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# DISAGREEMENTS IN ANAPHORIC ANNOTATION

<http://www.dali-ambiguity.org>



# Disagreements in anaphora (and other aspects of language interpretation)



# Anaphora (AKA coreference)

So she [Alice] was considering in her own mind (as well as she could, for the hot day made her feel very sleepy and stupid), whether the pleasure of making a daisy-chain would be worth the trouble of getting up and picking the daisies, when suddenly **a White Rabbit with pink eyes** ran close by her.

There was nothing so VERY remarkable in that; nor did Alice think it so VERY much out of the way to hear **the Rabbit** say to **itself**, 'Oh dear! Oh dear! **I** shall be late!' (when she thought it over afterwards, it occurred to her that she ought to have wondered at this, but at the time it all seemed quite natural); but when **the Rabbit** actually **TOOK A WATCH OUT OF ITS WAISTCOAT-POCKET**, and looked at it, and then hurried on, Alice started to her feet, for it flashed across her mind that she had never before seen a rabbit with either a waistcoat-pocket, or a watch to take out of it, and burning with curiosity, she ran across the field after **it**, and fortunately was just in time to see **it** pop down a large rabbit-hole under the hedge.

# Building NLP models from annotated corpora

- Use **TRADITIONAL CORPUS ANNOTATION / CROWDSOURCING** to create a **GOLD STANDARD** that can be used to train supervised models for various tasks
- This is done by collecting multiple annotations (typically 2-5) and going through **RECONCILIATION** whenever there are multiple interpretations
- **DISAGREEMENT** between coders (measured using coefficients of agreement such as  $\kappa$  or  $\alpha$ ) viewed as a serious problem, to be addressed by revising the coding scheme or training coders to death
- Yet there are very many types of NLP annotation where **DISAGREEMENT IS RIFE** (wordsense, sentiment, discourse)

# Ambiguity in anaphora

15.12 M: we're gonna take the engine E3

15.13 : and shove it over to Corning

15.14 : hook [it] up to [the tanker car]

15.15 : \_and\_

15.16 : send **it** back to Elmira

(from the TRAINS-91 dialogues collected at the University of Rochester)

# Ambiguity: What antecedent? (Poesio & Vieira, 1998)

About 160 workers at *a factory* that made paper for the Kent filters were exposed to asbestos in the 1950s.

*Areas of the factory* were particularly dusty where the crocidolite was used.

Workers dumped large burlap sacks of the imported material into a huge bin, poured in cotton and acetate fibers and mechanically mixed the dry fibers in a process used to make filters.

Workers described "clouds of blue dust" that hung over *parts of the factory*,

even though exhaust fans ventilated the area.

# Ambiguity: DISCOURSE NEW or DISCOURSE OLD?

(Poesio, 2004)

What is in **your cream**

**Dermovate Cream** is one of a group of medicines called topical steroids.

"Topical" means they are put on the skin. Topical steroids reduce the redness and itchiness of certain skin problems.

# Ambiguity: EXPLETIVES

'I beg your pardon!' said the Mouse, frowning, but very politely: 'Did you speak?'

'Not I!' said the Lory hastily.

'I thought you did,' said the Mouse. '--I proceed. "Edwin and Morcar, the earls of Mercia and Northumbria, declared for him: and even Stigand, the patriotic archbishop of Canterbury, found **it** advisable--"'

'Found **WHAT?**' said the Duck.

'Found **IT**,' the Mouse replied rather crossly: 'of course you know what "it" means.'



# More evidence of disagreement raising from ambiguity

- For anaphora
  - **Versley 2008**: Analysis of disagreements among annotators in the Tüba/DZ corpus
    - Formulation of the **DOT-OBJECT hypothesis**
  - **Recasens et al 2011**: Analysis of disagreements among annotators in (a subset of) the ANCORA and the ONTONOTES corpus
    - The **NEAR-IDENTITY hypothesis**
- **Wordsense: Passonneau et al, 2012**
  - Analysis of disagreements among annotators in the wordsense annotation of the MASC corpus
  - Up to 60% disagreement with verbs like *help*
- **POS tagging: Plank et al, 2014**

# Facets (Versley, 2008)

As a lawyer in Boston, [1 John Travolta] sues two businesses that he holds responsible for eight children having died of leukemia.

At first, [2 the calculating career lawyer] only scents the high amount of compensation (. . .).

A court drama, environmental thriller and great actors' cinema, in which [3 Travolta] and his antagonist Robert Duvall reach top form.

# Near-identity cases (Recasens et al, 2011)

“**[Your father]**<sub>a</sub> was the greatest, but **[he]** was also one of us,” commented an anonymous old lady while she was shaking Alessandro’s hand—**[Gassman]**<sub>a</sub>’s best known son.

“I will miss **[the actor]**<sub>a1</sub>, but I will be lacking **[my father]**<sub>a</sub> especially,” he said.

“On homecoming night **[Postville]** feels like Hometown, USA . . . For those who prefer **[the old Postville]**, Mayor John Hyman has a simple answer.



# Collecting the data

# Explicit and implicit disagreements

19.10: we need to get the bananas to Corning by 3

19.11: uh

19.12: *maybe* it's gonna be faster if we

19.13: send E1

19.14: E1's boxcar picks up at Dansville

19.15: instead of going back to Avon

19.16: have it go on to Corning

19.17: uh pick up the tanker get the oranges send them to Elmira

19.18: cause that's gonna be the longest thing

Key: Full agreement One outlier Implicit Explicit

# Collecting disagreements online

The screenshot shows the homepage of the Phrase Detectives website. The browser address bar displays 'anawiki.essex.ac.uk/phrasedetectives/index.php'. The page features a central logo with a detective character and the text 'PHRASE DETECTIVES'. On the left, there is a 'LOGIN' section with fields for 'USERNAME' and 'PASSWORD', a 'Login' button, and a 'Register here' link. Below the login section are social media links for Facebook (37 people like this) and buttons for 'Instructions' and 'FAQ'. A 'SHARE THIS' section with social media icons is also present. The main content area includes a 'Welcome to Phrase Detectives' message, a '321 docs completed' announcement, and 'Quick instructions'. On the right, there are sections for 'TOPSCORES' (This Week and This Month), a 'LEADERBOARD' table, and a 'Detective's Bulletin' with dates '30 Nov 10' and '01 Nov 10'. The 'LEADERBOARD' table lists players and their scores for Week, Month, and Alltime.

| WEEK          | MONTH | ALLTIME |
|---------------|-------|---------|
| papillon      | 1417  | 1417    |
| norma         | 904   | 904     |
| JMS           | 842   | 842     |
| JRS           | 501   | 501     |
| kazzade       | 310   | 310     |
| darraghsdream | 296   | 296     |
| JCS           | 203   | 203     |
| julia3164     | 157   | 157     |
| domma17       | 103   | 103     |
| gully         | 87    | 87      |
| johnnickel    | 72    | 72      |
| poppyseed     | 66    | 66      |
| prufrax       | 61    | 61      |
| KULIKOV       | 55    | 55      |
| roorrea       | 45    | 45      |
| Grammar       | 42    | 42      |
| axnicho       | 39    | 39      |
| VB            | 38    | 38      |
| crazyhorse    | 27    | 27      |
| livio.robald  | 27    | 27      |

www.phrasedetectives.org

# Gamifying annotation

- **Find The Culprit** (Annotation)  
User must identify the closest antecedent of a markable if it is anaphoric
- **Detectives Conference** (Validation)  
User must agree/disagree with a coreference relation entered by another user



# Find the Culprit

(aka Annotation Mode)

## The Count of Monte Cristo

Having arrived before the Pont du Gard, the horse stopped, but whether for his own pleasure or that of his rider would have been difficult to say. However that might have been, the priest, dismounting, led his steed by the bridle in search of some place to which he could secure him. Availing himself of a handle that projected from a half-fallen door, he tied the animal safely and having drawn a red cotton handkerchief, from **his** pocket, wiped away the perspiration that streamed from **his** brow, then, advancing to the door, struck thrice with the end of his iron-shod stick.



Not mentioned before!



Skip this one

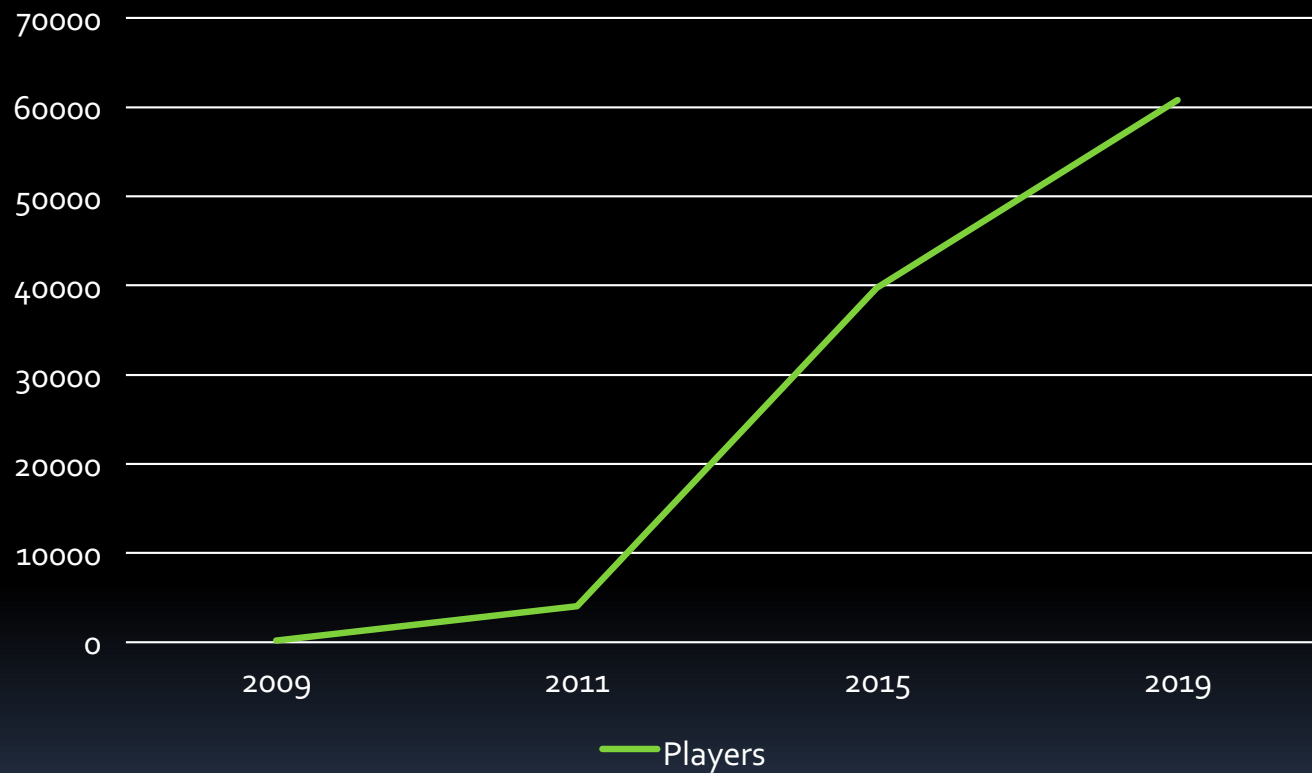
Found it!





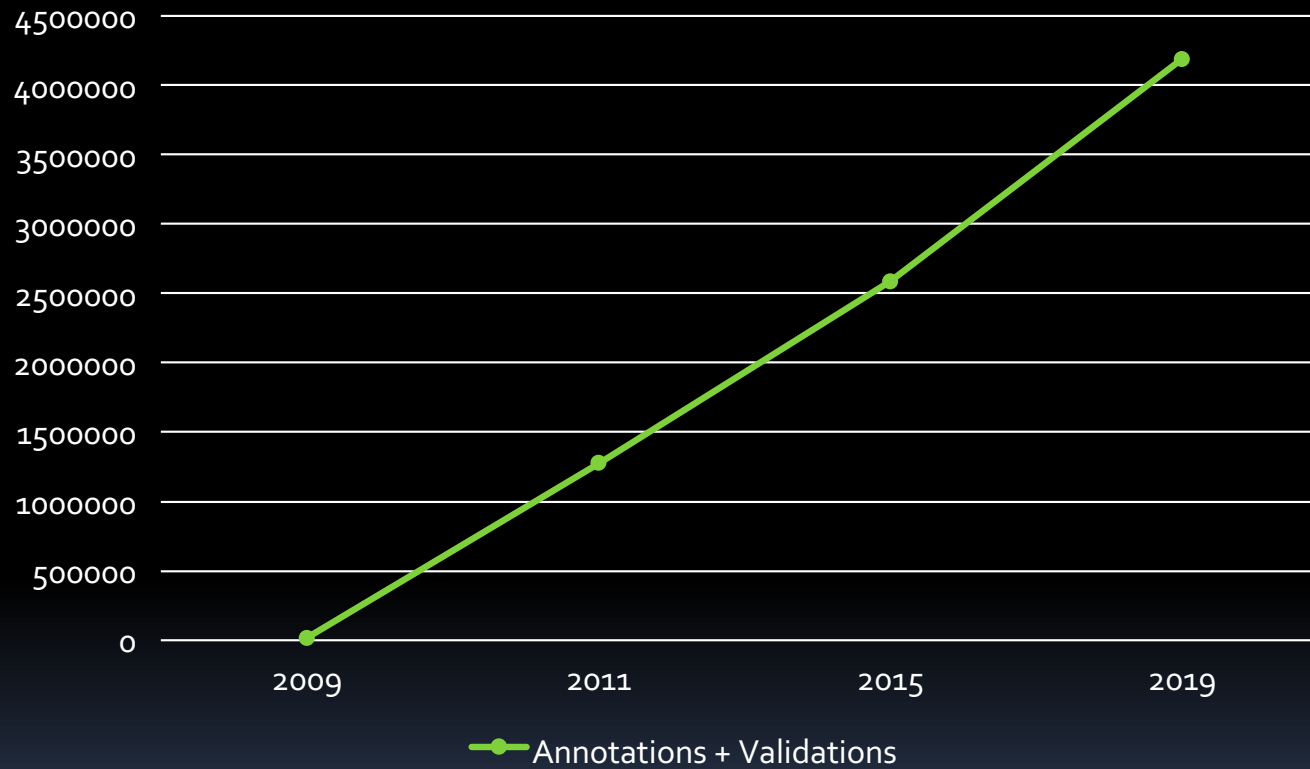
# Number of players

Players

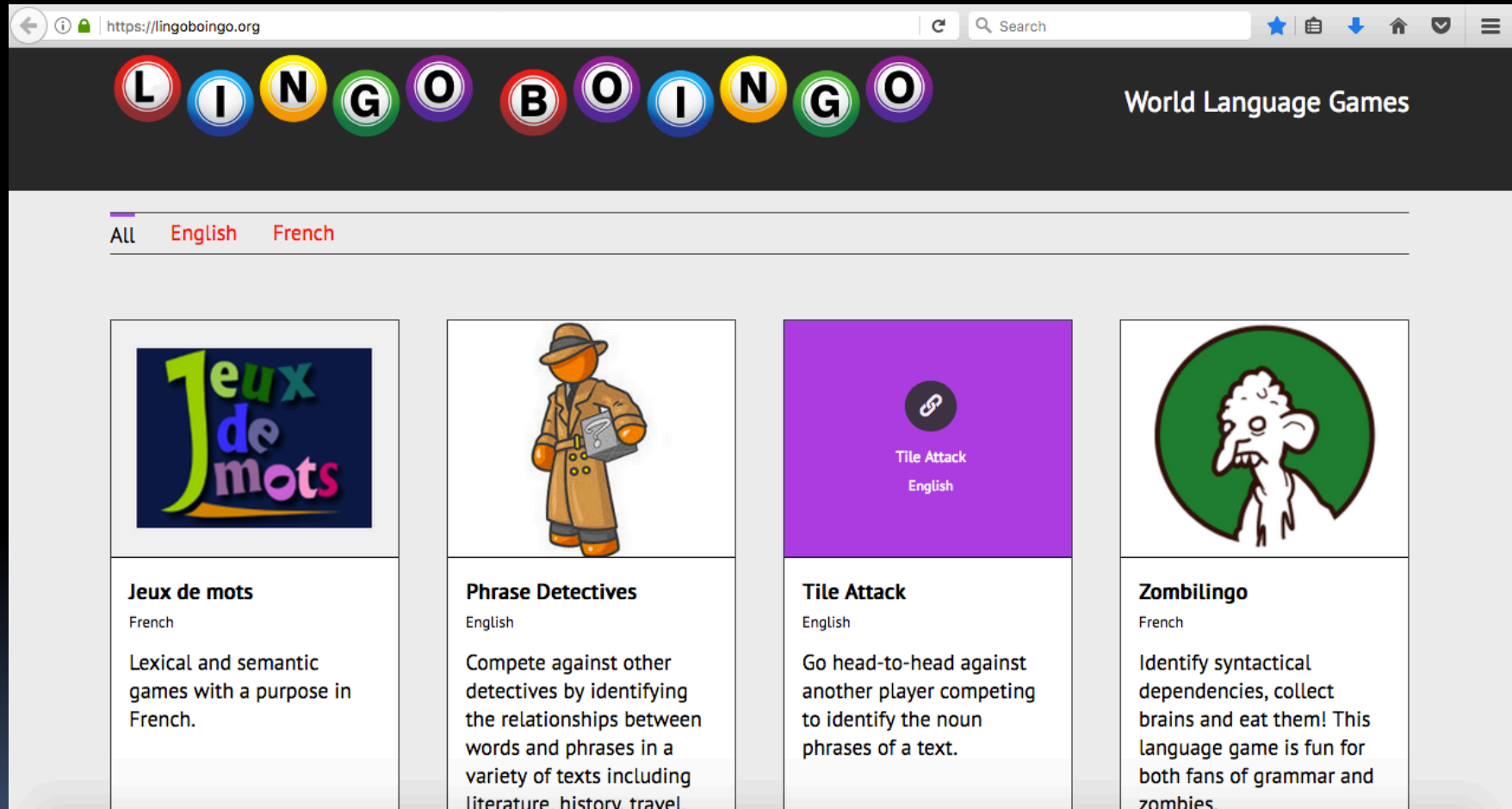


# Number of judgments

## Annotations + Validations



# LingoBoingo



The screenshot shows the LingoBoingo website interface. At the top, the URL is <https://lingoboingo.org>. The navigation bar features the LingoBoingo logo, which consists of the letters L, I, N, G, O, B, O, I, N, G, O in colorful circles, and the text "World Language Games". Below the navigation bar, there are tabs for "All", "English", and "French". The main content area displays four game cards:

- Jeux de mots** (French): Lexical and semantic games with a purpose in French.
- Phrase Detectives** (English): Compete against other detectives by identifying the relationships between words and phrases in a variety of texts including literature, history, travel.
- Tile Attack** (English): Go head-to-head against another player competing to identify the noun phrases of a text.
- Zombilingo** (French): Identify syntactical dependencies, collect brains and eat them! This language game is fun for both fans of grammar and zombies.



# The Phrase Detectives Corpus

# The Phrase Detectives Corpus

- Data:
  - 1.2M words total, of which around 330K totally annotated
  - About 50% Wikipedia pages, 50% fiction
- Markable scheme:
  - Around 25 judgments per markable on average
  - Judgments:
    - NR/DN/DO
    - For DO, antecedent
- Phrase Detectives 1 (with gold annotation) released via LDC in 2016
- Phrase Detectives 2 just released

# PD corpus: annotation scheme

| Type                     | Example  | ONTONOTES             | PRECO | ARRAU    | Present corpus |
|--------------------------|--|-----------------------|-------|----------|----------------|
| predicative NPs          | [John] is <u>a teacher</u><br>[John, <u>a teacher</u> ]                      | Pred                  | Coref | Pred     | Pred           |
| singletons               |  | No                    | Yes   | Yes      | Yes            |
| expletives               | <u>It's</u> five o'clock   | No                    | No    | Yes      | Yes            |
| split antecedent plurals | [John] met [Mary]<br>and <u>they</u> ...                                     | No                    | No    | Yes      | Yes            |
| generic mentions         | [Parents] are usually busy.<br><u>Parents</u> should get involved            | Only with<br>pronouns | Yes   | Yes      | Yes            |
| event anaphora           | Sales [grew] 10%.<br><u>This growth</u> is exciting                          | Yes                   | No    | Yes      | No             |
| ambiguity                | <u>Hook up</u> [the engine]<br>to [the boxcar]<br>and send <u>it</u> to Avon | No                    | No    | Explicit | Implicit       |


# PD2: Size

|                     |           | Docs | Tokens | Markables      |
|---------------------|-----------|------|--------|----------------|
| $C_{\text{gold}}$   | Gutenberg | 5    | 7536   | 1947 (1392)    |
|                     | Wikipedia | 35   | 15287  | 3957 (1355)    |
|                     | GNOME     | 5    | 989    | 274 (96)       |
|                     | Subtotal  | 45   | 23812  | 6178 (2843)    |
| $C_{\text{silver}}$ | Gutenberg | 145  | 158739 | 41989 (26364)  |
|                     | Wikipedia | 350  | 218308 | 57678 (19444)  |
|                     | Other     | 2    | 7294   | 2126 (1339)    |
|                     | Subtotal  | 497  | 384341 | 101793 (47147) |
| All                 | Total     | 542  | 408153 | 107971 (49990) |

## PD2: Number of judgments

- 2,235,664 judgments from 425 1958 players, of which
  - 1,358,559 annotations and
  - 426 867,844 validations.
- On average, 20.6 judgments per markable
- Compare:
  - About 600K judgments for Ontonotes (~ 3 per markable)
  - About 10M judgments for PRECO (also ~ 3 per markable)





# Assigning a probability to interpretations

# Bayesian models of annotation

- The problem of reaching a conclusion on the basis of judgments by separate experts that may often be in disagreement is a longstanding one in epidemiology
- A number of techniques developed to aggregate these judgments
- A particularly popular approach is to use BAYESIAN MODELS OF ANNOTATION
  - Dawid and Skene 1979 (also used by Passonneau & Carpenter 2014)
  - Carpenter (2008)
  - Raykar et al 2010
  - Hovy et al, 2013

# Bayesian Models of Annotation

- A Bayesian model of annotation specifies the probability of a particular label on the basis of PARAMETERS specifying the behavior of the annotators, the prevalence of the labels, etc
- In Bayesian models, these parameters are specified in terms of PROBABILITY DISTRIBUTIONS

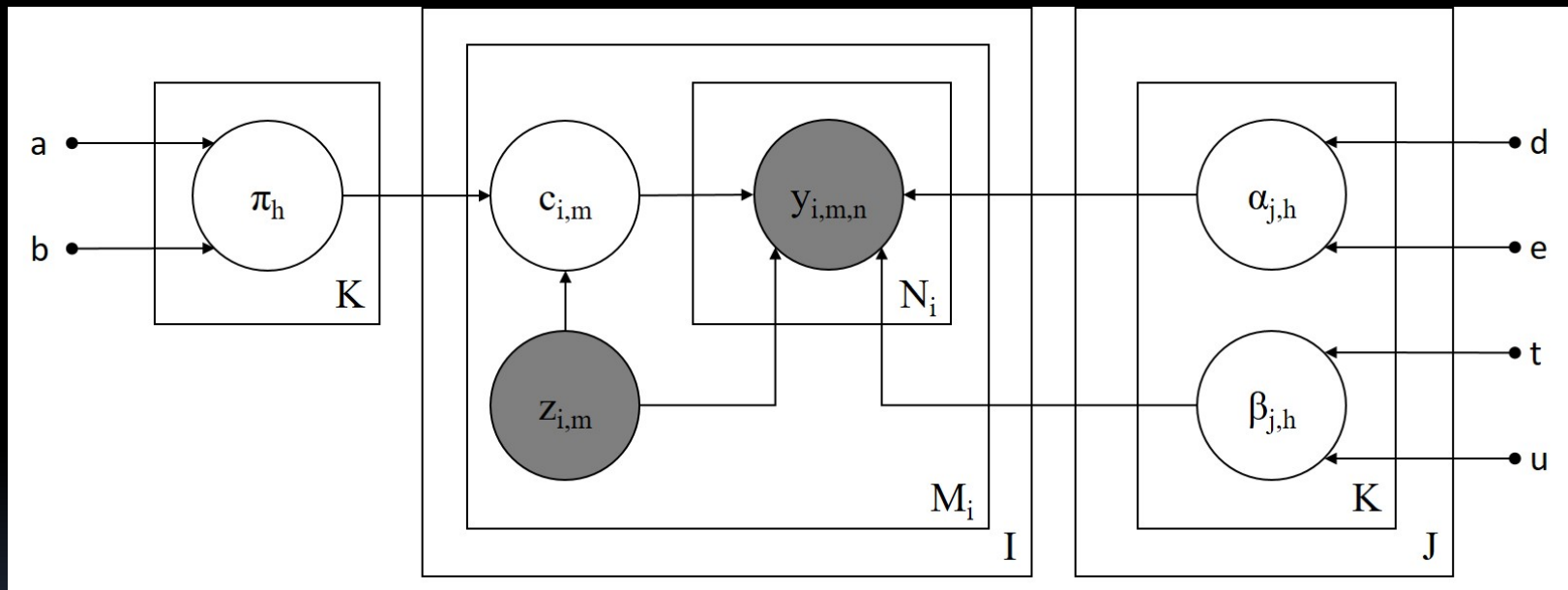
# Comparing Bayesian Annotation Models

- Implemented in Stan (<http://mc-stan.org/>) some of the BAMs best-known in computational linguistics (Dawid & Skene, MACE, Carpenter's four models) & compared them on PD Gold data
- Evaluation metrics:
  - Accuracy
  - Annotator accuracy
  - Item difficulty
- The PD data are unique in a number of ways
  - Lots of judgments
  - Different types of noise from crowdsourcing
  - Gold info about spammers
- Paun et al, 2018. Comparing Bayesian Models of Annotation. *Transactions of the ACL*

# Mention Pair Annotation (MPA)

- No existing BAM however can work with ANAPHORIC information, in which the 'labels' are not a discrete set, but coreference chains
- Our first model, called MPA, is a generative model of the process of linking mention pairs
- On the Phrase Detectives Data, it achieves an accuracy of 91.43% (as opposed to 84% for Majority Voting)
- Paun et al, 2018. A probabilistic annotation model for crowdsourcing coreference. *Proc. Of EMNLP*.

# MPA





# Anaphora resolution with PD 2

# Methods

- The most likely (SILVER) labels extracted via MPA can be used to train CONLL-style coreference systems (if singletons and NR markables are ignored) or systems carrying out the full anaphora task
- For the second task, the Extended Coreference Score developed by Moosavi for the CRAC 2018 Shared Task can be used (Poesio et al, 2018)
- Two systems were trained and evaluated:
  - The state-of-the-art Lee et al 2018 system
  - Our own cluster ranking model (Yu et al, submitted)



# Results on the CONLL task and with singletons

| Singletons | Method             | MUC         |             |             | BCUB        |             |             | CEAFE       |             |             | Avg.<br>F1  |
|------------|--------------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|-------------|
|            |                    | P           | R           | F1          | P           | R           | F1          | P           | R           | F1          |             |
| Included   | Our Model          | 79.3        | 72.5        | 75.7        | 72.1        | 69.3        | 70.7        | 70.5        | 73.2        | 71.8        | 72.7        |
| Excluded   | Our Model          | 79.3        | <b>72.5</b> | <b>75.7</b> | 58.3        | 52.4        | <b>55.2</b> | <b>58.3</b> | <b>49.5</b> | <b>53.5</b> | <b>61.5</b> |
|            | Our Model*         | 77.8        | 71.8        | 74.6        | 55.4        | <b>53.7</b> | 54.6        | 56.2        | 49.0        | 52.4        | 60.5        |
|            | Lee et al. (2018)* | <b>80.8</b> | 66.1        | 72.7        | <b>63.3</b> | 45.1        | 52.7        | 56.7        | 44.7        | 50.0        | 58.5        |

# Results with NR markables

|                 | P    | R    | F1   |
|-----------------|------|------|------|
| Non-referring   | 55.2 | 54.0 | 54.6 |
| Expletives      | 62.3 | 86.0 | 72.3 |
| Predicative NPs | 49.7 | 47.7 | 48.7 |



# Ambiguity in the PD corpus



# Raw disagreements

|                 | 1     | 2     | 3     | 4    | 5    | 6    | 7    |
|-----------------|-------|-------|-------|------|------|------|------|
| PD <sub>G</sub> | 38.8% | 30.6% | 18.5% | 7.3% | 2.5% | 1%   | 0.6% |
| PD <sub>C</sub> | 36%   | 30%   | 19%   | 8.8% | 3.8% | 1.8% | 0.8% |

Total number of markables in  
PD<sub>C</sub>: 108,013  
Total number of markables  
with no disagreements: 38579

61.2% of markables in PD<sub>G</sub>,  
and 64% in PD<sub>C</sub>, has more  
than 1 interpretation

# An example of disagreement

RB ne75965

The day came that had been fixed for the marriage. The bridegroom arrived and also a large company of guests, for the miller had taken care to invite all his friends and relations. As [they] sat at the feast, each guest in turn was asked to tell a tale; the bride sat still and did not say a word.

DO ne75948 {for the miller had taken care to invite [all his friends and relations]} (11,3,1,13),  
DO ne75945 {a large company of [guests]} (2,2,2,2),  
DN (10,3,1,12),  
DO ne75936 {the girl}, ne75942 {[the bridegroom]}, ne75945, ne75948 (1,1,3,-1),  
DO ne75942, ne75946 {[the miller]}, ne75948 (2,2,2,2),  
DO ne759370001 {[her]}, ne75942 {[the bridegroom]}, ne75945 {[the large company of guests]},  
ne759490001 {his (the miller)} (1,0,4,-3,e2,e18),  
DO ne75942 ne75948 ne759370001 ne75946 (1,3,1,3),  
DO ne75942 ne75948 ne759370001 (1,2,2,1),  
DO ne75942 ne759370001 ne75945 (1,0,4,-3),  
DO ne75948 ne75946 (2,1,3,0),  
DO ne75942 ne75948 ne759370001 ne75945 ne75946 (2,1,3,0),  
DO ne75936 ne75937 {her father aka the miller} ne75942 ne75948 (1,0,4,-3)

+ 2 not\_selectable, 3 skips

**81 A+V, 5 comments skips, Total: 86 judgments**

# Not all disagreements are due to ambiguity

- Pradhan et al (2012): The analysis of the around 20,000 mentions on which there was disagreement in the ONTONOTES coreference annotation suggests that reasons include
  - Ambiguity proper ('unclear interpretation' or 'disagreements on reference') (30% of disagreements, 7% of all mentions)
  - Annotator error (25% of the cases of disagreement)
  - Limitations of the coding scheme (36.5% of all disagreements)
  - Interface limitations (7.5% of all disagreements)

# Interface limitations in PD

- Interface limitations

- **DDEIX**: ne75896

The old woman then mixed a sleeping draught with their wine, and before long they were all lying on the floor of the cellar, fast asleep and snoring. As soon as the girl was assured of **this**,

DN (15, 2, 2, 15),

DO ne75894 {[fast asleep] and snoring} ?? (2, 1, 3, 0),

DO ne75895 {[the girl]} ?? (1, 0, 4, -3),

DO ne75908 ?? {they were all lying on [the floor] in the cellar} (3, 0, 4, -1),

DO ne75889 {they} ne75890 ?? (1, 2, 2, 1)

3 skips

Note in this case we also have the type of ambiguity with DDEIX discussed in Poesio et al 2003, 2006

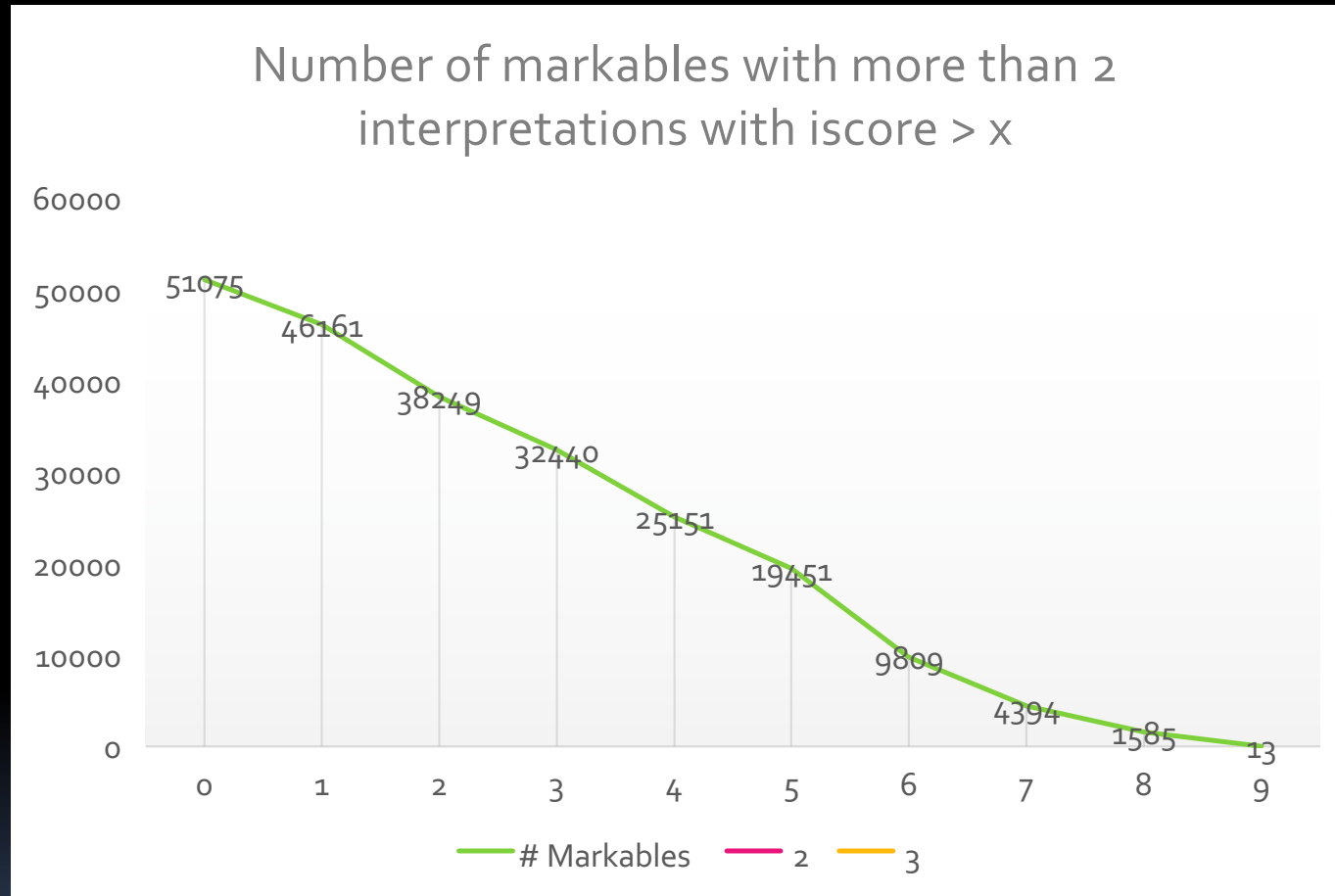
# The validation filter

An interpretation can be 'scored' by counting the number of players who produce / agree with it, and subtracting the number of players who disagree with it

$$\text{ISCORE}_i = \text{ANN}_i + \text{AGR}_i - \text{DISAGR}_i$$



# Filtering using validation



# A second filter: MPA


|                                   | .5     | .3     | .1     |
|-----------------------------------|--------|--------|--------|
| #markables                        | 104194 | 106042 | 106857 |
| #mentions with more than one int. | 2587   | 5263   | 10283  |
| Highest number of int.            | 6      | 5      | 6      |

# A second filter: MPA

|        | 0    | 1     | 2    | 0 or 2 |
|--------|------|-------|------|--------|
| $PD_G$ | 2.3% | 93.4% | 4.3% | 6.6%   |
| $PD_C$ | 3.5% | 94%   | 2.4% | 5.9%   |



# MPA and ambiguity

- The questions:
    - What types of ambiguity are there?
    - Which cases of ambiguity are correctly predicted by MPA?
    - Which cases of ambiguity are not caught by MPA, if any?
- 

# An analysis of disagreements in the PD2 corpus

- Chosen a few docs from PD<sub>G</sub>
  - So far completely analyzed two Gutenberg docs:
    - Little Red Riding Cap (Grimm)
    - The Robber Bridegroom (Grimm)
- Labelled the disagreements as indicating
  - Ambiguity (definitely, possibly)
  - Cheating/Misunderstanding
  - Spurious ambiguity
  - Interface Problems (with attempt at classification)

# Plurals 2: bare plurals

LRC ne7546

'Little Red-Cap raised her eyes, and when she saw the sunbeams dancing here and there through the trees, and pretty flowers growing everywhere, she thought: 'Suppose I take grandmother a fresh nosegay; that would please her too. It is so early in the day that I shall still get there in good time'; and so she ran from the path into the wood to look for **[flowers]**

DN (5,1,3,3,e18),

DO ne7532 {and [pretty flowers] growing everywhere ... } (8,2,2,8,e2),

DO ne7536 {Suppose I take grandmother [a fresh nosegay]} (6,2,2,6),

PR ne7536 ?? (1,1,3,-1),

PR ?? (1,1,3,-1),

DO ne7537 {[that] would please her too} ?? (1,3,1,3)

# Plurals 3: `we' and `you'

RB ne75698 (MPA: none)

And you, my love,' said the bridegroom, turning to her,  
'is there no tale you know? Tell us something.' 'I will tell  
**[you]** a dream, then,' said the bride.

DO ne75965 (9,3,1,11,e2,e18),

DO ne75960 (4,2,2,4),

DN (2,0,4,-2)

# Additional sources of ambiguity: paths

Her betrothed only replied, 'You must come and see me next Sunday; I have already invited guests for that day, and that you may not mistake the way, I will strew ashes along **the path**.'

When Sunday came, and it was time for the girl to start, a feeling of dread came over her which she could not explain, and that she might be able to find [**her path**] again,

DN (6,1,3,4,e18),

DO ne75663 (8,1,3,6,e2)



# Ambiguity: REFERRING or NON-REFERRING?

There was nothing so VERY remarkable in that; nor did Alice think **it** so VERY much out of the way to hear the Rabbit say to itself, 'Oh dear! Oh dear! I shall be late!' (when she thought **it** over afterwards, **it** occurred to her that she ought to have wondered at this, but at the time **it** all seemed quite natural); but when the Rabbit actually TOOK A WATCH OUT OF ITS WAISTCOAT-POCKET, and looked at it, and then hurried on, Alice started to her feet, for it flashed across her mind that she had never before seen a rabbit with either a waistcoat-pocket, or a watch to take out of it, and burning with curiosity, she ran across the field after it, and fortunately was just in time to see it pop down a large rabbit-hole under the hedge.

# Ambiguity: DN / DO

The rooms were carefully examined, and results all pointed to an abominable crime. The front room was plainly furnished as a sitting-room and led into a small bedroom, which looked out upon the back of one of the wharves. Between the wharf and **the bedroom window** is a narrow strip, which is dry at low tide but is covered at high tide with at least four and a half feet of water. The bedroom window was a broad one and opened from below. On examination traces of blood were to be seen upon the windowsill, and several scattered drops were visible upon the wooden floor of the bedroom. Thrust away behind a curtain in the front room were all the clothes of Mr. Neville St. Clair, with the exception of his coat. His boots, his socks, his hat, and his watch -- all were there. There were no signs of violence upon any of these garments, and there were no other traces of Mr. Neville St. Clair. Out of **the window** he must apparently have gone





# `DN' when retelling a story as a dream

I went alone through [ne75972 a forest ] and came at last to  
[ne75974 a house] ....

DN (9,0,0,9,e2)


3 out\_of\_context\_window, 3 skips

# Preliminary figures

|         | Total | Dis   | GA         | ICP        |
|---------|-------|-------|------------|------------|
| LRC     | 401   | 79.1% | 28 (7%)    | 31 (7.7%)  |
| RG      | 464   | 68.3% | 52 (11.2%) | 60 (12.9%) |
| Average | 633   | 73.7% | 9.1%       | 10.3%      |

# MPA as ambiguity detector

- MPA is good at
  - Catching misunderstandings
  - Catching spurious ambiguity
- But not as good as ambiguity detector:
  - R: ~ 20%
  - P: ~ 50%



An hypothesis about  
justified and unjustified  
ambiguity

# Previous theories of 'unproblematic' ambiguity

- Poesio et al (1999, 2003, 2006)
  - **JUSTIFIED SLOPPINESS**: 'ambiguous' references considered felicitous when candidate antecedents form a **MEREOLOGICAL STRUCTURE**
- Versley (2008)
  - **GENERALIZED JUSTIFIED SLOPPINESS**: ambiguous references felicitous when antecedents part of a **DOT OBJECT** in the sense of Pustejovsky and Asher
- Recasens et al (2010, 2012, 2014)
  - **QUASI-COREFERENCE**: coreference relation is a **CONTINUUM** between **IDENTITY** and **NON-IDENTITY**


# Some additional evidence

- Frazier and Rayner (1990) and subsequent work on LEXICAL POLYSEMY: interpretation of polysemy different from interpretation of homonymy in that initial interpretation is not completely resolved (today we would say: **UNDERSPECIFIED**)
- The mereological cases cannot really be viewed as dot-objects
- Recasens et al 2014: identity, near-identity and not-identity **NOT A CONTINUUM**



# Preliminary new theory

- **UNDERSPECIFICATION HYPOTHESIS:**
  - Ambiguity is not problematic if the interpretations are part of an **UNDERSPECIFIED STRUCTURE**
    - But: mereological structure / dot objects are **DISTINCT** types of underspecified interpretation
- **There are cases of UNJUSTIFIED SLOPPINESS**
  - E.g., references to plans, areas
  - More similar to **GOOD-ENOUGH** cases (Ferreira et al)



# Using information about disagreement in anaphora resolution

# Previous work: using disagreement to filter

- Reidsma & Carletta (2008) and Beigman-Klebanov & Beigman (2009): use NOISE MODELS to exclude 'hard cases' from training
- The CrowdTruth project (Arroyo & Welty, 2014): DISAGREEMENT IS SIGNAL
  - Arroyo & Welty, 2013: use disagreement information to filter workers / sentences for relation extraction
  - See also Inel et al, 2014, 2017; Dumitrache et al, 2017, 2018
  - <http://www.crowdtruth.org>

# Previous work: train with a probabilistic model

- Plank et al (2014): develop a loss function such that weight update is discounted by a factor depending on disagreement on an item



# Conclusions

- Between 10% (written text, not considering discourse deixis) and 30% (spoken language, with deixis) of nominal expressions in language could be anaphorically interpreted in different ways
- This suggests that the assumption that each such expression has a unique 'gold' interpretation is only a convenient idealization
- We are developing (freely available) resources that will allow ourselves NLP researchers to train models that do not make that assumption

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