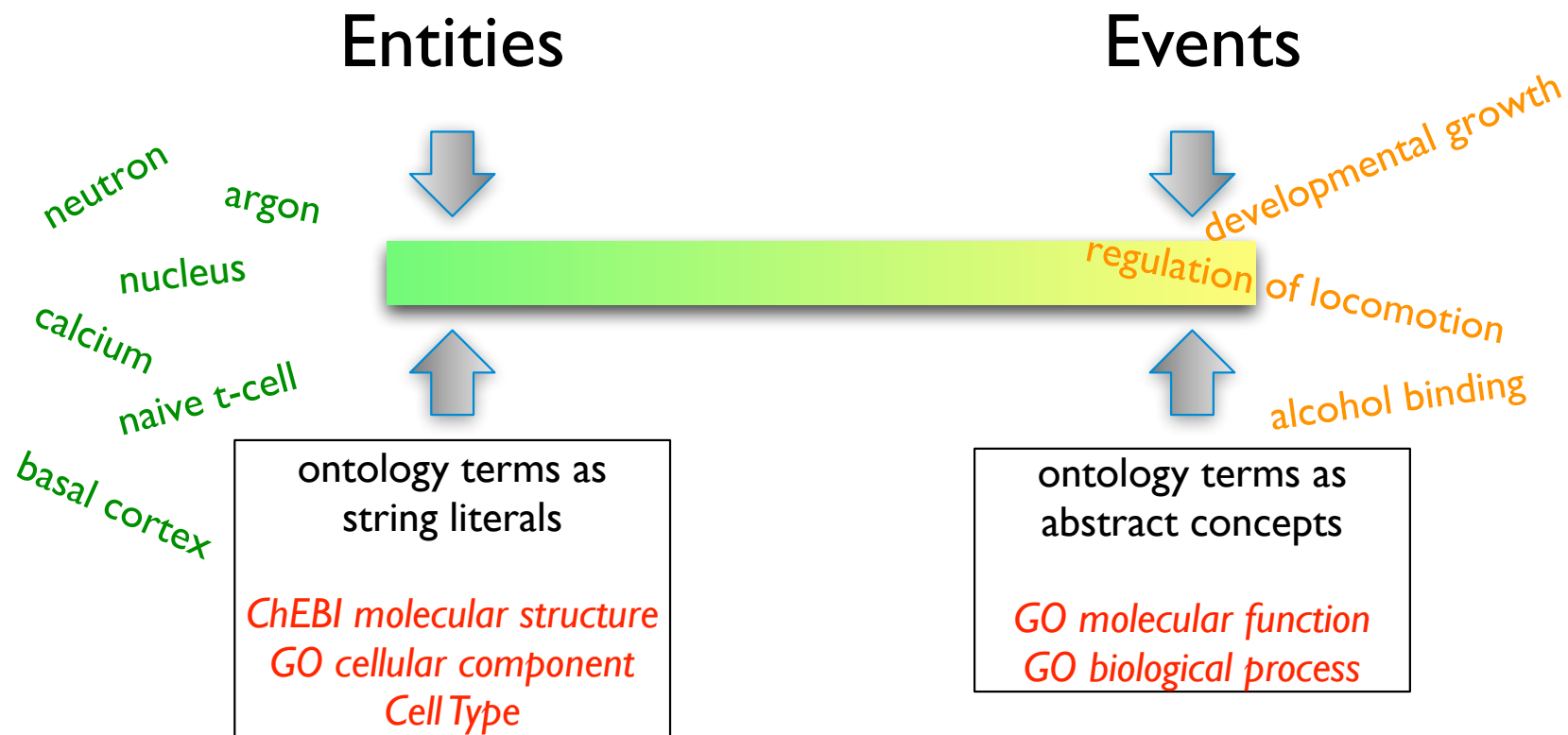
- 1,330 peer-reviewed gene-related databases in 2011 NAR db issue
- Over 20 million PubMed entries ($> 2,200/\text{day}$)
- Breakdown of disciplinary boundaries makes more of it relevant to each of us
- “Like drinking from a firehose” – Jim Ostell

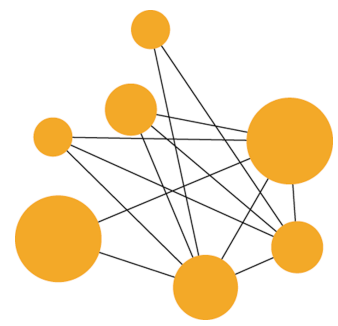




Language processing of Biomedical texts

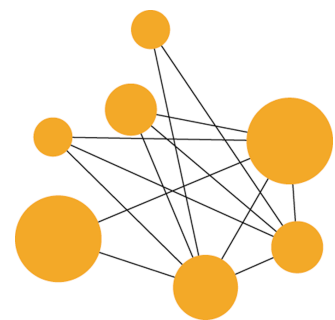
- Tools that support identification, indexing, and extraction of biological concepts





Scientific Publishing & Semantics

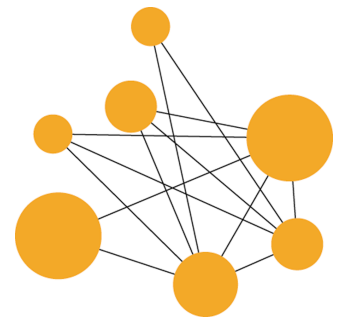
- Content enrichment
- Direct access to (relevant) external data
- Structured digital abstracts
- Enables
 - Interactivity
 - targeted searches
 - relevance linking
 - formalizing content; actionable data



Making BioNLP relevant

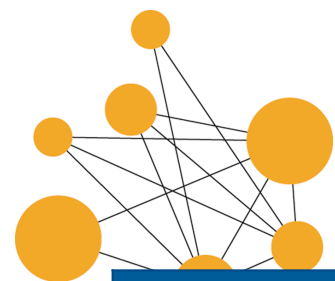
- Recognition of OBO terms, relations
- CRAFT corpus (first release very soon)

The significance of the **interaction** between **DAZAP1** and **DAZL/DAZ** remains to be defined. These **proteins** may act together to facilitate the **expression** of a **set of genes** in germ **cells**. For example, **DAZAP1** could be involved in the **transport of the mRNAs** of the target **genes** of **DAZL**. Alternatively, **DAZL** and **DAZAP1** may act antagonistically to **regulate** the timing and the level of **expression**. Such an antagonistic **interaction** between two **interacting RNA-binding proteins** is exemplified by the **neuron-specific nuclear RNA-binding protein**, **Nova-1**. **Nova-1** **regulates** the alternative **splicing of the pre-mRNAs** encoding **neuronal inhibitory glycine receptor $\alpha 2$ (GlyR $\alpha 2$)** [23]. The ability of **Nova-1** to activate **exon** selection in **neurons** is antagonized by a second **RNA-binding protein**, **brPTB (brain-enriched polypyrimidine tract-binding protein)**, which **interacts** with **Nova-1** and inhibits its function [24]. **DAZAP1** could function in a similar manner by **binding** to **DAZL** and inhibiting its function. Comparing the phenotypes of **Dazl1** and **Dazap1** single and double knock-out **mice** may provide some clues to the significance of their **interaction**. **Dazl1** knock-out **mice** have already been generated and studied [6]. The **spermatogenic** defect in the male becomes apparent only after day 7 post partum when the germ **cells** are committing to **meiosis** (H. Cooke, personal communication). The **genomic** structure of **Dazap1**, delineated here, should facilitate the generating of **Dazap1** null **mutation**.



THE LINGUISTIC CHALLENGES

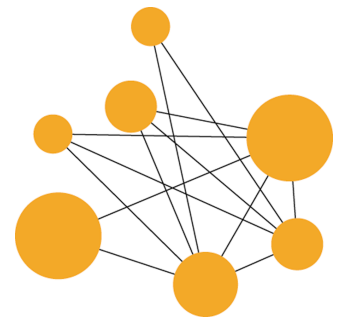
of BioNLP



Verbs in Biomedical Text

Biomedical	BNC
show	do
suggest	say
use	make
indicate	go
contain	see
describe	take
express	get
bind	know
require	come
observe	give
find	think
determine	use
demonstrate	find
perform	look
induce	want

- Verb usage differs significantly from general English
- Domain-specific verbs: *phosphorylate, ubiquitinate*
- Verbs that have a domain-specific sense: *express, regulate, signal, transcribe*



Arity

K. Bretonnel Cohen and Lawrence Hunter (2006). A critical review of PASBio's argument structures for biomedical verbs. *BMC Bioinformatics* 7(Suppl. 3):S5.



Most research on biomedical semantics

- All relationships binary
 - Protein/protein
 - Drug/gene
 - Drug/disease
 - Drug/effect

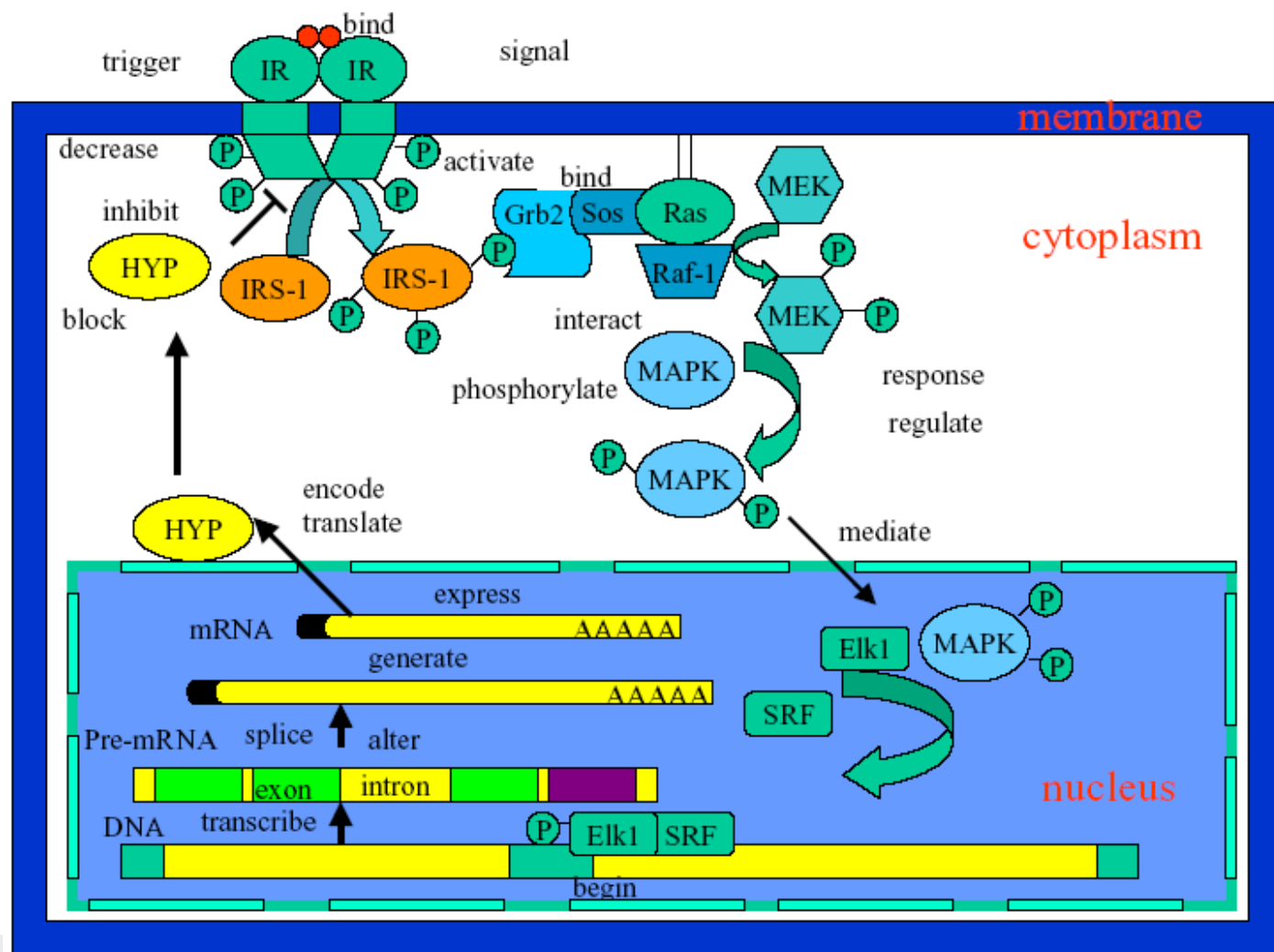


PASBio (Predicate-Argument Structures for Biology) Project

- Described in Wattarujeekrit, Shah, and Collier (2004)
- Set of 29 verbs, 34 predicates with associated argument structures
- 10 annotated examples each
- Publicly available



Predicate selection for PASBio:gene expression; regulation; signalling





PASBio findings

- Overlap between domain-specific and “General English” semantics is low
- Biological domain has more “core arguments,” fewer “adjuncts”



PASBio findings

- 9/29 didn't occur in PropBank or had different sense
- 45% (9/20) had more arguments
- 25% (5/20) had fewer arguments
- 30% (6/20) had same number



Native speaker intuition behind this

- I don't believe what you tell me unless I know when, where, at what pH, at what temperature...
- Consequence: weak distinction between core arguments and adjuncts

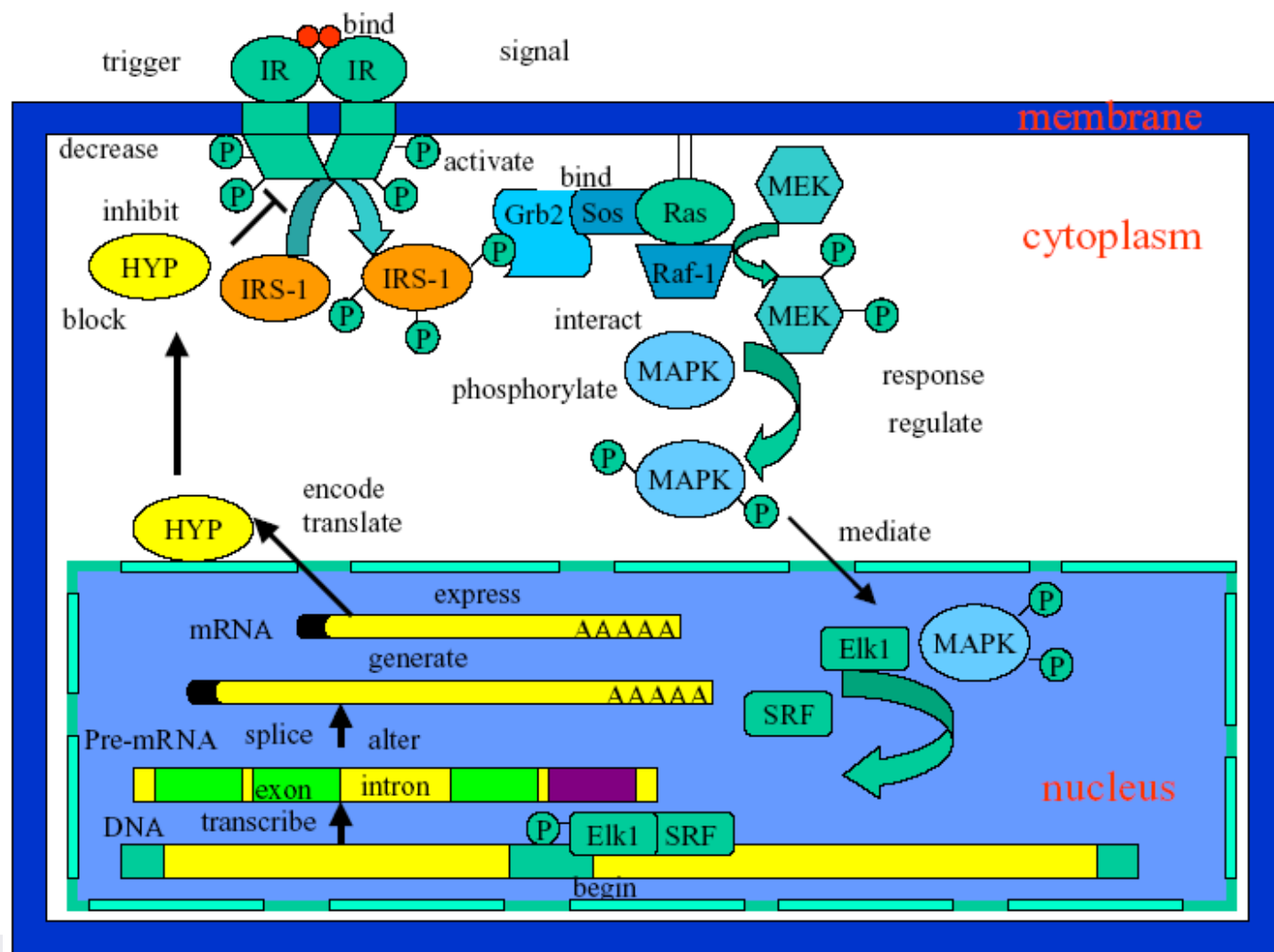


Example representation: *express*

- Arg1: named entity being expressed (gene or gene product)
- Arg2: property of the existing name entity
- Arg3: location referring to organelle, cell or tissue



Expression





Are these arguments to a predicate...

- Example Arg2s:

- two mRNA isoforms of 2.4 and 4.0 kb
- 2.0 and 2.4 kilobases in length

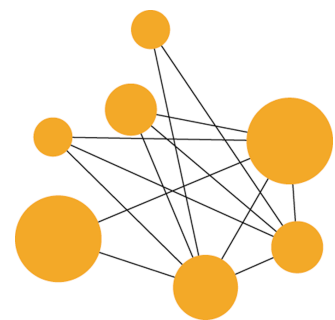
Two equally abundant mRNAs for il8ra, 2.0 and 2.4 kilobases in length, are expressed in neutrophils and arise from usage of two alternative polyadenylation signals.



...or slots in a frame?

- PUNDIT:
 - Customer
 - Symptoms
 - Actions taken
 - Success or failure

*Note: frame slots inferred from Palmer et al.
(1986)*

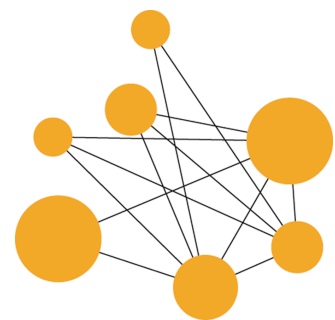


Subcategorization frames

- Help to derive the correct interpretation
- e.g. resolve attachment ambiguities:

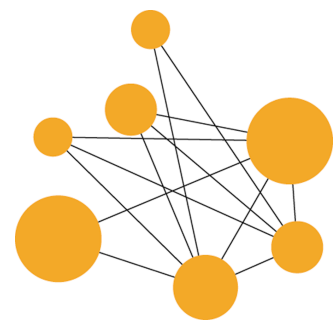
assessing ubiquitin expression in infected mice brains

two poly(A)+ RNAs transcribed from the opposite strand of the upstream flanking regions lacked ...



Subcategorization frame learning

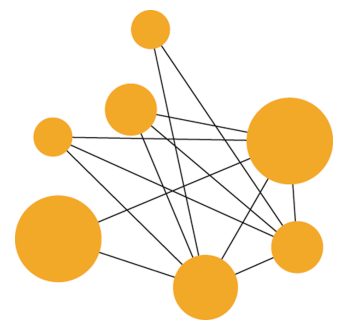
- Goal: acquire subcategorization frames from a corpus
- utilize dependency relations derived from a statistical parser
- map from sets of dependency relations to a SCF via manually developed (unification) rules
- rules defined for COMLEX, ANLT, NOMLEX frames



Biomedical verb classification

- Induce lexical classes from corpora
 - Build an inventory of subcategorization frames for each verb
 - Cluster verbs according to shared subcat frames

It	indicate	that	protein	activate	gene
	suggests demonstrates indicates implies		p53 TP53 DMP53 ...	activates up-regulates induces stimulates	WAF1 CIP1 p21 ...



Nominalization and Alternations in Biomedical Text

K. Bretonnel Cohen, Martha Palmer, and Lawrence Hunter (2008). Nominalization and alternations in biomedical language. *PLoS ONE* 3(9).



Goals of the study

- Characterize behavior of nominalizations in biomedical text
- Determine implications for system-building



Definitions

- Nominalization: noun derived from a verb
 - Verbal nominalization: *activation, inhibition, induction*
 - Argument nominalization: *activator, inhibitor, inducer, mutant*



Nominalizations are dominant in biomedical texts

Predicate	Nominalization	All verb forms
Express	2,909	1,233
Develop	1,408	597
Analyze	1,565	364
Observe	185	809
Differentiate	737	166
Describe	10	621
Compare	185	668
Lose	556	74
Perform	86	599
Form	533	511

Data from
CRAFT corpus



Relevant points for text mining

- Nominalizations are an obvious route for scaling up recall
- Nominalizations are more difficult to handle than verbs...
- ...but can yield higher precision (Cohen et al. 2008)



Definitions

- Argument: participant in or qualifier of the action of the predicate

Argument	
Arg0	Causer of increase
Arg1	Thing increasing
Arg2	Amount increased by
Arg3	Start point
Arg4	End point

Our representation of this predicate is the same as PropBank's.
[doi:10.1371/journal.pone.0003158.t001](https://doi.org/10.1371/journal.pone.0003158.t001)



Definitions

Argument

Arg0	Causer of increase
------	--------------------

Arg1	Thing increasing
------	------------------

Arg2	Amount increased by
------	---------------------

Arg3	Start point
------	-------------

Arg4	End point
------	-----------

Our representation of this predicate is the s
doi:10.1371/journal.pone.0003158.t001

Do870, an azole antifungal agent[Arg0], produced
dose-related **increases** in *total cytochrome P450*
and aldrin epoxidase[Arg1]



Definitions

Argument

Arg0	Causer of increase
Arg1	Thing increasing
Arg2	Amount increased by
Arg3	Start point
Arg4	End point

Our representation of this predicate is the s
doi:10.1371/journal.pone.0003158.t001

Increase in phosphorylation of APP[Arg1] by overexpression of the nerve growth factor receptor Trk A[Arg0]



Definitions

- 2 arguments: *Activate*
 - Arg0: Activator
 - Arg1: Activatee
- 3 arguments: *Inhibit*
 - Arg0: Inhibitor
 - Arg1: Inhibitee
 - Arg2: Amount of inhibition



Definitions

- Alternation: variations in the surface syntactic form of predicates and their arguments
 - Active/passive
 - X phosphorylates Y
 - Y is phosphorylated by X
 - Transitive/intransitive
 - X decreases Y
 - Y decreases



Alternations of nominalizations: positions of arguments

- Any combination of the set of positions for each argument of a nominalization
 - Pre-nominal: phenobarbital **induction**, trkA **expression**
 - Post-nominal: **increases** of oxygen
 - No argument present: **Induction** followed a slower kinetic...
 - Noun-phrase-external: this enzyme can undergo **activation**



Pre-nominal arguments

- Agent (Arg0)
 - cytochrome(s) P-450 **mediation**
 - interferon-gamma **inhibition** of VSV replication
 - Phenobarbital **treatment**
- Patient (Arg1, \approx logical object)
 - trkA **expression**
 - agonist **association**
 - cancer **treatment**



Noun-phrase-external arguments

- EWS/FLI-1 antagonists induce growth **inhibition** of Ewing tumor cells
 - Support verb links agent to noun phrase
- potency of sertraline for dopamine reuptake **inhibition**
 - Transparent noun
- Phenobarbital (PB) has long been known as an inducer of drug-metabolizing enzymes in liver, but the molecular mechanism underlying this **induction** is still poorly understood
 - Event coreference



Alternations of nominalizations

- **activation** of molecular oxygen by alkaline hemin
 - Argo post-nominal, Arg1 post-nominal
- K(ATP) **activation** by cromakalim
 - Argo post-nominal, Arg1 pre-nominal
- Mutational **activation** of the ras genes
 - Argo pre-nominal, Arg1 post-nominal



Previous work on nominalizations in the biomedical domain

- Ono et al. (2001): *interaction, association, complex, and binding*
- Pustejovsky et al. (2002): *inhibition and inhibitor*
- Hu et al. (2005), Narayanaswamy et al. (2005), Yuan et al. (2006): *phosphorylation*
- Lots of early work by Zellig Harris, the Linguistic String Project, other workers in sublanguage model



Prediction investigated

- Within scientific language, we should expect a limited variety of alternations



Previous work on nominalizations in the biomedical domain

- GENESCENE: tackles all verbal nominalizations
 - Arguments recognized only if following nominalization and preceded by *of*, *in*, or *by*

Leroy and Chen (2002), Leroy et al. (2003),
Leroy and Chen (2005)



- A sample predicate for which the three prepositions *of*, *in*, and *by* are insufficient for capturing all arguments.

Argument		Associated prepositions
Arg0	Causer of increase	<i>after, by, during, in, of</i>
Arg1	Thing increasing	<i>in, for, of, with</i>
Arg2	Amount increased by	<i>by, in, of, up, with</i>
Arg3	Start point	<i>From</i>
Arg4	End point	<i>to, with</i>

Our representation of this predicate is the same as PropBank's.

[doi:10.1371/journal.pone.0003158.t001](https://doi.org/10.1371/journal.pone.0003158.t001)



Materials and methods

- Release 0.9 of the PennBioIE corpus (collection of abstracts of journal articles, annotated with parts of speech, syntactic structure, and entities)



Materials and methods

- Marked arguments for 746 tokens of nominalizations of the 10 most common verbs
- Second annotator marked 15% of these to calculate interannotator agreement (87.5%)

Text source: 4_source_file_852_35168.src

7-Ethoxyresorufin (a selective and competitive blocker of CYP 1A isozyme) blunted ACh and histamine mediated EDHF responses but did not alter vasodilation initiated through K⁺ channel activation by either cromakalim or NS-1619, or through the nitric oxide-cGMP pathway (sodium nitroprusside).

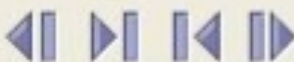
comments

Filter - Empty Filter

filter selection:



span edit:



annotations



- ☐ K⁺ channel (Pre-predicate Patient)
- ☒ activation (Activate)
- ☐ by either cromakalim or NS-1619 (Post-predicate Agent)

annotated class

● Activate

slots of annotated class (2 values)

Arg0 Activator



☒ by either cromakalim or NS-1619 (Post-predicate Agent)

Arg1 Activee



☒ K⁺ channel (Pre-predicate Patient)



Result 1: attested alternations are extraordinarily diverse

- *Inhibition*, a 3-argument predicate—
Arguments 0 and 1 only shown

		Arg0			
		Pre	Post	Ext	Abs
Arg1	Pre	–	2	8	4
	Post	1	15	16	26
	Ext	1	3	5	1
	Abs	3	2	2	6

Data is combined from both parts of the BioIE corpus. 24/64 possible patterns are attested in 95 tokens (5 can't-tell).
doi:10.1371/journal.pone.0003158.t032



Results for 2-argument verbs

	Alternations	Tokens	X	attested/possible	type/token
<i>expression</i>	6	97	4	0.375	0.062
<i>mediation</i>	2	2	2	0.124	1.0
<i>containment</i>	1	1	0	.063	1.0
<i>activation</i>	14	91	9	0.875	0.154

The maximum number possible is 4^2 . Data is given for the full BioIE corpus. The column labelled *tokens* shows the number of tokens for which no argument was labelled “can’t tell.” The column labelled *X* shows the number of tokens with at least one argument labelled “can’t tell.”

doi:10.1371/journal.pone.0003158.t014



Results for 3-argument verbs

	Alternations	Tokens	X	attested/ possible	type/token
<i>Inhibition</i>	24	95	5	0.375	0.253
<i>Induction</i>	19	92	8	0.297	0.21
<i>association.01</i>	5	8	0	0.078	0.625
<i>association.02</i>	10	78	1	0.156	0.128
<i>treatment.04</i>	9	58	7	0.141	0.155

The maximum number possible is 4^3 . Data is given for the full BioIE corpus. The column labelled *tokens* shows the number of tokens for which no argument was labelled “can’t tell.” The column labelled *X* shows the number of tokens with at least one argument labelled “can’t tell.”

doi:10.1371/journal.pone.0003158.t015



Result 2: syntactic positions

- Most common syntactic positions for each semantic role:

Semantic role	Total	Most common syntactic positions
Arg0	570	Absent (378), NP-external (82), Post-nominal (64), Pre-nominal (46)
Arg1	612	Post-nominal (341), Pre-nominal (124), Absent (79), NP-external (68)

See Tables 43 and 44 for the raw data.
doi:10.1371/journal.pone.0003158.t025



Result 3: semantic roles

- Most frequent semantic roles for each syntactic position:

Position	Total		
Pre-nominal	Arg1 (124)	Arg0 (51)	175
Post-nominal	Arg1 (341)	Arg0 (107)	448
NP-external	Arg0 (85)	Arg1 (68)	153
Absent	Arg0 (378)	Arg1 (79)	461

Only Args 0 and 1 are indicated. *Association.02,03* are omitted. See Tables 43 and 44 for the raw data.

[doi:10.1371/journal.pone.0003158.t026](https://doi.org/10.1371/journal.pone.0003158.t026)



Implications for system-building

- Distinction between absent and noun-phrase-external arguments is crucial and difficult, and finite state approaches will not suffice; merging data from different clauses and sentences may be useful
- Pre-nominal arguments are undergoer by ratio of 2.5:1
- For predicates with agent and patient, post/post and pre/post patterns predominate, but others are common as well



What can be done?

- External arguments:
 - semantic role labelling approach
 - ...but, very important to recognize the absent/external distinction, especially with machine learning
 - pattern-based approach
 - ...but, approaches to external arguments (RLIMS-P) are so far very predicate-specific



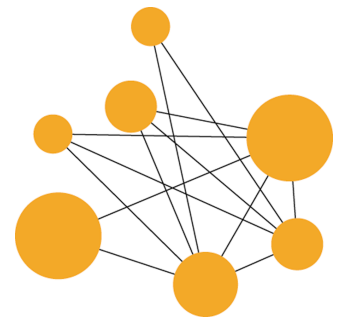
What can be done?

- Pre-nominal arguments:
 - apply heuristic that we have identified based on distributional characteristics
 - for most frequent nominalizations, manual encoding may be tractable



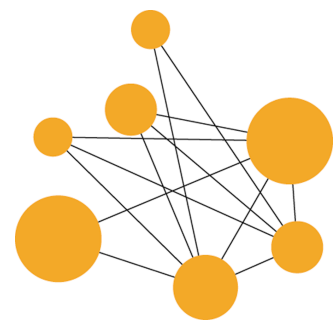
Future analysis

- Can identity of pre-predicate arguments be characterized on a per-predicate basis?
 - At minimum will require word sense disambiguation (phenobarbital treatment/ cancer treatment)
- Can pre-predicate arguments be characterized by semantic class?



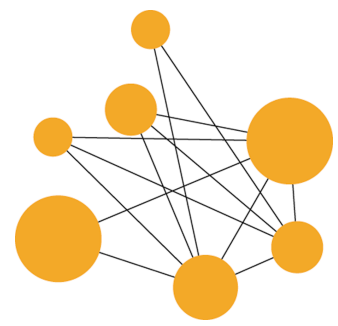
INFORMATION EXTRACTION

technology for BioNLP

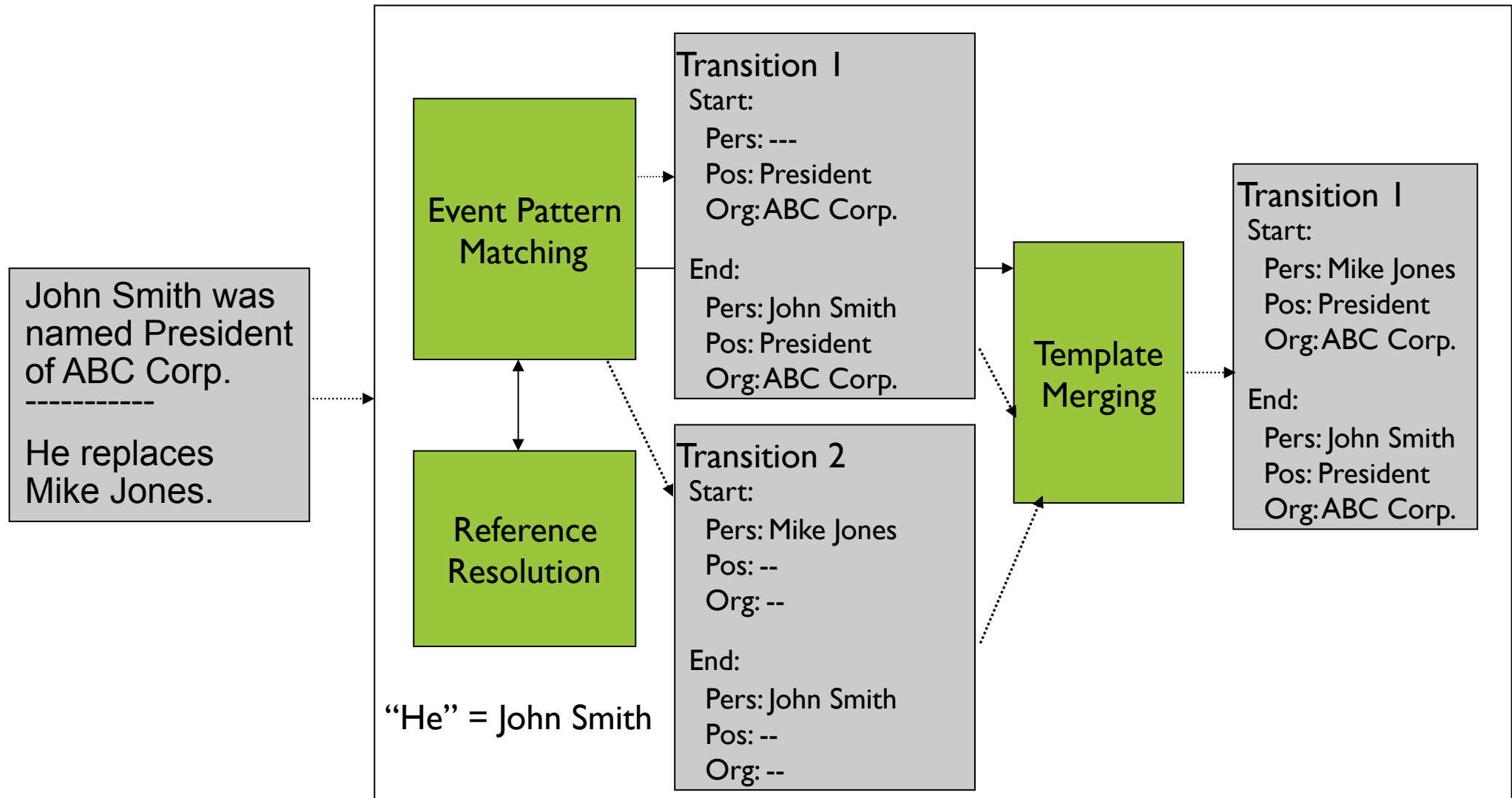


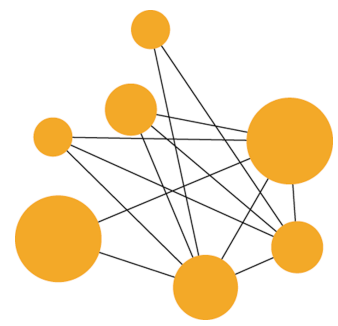
Information Extraction

- Algorithms that
 - automatically extract structured information from unstructured (natural language) text
 - aim to identify entities and events of interest
 - utilize natural language processing
 - (linguistic) rule-based
 - machine learning



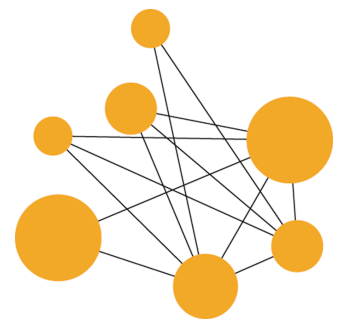
Information Extraction (MUC example)



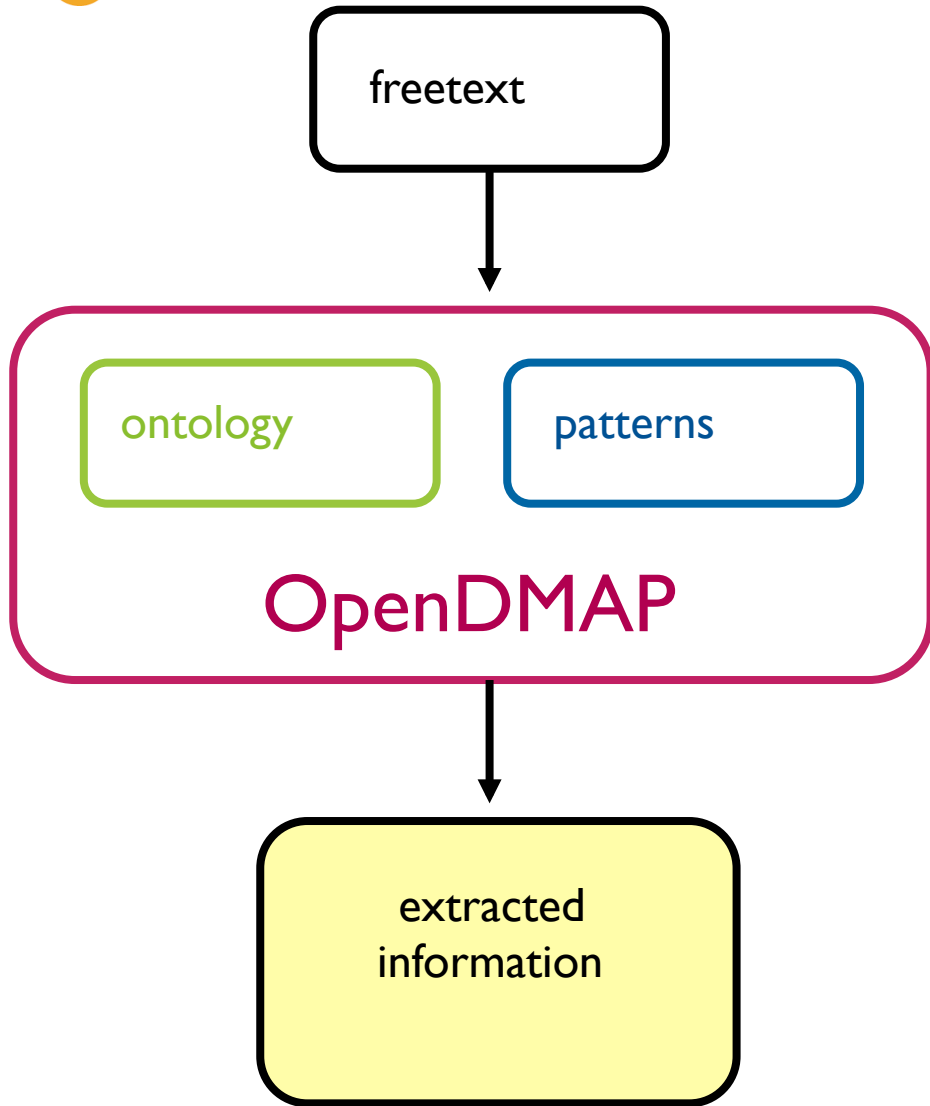


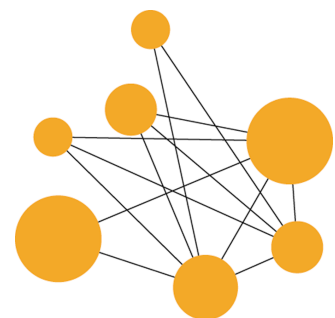
OpenDMAP extracts typed relations from the literature

- Concept recognition tool
 - Connect ontological terms to literature instances
 - Built on Protégé knowledge representation system
- Language patterns associated with concepts and slots
 - Patterns can contain text literals, other concepts, constraints (conceptual or syntactic), ordering information, or outputs of other processing.
 - Linked to many text analysis engines via UIMA
- Best performance in BioCreative II IPS task
- >500,000 instances of three predicates (with arguments) extracted from Medline Abstracts
- [Hunter, et al., 2008] <http://bionlp.sourceforge.net>



OpenDMAP





OpenDMAP

freetext

Cyclin E2 interacts with Cdk2 in a functional kinase complex.

ontology

patterns

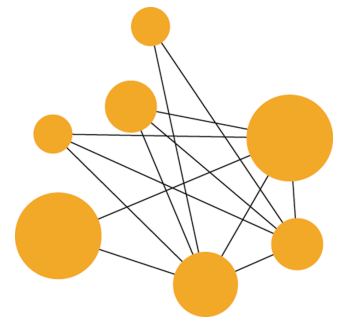
OpenDMAP

<ontology>

Protein protein interaction :=
[int1] interacts with [int2]

extracted
information

protein protein interaction:
interactor1: cyclin E2
interactor2: cdk2



OpenDMAP

PROTÉGÉ ONTOLOGY

CLASS: protein protein interaction

SLOT: interactor1

TYPE: molecule

SLOT: interactor2

TYPE: molecule

PATTERNS

{c-interact} := [interactor1] interacts with [interactor2]

{c-interact} := [interactor1] is bound by [interactor2]

...

OpenDMAP

CLASS BROWSER

For Project: ● generif

Class Hierarchy

- :THING
- ▶ ● :SYSTEM-CLASS
- ▼ ● c-object
 - ▼ ● c-molecule
 - c-protein
 - c-protein-receptor
 - c-rna
 - c-dna
 - c-cell-type
 - c-cell-part
 - c-cell-line
 - ▼ ● c-bioprocess
 - ▶ ● c-transport
 - c-activate
 - c-grow
 - c-anchor
 - c-interact
 - c-interact-neo

CLASS EDITOR

For Class: ● c-interact (instance of :STANDARD-CLASS)

Name

c-interact

Documentation

Role

Concrete ●

Template Slots



















Name	Cardinality	Type
■ interactor1	single	Instance of c-molecule
■ interactor2	single	Instance of c-molecule
■ :NAME	single	String

Classes Slots Forms Instances Queries


CLASS BROWSER

For Project:  generif

Class Hierarchy

-  :THING
-  :SYSTEM-CLASS
-  c-object
 -  c-molecule
 -  c-protein
 -  c-protein-receptor
 -  c-rna
 -  c-dna
 -  c-cell-type
 -  c-cell-part
 -  c-cell-line
 -  c-bioprocess
 -  c-transport
 -  c-activate
 -  c-grow
 -  c-anchor
 -  c-interact
 -  c-interact-neo

CLASS EDITOR

For Class:  c-interact (instance of :STANDARD-CLASS)

Name




c-interact

Documentation

Role

Concrete 

Template Slots

Name	Cardinality	Type
 interactor1	single	Instance of c-molecule
 interactor2	single	Instance of c-molecule
 :NAME	single	String

CLASS BROWSER

For Project: ● generif

Class Hierarchy

- :THING
- ▶ ● :SYSTEM-CLASS
- ▼ ● c-object
 - ▼ ● c-molecule
 - c-protein
 - c-protein-receptor
 - c-rna
 - c-dna
 - c-cell-type
 - c-cell-part
 - c-cell-line
 - ▼ ● c-bioprocess
 - ▶ ● c-transport
 - c-activate
 - c-grow
 - c-anchor
 - c-interact
 - c-interact-neo

CLASS EDITOR

For Class: ● c-interact (instance of :STANDARD-CLASS)

Name

c-interact

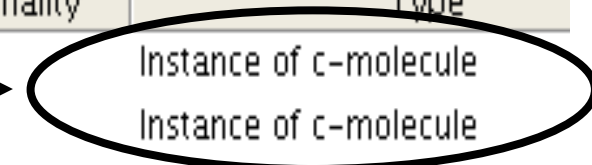
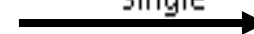
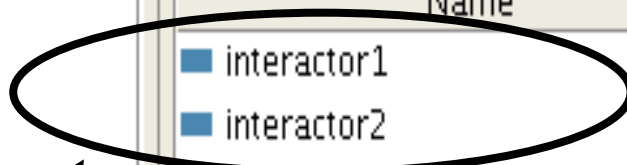
Documentation

Role

Concrete ●

Template Slots

Name	Cardinality	Type
■ interactor1	single	Instance of c-molecule
■ interactor2	single	Instance of c-molecule
■ :NAME	single	String



Classes Slots Forms Instances Queries

CLASS BROWSER

For Project:  generif

Class Hierarchy

- :THING
- ▶ ● :SYSTEM-CLASS
- ▼ ● c-object
 - c-molecule
 - c-protein
 - c-protein-receptor
 - c-rna
 - c-dna
 - c-cell-type
 - c-cell-part
 - c-cell-line
- ▼ ● c-bioprocess
- ▶ ● c-transport
- c-activate
- c-grow
- c-anchor
- c-interact
- c-interact-neo

CLASS EDITOR

For Class: ● c-interact (instance of :STANDARD-CLASS)

Name

c-interact

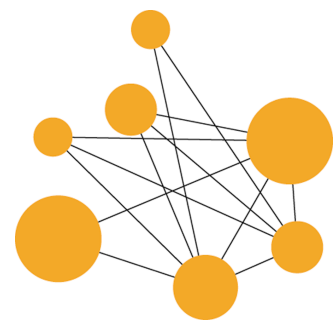
Documentation

Role

Concrete ●

Template Slots

Name	Cardinality	Type
interactor1	single	Instance of c-molecule
interactor2	single	Instance of c-molecule
:NAME	single	String



BioCreative Example

- Some BioCreative patterns for *interact*

`{c-interact} := [interactor1] {w-is} {w-interact-verb1} {w-preposition} the?
[interactor2];`

`{w-is} := is, are, was, were;`

`{w-interact-verb1} := co-immunoprecipitate, co-immunoprecipitates, co-immunoprecipitated, co-localize, co-localizes, co-localized;`

`{w-preposition} := among, between, by, of, with, to;`

- Matched text:

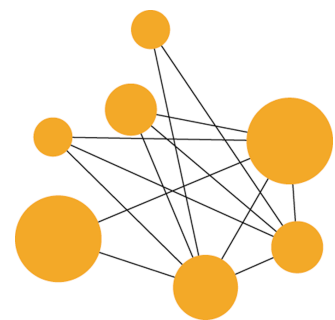
PMID 16494873, SENT_ID 16494873_114

Upon precipitation of the SOX10 protein with anti-HA antibody, Western blot detection revealed expression of UBC9-V5 (25 kDa) in the sample (Fig. 1, line 6), indicating that **{UBC9 was co-immunoprecipitated with SOX10}**.

INTERACTOR_1: UBC9 resolved to UniprotID: UBC9_RAT

INTERACTOR_2: SOX10 resolved to UniProtID: SOX10_RAT

{c-interact} := [UBC9_RAT]_{interactor_1}, [SOX10_RAT]_{interactor_2}

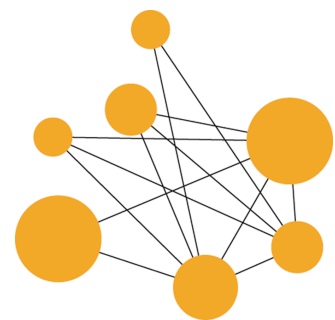


BioCreative Results

- 359 full-text articles in the test set
- 385 interaction assertions produced
- Performance averaged per article (to avoid dominance of a few assertion-heavy articles)

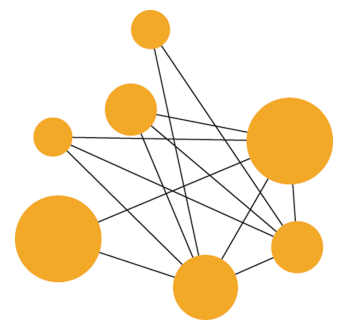
P = 0.39, R = 0.31, F = 0.29

- Best result in the evaluation!
 - F score 10% higher than next-scoring system
 - F score > 3 standard deviations above mean
 - Recall 20% higher than next-scoring system



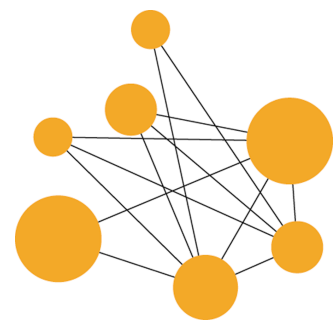
BioCreative conclusions

- Information extraction in biomedical text is hard
 - Linguistic variability in how concepts are expressed
 - Complex concepts with multiple “slots”
- OpenDMAP advances the state of the art
 - Use of an ontology grounds the search for information
 - Flexibility of the pattern language to incorporate constraints at different levels (conceptual, lexical, word order, linguistic)



Integrating background knowledge

- Can improve OpenDMAP precision with minimal cost to recall
 - Take advantage of background knowledge
 - Tighten constraints on slot fillers in the ontology
 - No change to existing patterns
- Proof of concept:
 - Distinguish among several types of protein activation (enzyme and receptor) in GeneRIFs
 - Utilize Gene Ontology annotations



Refining selectional restrictions

enzyme activator activity

activating entity: protein

activated entity: protein - catalytic activity

receptor activator activity

activating entity: protein

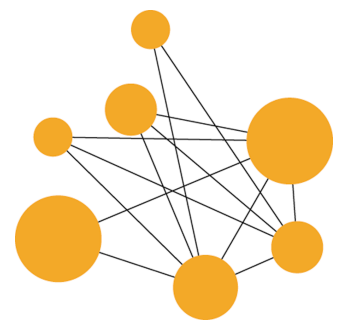
activated entity: protein - receptor activity

TP: [GeneRIF I04I55]

an ER stress induces the activation of [caspase-12_{protein} - catalytic activity]_{activated_entity} via [caspase-3_{protein}]_{activator}

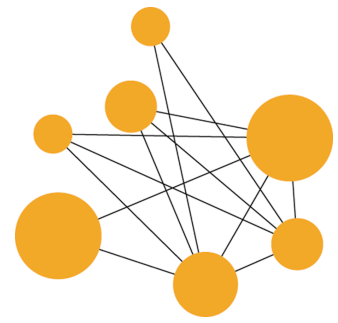
prevented FP: [GeneRIF I05594]

factor Xa can induce mesangial cell proliferation through the activation of ERK_{protein} via PAR2_{protein} in mesangial cells



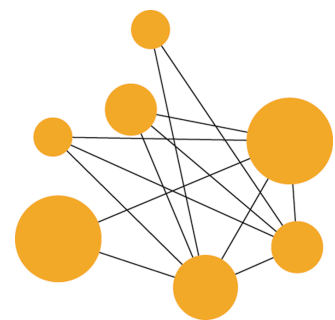
Results

		Original	Additional Memory	Difference
Enzyme Events	Precision	0.24	0.37	0.13
	Recall	0.27	0.20	-0.07
	F-measure	0.26	0.26	0.00
Receptor Events	Precision	0.08	0.34	0.26
	Recall	0.17	0.12	-0.05
	F-measure	0.11	0.18	0.07
Total	Precision	0.16	0.36	0.20
	Recall	0.24	0.18	-0.06
	F-measure	0.19	0.24	0.05



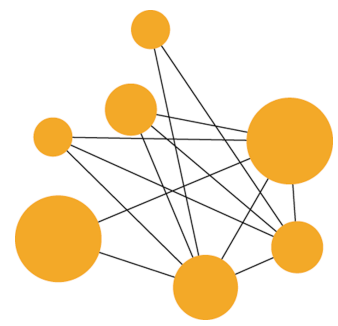
LEXICAL RESOURCES

for BioNLP



The importance of lexical resources

- Need to characterize the linguistic behavior of terms to establish word meaning in context
 - Morphosyntactic behavior
 - inflectional patterns
 - part of speech
 - argument structure
 - Semantic information
- Need to recognize different terms that express the same or closely related meanings
 - To support database integration, multi-database querying
 - To enable generalization of information extraction templates
 - To support general text understanding and meaning analysis (e.g. semantic reasoning over text or during text processing)



Unified Medical Language System

Metathesaurus

1 million+
biomedical
concepts
from over 100
sources

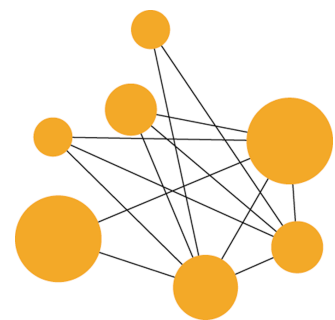
Semantic Network

135 broad
categories and
54 **relationships**
between them

SPECIALIST Lexicon +Tools

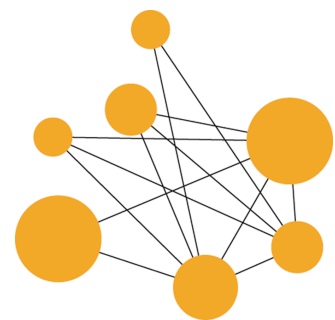
lexical
information and
programs for
**language
processing**

3 Knowledge Sources
used separately or together



Metathesaurus

- 100+ general and specialized biomedical vocabularies
- 17 languages (63% English)
- 1 million+ concepts; 6 million+ names
- 100K+ relationships (hierarchical, semantic, statistical and mapping relationships)
- Distributed in a common electronic format



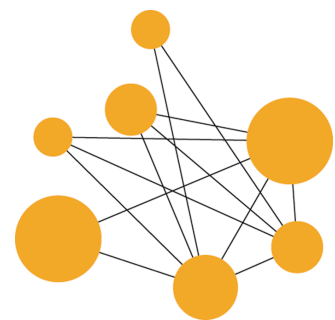
Metathesaurus Concepts

- Synonymous terms clustered into a concept
- Unique identifier (CUI) is assigned
- Source information preserved

Addison's disease	SNOMED CT	PT	363732003
Addison's Disease	MedlinePlus	PT	T1233
Addison Disease	MeSH	PT	D000224
Primary Adrenal Insufficiency	MeSH	EN	D000224
Primary hypoadreanlism syndrome,Addison	MedDRA	LT	I0036696

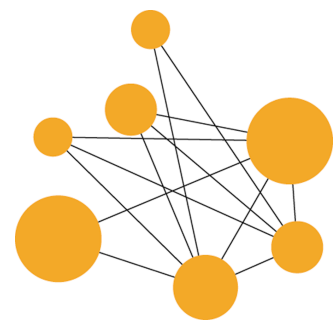
C0001403

Addison's disease

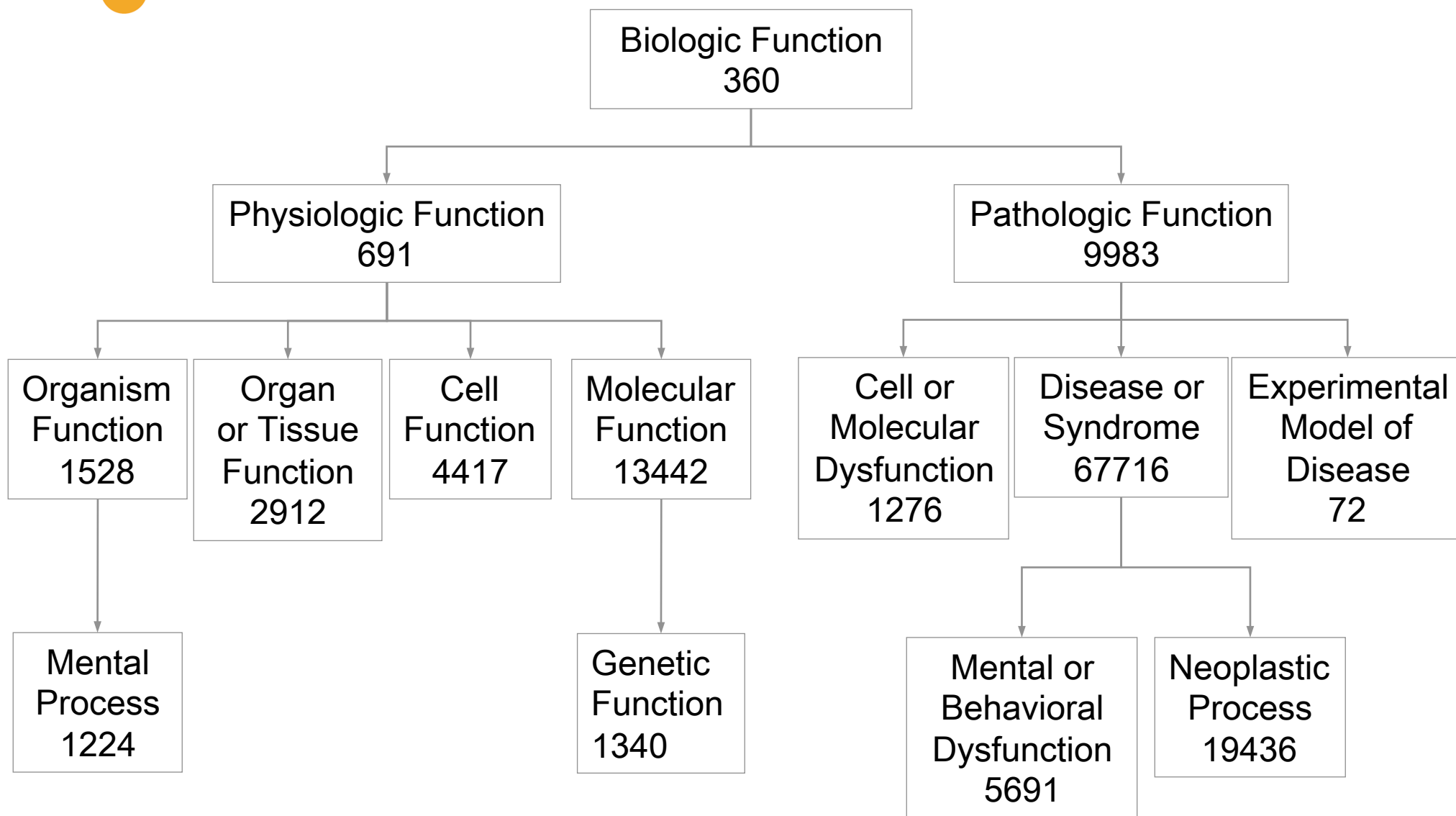


Semantic Network

- 135 Semantic Types
 - Broad subject categories in 2 hierarchies
 - Assigned to all Metathesaurus concepts
- 54 Semantic Relationships
 - Useful, important links between Types
 - Hierarchical “isa” and associative relations
- Categorize the Metathesaurus
- Enhance meaning of concepts



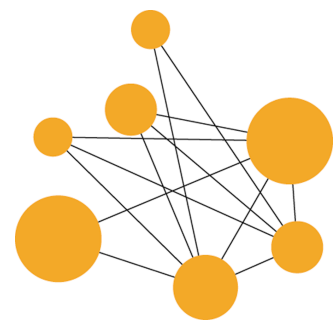
“Biologic Function” hierarchy





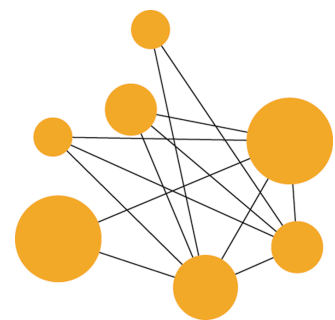
Semantic Relations

- Disease or Syndrome **associated with** Finding
- Disease or Syndrome **result of** Pathologic Function
- Body Part, Organ, or Organ Component **location of** Disease or Syndrome
- Hormone **affects** Disease or Syndrome
Hormone **causes** Disease or Syndrome
Hormone **complicates** Disease or Syndrome



SPECIALIST Lexicon

- English lexicon of 300K+ common words and biomedical terms
- Lexical records encode information on:
 - Syntax
 - Morphology
 - Orthography
- Used with associated lexical tools
 - in Metathesaurus production
 - in natural language processing applications



SPECIALIST Lexical Entry

{base=disease

entry=E0023270

cat=noun

variants=reg

variants=uncount

compl=pphr(of,np|bone|)

compl=pphr(of,np|breast|)

compl=pphr(of,np|liver|)

compl=pphr(of,np|ovary|))}

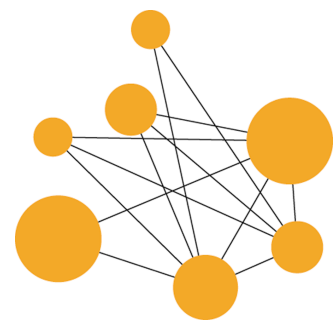
Base form

Unique identifier

Part of speech

Lexical variants

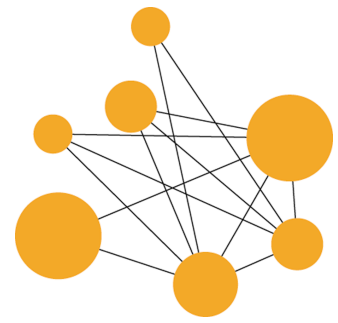
**Prepositional
phrase
complements**



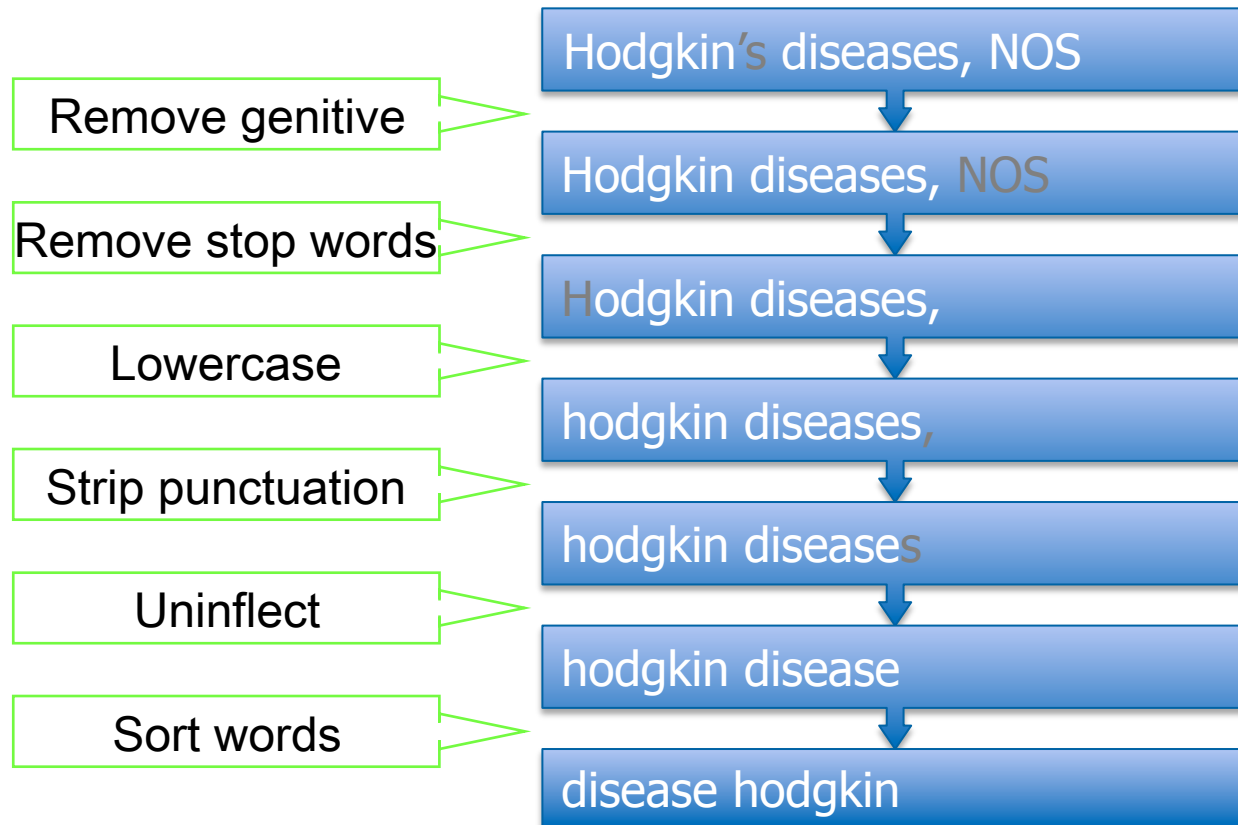
Lexical Tools

- Manage lexical variation in biomedical terminologies and text
- Used separately or with SPECIALIST Lexicon
- Perform transformations selected and ordered by users
- 3 primary programs: normalizer, word index generator, lexical variant generator

<http://umlslex.nlm.nih.gov/lvg/current/>



Normalization I



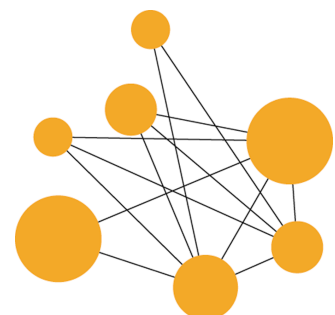


Normalization 2

Hodgkin Disease
HODGKINS DISEASE
Hodgkin's Disease
Disease, Hodgkin's
Hodgkin's, disease
HODGKIN'S DISEASE
Hodgkin's disease
Hodgkins Disease
Hodgkin's disease NOS
Hodgkin's disease, NOS
Disease, Hodgkins
Diseases, Hodgkins
Hodgkins Diseases
Hodgkins disease
hodgkin's disease
Disease, Hodgkin

normalize

disease hodgkin



BioFrameNet

Andrew Dolbey, PhD dissertation, 2009, BioFrameNet: a FrameNet Extension to the Domain of Molecular Biology

Transport Valence Pattern

Transport_destination	Transported_Entity	Transporting_Entity
PP[to]	NP	CNI
Dep	Obj	--

GRIF 82174 EntrezGene ID: 66013 [symbol:Arhgef9] PMID: 1521530

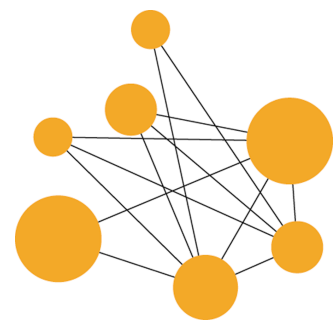
TRANSLOCATES gephyrin to submembrane microaggregates CNI

Transport Valence Pattern

Transport_destination	Transported_Entity	Transporting_Entity
PP[to]	NP	CNI
Dep	Ext (subj)	--

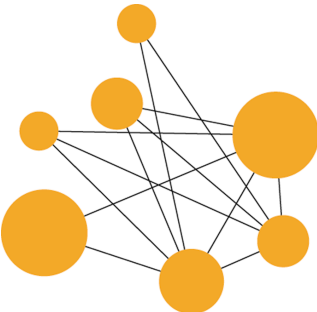
GRIF 72788 EntrezGene ID: 654817 [symbol:NCF1C] PMID: 1285569

p47phox is TRANSLOCATED to membrane ruffles through a VEGF-WAVE1 pathway CNI



ELDA BioLexicon

- Subcategorization frames + Event frames
 - Roles: agent, theme, manner, instrument, destination, condition, rate, descriptive agent, descriptive theme, purpose, *location*, *temporal*
 - Used annotations in Gene Regulation corpus and thematic hierarchies to guide linking
 - Result: 668 event frames for 168 verbs



BioVerbNet

preliminary effort

BioVerbNet v0.1

[VIEW](#) OR [MAN](#)

structural_modification-1

Members: 22, Frames: 4

[POST COMMENT](#)

CLASS HIERARCHY

STRUCTURAL_MODIFICATION-1*
NO SUBCLASSES

HYDROGENATE
HYDROXYLATE
HYPER-PHOSPHORYLATE
METHYLATE
MYRISTOYLATE
PALMITOYLATE

PALMYTOYLATE
PHOSPHORYLATE
POLYUBIQUITINATE
PRENYLATE
PROTANATE
SULFATION

SULFATION
SULPHATION
SUMOYLATE
UBIQUITINATE

nown about the substrates for ERK5 in vivo , however it has been suggested to phosphorylate connexin 43 [11] and the transcription factor MEF2C [12 - 14] ."

TION(DURING(E), CAUSE, PATIENT)

ined , we observed an increase in tyrosine phosphorylation in response to ligand (Figure 10A) ."

TION(DURING(E), CAUSE, ?PATIENT) **?PREP**(DURING(E), ?PATIENT, LOCATION)

g , the PDGFR² dimerizes and is autophosphorylated on as many as 13 cytoplasmic tyrosine residues ."

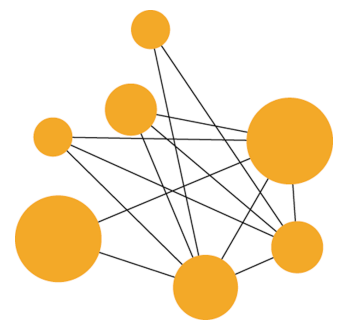
AT} LOCATION

TION(DURING(E), CAUSE, PATIENT) **PREP**(DURING(E), PATIENT, LOCATION)

ither MEF2C has functions which are independent of its phosphorylation by ERK5 in vivo at this developmental stage , or that other kinases such as p38 can also phosphorylate the same sites on MEF2C as

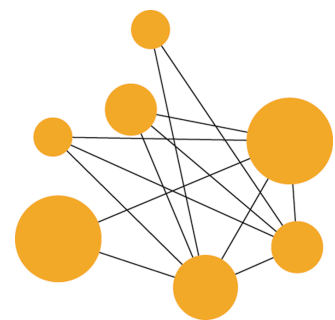
ON AT} PATIENT

TION(DURING(E), CAUSE, PATIENT) **PREP**(DURING(E), PATIENT, LOCATION)



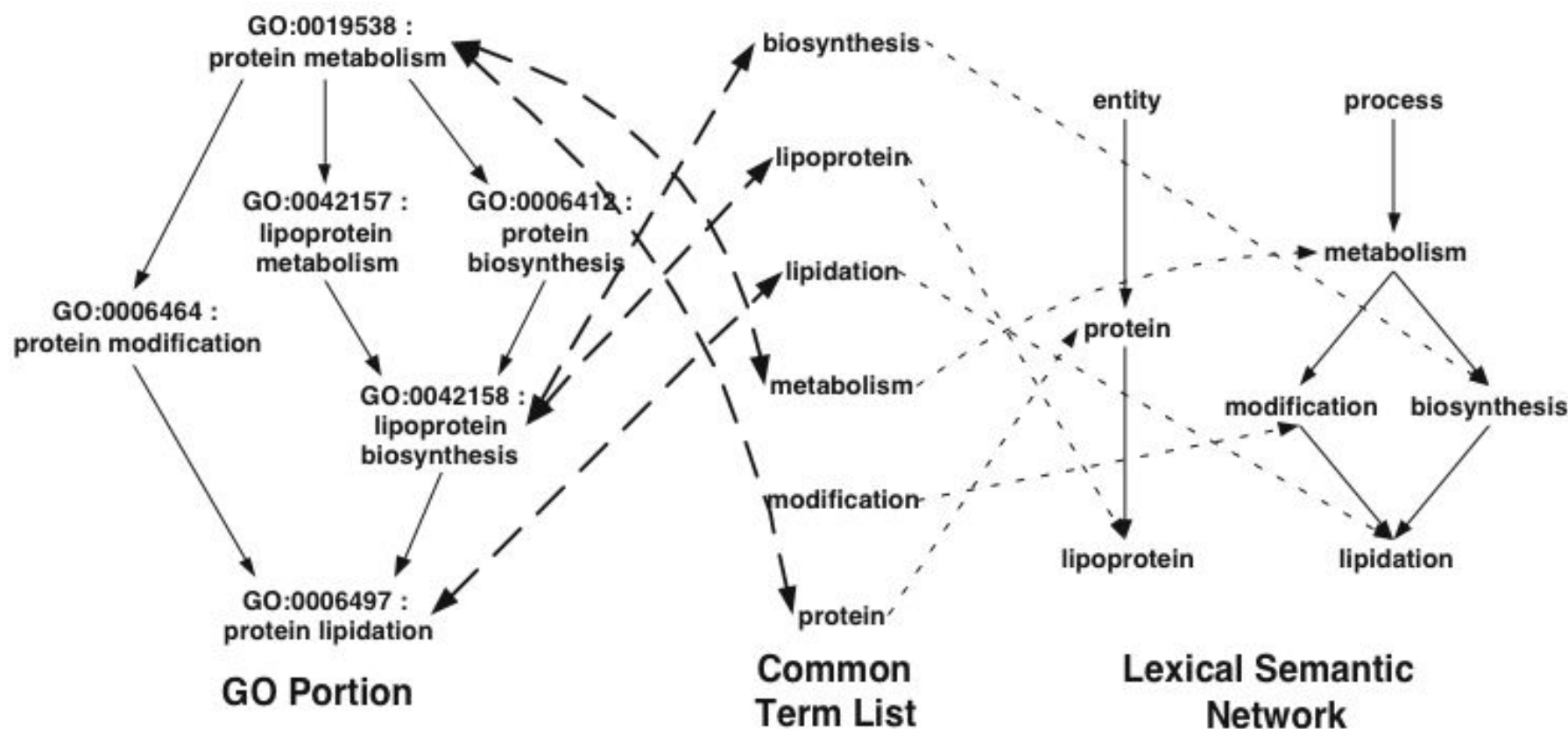
BioLemmatizer

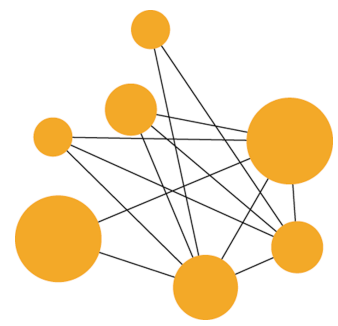
- New tool under development in our group
- Based on the MorphAdorner tool, using specialized lexicons
- Given {token, POS}
- Produce base form for token
- (default behavior for token w/out POS)



GO as a lexical semantic resource

- The Gene Ontology represents semantic relationships (is_a, part_of) between biological phrases representing molecular functions/processes
- Utilize the structure of the GO and lexical correspondences to infer relationships at the term level from relationships between phrases





Inferring Lexical Relations from GO

Parallel rule:

vanillin metabolism *isa* aldehyde metabolism \Rightarrow
vanillin *isa* aldehyde

lipoprotein biosynthesis *isa* lipoprotein metabolism \Rightarrow
biosynthesis *isa* metabolism

Modifier rule: blocking rule for modifiers

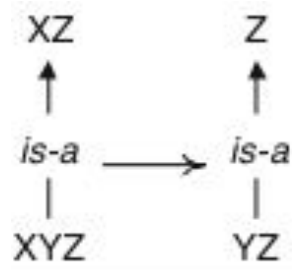
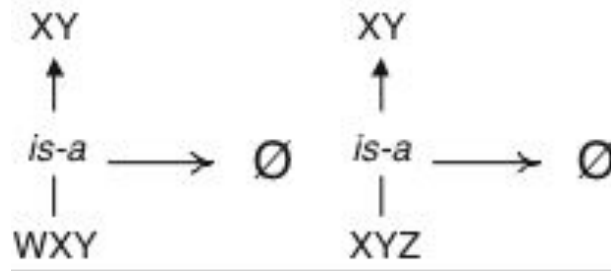
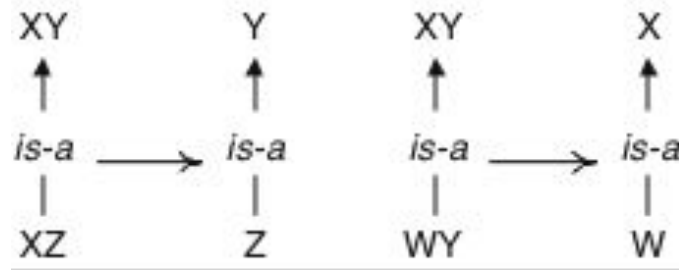
Positive gravitactic behavior *isa* gravitactic behavior $\Rightarrow \emptyset$

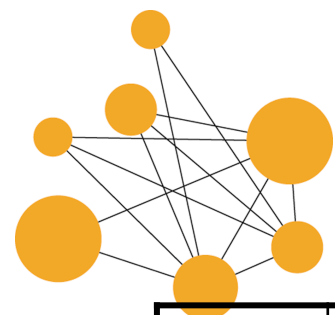
Larval feeding behavior (sensu insecta) *isa* Larval feeding
behavior $\Rightarrow \emptyset$

Insertion rule: right-branching heuristic

adult feeding behavior *isa* adult behavior \Rightarrow
feeding behavior *isa* behavior

chemosensory jump behavior *isa* chemosensory behavior
 \Rightarrow jump behavior *isa* behavior



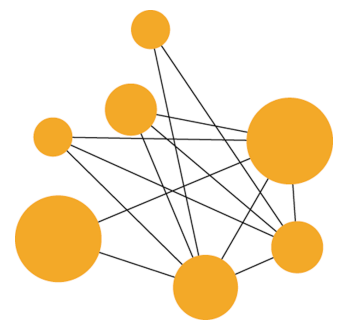


Relations inferred (with counts)

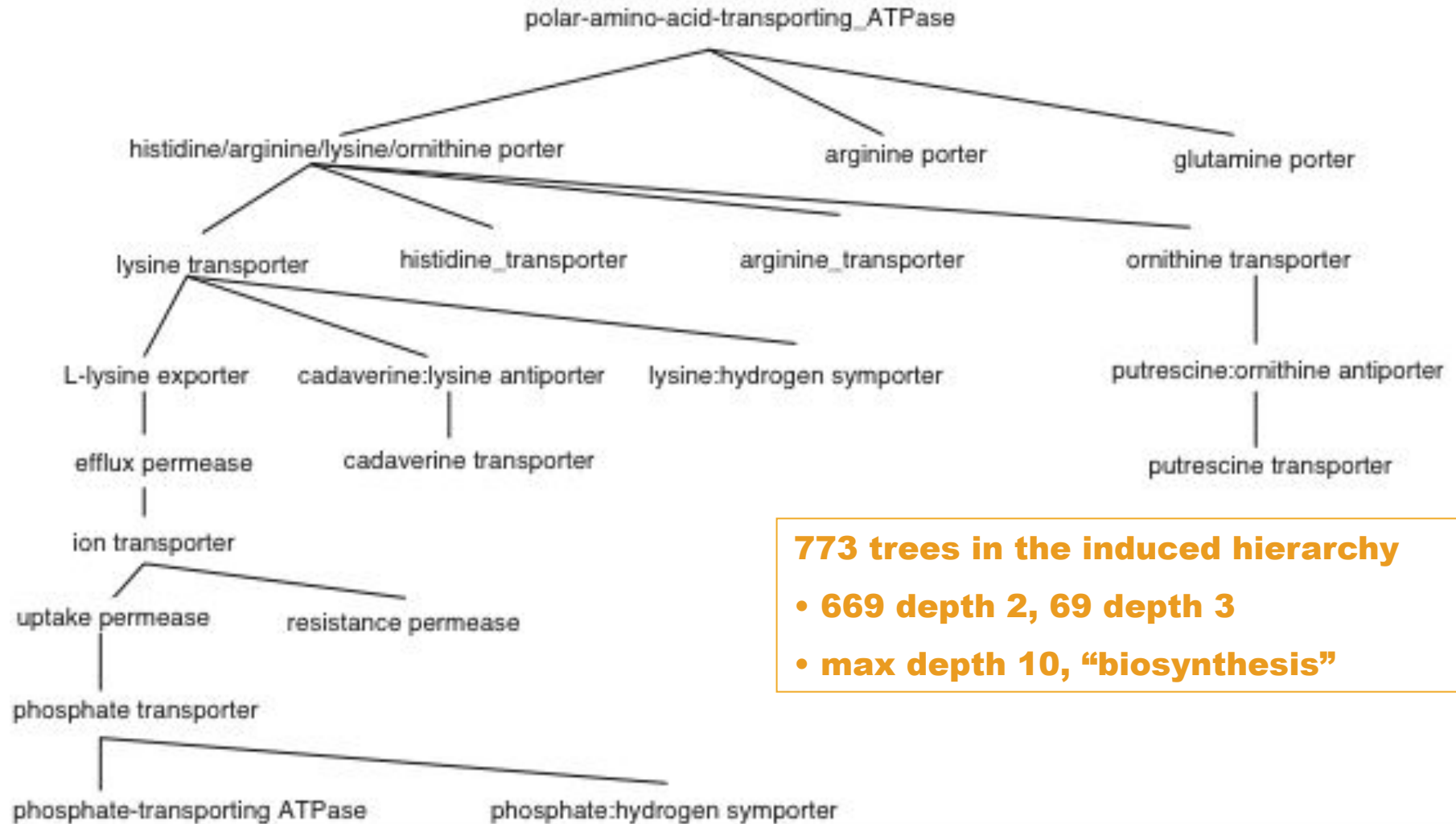
581	biosynthesis isa metabolism
577	catabolism isa metabolism
44	receptor isa binding
38	deoxyribonucleoside isa nucleoside
35	ribonucleoside isa nucleoside
33	permease isa transporter
27	Saccharomyces isa Fungi
22	porter isa transporter
15	oxidation isa metabolism
14	tRNA isa RNA

14	inhibitor isa regulator
13	ribonucleotide isa nucleotide
11	proliferation isa activation
11	differentiation isa activation
11	deoxyribonucleotide isa nucleotide
10	rRNA isa RNA
10	mRNA isa RNA
9	snRNA isa RNA
8	modification isa metabolism
8	methylation isa modification

6,364 unique relations inferred; only 70 already exist in the GO
3,270/6,589 unique labels inferred that do not occur in the GO
as terms

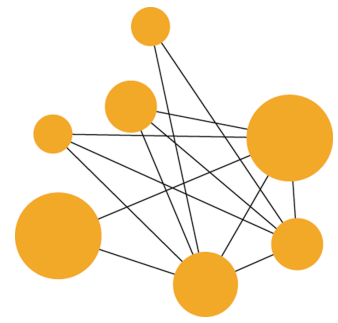


A portion of the induced network



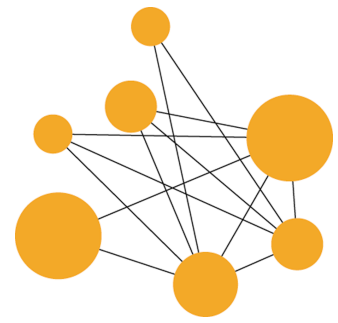
773 trees in the induced hierarchy

- 669 depth 2, 69 depth 3
- max depth 10, “biosynthesis”

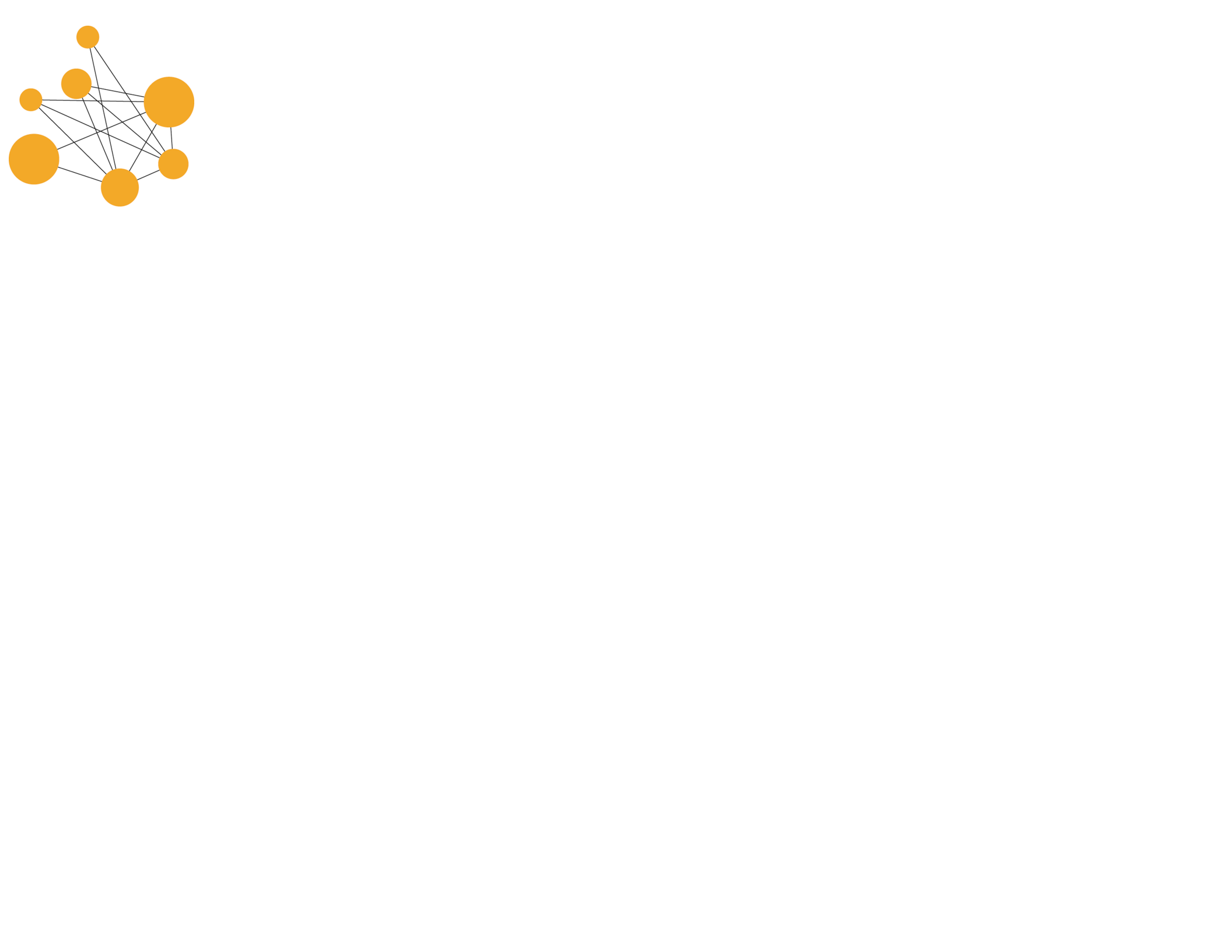


Acknowledgements

- Kevin B. Cohen
- NIH/NLM and NSF grants to me and to Larry Hunter
- Other members of the Center for Computational Pharmacology at University of Colorado Anschutz Medical Campus (Denver)



- **THANK YOU!**





- disk drive (was) down (at) 11/16-2305.
- (has) select lock.
- spindle motor is bad.
- (is) awp spindle motor.
- (disk drive was) up (at) 11/17-1236.
- replaced spindle motor.



Biomedical verb semantics

- Semantic Network (NLM)
 - Can relate objects in an ontology
- Friedman et al. (2002)
 - Complex embedding
- McDonald et al. (2005)
 - Arity > binary
- LSAT
 - Applied PASBio PAS to information extraction
- Kogan et al. (2005)
 - General & medical domains require different PAS



Some recent history

- 2004: PASBio publication (BMC Bioinformatics)
- 2005: Extension of PASBio to medical predicates (Kogan et al., AMIA)