Content Realization #2

LING 573 — Systems & Applications May 10, 2018





Begin Recording





Announcements:

- Please include your condor file as ../src/D3.cmd
 - Helps with ease of grading having consistent location.





Miscellanea





I Feel Personally Attacked By XKCD

	COMMENT	DATE
Q	CREATED MAIN LOOP & TIMING CONTROL	14 HOURS AGO
φ	ENABLED CONFIG FILE PARSING	9 HOURS AGO
φ	MISC BUGFIXES	5 HOURS AGO
0	CODE ADDITIONS/EDITS	4 HOURS AGO
Q	MORE CODE	4 HOURS AGO
0	HERE HAVE CODE	4 HOURS AGO
	AAAAAAAA	3 HOURS AGO
þ	ADKFJ5UKDFJ5OKLFJ	3 HOURS AGO
φ	MY HANDS ARE TYPING WORDS	2 HOURS AGO
ļ þ	HAAAAAAANDS	2 HOURS AGO

AS A PROJECT DRAGS ON, MY GIT COMMIT MESSAGES GET LESS AND LESS INFORMATIVE

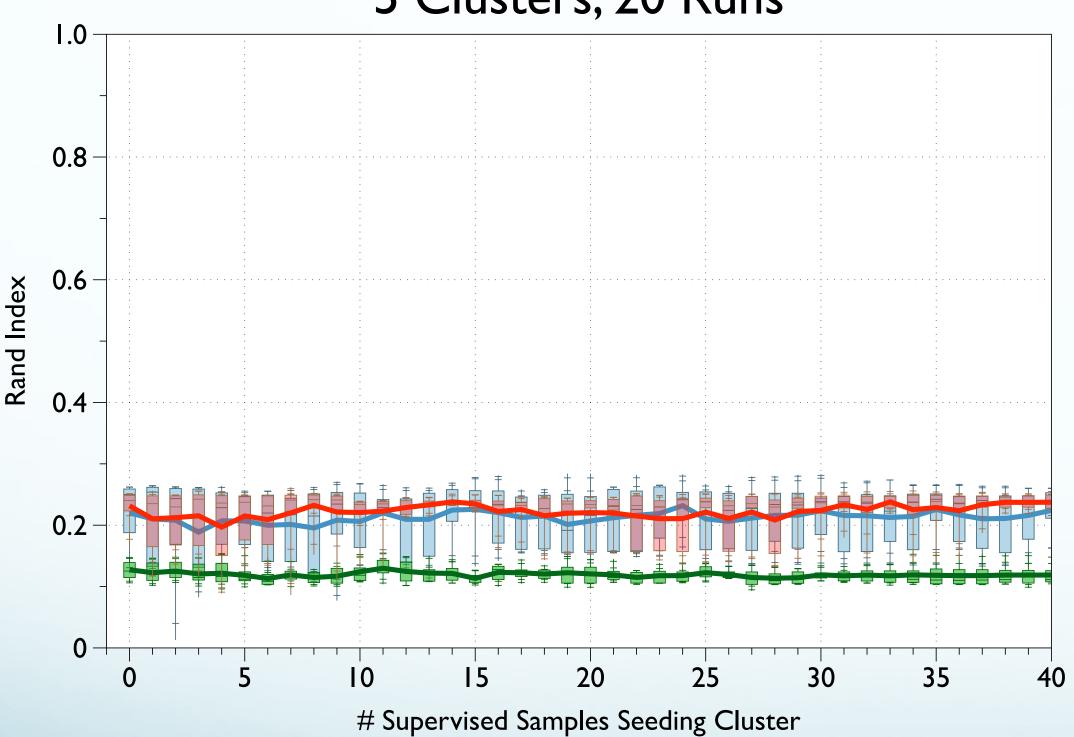
Merge branch 'asdfasjkfdlas/alkdjf' into sdkjfls-final



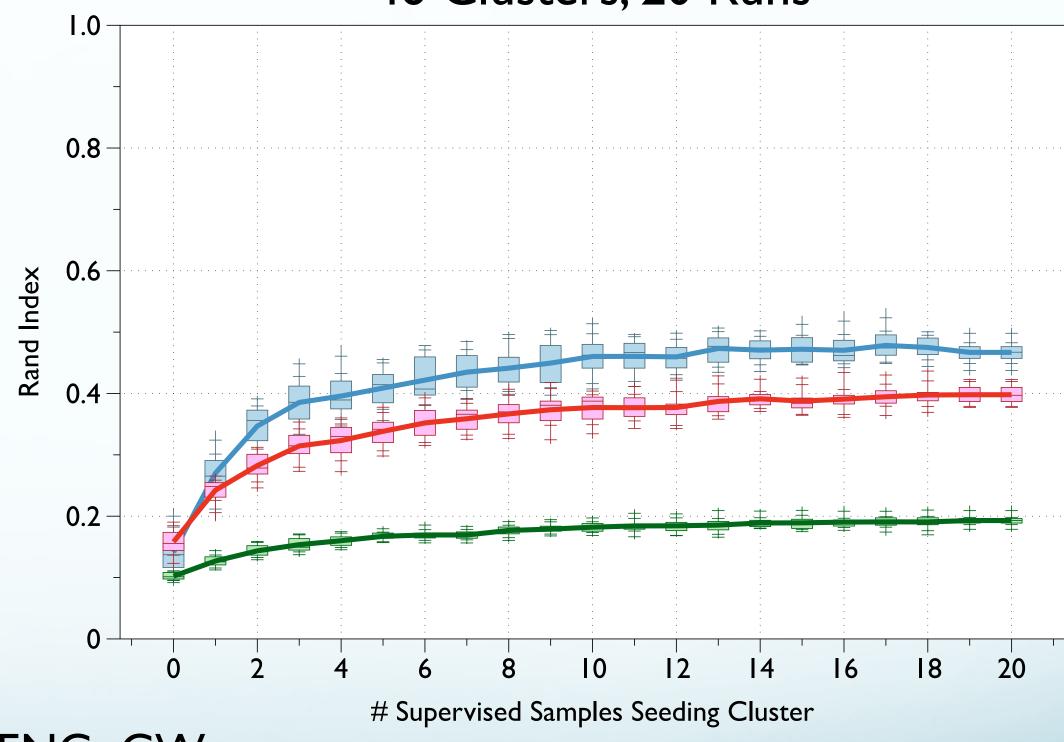


The Continuing Saga of Embeddings

KMeans Clustering on TAC Categories 5 Clusters, 20 Runs



KMeans Clustering on TAC Topics 46 Clusters, 20 Runs



= 100M ENG_GW = 10M ENG_GW

= IM ENG_GW



The Continuing Saga of Embeddings

- gensim Doc2Vec embeddings on IM, I0M, I00M:
 - /dropbox/17-18/573/other_resources/word_embeddings/eng_gw
- IB, full set coming soon.
- Obviously, YMMV





Roadmap / Overview

So far:

- We've discussed Linguistic Quality
- How to define the particular issues that affect summarization system
- Examined a system meant to (re)rank output based on linguistic quality

Now:

- Improving references
 - Contributes to better readability,
- Introduction to compression
- Can help with long, run-on sentences improve included content





Content Realization: Reference Resolution





NP-Rewrite

(Nenkova, 2008)

- Intuition:
 - Referring expressions are common sources of errors
 - References to people prevalent in news data, summaries
 - Targeted rewriting can improve readability
- Other intuitions:
 - Some entities are more important than others
 - Important entities may need less reintroduction





Challenges

- Lack of training data:
 - No summary data labeled for information status
- Readers sensitive to referring expressions
 - Prior work on NP rewriting has shown mixed results
 - Some improvement, some failures
- Relies on coreference
 - May contain errors





NP Rewrite Very Good Example

• While the British government defended the arrest, it took no stand on extradition of Pinochet to Spain, leaving it to the courts.

• While the British government defended the arrest in London of former Chilean dictator Augusto Pinochet, it took no stand on extradition of Pinochet to Spain, leaving it to British courts.





NP Rewrite Mixed Example

Duisenberg has said growth in the euro area countries next year will be about
 2.5 percent, lower than the 3 percent predicted earlier.

• Wim Duisenberg, the head of the new European Central Bank, has said growth in the euro area countries next year will be about 2.5 percent, lower than just I percent in the euro-zone unemployment predicted earlier.





NP-Rewrite: Approach

(Nenkova, 2008)

I. Estimate "importance" of each content word by MLE

$$p(w_i) = \frac{n_i}{N}$$

- 2. Estimate weight for sentence by sum of p(w)s
- 3. Select sentence with highest weight
 - a. Find all NPs in this sentence
 - b. Find highest-weighted coreferring NPs across all documents
 - c. Replace NPs in this sentence with highest-weighted NP





Results

- Not significant improvement
 - 0.4039 0.4169 on Pyramid
 - Out of 50 test cases:
 - 22 worse
 - 23 better
 - 5 the same





Discussion

Problems:

- Extremely basic content selection approach
- Relies on coreference difficult task to perform automatically

Possible Improvements:

Perhaps similar method could be used with more advanced content selection





Information Status Distinctions:

(Siddarthan et al 2011)

Intuition:

- Discourse and background world knowledge are proxies for "Information Status"
- Referring expressions should be conditioned upon their information status
- Information status constrains realization

Approach:

- Automatically identify information status
- Use IS to guide rule-based generation of referring expressions





Information Status Distinctions:

(Siddarthan et al 2011)

- discourse-new vs. discourse-old
 - Is this first mention of person, or subsequent mention?
- hearer-new vs. hearer-old
 - Person less well known vs. likely very well known
 - Don't waste space describing well-known individuals
- Major vs. Minor Character
 - Salience of the person in the news story





Information Status Distinctions:

(Siddarthan et al 2011)

- Research Question:
 - Is it possible to infer these distinctions via modeling?
 - Apply this model to predict information status?





Corpus Analysis

- Assess relation between:
 - Information status and referring expressions
- Used data from DUC 2001–2003
- NER, Coreference automatic

		Discourse-new	Discourse-old
Name Form	Full name	0.97	0.08
	Surname only	0.02	0.87
	Other (e.g., Britney, JLo)	0.01	0.05
Pre-Modification	Any	0.51	0.21
	None	0.49	0.79
Post-Modification	None	0.60	0.89
	Apposition	0.25	0.04
	Relative clause	0.07	0.03
	Other	0.08	0.04
Any Modification	Some Modification	0.76	0.30
(Either Pre- or Post-)	No Modification	0.24	0.70

Siddarthan et al, 2011 p. 818, Table 1





Generating Discourse-New

- I. IF NP head is person name,
 - a. IF appears with pre-modifier, THEN
 - i. insert longest pre-modifier + full name
 - b. ELSE IF appears with apposition modifier THEN
 - i. add apposition to reference
- 2. ELSE don't rewrite





Generating Discourse-Old

1. Use surname only, remove all pre/post modifiers.





Example Rewrite

- Honecker has come under investigation for charges of corruption and living in luxury at the cost of the state. Former East German leader Erich Honecker may be moved to a monastery to protect him from a possible lynching by enraged citizens. As protests gathered strength last fall, Erich Honecker, East Germany's longtime orthodox leader "lost touch with reality," according to the man who succeeded him as Communist leader only to be ousted later. Ousted East German leader Erich Honecker, who is expected to be indicted for high treason, was arrested Monday morning upon release from a hospital and taken to prison.
- Former East German leader Erich Honecker has come under investigation for charges of corruption and living in luxury at the cost of the state. Honecker may be moved to a monastery to protect him from a possible lynching by enraged citizens. As protests gathered strength last fall, Honecker "lost touch with reality," according to the man who succeeded him as Communist leader only to be ousted later. Honecker, who is expected to be indicted for high treason, was arrested Monday morning upon release from a hospital and taken to prison.





Detecting Features

- discourse-new vs. discourse-old:
 - Obvious from summary order
- Major vs. Minor and hearer-new vs. hearer-old?
 - How to establish hearer or major/minor status?
 - Train on mentions in model summaries





Detecting Features

- hearer-new vs. hearer-old
 - Infer familiarity via whether model summaries include:
 - title/role+surname or unmodified fullname
- Major vs. Minor:
 - Referred to by name in some human summary of topic vs. never referenced
 - 258 major/3926 minor by data





Classifier Training: Features

• frequency:

- "rf•idf" for references (number of references, total docs w/reference)
- Proportion of first mentions containing full name
- Number times person referenced after first mention

lexical

- Probability of first reference by bigram
- Top 20 high frequency description words present in initial references (est. from newswire)





Classifier Training: Features

syntactic

- # appositives or rel. clauses attaching to initial reference
- # times appositives used
- # times rel. clauses used
- # copula constructions involving referent

classifiers:

Used SVM, Decision Trees





Classifier Results

System	All Features	Majority Class Baseline
hearer-new	0.75	0.70
hearer-old	0.76	0.00
major	0.60	0.00
minor	0.98	0.97

• Significantly better than baseline in all cases





Applying Classifier Decision

IF discourse-new AND NP head is person name

IF minor

exclude name, use only role, modifiers, etc

IF MAJOR and hearer-old

Include name and role/temporal (only)

IF MAJOR and hearer-new

Include name and role/temporal

Also include affiliation, post-mod (classifier)

IF discourse-old:

Surname ONLY





Evaluation

- Created (nearly) deterministic rule set
 - Based on information status classification
 - To rewrite referring expressions in extractive summaries
- Evaluated in paired preference tests
 - Compared original extractive and rewritten summaries
- Where a preference was expressed
 - Rewritten summaries rated as more coherent
 - Extractive rated as more informative
 - Why? Rewrite rules generally shrink rather than add content





Discussion

• Pros:

- Intuitive, interpretable model
- Solid results: ~0.75 accuracy, higher if humans agree
- Often preferred to extract

Cons:

- Limited: only applies to person names
- Error propagation: coreference, NP extraction
- Ignores other aspects of realization, i.e. length





Sentence Compression





Sentences Compression Motivation

- Summaries are short!
- Maximizing information means removing extraneous information.
- Former East German leader Erich Honecker, who "lost touch with reality,"...
 - → Honecker...
 - ...saving of ten words!
 - Remove anaphora
 - Remove relative clause





Sentences Compression Motivation

- As U.S. Court of Appeals Judge Sonia Sotomayor made her Senate debut with a series of private meetings, Republicans said they would prefer holding hearings on her nomination in September, which could cloud the speedy summertime confirmation Obama wants.
- In this sentence, there are:
 - Discussion of the series of meetings
 - Timing of meetings
 - Political opinion of meetings





Compression Approaches

- At a high level:
 - Find ways to remove portions of sentences
 - ...while maintaining grammaticality!
- Approaches:
 - Heuristc / Rule-Based
 - Machine Learning Approaches
 - Integrated with Selection / Optimization





Overview of Approaches

- Heuristic approaches
 - Deep vs. Shallow processing
 - Information- vs. readability-oriented
- Machine-learning approaches
 - Sequence models: HMM, CRF
 - Deep vs. Shallow information
- Integration with selection
 - Pre/post-processing: Candidate selection: heuristic/learned





Common to Many Approaches

- Start with syntactic parse
 - Way to represent sentence as subcomponents
 - Can learn or write rules
 - For which parse rules / subtrees should be maintained or not





Form	CLASSY	ISCI	UMd	SumBasic+	Cornell
Initial Adverbials	Y	M	Y	Y	Y
Initial Conj	Y		Y	Y	
Gerund Phrases	Y	M	М	Y	M
Relative clause appositives	Y		M	Y	Y
Other adv	Y				
Numeric: ages,	Y				
Junk (byline, edit)	Y				Y
Attributives	Y	Y		Y	Y
Manner modifiers	M	Y	М		Y
Temporal modifiers	M	Y	Υ		Y
POS: det, that, MD			Y		
XP over XP			Υ		
PPs (w/, w/o constraint)			Y		
Preposed Adjuncts			Υ		
SBARs			Υ		M
Conjuncts			Υ		
Content in parentheses		Y			Y





After Presentations:

• Deeper dive into sentence compression approaches





Deliverable #3

- Code Due Monday, I I:00pm PT
- Report Due Thursday, I I:00pm PT
- Presentations Due Ipm day of presentations
 - Presentation Time Slots:
 - Will be randomized—will post this afternoon
 - Will reverse order for D4.
- Note!
 - Please include your condor submit file in ../src/D3.cmd.





Deliverable #4

- Final system
 - Continue system improvement
 - Add content realization
- Evaluation:
 - Devtest (2010)
 - Evaltest (2011)
 - New blind test



