# Semantic Role Labeling Tutorial NAACL, June 9, 2013

Part I: Martha Palmer, University of Colorado Part 2: Shumin Wu, University of Colorado

Part 3: Ivan Titov, Universität des Saarlandes







### Outline

Part I
 Linguistic Background, Resources, Annotation
 Martha Palmer, University of Colorado

Part 2
 Supervised Semantic Role Labeling and Leveraging
 Parallel PropBanks
 Shumin Wu, University of Colorado

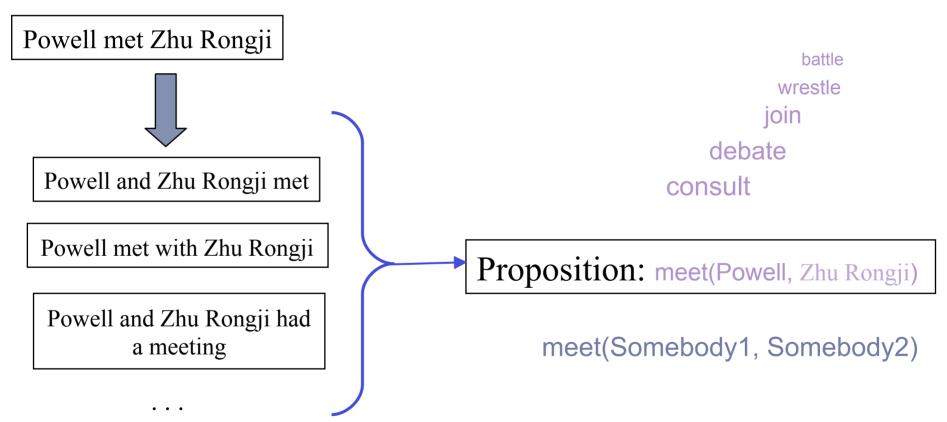
Part 3

Semi-, unsupervised and cross-lingual approaches

Ivan Titov, Universität des Saarlandes, Universteit van Amsterdam



# Motivation: From Sentences to Propositions Who did what to whom, when, where and how?



When Powell met Zhu Rongji on Thursday they discussed the return of the spy plane.

meet(Powell, Zhu) discuss([Powell, Zhu], return(X, plane))



## Capturing semantic roles

#### **SUBJ**

Dan broke [ the laser pointer.]

#### **SUBJ**

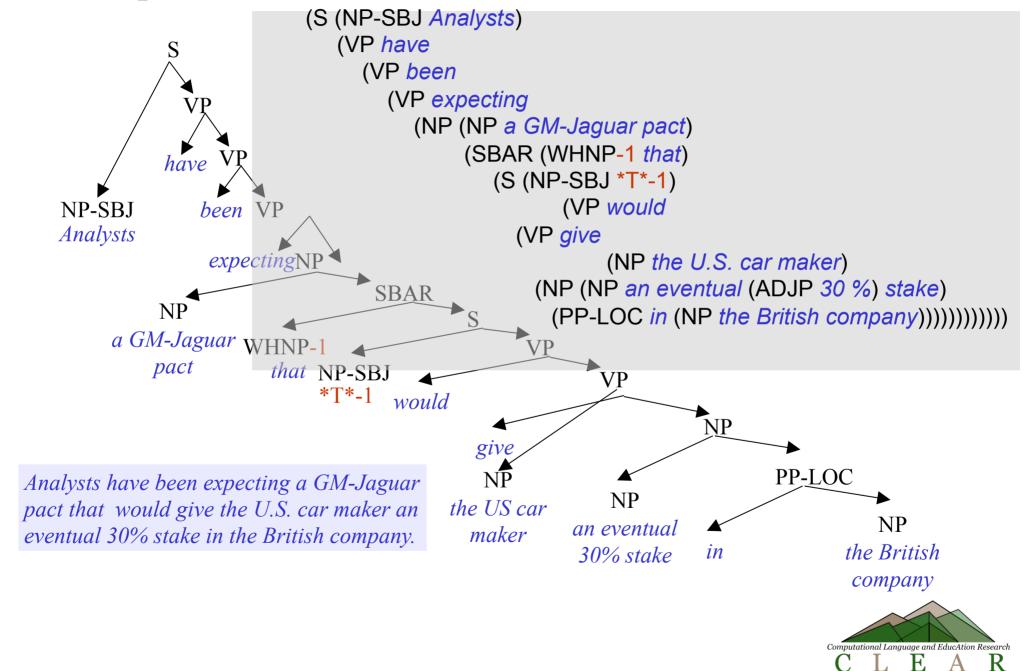
▶ [The windows] were broken by the hurricane.

#### SUBJ

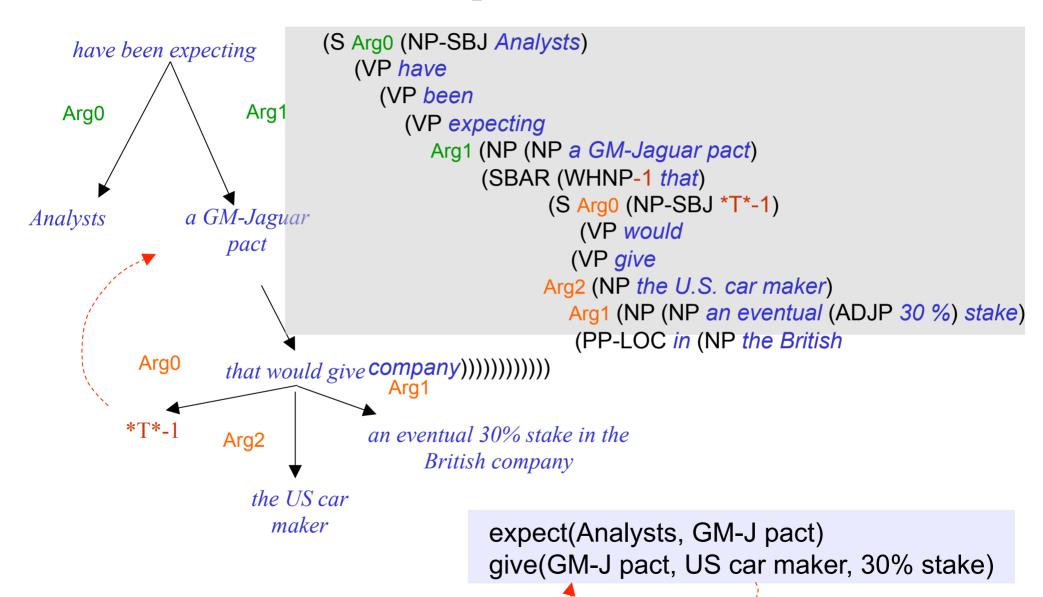
▶ [The vase] broke into pieces when it toppled over.



## PropBank - A TreeBanked Sentence



## The same sentence, PropBanked



## **SRL** Questions

- Why Arg0 and Arg1?
- What about nouns and adjectives?
- What about other languages?
- How does PropBank relate to VerbNet and FrameNet?
- Will we ever get past the WSJ?
- ▶ How do SRL systems get trained?
- Can this be done without training data?
- . . . . . .



## Why Arg0? Linguistic Background and Resources

- ▶ Fillmore Cases
  - Useful generalizations, fewer sense distinctions,
- Dowty Proto-typical Agents and Patients
  - A bag of "agentive" entailments
  - PropBank
- Levin Verb classes based on syntax
  - Syntactic behavior is a reflection of the underlying semantics
  - VerbNet
- Back to Fillmore and FrameNet
- SemLink
- ▶ PropBank → AMR



## Linguistic Background: Case Theory,

The Case for Case, Charles J. Fillmore

- Case relations occur in deep-structure
  - Surface-structure cases are derived
- A sentence is a verb + one or more NPs
  - Each NP has a deep-structure case
    - ► A(gentive)
    - ► l(nstrumental)
    - ▶ D(ative) recipient
    - ► F(actitive) result
    - ▶ L(ocative)
    - ▶ O(bjective) affected object, theme
  - Subject is no more important than Object
    - Subject/Object are surface structure



## Case Theory Benefits - Generalizations

#### Fewer tokens

- Fewer verb senses
- ▶ E.g. cook/bake [ \_\_O(A)] covers
  - Mother is cooking/baking the potatoes
  - The potatoes are cooking/baking.
  - Mother is cooking/baking.

## Fewer types

- "Different" verbs may be the same semantically, but with different subject selection preferences
- ▶ E.g. like and please are both [ \_\_O+D]
- Great, let's do it!



## Oops, problems with Cases/Thematic Roles

- How many and what are they?
- Fragmentation: 4 Agent subtypes? (Cruse, 1973)
  - The <u>sun</u> melted the ice./<u>This clothes dryer doesn't dry clothes well</u>
- Ambiguity: Andrews (1985)
  - Argument/adjunct distinctions Extent?
  - The kitten licked my fingers. Patient or Theme?
- $\Theta ext{-}$ Criterion (GB Theory): each NP of predicate in lexicon assigned unique θ-role (Chomsky 1981).

```
[Agent (or Source) Esau] sold [Theme his birthright] [Goal to Jacob] for a bowl of porridge.
```

```
[Goal Esau] sold his birthright
[Source to Jacob] for a [Theme bowl of porridge].

Jackendoff
```



# Thematic Proto-Roles and Argument Selection, David Dowty, 1991

Role definitions have to be determined verb by verb, and with respect to the other roles

- Event-dependent Proto-roles introduced
  - Proto-Agent
  - Proto-Patient
- Prototypes based on shared entailments



## Proto-Agent- the *mother*

### Properties

- Volitional involvement in event or state
- Sentience (and/or perception)
- Causes an event or change of state in another participant
- Movement (relative to position of another participant)
- (exists independently of event named)
  - \*may be discourse pragmatic

### Proto-Patient – the *cake*

### Properties:

- Undergoes change of state
- Incremental theme
- Causally affected by another participant
- Stationary relative to movement of another participant
- (does not exist independently of the event, or at all)
- \*may be discourse pragmatic



## Argument Selection Principle

- For 2 or 3 place predicates
- Based on empirical count (total # of entailments for each role).
  - ▶ Greatest number of Proto-Agent entailments → Subject;
  - ▶ greatest number of Proto-Patient entailments → Direct Object.
- Alternation predicted if number of entailments for each role similar (non-discreteness).

```
[Mother AGENT] baked a cake.
[The cake PATIENT] baked.
```



# PropBank Semantic Role Labels – based on Dowty's Proto-roles

## PropBank Frame for break:

Frameset break.0 I "break, cause to not be whole":

Arg0: breaker

Argl: thing broken

Arg2: instrument

Arg3: pieces

## Why numbered arguments?

- Lack of consensus concerning semantic role labels
- Numbers correspond to verb-specific labels
- Arg0 Proto-Agent, and Arg1 Proto-Patient, (Dowty, 1991)
- Args 2-5 are highly variable and overloaded poor performance



# PropBank seeks to provide consistent argument labels across different syntactic realizations

### Uuuuuusually...

- Arg0 = agent, experiencer
- Argl = patient, theme
- Arg2 = benefactive / instrument / attribute / end state
- Arg3 = start point / benefactive / instrument / attribute
- Arg4 = end point

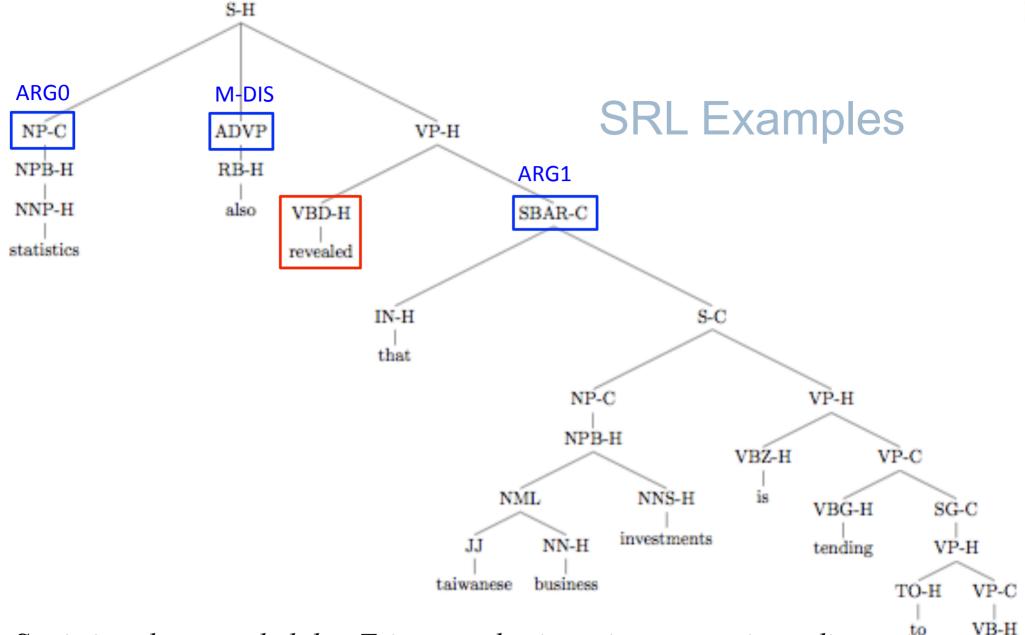


# PropBank seeks to assign functional tags to all modifiers or adjuncts to the verb

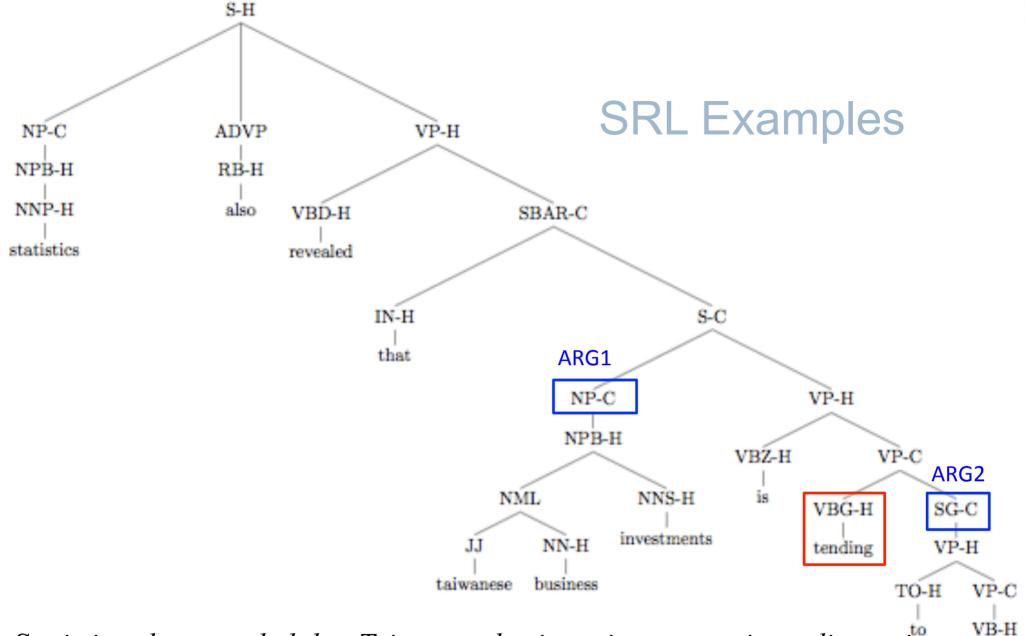
## Variety of ArgM's:

- TMP when? yesterday, 5pm on Saturday, recently
- LOC where? in the living room, on the newspaper
- DIR where to/from? down, from Antartica
- MNR how? quickly, with much enthusiasm
- PRP/CAU -why? because ..., so that ...
- REC himself, themselves, each other
- GOL end point of motion, transfer verbs? To the floor, to Judy
- ADV hodge-podge, miscellaneous, "nothing-fits!"
- PRD this argument refers to or modifies another: ...ate the meat <u>raw</u>

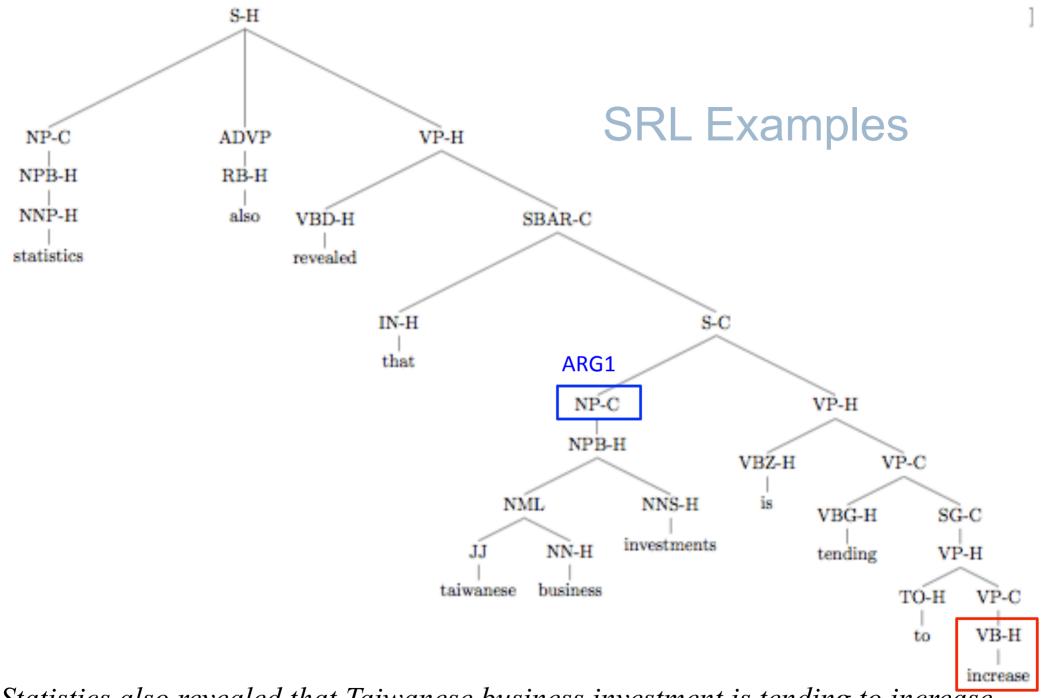




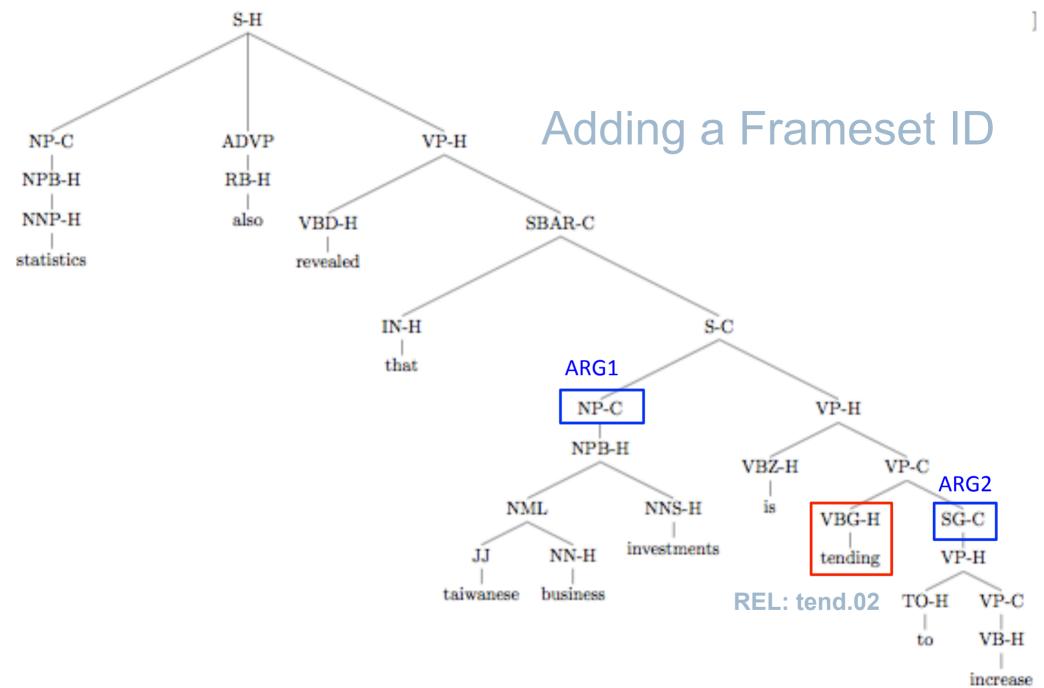








Computational Language and EducAtion Research



Computational Language and EducAtion Research

# Why do we need Frameset ID's?

PropBank Frames Files: tend.01, care for

### Roles:

Arg0: tender

Arg1: thing tended (to)

Example: John tends to the needs of his patrons.

Arg0: John

REL: tend

Arg1: the needs of his patrons



# Sense distinctions in PropBank – coarse-grained

PropBank - Frames Files: tend.02, have a tendency

### Roles:

Arg1: Theme

Arg2: Attribute

Example: The cost, or premium, tends to get fat in times of crisis.

Arg1: The cost, or premium

REL: tend

Arg2: to get fat in times of crisis.

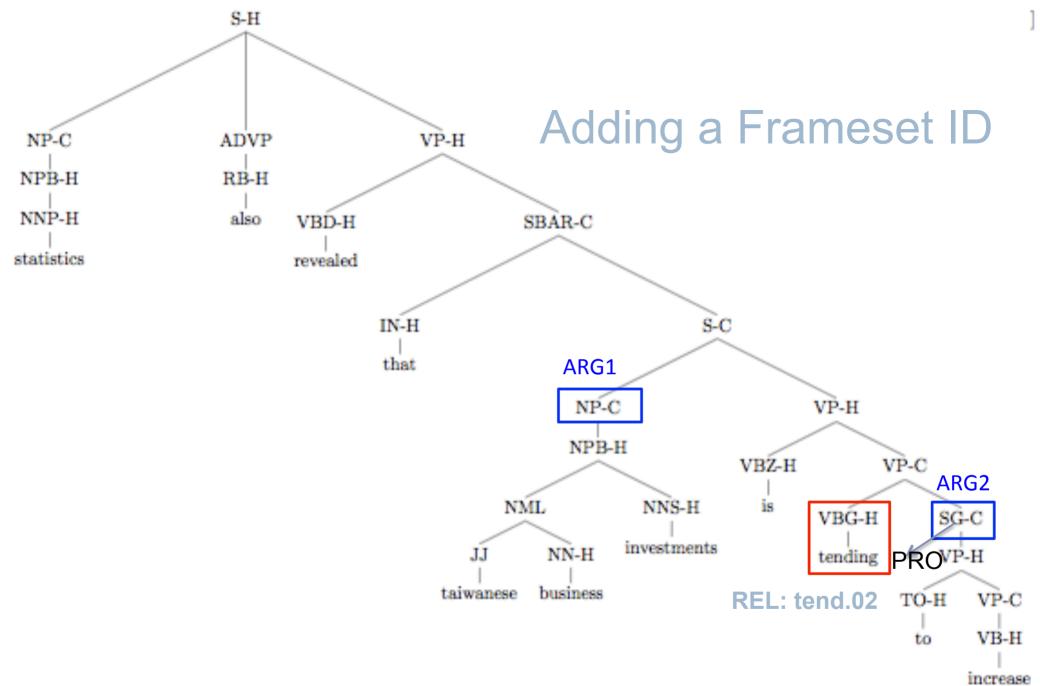


## Visual Example: traces BASED on Jubilee

Choi, et. al., NAACL-10 Demo

```
SBAR
   WHNP-1
      WP What
      NP-SB| 2:1-ARGO
        PRP I
        VBP have
           VBN experienced 4:0-rel
           NP 5:1-ARG1
              -NONE- *T*-1
            NP-ADV 6:1-ARGM-REC
              PRP myself
CC or
```





### Actual data for leave

Leave .01 "move away from" Arg0 rel Arg1 Arg3 Leave .02 "give" Arg0 rel Arg1 Arg2

sub-ARG0 obj-ARG1 44
sub-ARG0 20
sub-ARG0 NP-ARG1-with obj-ARG2 17
sub-ARG0 sub-ARG2 ADJP-ARG3-PRD 10
sub-ARG0 sub-ARG1 ADJP-ARG3-PRD 6
sub-ARG0 sub-ARG1 VP-ARG3-PRD 5
NP-ARG1-with obj-ARG2 4
obj-ARG1 3
sub-ARG0 sub-ARG2 VP-ARG3-PRD 3



## Annotation procedure, WSJ PropBank

Palmer, et. al., 2005

- PTB II Extraction of all sentences with given verb
- Create Frame File for that verb Paul Kingsbury
  - (3100+ lemmas, 4400 framesets, 118K predicates)
  - Over 300 created automatically via VerbNet
- First pass: Automatic tagging (Joseph Rosenzweig)
  - http://www.cis.upenn.edu/~josephr/TIDES/index.html#lexicon
- Second pass: Double blind hand correction

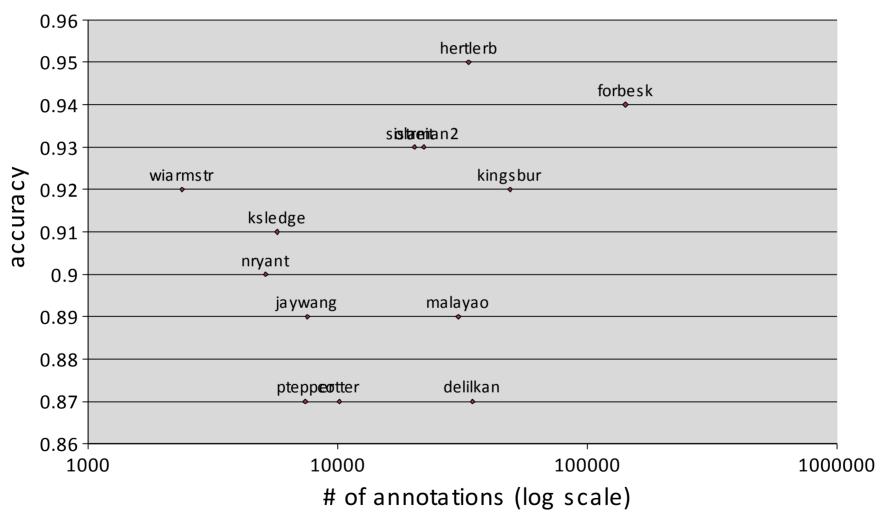
Paul Kingsbury

- ▶ Tagging tool highlights discrepancies Scott Cotton
- Third pass: Solomonization (adjudication)
  - Betsy Klipple, Olga Babko-Malaya



# Annotator accuracy – ITA 84%

## Annotator Accuracy-primary labels only





## **SRL** Questions

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## A Preliminary Classification of English Verbs, Beth Levin

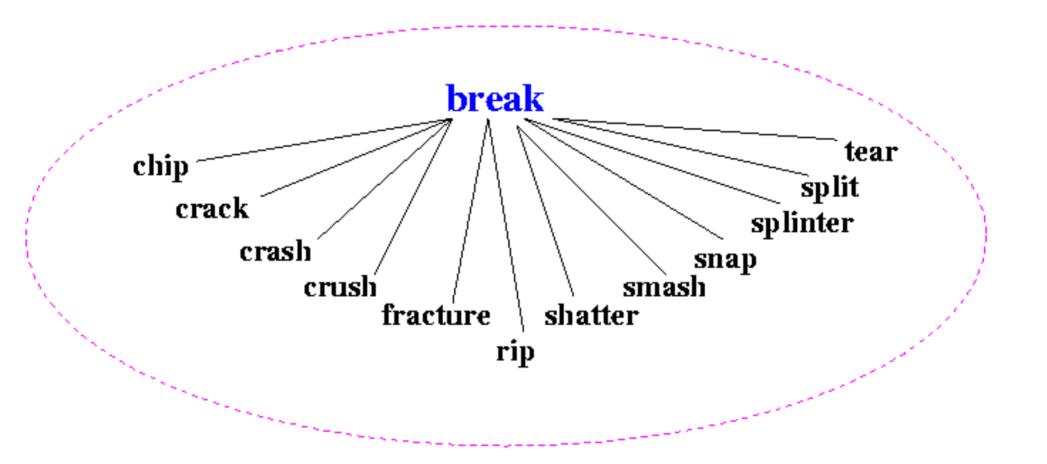
- Based on diathesis alternations
  - The range of syntactic variations for a class of verbs is a reflection of the underlying semantics
- ▶ 47 top level classes, 193 second and third level, 3100 verbs
- ▶ Based on pairs of syntactic frames.

  John broke the jar. / Jars break easily. / The jar broke. /\*John broke at the jar.

  John cut the bread. / Bread cuts easily. / \*The bread cut/John cut at the bread..
- Reflect underlying semantic components contact, directed motion, exertion of force, change of state
- > Synonyms, syntactic patterns (conative), relations



## Break Levin class - Change-of-state



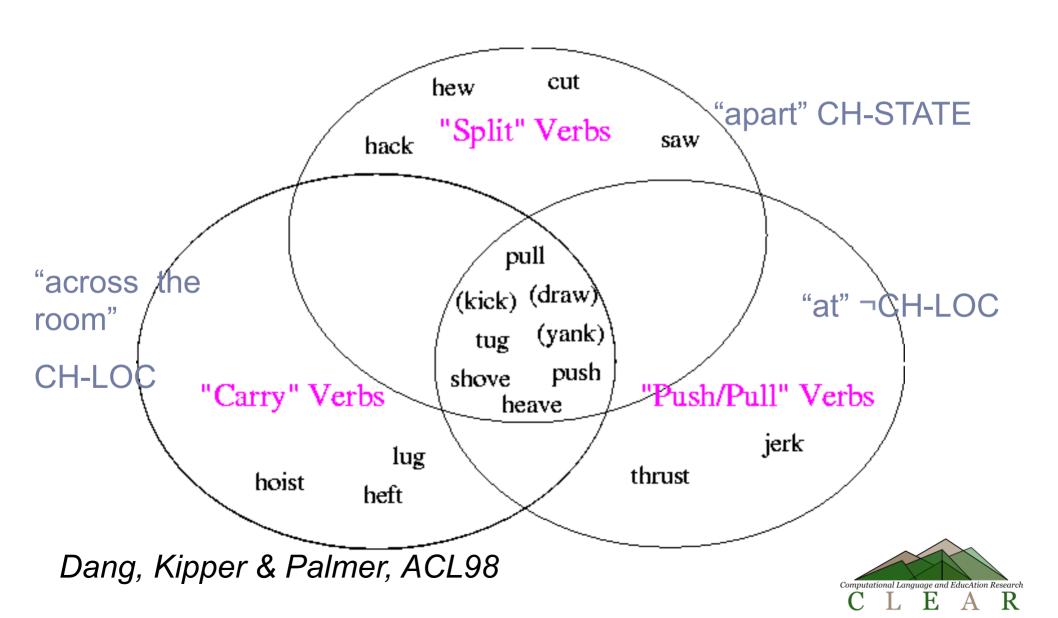


### Confusions in Levin classes?

- Not semantically homogenous
  - **{braid**, clip, file, powder, pluck, etc...}
- Multiple class listings
  - homonymy or polysemy?
- Alternation contradictions?
  - Carry verbs disallow the Conative, but include
  - {push,pull,shove,kick,draw,yank,tug}
  - also in Push/pull class, does take the Conative



### Intersective Levin Classes



### Intersective Levin Classes

- More syntactically and semantically coherent
  - sets of syntactic patterns
  - explicit semantic components
  - relations between senses



verbs.colorado.edu/verb-index/index.php



# VerbNet – Karin Kipper Schuler

#### Class entries:

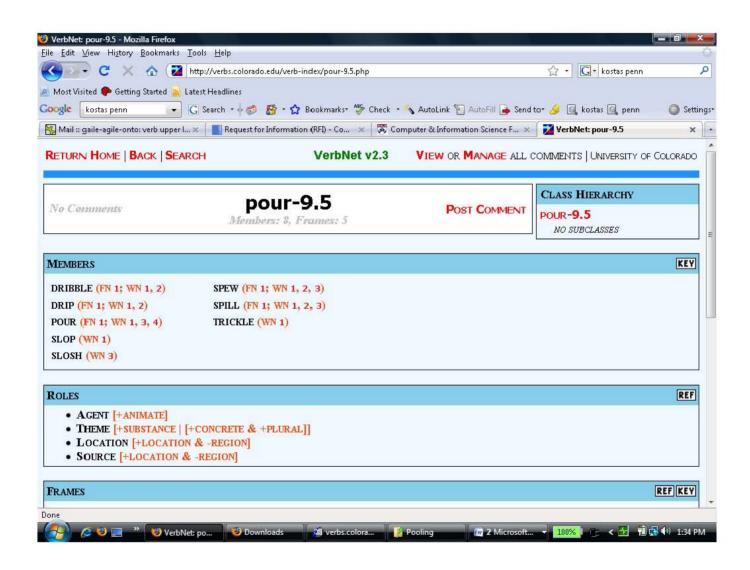
- Capture generalizations about verb behavior
- Organized hierarchically
- Members have common semantic elements, semantic roles (28) and syntactic frames

#### Verb entries:

- Refer to a set of classes (different senses)
- each class member linked to WN synset(s) and FrameNet frames
- Currently 6300 verbs
- Adding Constructions

Hwang, et.al, NAACL-HLT Construction Workshop, 2010 Bonial, et. al., ACL RELMS Workshop, 2011

### VerbNet example – Pour-9.5



# FrameNet, Chuck Filmore

- > The **lexical unit** (Cruse 1986), a pairing of a word with a sense (or a FrameNet frame.)
- In one of its senses, the verb *observe* evokes a frame called **Compliance**: this frame concerns people's responses to norms, rules or practices.
  - Our family observes the Jewish dietary laws.
  - You have to **observe** the rules or you'll be penalized.



#### The FrameNet Product – ADD STATS

#### The FrameNet database constitutes

- a set of frame descriptions
  - Frames, Frame Elements, Valence Possibilities
- a set of corpus examples annotated with respect to the frame elements of the frame evoked by each lexical unit
- lexical entries, including definitions and displays of the combinatory possibilities of each lexical unit, as automatically derived from the annotations
- a display of **frame-to-frame relations**, showing how some frames are elaborations of others, or are components of other frames.



### Frame Elements for Compliance

# The Frame Elements that figure in the Compliance frame are called

- Norm (the rule, practice or convention)
- Protagonist (the person[s] reacting to the Norm)
- Act (something done by the Protagonist that is evaluated in terms of the Norm)
- State\_of\_affairs (a situation evaluated in terms of the Norm)



- You do a whole frame for just observe?
- No. There are other Compliance words too.

V - adhere, comply, conform, follow, heed, obey, submit, ...;

#### **AND NOT ONLY VERBS**

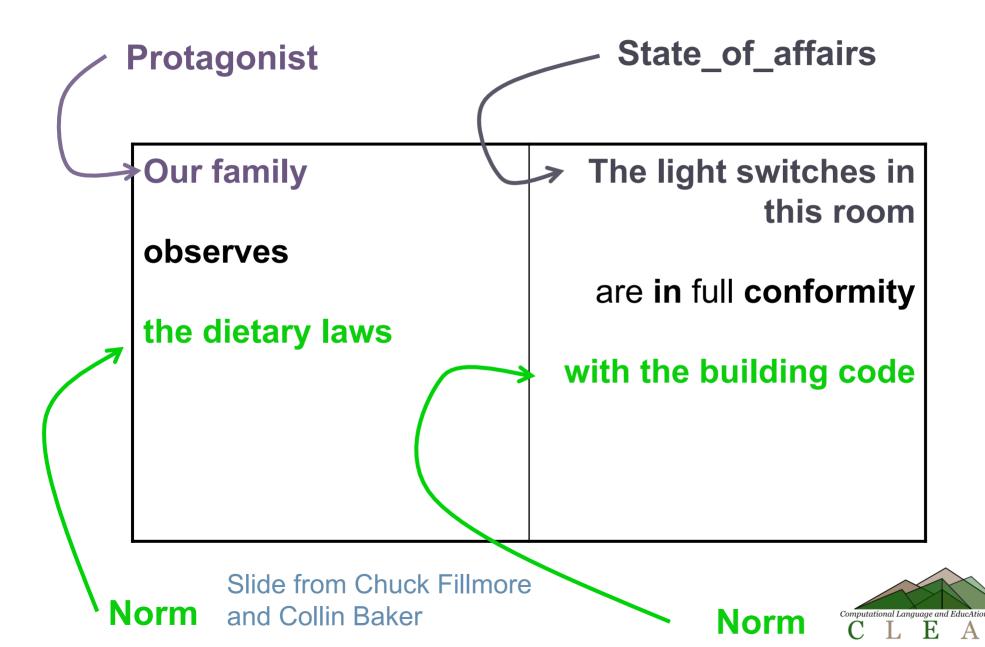
- N adherence, compliance, conformity, obedience, observance, ...;
- A compliant, obedient, ...;
- PP in compliance with, in conformity to, ...;

#### AND NOT ONLY WORDS FOR POSITIVE RESPONSES TO NORMS

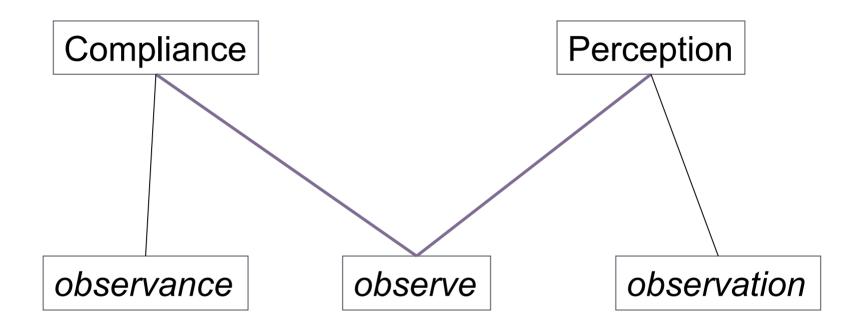
- V break, disobey, flout, transgress, violate ,...;
- N breach, disobedience, transgression, violation,...;
- PP in violation of, in breach of, ...



# Tagging Compliance sentences



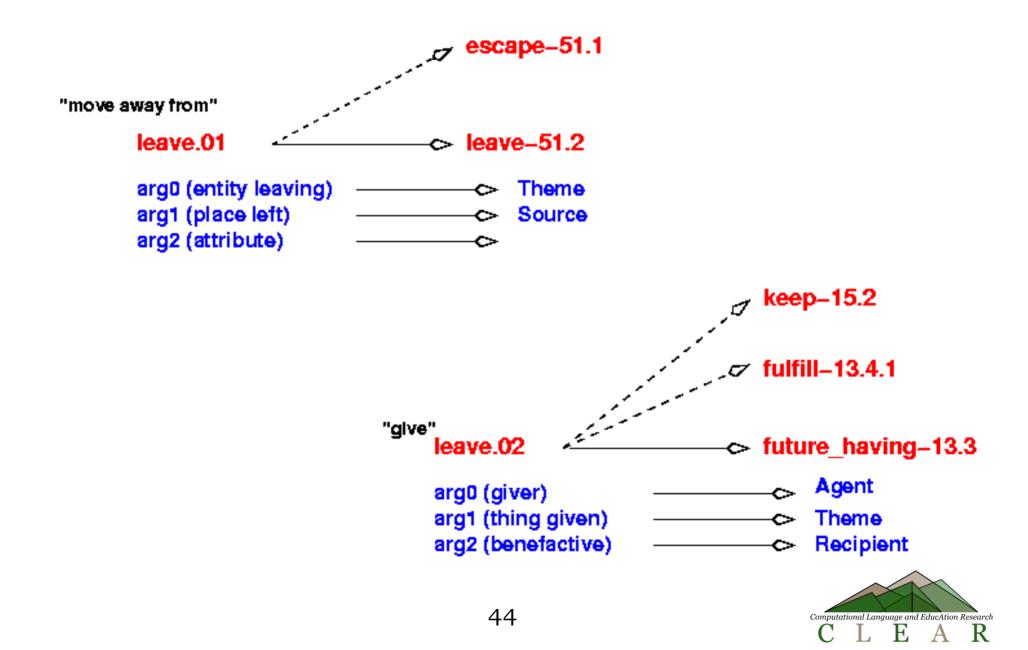
#### words, frames, lexical units



2 lexical units sharing same form: Compliance.observe, Perception.observe



# Mapping from PB to VerbNet - SemLink http://verbs.colorado.edu/semlink



# Mapping from PropBank to VerbNet (similar mapping for PB-FrameNet) - SemLink

Frameset id =	Sense =	VerbNet class =
leave.02	give	future-having 13.3
Arg0	Giver	Agent/Donor*
Arg1	Thing given	Theme
Arg2	Benefactive	Recipient

\*FrameNet Label

Baker, Fillmore, & Lowe, COLING/ACL-98 Fillmore & Baker, WordNetWKSHP, 2001



#### PropBank/FrameNet - SemLink

Sell Buy Arg0: seller Arg0: buyer Arg1: goods Arg1: goods Arg2: seller Arg2: buyer Arg3: rate Arg3: rate Arg4: payment Arg4: payment

More generic, more neutral – maps readily to VN,TR

Rambow, et al, PMLB03

E

# Can SemLink improve Generalization?

- After PropBank, SRL improved from 77% to 88% Automatic parses, 81% F, Brown corpus, 68%
- Overloaded Arg2-Arg5
  - PB: verb-by-verb
  - VerbNet: same thematic roles across verbs

#### Example

- ▶ Rudolph Agnew,..., was named [ARG2 {Predicate} a nonexecutive director of this British industrial conglomerate.]
- ....the latest results appear in today's New England Journal of Medicine, a forum likely to bring new attention [ARG2 {Destination} to the problem.]
- Use VerbNet as a bridge to merge PB and FN and expand the Size and Variety of the Training



# VerbNet - Arg2 groupings; (Total count 11068)

Group1 (43.93%)	Group2 (14.74%)	Group3 (32.13%)	Group4 (6.81%)	Group5 (2.39%)
Recipient; Destination; Location; Source; Material; Beneficiary	Extent; Asset	Predicate; Attribute; Theme; Theme2; Theme1; Topic	Patient2; Product	Instrument; Actor2; Cause; Experiencer



#### Process

- Retrain the SRL tagger
  - Original: Arg[0-5,A,M]
  - ARG2 Grouping: Arg[0,2-5,A,M] Arg1-Group[1-6]
- Evaluation
  - ▶ WSJ [+6%]
  - ▶ Brown [+10%]
- More Coarse-grained or Fine-grained?
  - more specific: data more coherent, but more sparse
  - more general: consistency across verbs even for new domains?



## PropBank/VerbNet/FrameNet - SemLink

- Complementary resources
- Redundancy is harmless, may even be useful
- PropBank provides the best training data
- VerbNet provides the clearest links between syntax and semantics
- FrameNet provides the richest semantics
- Together they give us the most comprehensive coverage
- SemLink <a href="http://verbs.colorado.edu/semlink/">http://verbs.colorado.edu/semlink/</a>
  - WSJ, sense tags and SRL, mappings to VN and FN



# WSJ instance example from SemLink

```
Pierre Vinken, 61 years old, will join the board as a nonexecutive director Nov. 29.
```

```
nw/wsj/00/wsj_0001.parse

0 8 gold join-v 22.1-2-1 Cause_to_amalgamate join.01
0:2-ARG0=Agent;Agent
7:0-ARGM-MOD
8:0-rel
9:1-ARG1=Patient;Part_1
11:1-ARGM-PRD 15:1-ARGM-TMP
```



### Annotated Data – Current PropBank Status

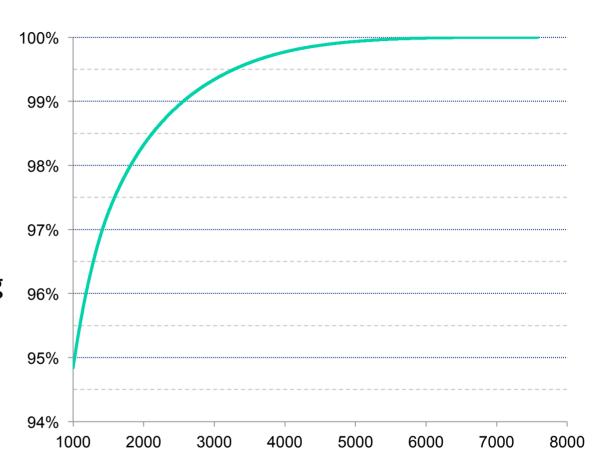
Pradhan, et.al., IJSC 2007, Albright, et. al., JAMIA, 2013, Palmer, et. al., ICON-09

- DARPA-GALE, OntoNotes 5.0
  - BBN, Brandeis, Colorado, Penn
  - Multilayer structure: NE, TB, PB, WS, Coref
  - Three languages: **English**, **Arabic**, Chinese
  - Several Genres (@ ≥ 200K): NW, BN, BC, WT
    - Close to 2M words @ language (less PB for Arabic)
  - Parallel data, E/C, E/A
  - PropBank frame coverage for rare verbs
  - Recent PropBank extensions
- ▶ Clinical Notes 400K available, goal is 700K
- ▶ Hindi/Urdu PropBank, 400K Hindi, 200K Urdu
- BOLT discussion forums, SMS, email, Egyptian



# PropBank Verb Frames Coverage

- ▶ The set of verbs is open
- But the distribution is highly skewed
- For English, the 1000 most frequent lemmas cover
   95% of the verbs in running text.
  - Graphs show counts over
     English Web data containing
     I 50 M verbs.





# Verb Frames Coverage By Language – Current Count of Senses (lexical units)

Language	Final Count	Estimated Coverage in Running Text
English	10,615*	99%
Chinese	24, 642	98%
Arabic	7,015	99%

Only 111 English adjectives



#### Included in OntoNotes 5.1: Extensions to PropBank

- Original annotation coverage:
  - PropBank: verbs; past participle adjectival modifiers
  - NomBank: relational and eventive nouns.
- Substantial gap now bridging
  - Uniform treatment of light verbs,
  - Additional predicative adjectives,
  - Eventive nouns



## Gaps in proposition coverage

- Event Coreference chains include nominalizations with and without light verbs
  - \*China has threatened to slap sanctions on American companies that sell arms to its rival Taiwan as part of a range of punitive actions Beijing is taking to protest the deal... 'China will make further judgments as appropriate,' Xinhua reported."
  - Light verb/nominalization examples: slap sanctions, taking actions, make judgments
- PropBank structures for eventive nouns
  - sanction(China, US companies),
  - act(China),
  - judge(China, US companies)



# English Noun and LVC annotation

- Example Noun: Decision
  - ► Roleset: Arg0: decider, Arg1: decision...
  - "...[your<sub>ARG0</sub>] [decision<sub>REL</sub>] [to say look I don't want to go through this anymore<sub>ARG1</sub>]"
- Example within an LVC: Make a decision

```
"...[the President<sub>ARG0</sub>] [made<sub>REL-LVB</sub>]
the [fundamentally correct<sub>ARGM-ADJ</sub>]
[decision<sub>REL</sub>] [to get on offense<sub>ARG1</sub>]"
```



# 2-pass annotation, post-processing

- China will make further judgments as appropriate.
- Verb REL: [make],
  - Arg0: China,
  - ArgPRX: further judgments as appropriate.
- ▶ Noun RELPRX: [judgment]
  - Arg0: China
  - ▶ ArgM-PRD: as appropriate
- ▶ Merged REL: RELPRX: [make] [judgment]
  - Arg0: China
  - ArgM-PRD: as appropriate



# Abstract Meaning Representations – AMR, Maximal Use of PropBank Frame Files,

Knight, et. al., LAW-2013

He was not aware of research on smokers of the Kent cigarettes.

To get to canonical concept, we stem to English verbs,

where PropBank arguments are best described.

General direction of stemming: adverb → adjective → noun → verb :op1 "Kent"))))))



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- Collaborators: Suzanne Stevenson, Annie Zaenen, Orin Hargraves



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