#### A UIMA-Based QA System (D3)

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# Approach

- UIMA text processing pipeline
  - DKPro suite of NLP modules
  - Custom query and answer processing modules
- MALLET for classification
- Indri for indexing and passage retrieval

# UIMA

- Unstructured Information Management Architecture
- Provides structure for a pipeline of text processing components
- CAS Common Analysis Structure
  - Stores original text and annotations (feature structures) produced by components
  - Annotations exist in the context of a type hierarchy

#### **Question Classification**

- Created UIMA wrapper for MALLET MaxEnt classifier
- Nine classes based on Li and Roth's taxonomy:
  - ABBR
  - ENTY
  - DESC
  - NUM
  - LOC
  - HUM
  - DATE
  - MANNER
  - **REASON**

(DATE was a subclass of NUM) (MANNER was a subclass of DESC) (REASON was a subclass of DESC)

#### **Question Classification**

- On initialization, the system checks for a serialized classifier file to load
  If it doesn't exist, the classifier is retrained on 5500 labeled questions from Li and Roth (2004)
- The only features used are unigrams, but results were fairly good
- The default MALLET stopword removal component removes question words, which led to low accuracy
  - Other than fixing this component, no other modifications to default MALLET preprocessing components were made

#### **Question Classification**

- Performed experiments with different classifiers and sets of classes
- Naive Bayes
  - Fine set: 0.520
  - Coarse (6 class) set: 0.748
- MaxEnt
  - Fine set: 0.766
  - Coarse (6 class) set: 0.844
  - Coarse (9 class) set: 0.862

#### **Question Sets**

- Anaphor resolution using context within question sets
  - Created a list of contextually relevant entities and their types
  - The list is cleared for each new question set
- Stanford NER DKPro module used to extract entities from question text
- Question classification combined with extracted answers
- For each anaphor in the question text, the most recently added entity that matches the anaphor's type is determined to be the referent
  - $\circ$  'he' and 'she' are assumed to refer to human entities
  - An annotation on the anaphor is added to the CAS for use in later modules

## **Query Reformulation**

- Question text is parsed using the Stanford factored parser DKPro module
  Adds constituent and dependency annotations to the CAS
- Reformulates query string using techniques from MULDER (Kwok et al., 2001)
  - Subject-Aux movement
    - "was the Crip gang started" > "the Crip gang was started"
  - Subject-Verb movement
    - strip wh-words from queries
  - Verb Conversion
    - change aux + infinitive to conjugated form

## Misc. Improvements

- New DKPro components integrated:
  - Chunker
  - Parser
  - NER
- Modified Indri queries to put more weight on named entities and chunks rather than unigrams

#### Issues

- DKPro Text Classifier module is still in development
  - Doesn't support using a trained classifier to label new instances...
- Needed to write a fair amount of wrapper code to use MALLET as a UIMA component
  - Some duplicated components, since MALLET uses its own preprocessing pipeline
- Manipulating the parse annotations in the CAS ended up being tricky, so the query reformulation module is not yet functional
- Answer extraction is still being improved to take advantage of better techniques and information from the question classification and anaphor resolution modules

#### Results

- MRR scores computed on TREC 2006 test questions:
  - Strict: 0.0438 (D2: 0.0176)
  - Lenient: 0.0927 (D2: 0.0510)
- Main next step is to get improved answer extraction modules functioning and replace the baseline modules

#### References

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