



## A Bootstrapping Approach for Stakeholder Identification in Public Comment Corpora

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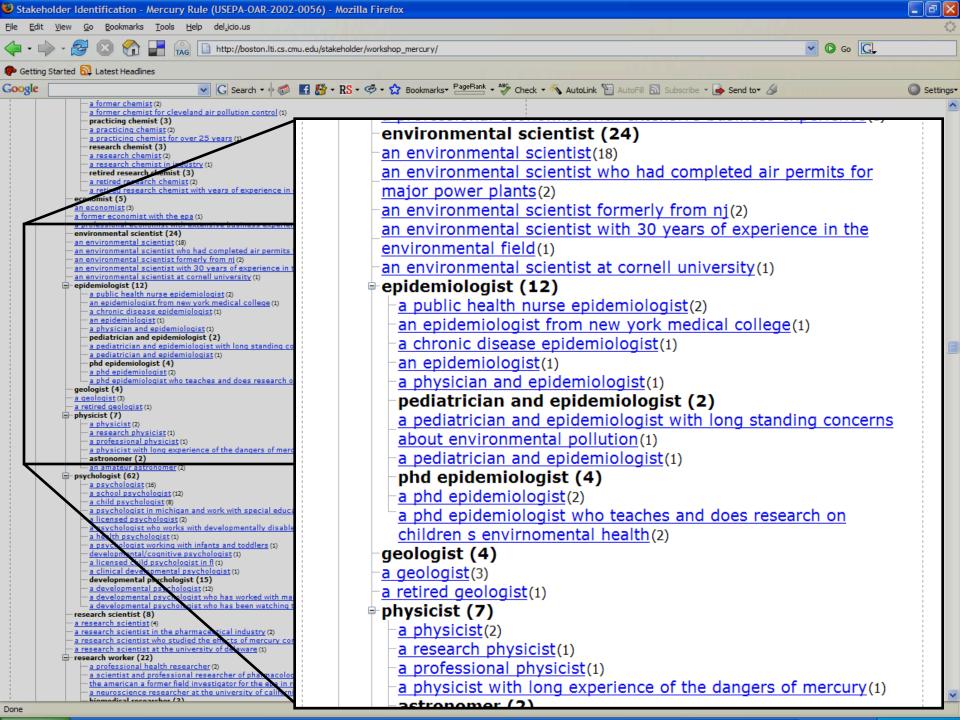
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#### Motivation

- Authors of public comments differ from one another in meaningful ways:
  - Level of impact
    - "I am a grandparent of children with environmentally aggravated illness."
    - "As a grandmother of 12 wonderful children, I urge you ..."
  - Expertise
    - "As a chemist that works in the area of fish analysis for mercury, I am concerned ..."
    - "Even as <u>a lay person</u>, I know that mercury is one of the most toxic substances known."
  - Geographical location
    - "I am <u>a resident of California</u> and someone who is aware ..."
    - "As a resident of a city with high mercury levels I am concerned about ..."
  - Level of representation
    - In EPA's Mercury Corpus, > 10% parents, of which 80% mothers
    - In DOI's Polar Bear Corpus, only 0.4% parents, but 77% wildlife advocates

### Motivation

- Government agencies may want to know who is represented in a corpus of public comments and may want to focus their attention on the subsets of comments from particular communities
- With only a search interface, this is difficult to do in large corpora
  - EPA's Mercury Corpus: 500,000 docs (12,249 stakeholders)
- Our goal is to provide:
  - a "bird's eye view" of the stakeholder landscape
  - a navigational aid



### Stakeholder definition

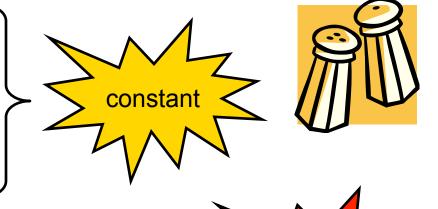
- Stakeholder = a group or community of which the author is a member (mentioned explicitly)
  - As a <u>former employee of a power company</u>, I know ...
  - As a woman of child-bearing age, I now have ...
  - I am <u>an avid fisherman</u> and I do not agree ...
- Sounds subjective?
  - Human agreement (f-measure): overlap = 0.70, exact = 0.53
  - Open question: what's the extent of the stakeholder mention?
    - "As <u>a person</u> that spent 2 two years recovering from mercury and arsenic poisoning, I am more than appalled that you ..."
    - "As a person that spent 2 two years recovering from mercury and arsenic poisoning, I am more than appalled that you ..."
- Stakeholder types not addressed
  - Impacted group (not necessarily the author) (e.g., children)
  - Entities being regulated (e.g., coal-fired power plants)

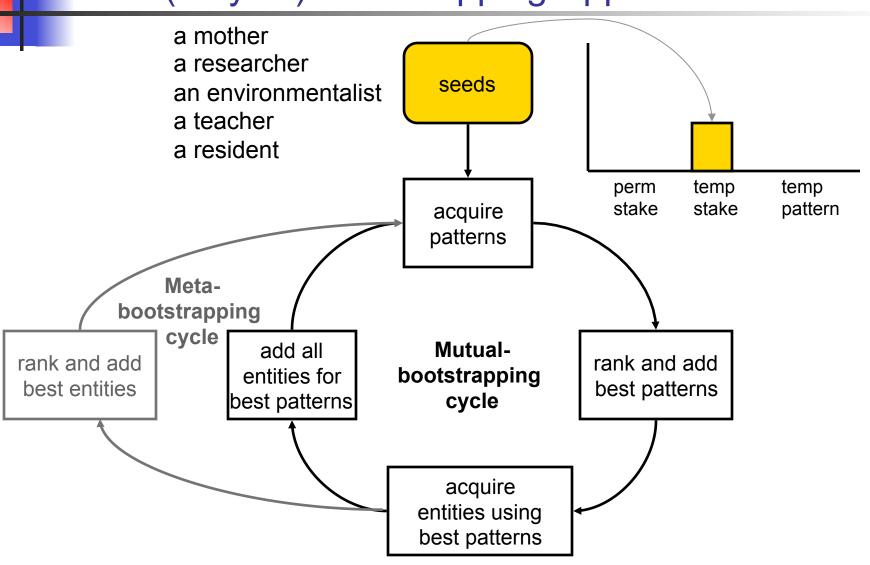
## Bootstrapping Approaches for Information Extraction: The General Idea

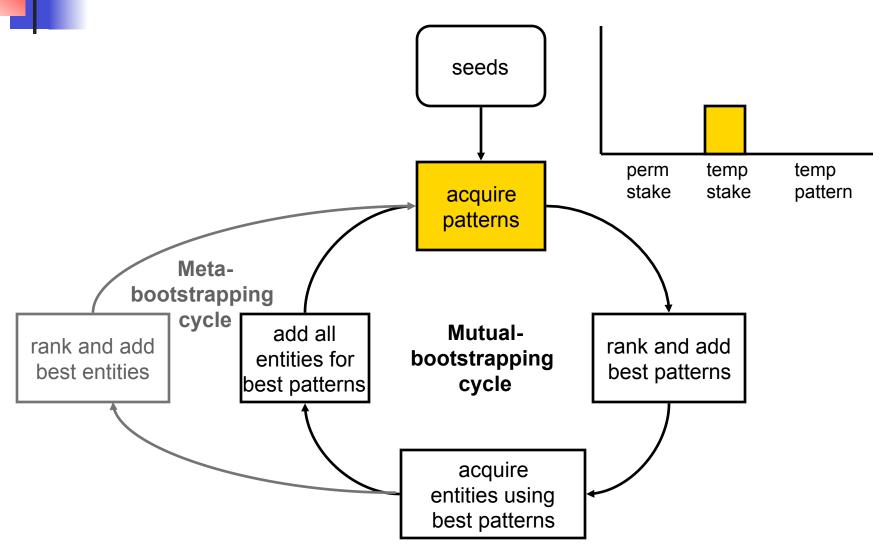
- Input: seeds
  - "a mother"
- Find all contexts where seeds occur
  - "As a mother, I worry about mercury pollution."
- Extract patterns from contexts
  - "As \_\_\_\_\_, I worry about mercury pollution."
- Score and keep only the best!!!
- Find other noun-phrases (NPs) that co-occur with patterns
  - "As a neurologist, I worry about mercury pollution."
- Score and keep only best!!!
- Add NPs to seeds
  - "a mother", "a neurologist"
- Repeat

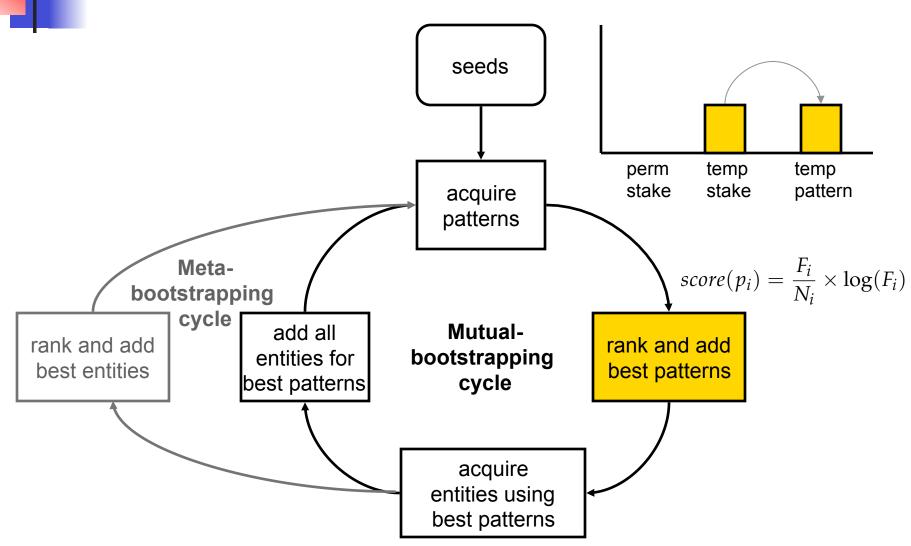
## Bootstrapping Approaches for Information Extraction: Ingredients

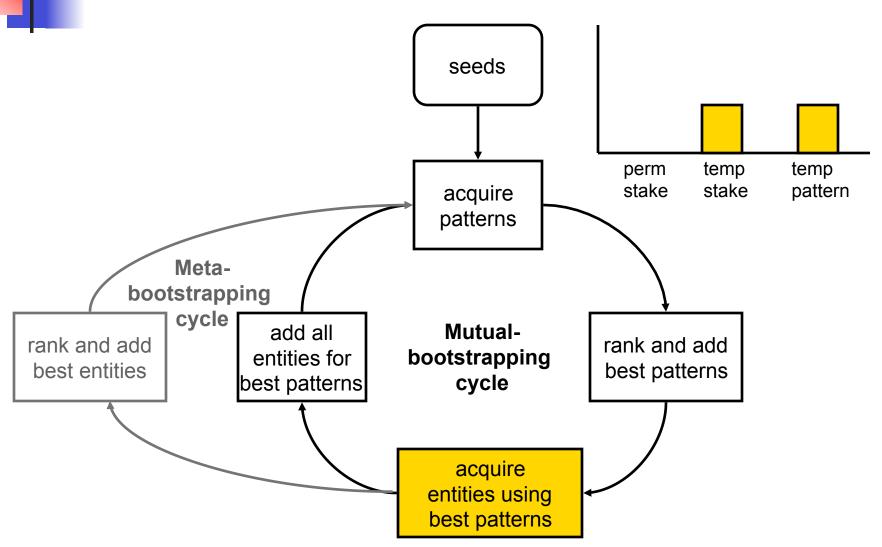
- The bootstrapping framework
- A scoring function for
  - Patterns
  - Entities
- A representation of context
  - "As \_\_\_\_\_, I worry about mercury pollution."
  - Definition: extraction pattern template (EPT)
  - What features of the context are the most informative for discriminating b/w target (stakeholder) and non-target (nonstakeholder) NPs?

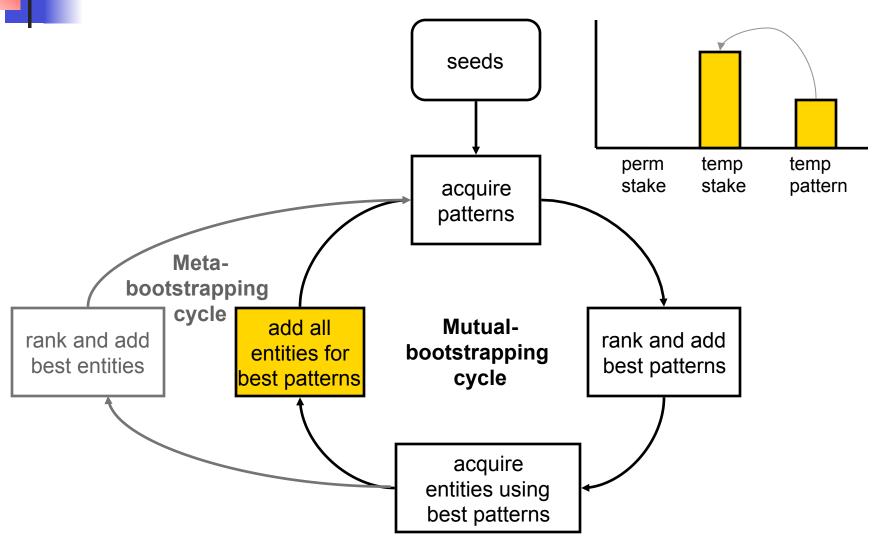


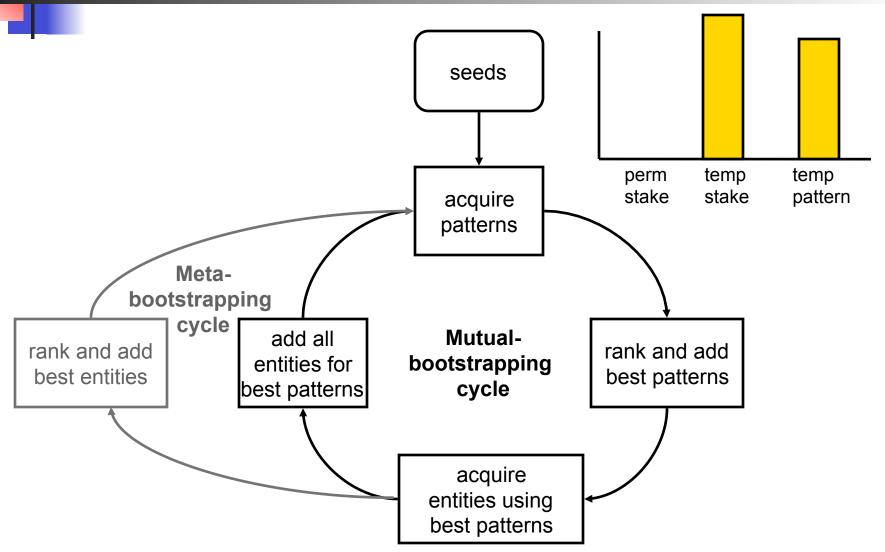


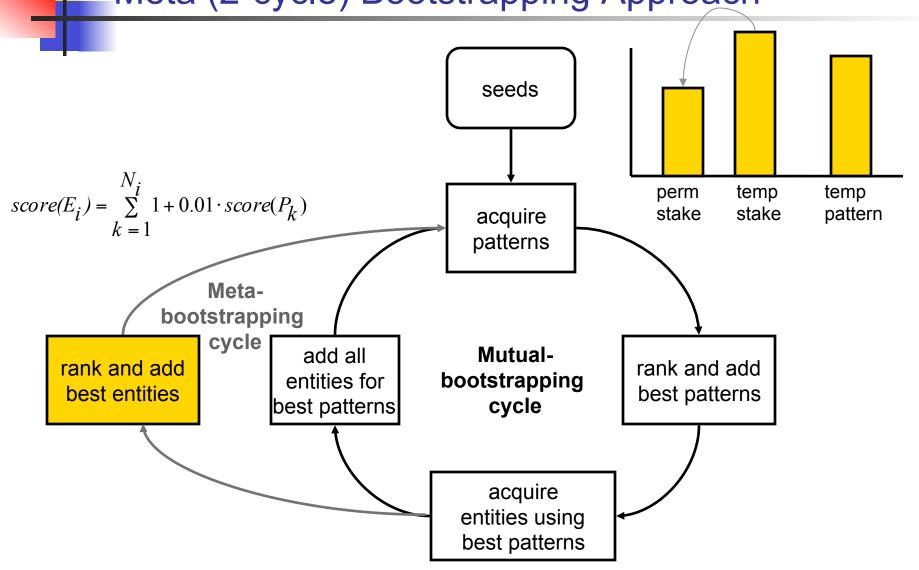


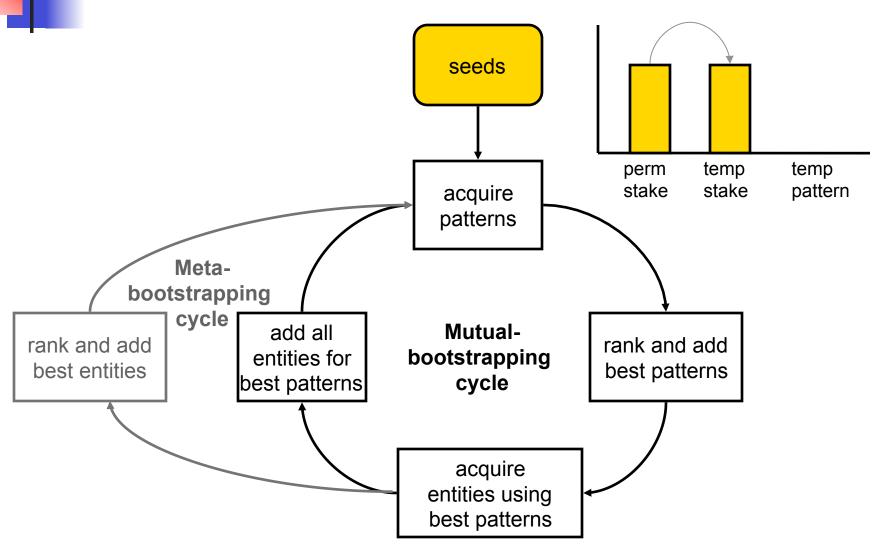












### 3 Extraction Patterns Templates

- 1) Surface-based patterns
  - $W_{L1} W_{L2} = W_{R1} W_{R2}$
  - W<sub>XY</sub> = {word,pos}
- 2) WordNet- (WN) based patterns
  - $W_{L1} W_{L2} = W_{R1} W_{R2}$
  - *W*<sub>XY</sub> = {**pos**}, *except if...* 
    - (1) part-of-speech( $W_{XY}$ ) = pronoun to avoid conflating 1<sup>st</sup> & 2<sup>nd</sup> person pronouns
      - "You are the agency responsible for protecting the environment, and I ..."
    - (2) part-of-speech( $W_{LY}$ ) = verb, to avoid conflating verbs that refer to the author and verbs that refer to a 3rd entity
      - "I have a child that suffers from mercury contamination, and I ..."
  - Stakeholder subsumed by "person" in WordNet

### 3 Extraction Pattern Templates (cont'd)

#### 3) Sundance-based patterns

- Sundance information extraction (IE) engine (chunking, shallow parsing, pronoun resolution, etc.) Used in prior work to extract: person, building name, victim, weapon, location, company name, professional title, ...
- 17 templates (15 verb-centric, 2 noun-centric), e.g.:

- Main assumption: target class NPs occur as arguments of some verbs more than others (?)
  - For, "person", perhaps (e.g., subject of verbs like "feel", "think", "say")
  - But, for "stakeholders"?

### Heuristics (Surface- and WN-based patterns)

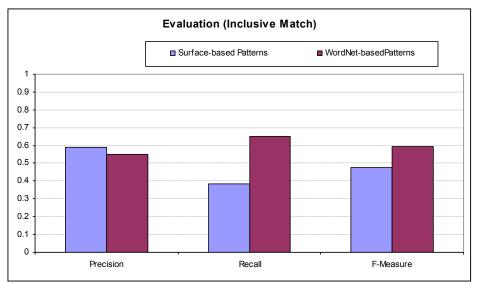
- NP-expansion (greedy NP chunker)
  - "I am a mother of two boys and I think that ...."
  - "I am a mother of two boys and I think ...."
- Head-NP querying
  - Long stakeholder NPs are rare
  - "a person who has children living by the infamous 4 corners power plant in New Mexico"
- List handling
  - "I am <u>a husband</u>, <u>a father</u>, <u>a teacher</u>, and <u>a concerned North</u> American."
- Adjective/Adverb padding
  - Allow optional adverb/adjective within learned patterns
  - "I am practically an advocate for the environment."

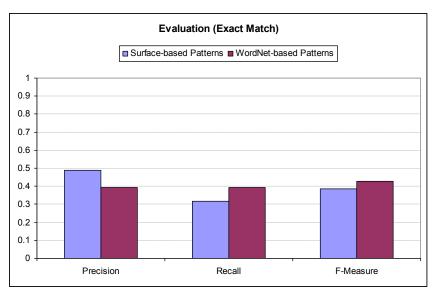
# Evaluation Methodology

- EPA's Mercury Corpus: > 500,000 documents
- Unsupervised Training
  - 120,000 documents, after duplicate detection
  - 5 seeds: a biologist, an environmentalist, a resident, a citizen, an American
  - Best 80 learned extraction patterns applied to test set
- Test set (annotated)
  - 1,020 documents
  - 60 stakeholders detected (about 1 in 20 documents)
- Measures: precision (P), recall (R), f-measure (F1):
  - Exact Match: reference NP = predicted NP
  - Inclusive Match: (reference NP within predicted NP) OR (predicted NP within reference NP)



### Results (Surface- & WN-based patterns)

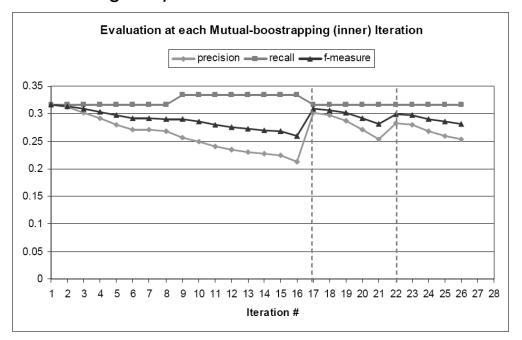




- Surface-based patterns suffer from low generalization, but have slightly higher precision
- WordNet-based patterns
  - Achieve higher recall by generalizing from the local context
  - Avoid loss in precision by imposing semantic constraint on extracted NP (NP is a "person")

### Results (Sundance-based patterns)

- None of the learned patterns occur in the test set
- Alternate evaluation
  - The system with seeded with all entities extracted by Sundance-based pattern "I am <dobj>"
    - best pattern on test set
  - Maximum advantage to perform well on this test set



### WN-based error analysis

- Effective extraction patterns follow heavy-tailed distribution
  - A few patterns extract many stakeholders and many patterns extract only a few stakeholders
- Bootstrapping algorithm rejects rare patterns due to low recall
- This is representation challenge
  - Local context should be overlooked in the presence of more meaningful long-distance evidence
  - "As a mother, the effect of mercury on children concerns me."
- Ambiguous contexts
  - "As a policy maker, I urge you to take immediate action ...."

### Sundance-based patterns error analysis

- Pronoun resolution
  - "As a [[former employee]<sub>NP</sub> of [the power industry] <sub>NP</sub>]<sub>NP</sub>, I know there ..."
  - "Being a [[child development]<sub>NP</sub> and [healthcare specialist] <sub>NP</sub>]<sub>NP</sub>, I am
    ..."
  - <subj> ActVp → I (NP antecedent) know/am
- Verb-centric assumption doesn't hold for stakeholders
  - Stakeholders do not occur exclusively as the subject/object of a verb set
    - "As <u>a mother</u>, I care about ..."
    - "This government cares only about ..."

### Conclusions

- Stakeholder mentions can be identified with about .60 precision/ recall in a bootstrapping framework.
- About 40% of the stakeholders are being missed
  - Rare contexts
    - "As a mother, the effects of mercury on children concerns me."
  - Ambiguous contexts
    - "As the agency responsible for the environment, I think ..."
  - Implicit mentions
    - "I live 5 miles away from a coal-fired power plant."
- What now?
  - Not committing to a single context representation
  - Organizing stakeholders (> 12,000 stakeholders in Mercury corpus)
  - Component technology for other text mining apps (e.g., sentiment analysis, identifying constructive comments)