

# Applications

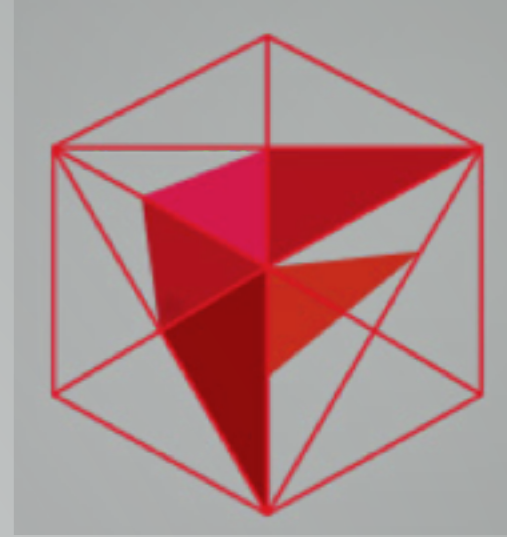
of FrameNet and ASRL

Collin F. Baker, ICSI

FrameNet Tutorial

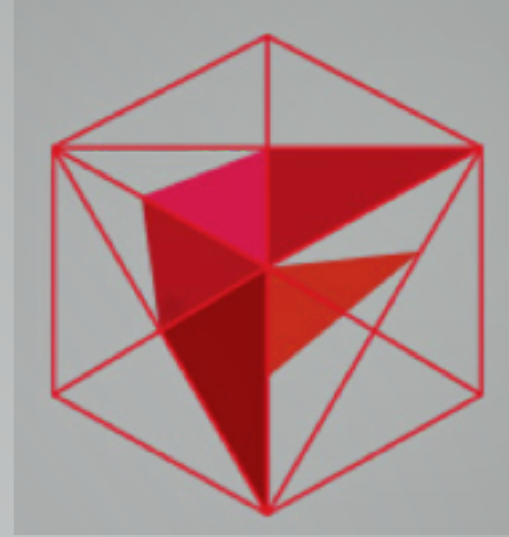
May 31, 2015

# FrameNet Brasil



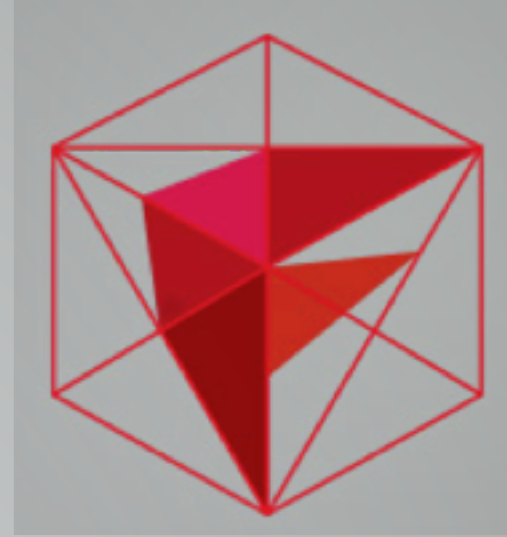
- Housed at Universidade Federal de Juiz de Fora, Minas Gerais, Brazil
- PIs: Maria Margarida Martins Salomão and Tiago Timponi Torrent

# FrameNet Brasil



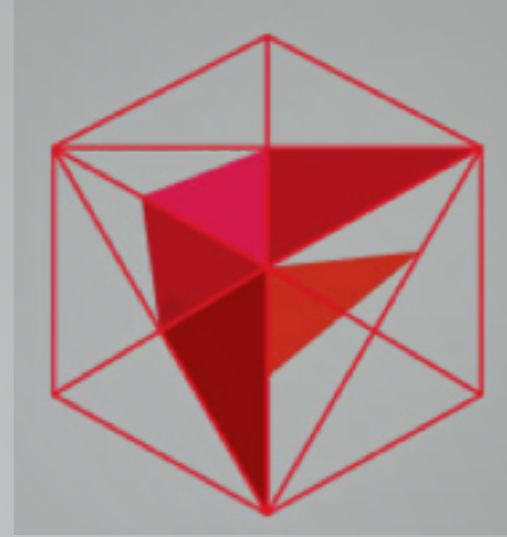
- Three projects:
  - Building Construction for Brazilian Portuguese
  - Copa 2014 (2014 Soccer World Cup)
  - Olympia (2016 Summer Olympics)

# FrameNet Brasil: Constructicon



- A repertoire of syntactic constructions, structures whose recognition, treatment and interpretation are key for Natural Language Processing
- Starting with 2 types of constructions:
  - *Para* + infinitive
  - Multiword quantifiers as example of constructions
    - comp. ling. "multiword expression" = Cog. Ling. "construction"?
    - clear division between work of lexicon and work of construction
    - full integration of both in parsing, NLP
    - Comparative conx. in EN, BR PT, SE (collab. with Swedish FN)

# FrameNet Brasil: Integrating metaphor



- model productive metaphors in Brazilian Portuguese, or import them from other databases (MetaNet)

- describe metaphorical conx., e.g., the Inceptive Aspect Construction:

Maria	rompeu	a chorar
Maria	break.PAST.3SG	to cry
Maria	broke	into tears

- create metonymy links
- infer the relations between schemas and the more specific frames via the network of relations

# Copa do Mundo 2014

(Salomão et al. 2011, 2013)



**FRAMENET**  
**BRASIL**  
COPA DO MUNDO

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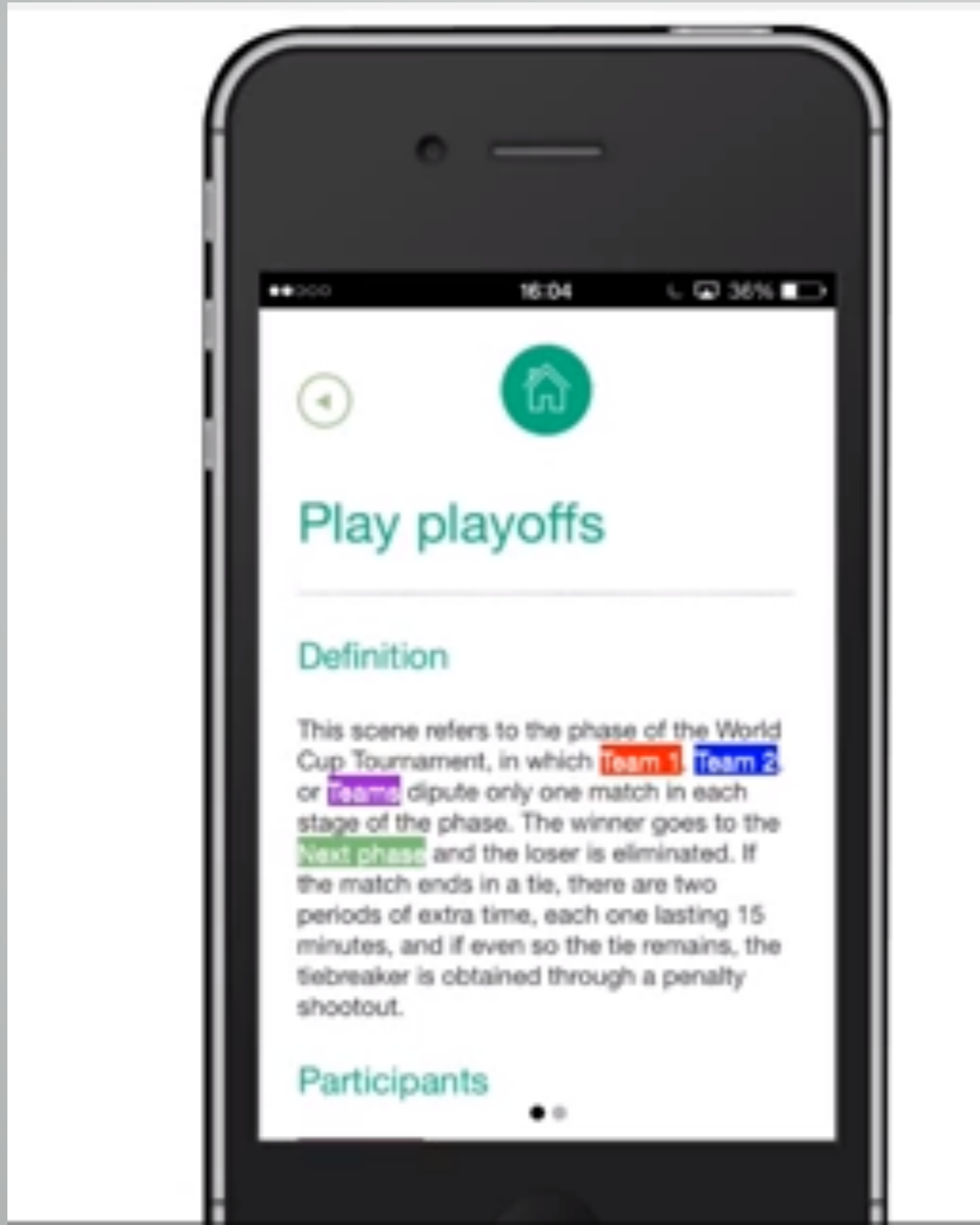
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- Website was used by tourists, tourism professionals, journalists and staff of 2014 World Cup in Brazil.

# Copa do Mundo

<https://youtu.be/Zi7vEIBzU68>



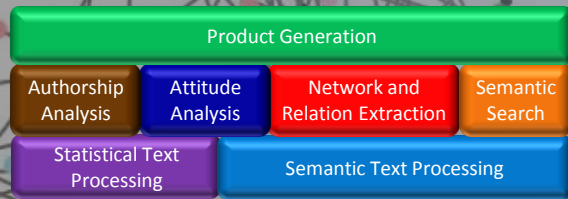
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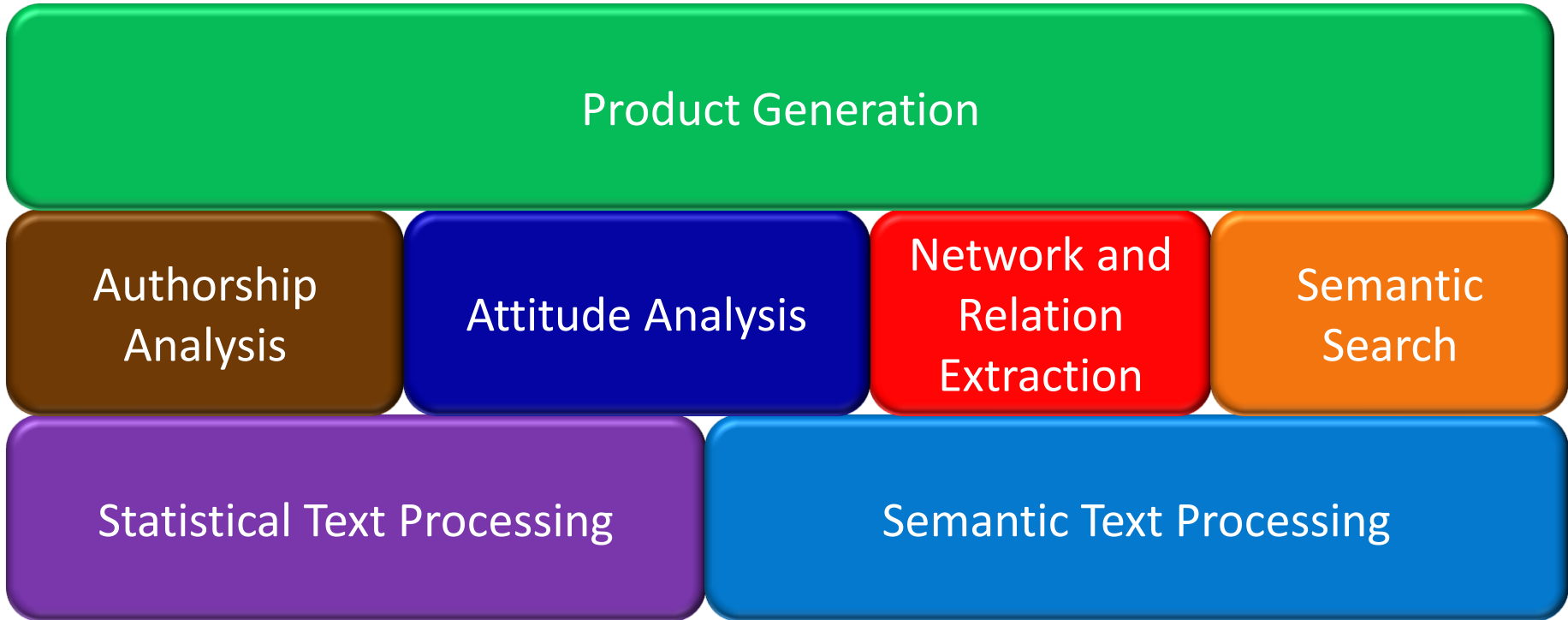
Olympics version  
with frames for all  
sports coming  
soon!

# Decisive Analytics Corporation

- Long-term collaboration with FN through a series of subcontracts, e.g. our current work on
  - Spatial relations
  - Negation, tense, mood and aspect
- E.g. of their products:
  - Network extraction
  - Attitude analysis
  - Semantic search

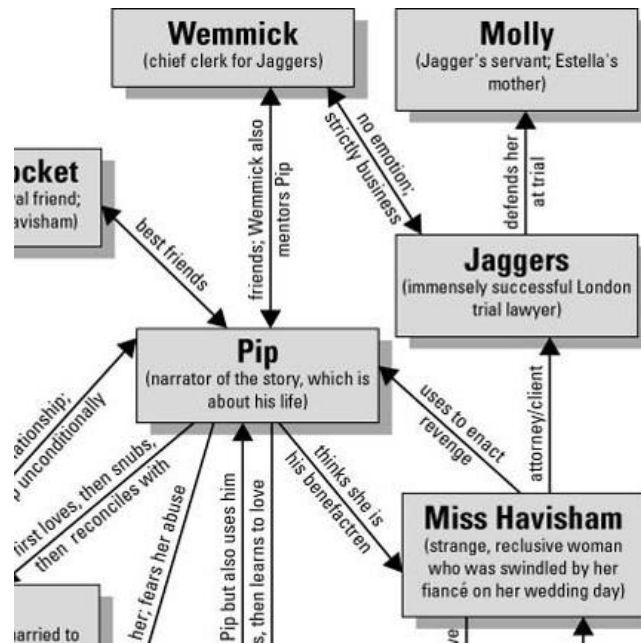
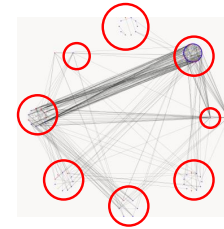


# Technology Profile



# Network Extraction and Refinement

- Network Extraction
  - *Disentangling the hairball of networks discovered in text*



# Motivation

- Gathering everything we've been told about how entities (Person, Organization, Location) are connected in the world
- Avoiding a abstraction to one-type-of-node/one-type-of-edge
- Avoiding the “hairball” problem



# Network and Relation Extraction

## Source Text:

Fenig Nabon smuggled the undocumented containers from the Shak Clan.

A source reported that Fenig Nabon bought a large quantity of merchandise form Talon Karrde.

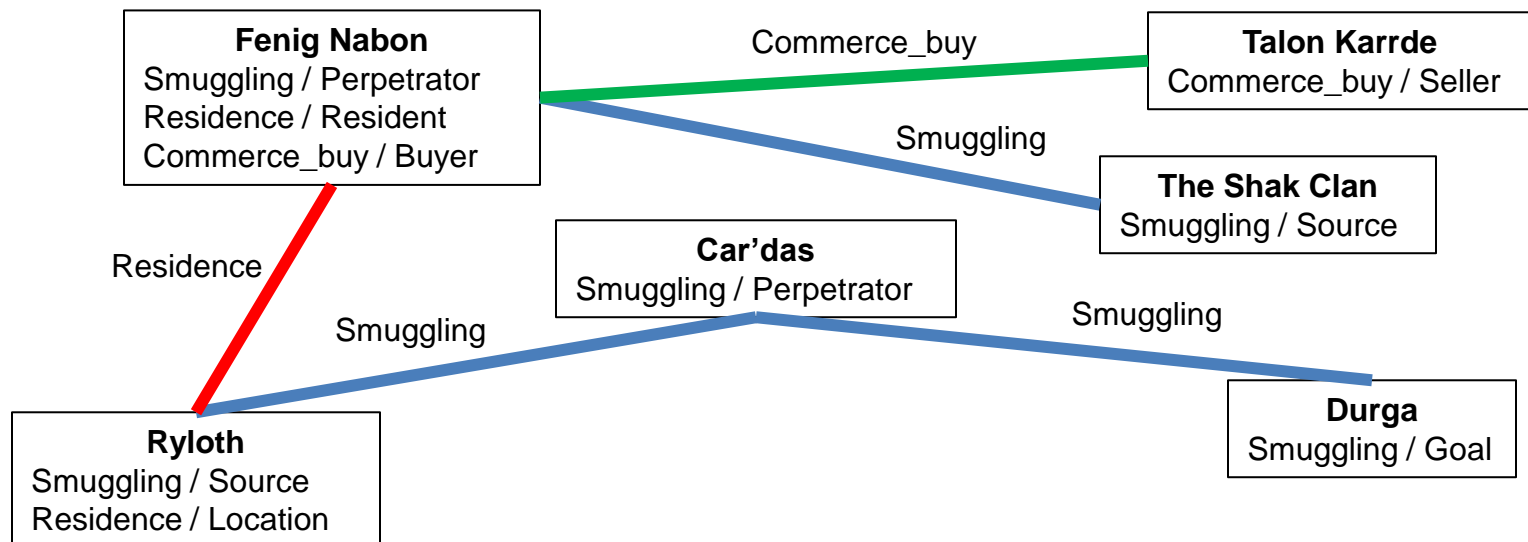
The Car'das smuggled a troupe of dancers form Ryloth to Durga.

Nabon resides in Ryloth.

## Semantic Labeling:

Labeled sentence	Frame	Roles
Fenig Nabon smuggled the undocumented containers from the Shak Clan.	smuggled → Smuggling	Perpetrator Goods Source
A source reported that Fenig Nabon bought a large quantity of merchandise from Talon Karrde.	bought → Commerce_buy	Buyer Goods Seller
The Car'das smuggled a troupe of dancers from Ryloth to Durga.	smuggled → Smuggling	Perpetrator Goods Source Goal
Nabon resides in Ryloth.	resides → Residence	Resident Location

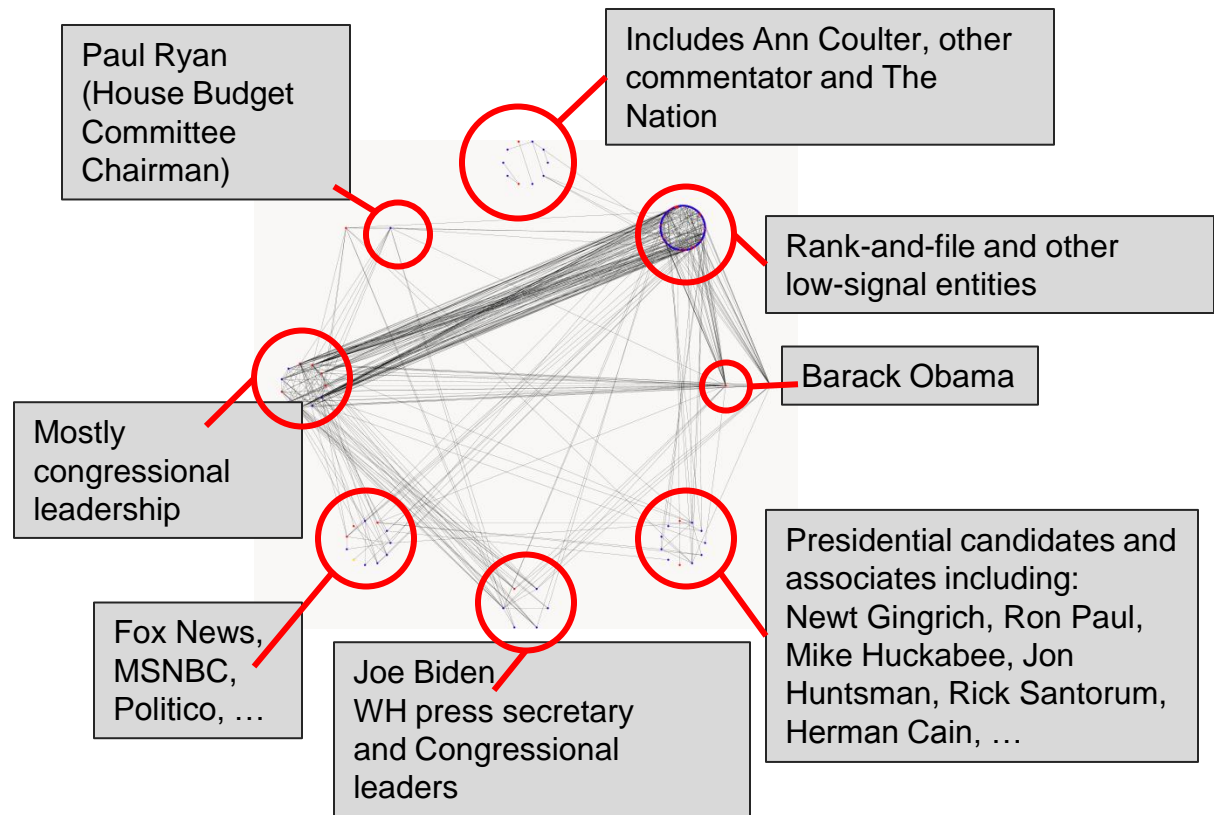
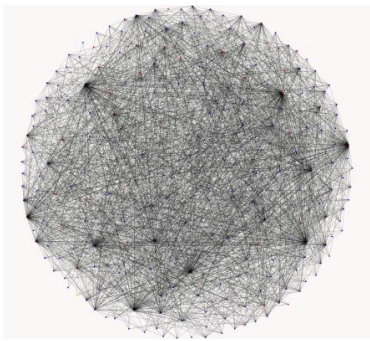
## Network:



# Organization Discovery

Relational modeling algorithms can identify the latent organization within a large, complex graph (e.g. solve the hairball problem)

Network automatically generated from corpus of articles about debt-ceiling debate

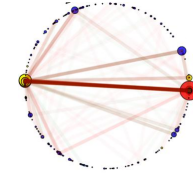


# Network Extraction: Conclusion

- An approach for going from frame labeled data to an entity network
- Filtering can focus analysis
- Relational modeling can reorganize a network into meaningful clusters based on frame data

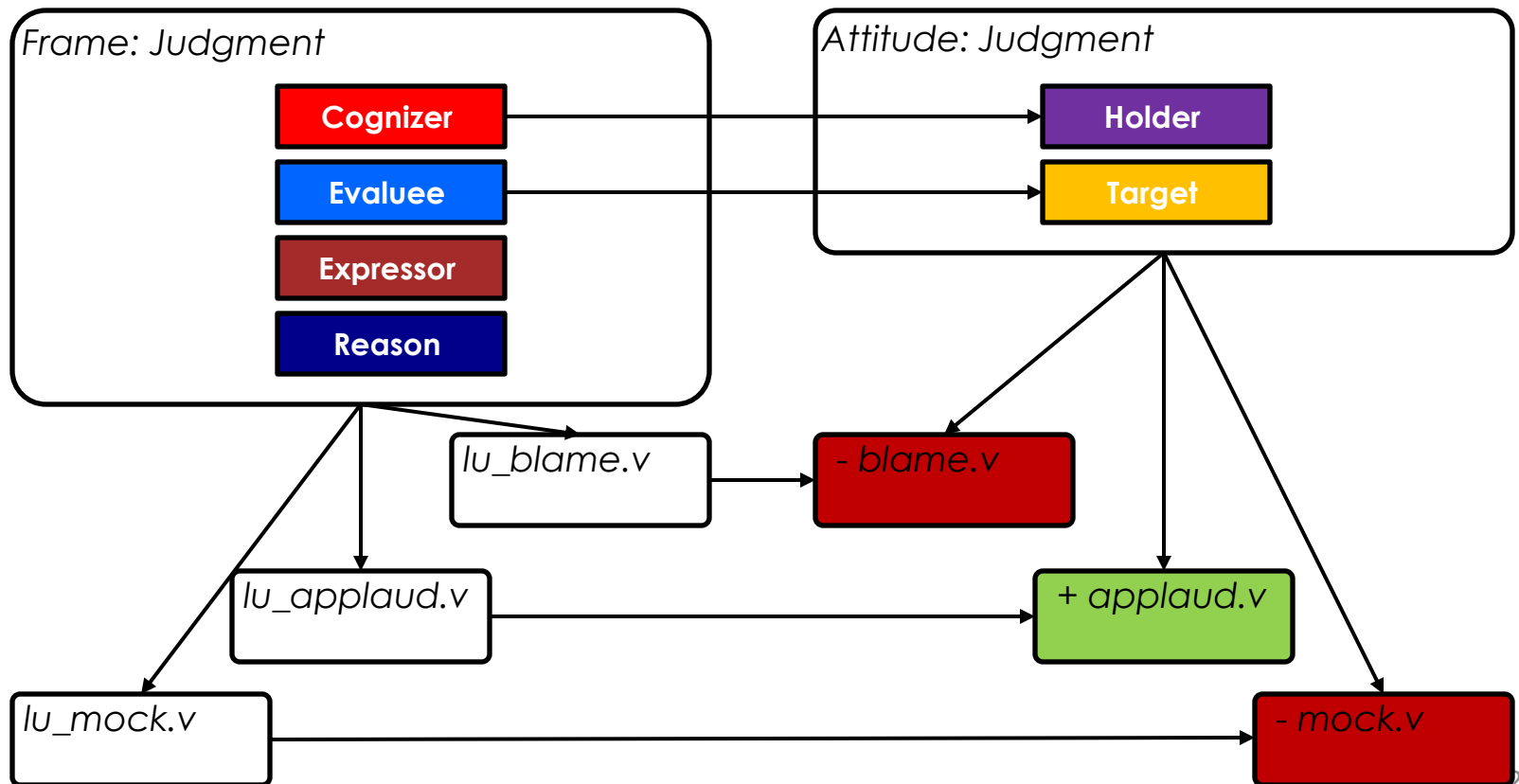
# Attitude Analysis

- Attitude Analysis
  - *Identifying Factions in Connected Entities*



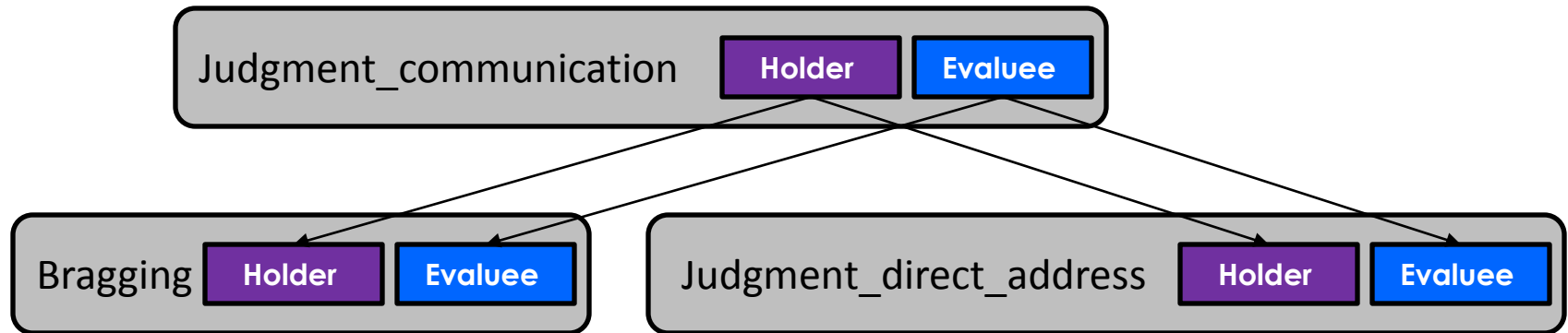
# How this was actually done

- Start with a semi-manual process mapping  
FrameNet to Attitudes

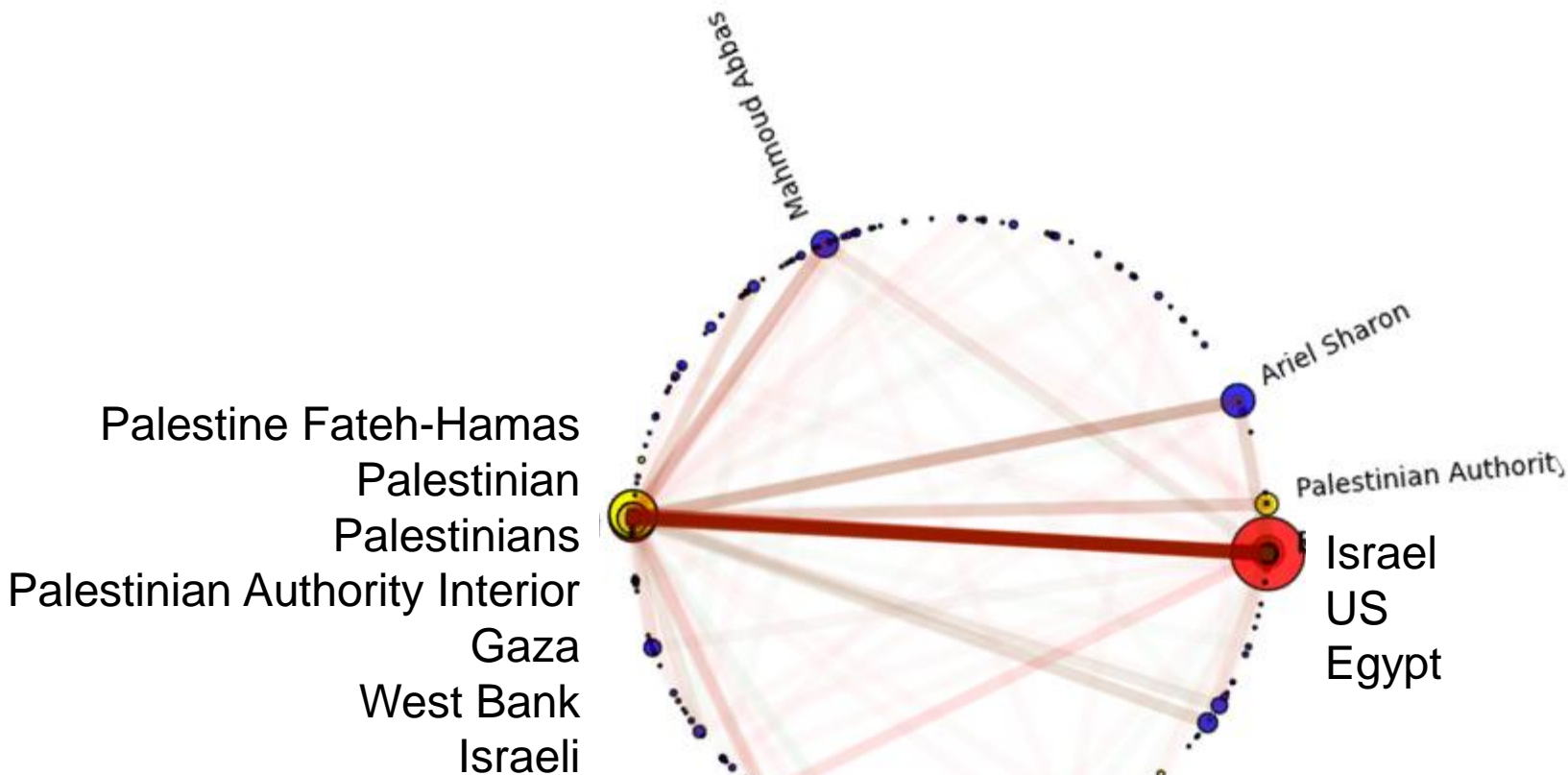


# The mapping process

- A semi-manual mapping process
  - Start with FrameNet and Polarity Lexicon
  - 4. Inheritance links to pre-populate Holder/Target mapping for Frame Elements



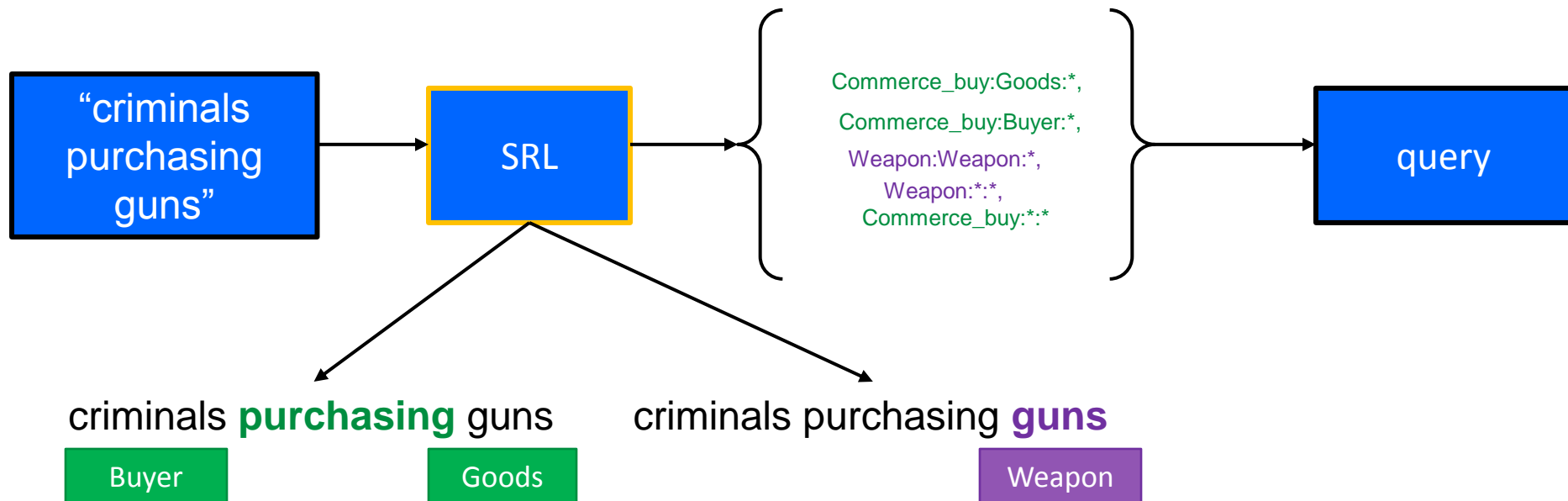
# How do different people feel about each other?



Measure	Fraction	Accuracy
Membership	10/12	83.3%
Implied Relations	25/35	71.4%

# Automatic Query Generation

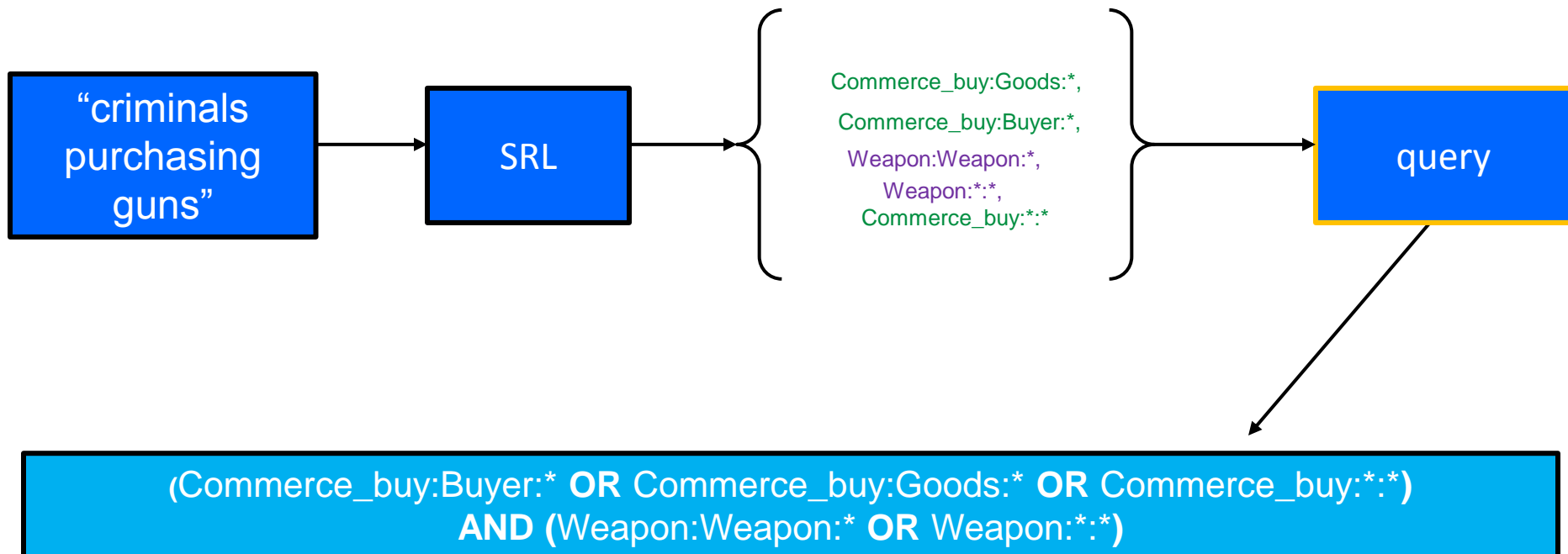
Take plain text input and generate a FrameNet-based query.



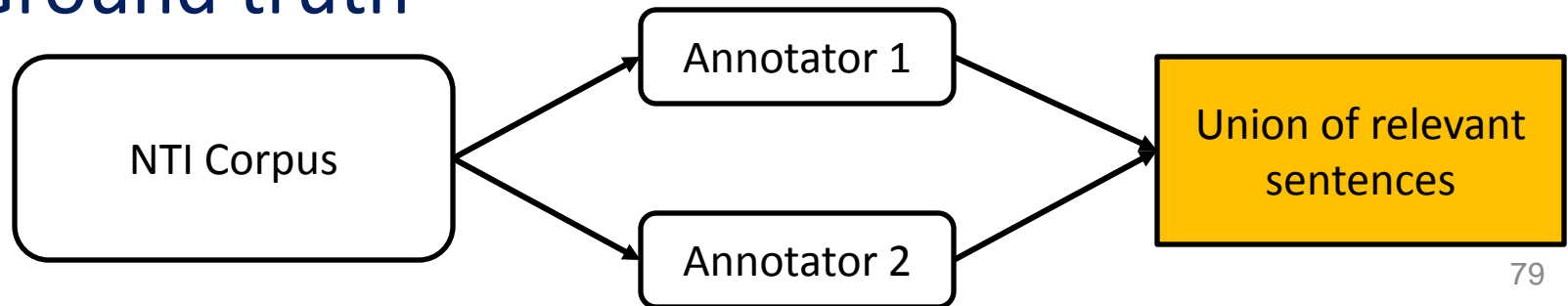


# Automatic Query Generation

Take plain text input and generate a FrameNet-based query.



- Hypothetical scenario
  - Draw links between countries that have exchanged resources (technology, money, supplies, people, etc.) with respect to chemical, nuclear, and biological warfare
- Dataset
  - Texts from the Nuclear Threat Initiative website
- Ground truth



# Moving around the FrameNet links

- **Ground truth**

**China is a key supplier of CW equipment and technologies to Iran according to US Department of Defense.**

**North Korea also allegedly provided more than 10 scientists to work on the Libyan missile program .**

**Other defensive material purchased by Iran includes respirators from Spain, protective gear from South Korea, and atropine autoinjectors from the Netherlands.**

# Moving around the FrameNet links

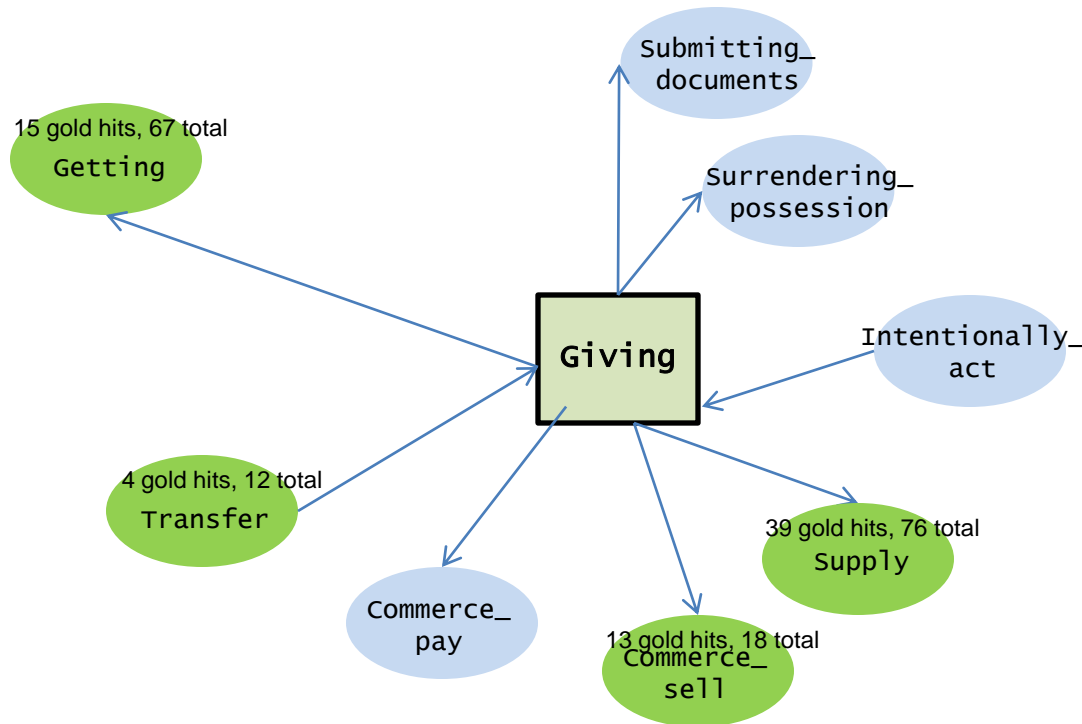
0 hops



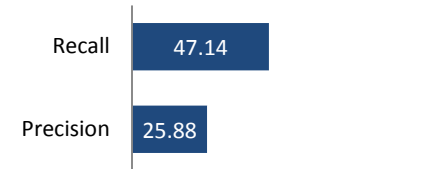
Giving

The **Giving** frame is central to the idea of resource exchange; however, a frame search for **Giving** with no hops does not find any relevant results, so the recall and precision metrics are both 0.

# Moving around the FrameNet links



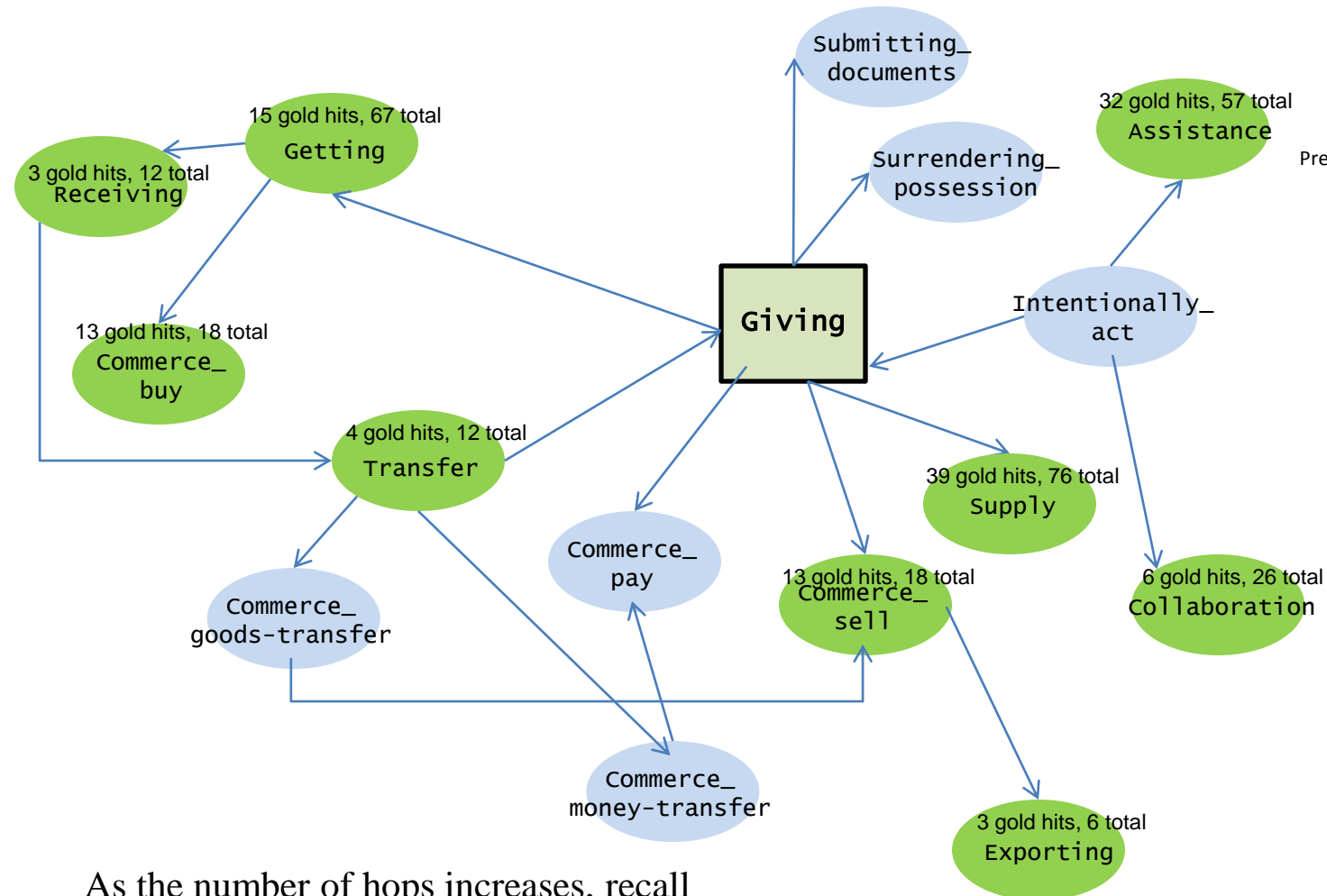
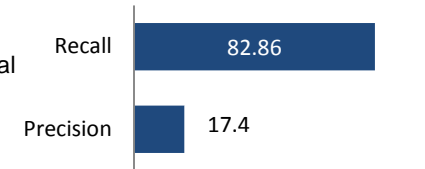
1 hop



Searching over all frames 1 hop from the original **Giving** frame returns almost half of all the relevant results.

# Moving around the FrameNet links

2 hops

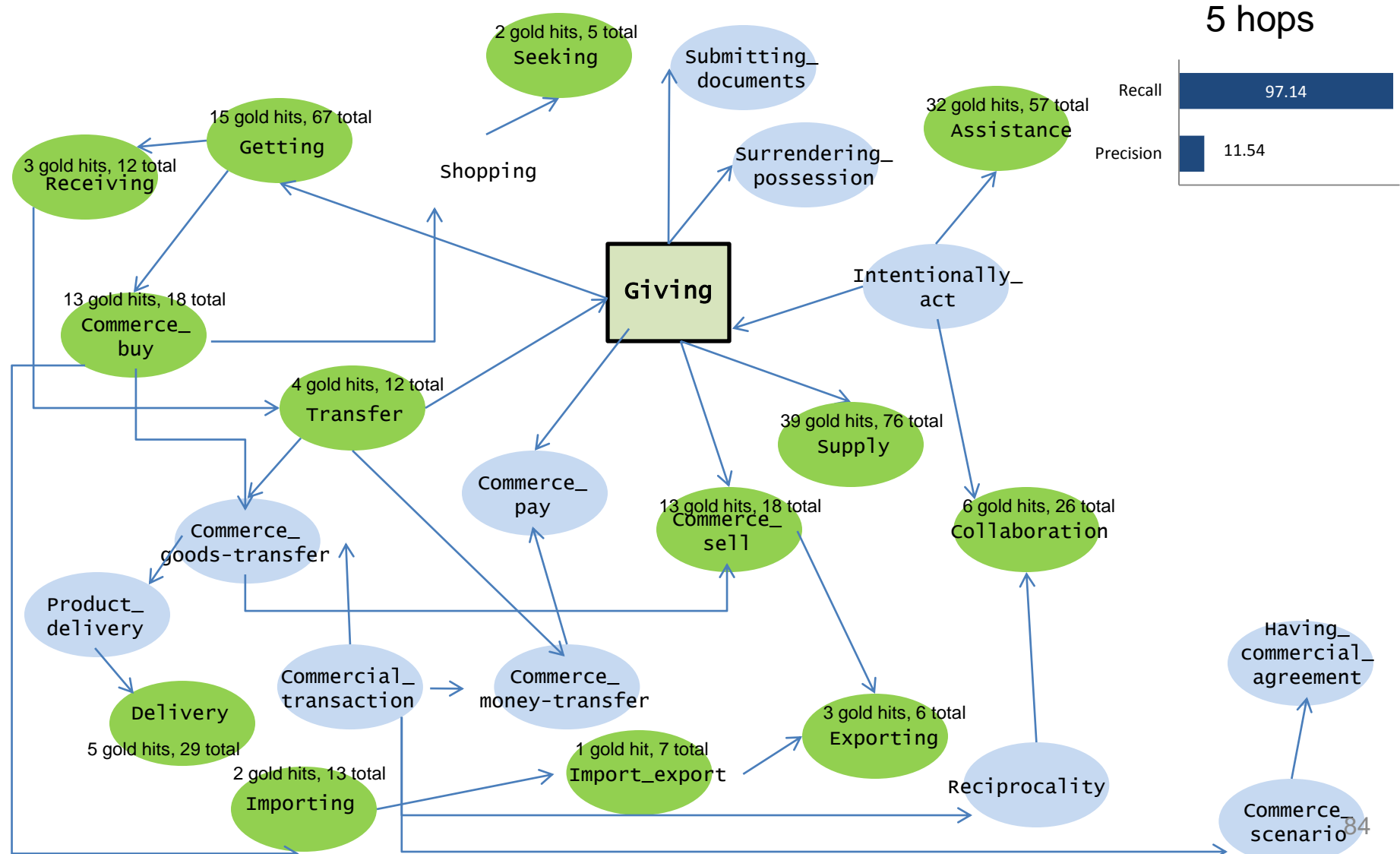


As the number of hops increases, recall increases and precision decreases.

# Moving around the FrameNet links

5 hops

Recall	97.14
Precision	11.54



# Semantic Search: Conclusion

- Ability to execute queries over frames, frame elements, and “terms”
- Terms can be abstracted too
- Queries may be generated from simple text
- Results come in several forms
- Many more features on the docket



Thank you!