

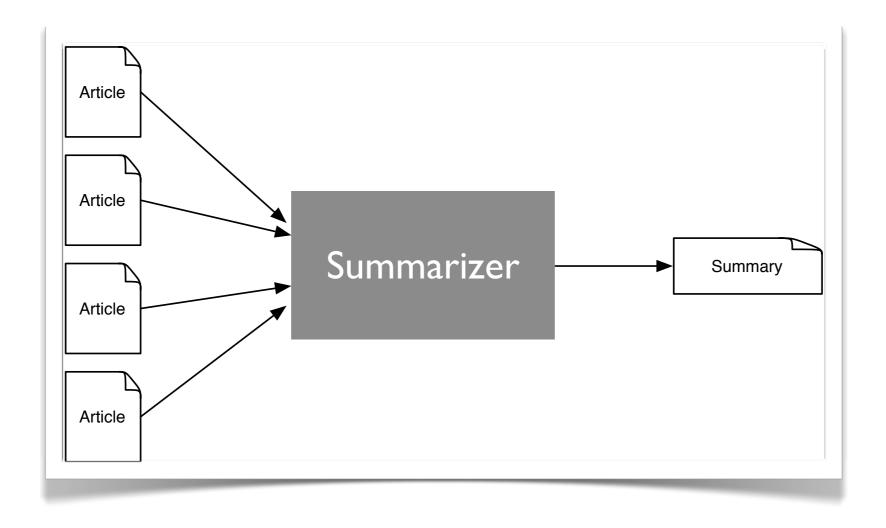


Exploiting Category-Specific Information

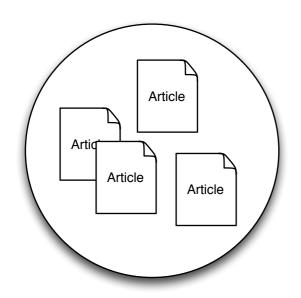
for Multi-Document Summarization

Jun-Ping Ng Praveen Bysani Ziheng Lin Min-Yen Kan Chew-Lim Tan

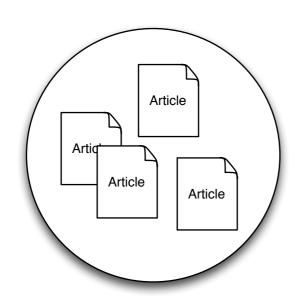
Multi-Document Summarization



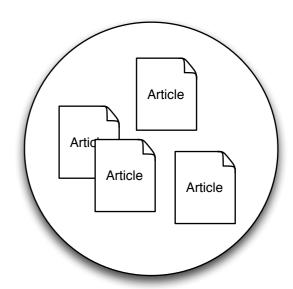
A Realistic Scenario



2012 US Presidential Election

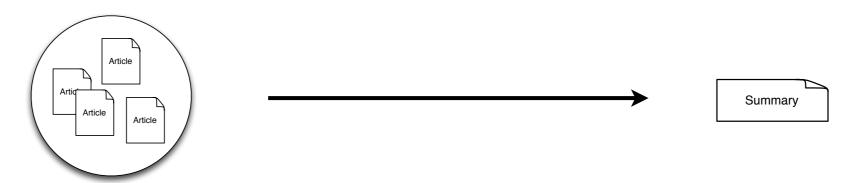


Hurricane Sandy



iPad Mini Release

Typical Summarizers



2012 US Presidential Election

Typical Summarizers



Hurricane Sandy

Article

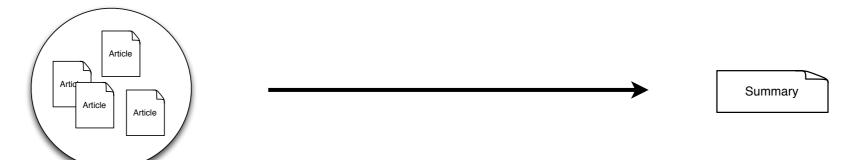
Typical Summarizers



2012 US Presidential Election

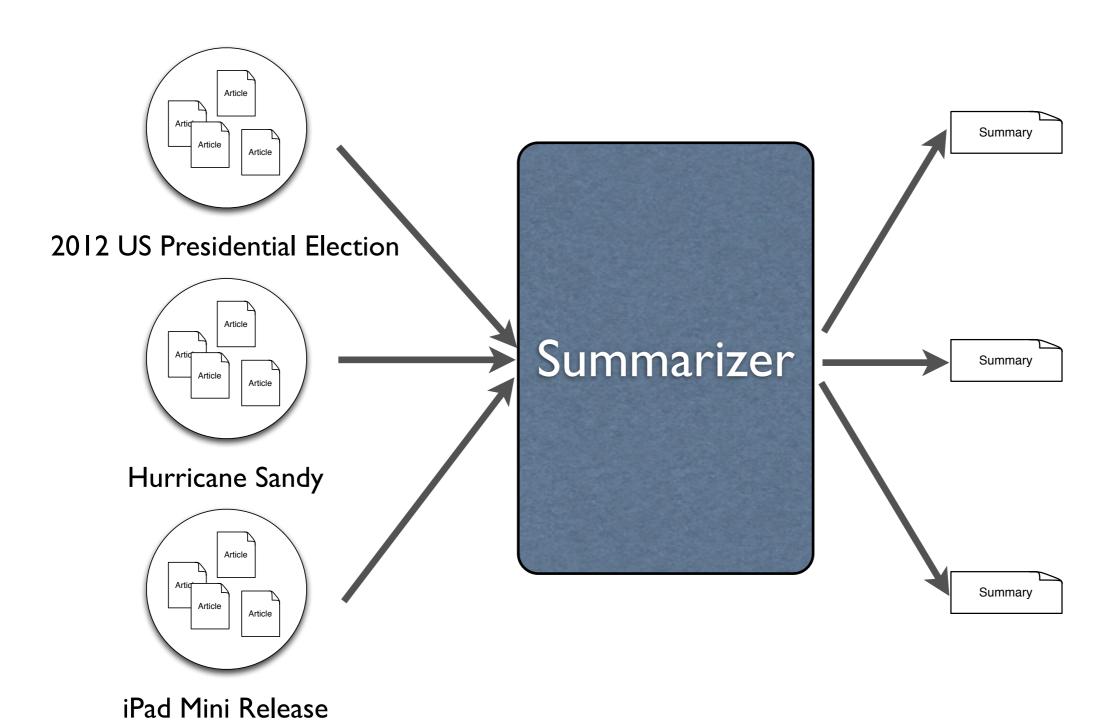


Hurricane Sandy



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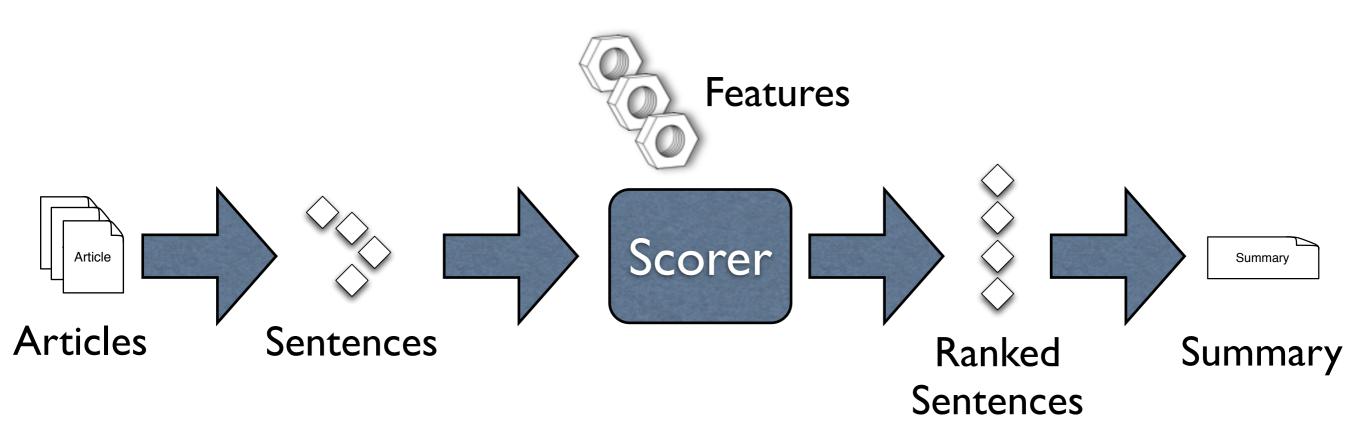
Key Insight



Outline

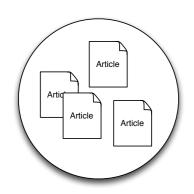
- Baseline Summarization Pipeline
- Category-Specific Features
- Experiments

Pipeline

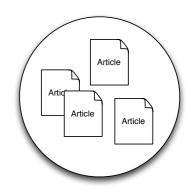


Generic Features

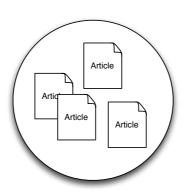
- Sentence position
- Sentence length
- Interpolated N-gram Document Frequency



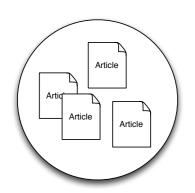
2012 US Presidential Election

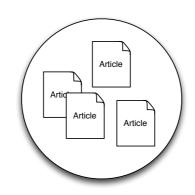


Hurricane Sandy

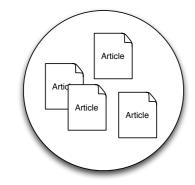


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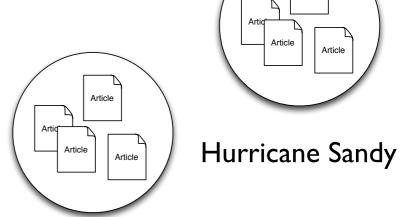




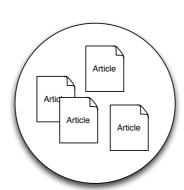
2012 US Presidential Election 2011 UK Election



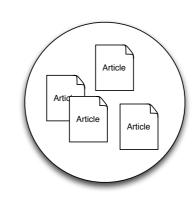
2012 Indian Assembly Elections



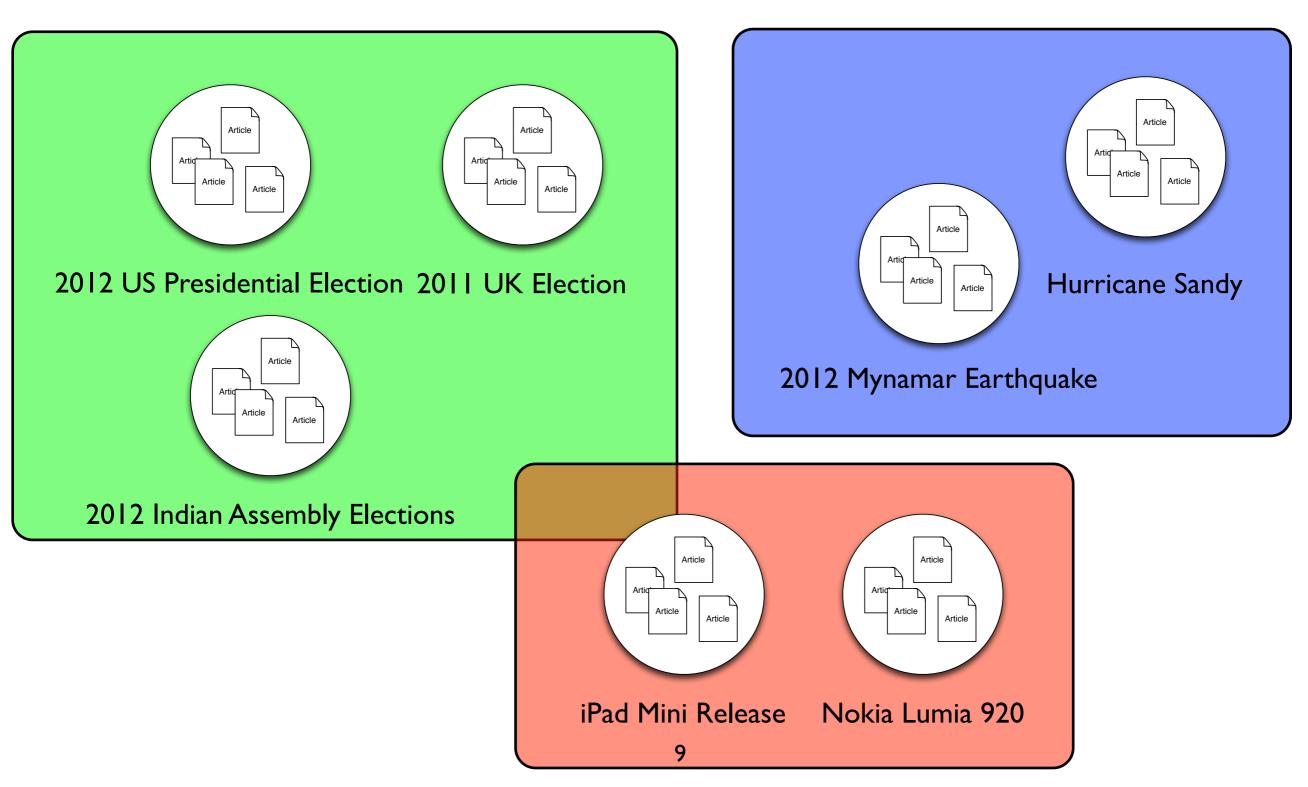
2012 Mynamar Earthquake

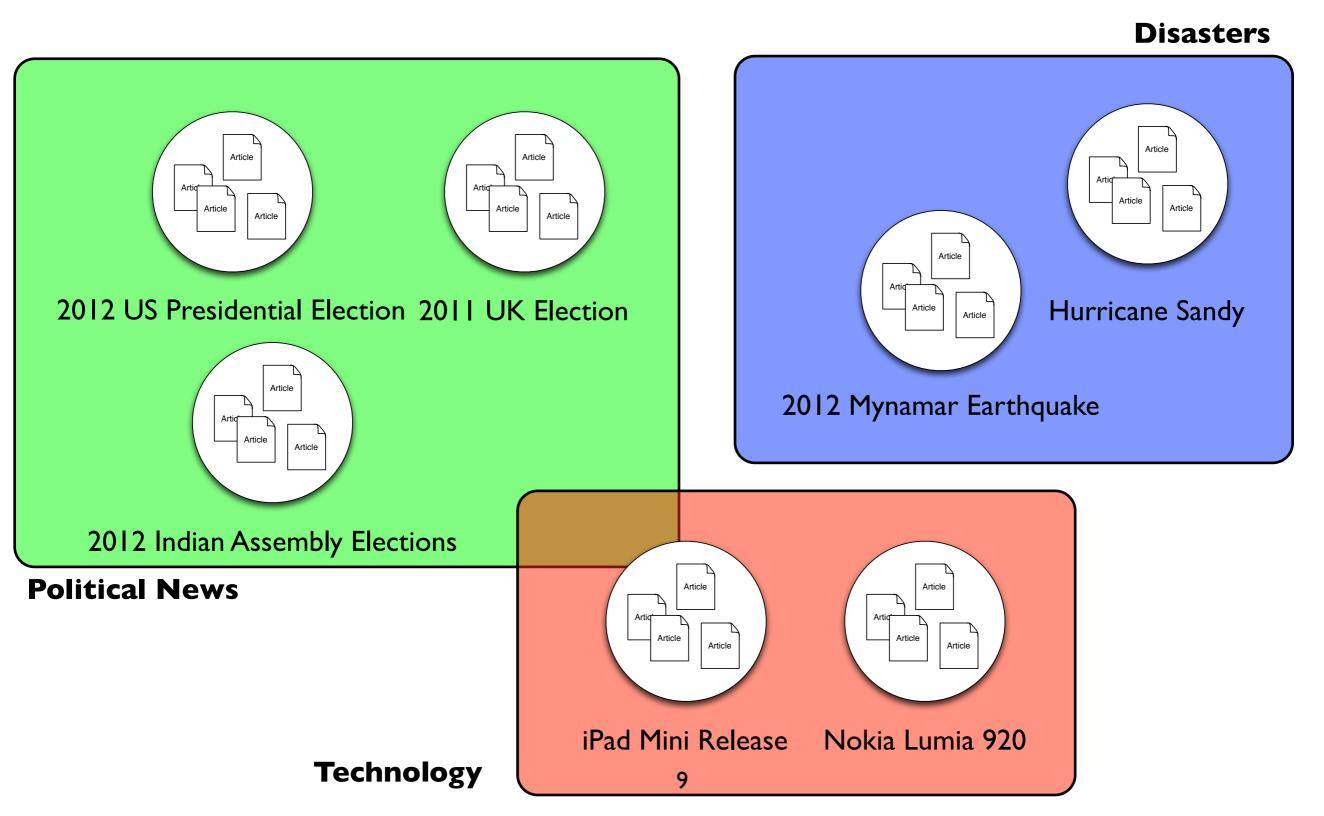


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Nokia Lumia 920



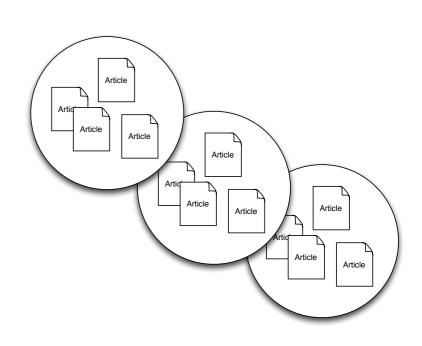


Data Set

- Made use of dataset from guided summarization task of TAC 2010/2011
- Sets of articles to be summarized grouped into topics
- Topics are grouped into categories

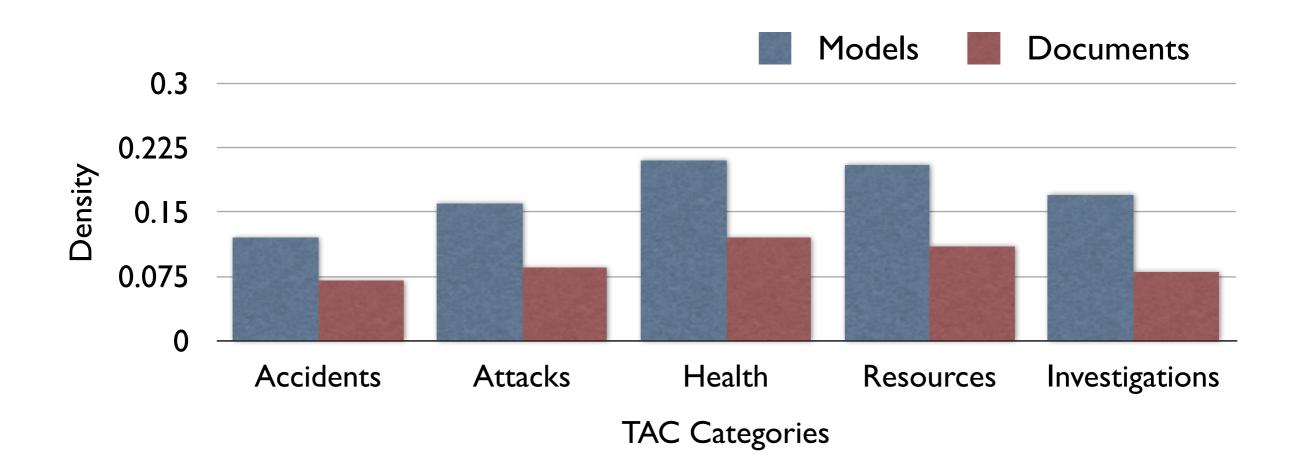
TAC Categories

- Accidents and Natural Disasters
- Attacks
- Health and Safety
- Endangered Resources
- Investigations and Trials



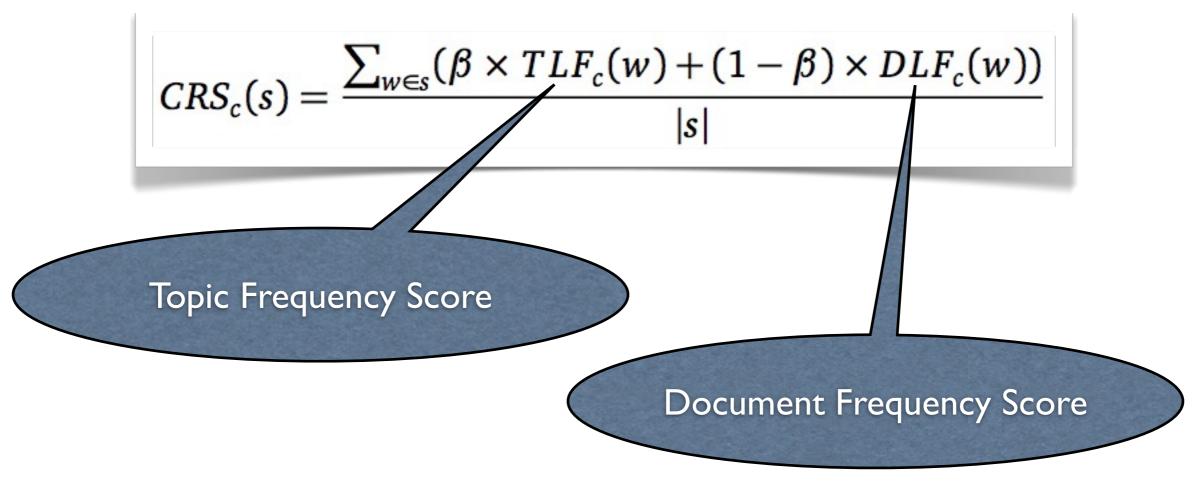
Justifying Our Intuition

 Words with high log-likelihood ratios are used more frequently inside model summaries



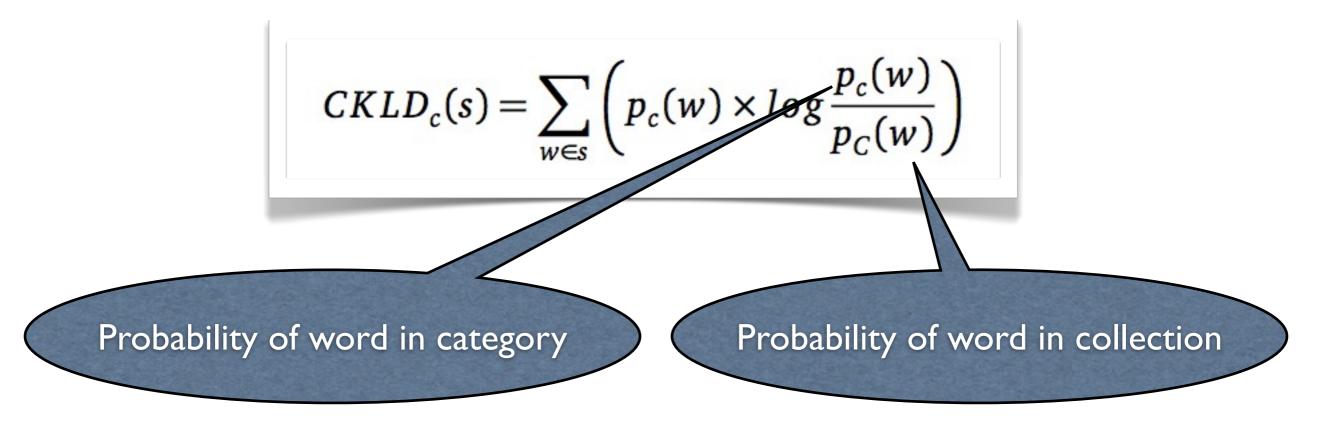
Category Relevance (CRS)

Importance of a word with respect to a category



Category KLD (CKLD)

 Measure of divergence of probability distribution of a word in a category, with respect to the whole collection



Top Words

CRS	CKLD
official	crane
people	bridge
report	construction
news	java
accident	people

For articles on "Accidents"

Summarization Results

 Category-specific features give significant performance enhancement

System	ROUGE-2	ROUGE-4
Generic + CRS + CKLD	0.13796	0.16808
Generic + CRS	0.13702	0.16788
Generic + CKLD	0.13525	0.16649
CLASSY	0.12780	0.15812
POLYCOM	0.12269	0.15974

^{*} ROUGE measures on TAC 2011 testing set

Less than Perfect Categorization

- In more realistic scenarios, perfect categorization is hard to come by
- Would our proposed CSI still be effective if we don't have perfect categorization?

Categorization Experiments

Perfect clustering information is not essential

System	ROUGE-2	p-value
* Perfect Categories	0.13796	-
EM Clustering	0.13647	0.154
X-Means Clustering	0.13546	0.117
K-Means Clustering	0.13569	0.365

Scores reported for 5 clusters

Analysis

Baseline

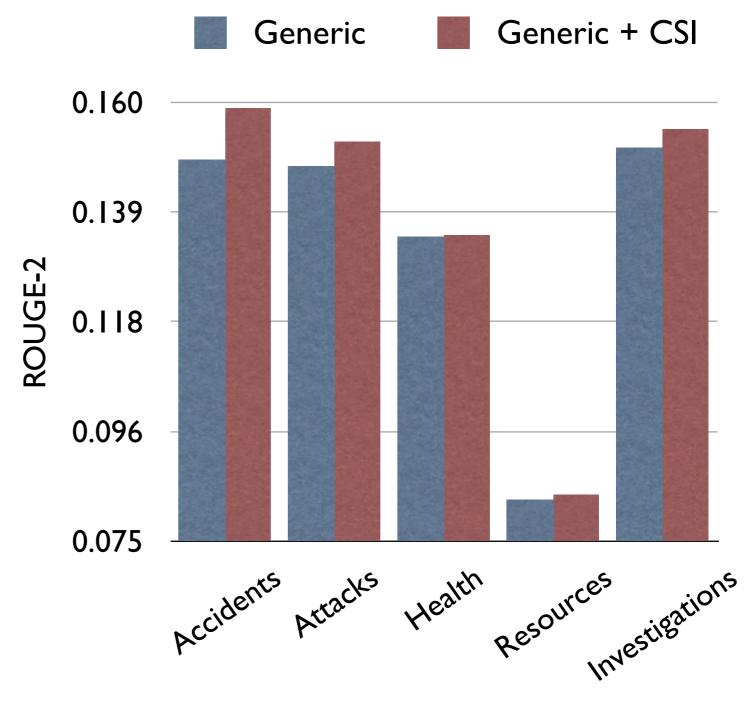
- The death toll could rise as thousands are still buried in debris and many are reported missing.
- Therefore, the relevant sectors and personnel should pay attention to disaster prevention.

Generic + CRS + CKLD

- + Chinese authorities did not detect any warning signs ahead of Monday's earthquake that killed more than 8,600 people.
- + Xinhua said 8,533 people had died in Sichuan alone, citing the local government.

Varying Difficulties

- Some categories are easier to improve on
- Categoryspecific features help more with the WHO, WHAT, WHEN, WHERE



Round Up

- Category-specific information (CSI) can help multi-document summarization
 - Effective in improving content selection
 - Robust and is useful even when less than ideal categorization results are available

Thank you!

Questions?

You can download the system here: http://wing.comp.nus.edu.sg/downloads/swing/