# SFU Q/A Summary Handler (SQuASH) for DUC-2006

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## Our Approach

(similar to DUC-2005)

#### 1) Annotate:

Perform syntactic and semantic analysis

#### 2) Extract:

- Content selection
- Optimization on DUC05 data

#### 3) Edit:

Improve the linguistic quality of summary

# **Annotation Components**

#### Syntactic Annotation

- Sentence Boundary Detection and Tokenization (Lingpipe)
- Part of Speech Tagging
- Parsing (Charniak)

#### Semantic Annotation

- Named Entity Recognition (Lingpipe)
- WordNet Relation Identification (WordNet)
- Semantic Role Labeling (Assert)

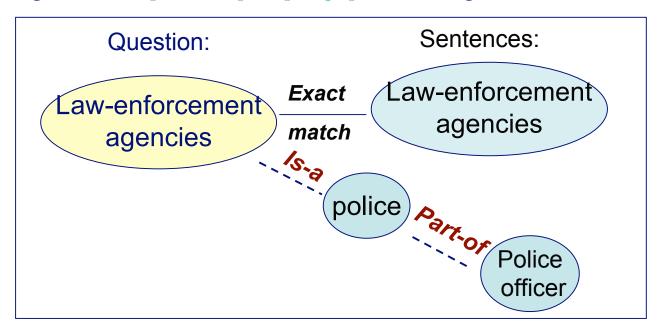
## WordNet Relation

#### Motivation

 Better identify relations between sentences and questions as well as relations between sentences

#### Example:

Q: What [sorts of] [law-enforcement] [tasks] are [dogs] being used for worldwide? What [law-enforcement agencies] are using dogs? What [breeds] of [dogs] are being used?



# Semantic Role Label (SRL)

#### Semantic Role

 The relationship that a syntactic constituent has with a predicate

#### Example

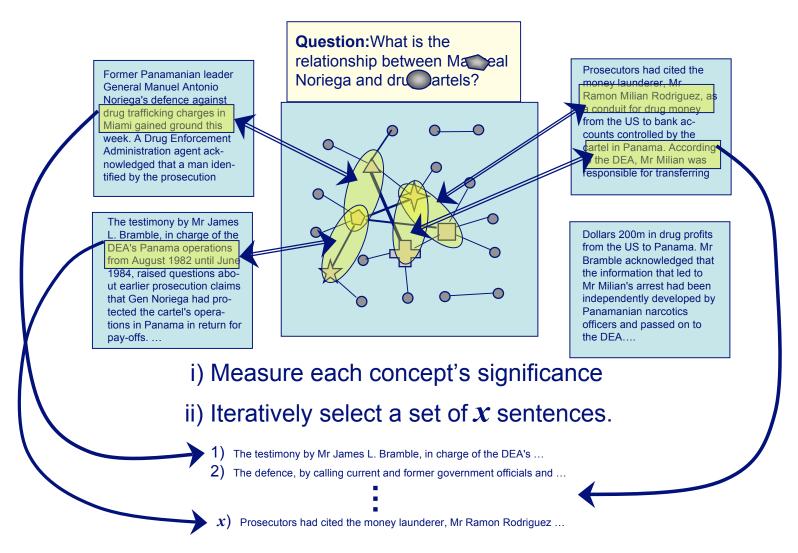
 Late buying gave the Paris Bourse a parachute after its free fall early in the day

```
[A_0 Late buying][ V gave][A_2 the Paris Bourse]
[A_1 a parachute] [AM-TMP after its free ...the day]
```

## SRL in SQuASH

- SRL in Sentence Extraction
  - Each clause (predicate and core arguments in SRL) gets a score
  - A sentence is selected by its single highest scoring clause
- SRL in Sentence Compression
  - Remove adjuncts (ARG-DIS, ARG-TMP)
  - Sentence syntactically correct without losing important semantic information
- Use ASSERT (Pradhan et al. 2004)
  - Precision 84%, F-score 79.4% on PropBank Corpus

## Sentence Extraction



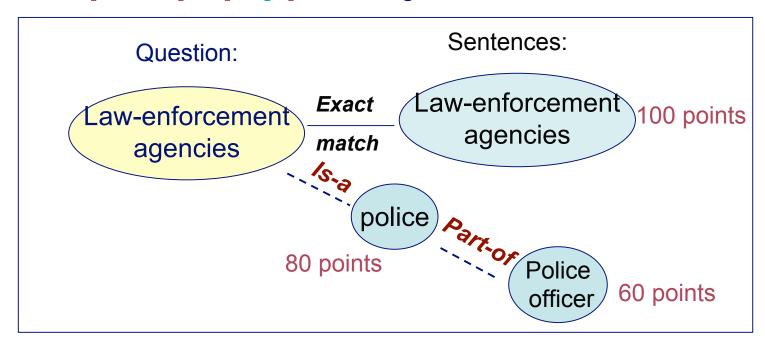
### **Extractor Features**

- Concept in clause appears in question
- Semantic relatedness of clause to question concept
- Number of Named Entities in clause
- Number of documents in which the concepts from a clause appear
- Number of sentences in which the concepts from a clause appear
- First Sentence bias

# Concept Relatedness

- The concept 'score' decreases as the distance of relatedness increases.
- Example:

Q: What [sorts of] [law-enforcement] [tasks] are [dogs] being used for worldwide? What [ law enforcement agencies ] are using dogs? What [breeds] of [dogs] are being used?



# Sentence Compression

- Motivation
  - Improves the ROUGE score on DUC05 data
  - Rouge 2 from 0.0649 to 0.0714
  - Rouge SU-4 from 0.1171 to 0.1284
- New handling of rules
  - Remove temporal and discourse conjunctives based on **Semantic Role Labels**: ARG-TMP and ARG-DIS
  - Remove sentences starting with pronouns excluding "it"

# Sentence Ordering (1)

#### Assumption

Semantically related sentences are ordered together [Barzilay, 2002]

#### Approach

- Semantic relatedness on the question level
  - Cluster sentences based on questions
- Semantic relatedness on the sentence level
  - Use WordNet Concept relations to calculate the lexical cohesion between sentences
- Pick the 1st sentence of the summary: Prefer the original 1st sentence

# Sentence Ordering (2)

#### Evaluation

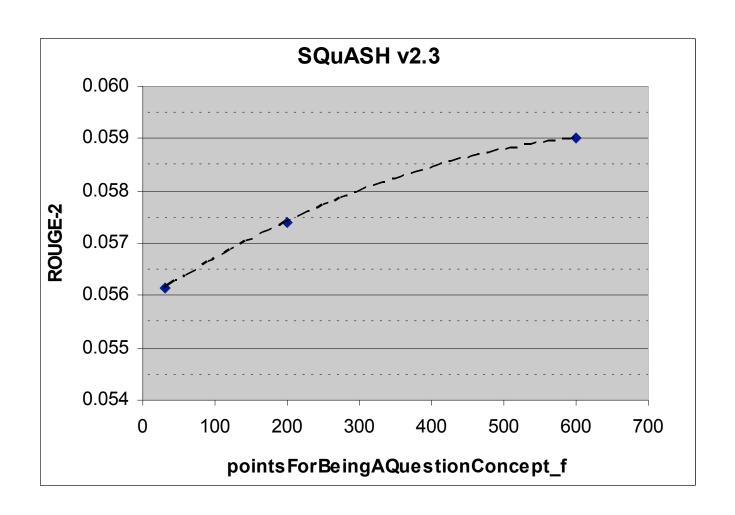
- compare summary with ordering and without ordering (a ranked sentence list from Extractor) on DUC 06 data
- Gather coherence judgments from 5 subjects
- Summary with ordering is better than summary without ordering

With Order	Good	Bad
W/o Order		
Good	8	12
Bad	20	6

# **Model Optimization**

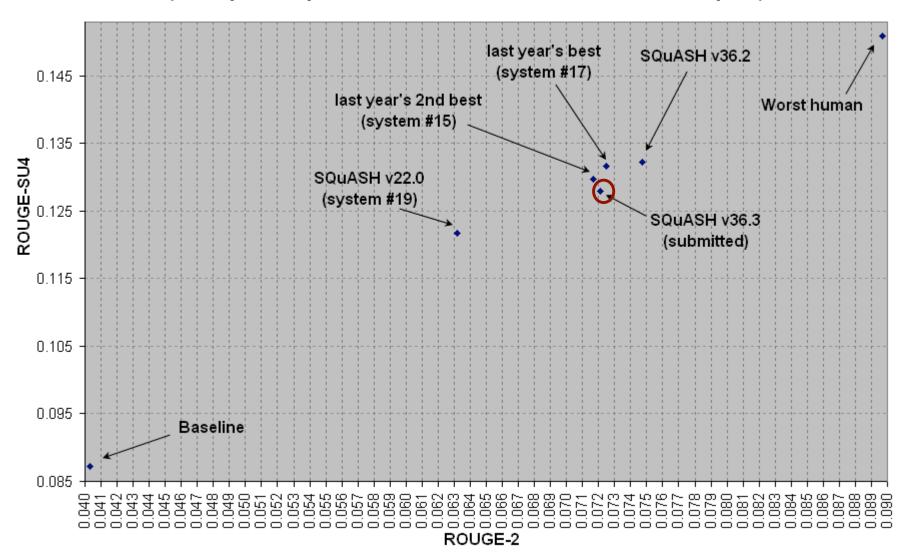
- Automatically tune and optimize the weights based on ROUGE-2 / SU-4 score on DUC-05 data
- Greedy hill-climbing technique
- The weight for each feature was optimized separately

# Sample of Optimization Exercise

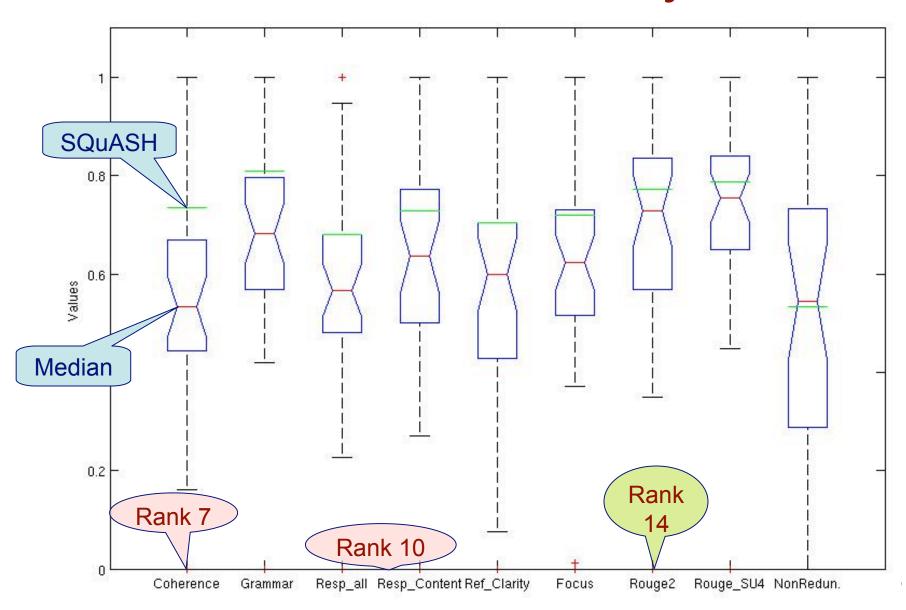


# System Optimization

ROUGE-2 vs ROUGE-SU4 performance relative to other systems (each point represents an eval over all 50 DUC-2005 topics)



# Performance Analysis



# Performance Analysis

 Correlation between content responsiveness and structure & coherence

Content		Structure & Coherence	
Responsiveness	Percent	Good (5,4)	Bad (2,1)
5	2.95%	51%	16%
4	15.76%	31%	38%
3	30.41%	16.25%	57.45%
2	36.06%	9%	67%
1	6.47%	4%	80%

## **Future Work**

- Annotator
  - Improve recall in Semantic Role Labeling
- Extractor
  - Discriminative model between the document text graphs and the human summaries
  - Link text graph vertices to an ontology.
     E.g. Person, Location, and organization ontology.

## **Future Work**

#### Editor

- Conduct the experiment study to test the hypothesis of relations between query and sentence ordering
- Incorporate the query in sentence compression and develop more sophisticated compression model

# Thank you!

