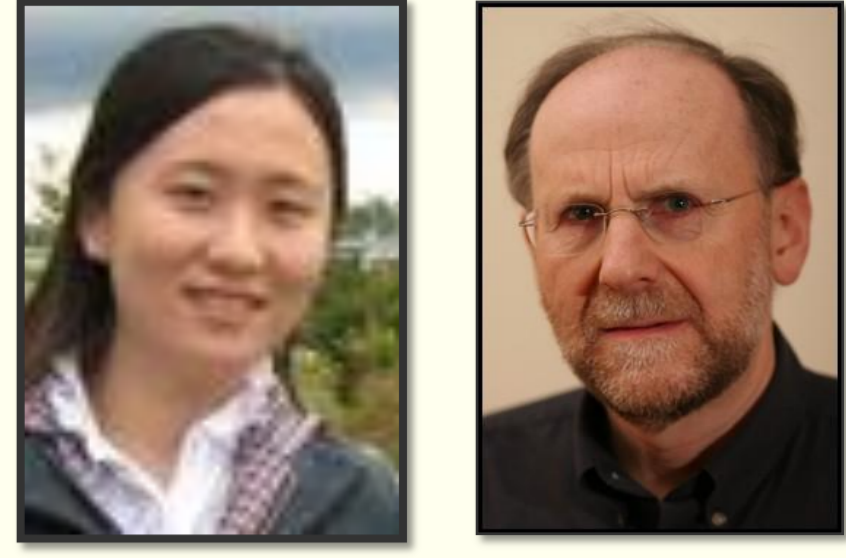


# Extending the Entity-based Coherence Model with Multiple Ranks

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## 1. Objective

- Extend Barzilay and Lapata (2008)'s **entity-based coherence model** by learning from more fine-grained coherence preferences.
- Assign **multiple ranks** to a set of permutations (not just the original **pairwise rankings**).
- Study the effect of the permutations used in training, and the effect of the coreference component used in entity extraction.
- Evaluate with **sentence ordering** and **summary coherence rating**, compared to B&L's original model.

## 2. Entity-based Local Coherence

### Entity-grid representation for a document $d$

	$e_1$	$e_2$	$e_3$	$e_4$	
$s_1$	-	X	X	-	The syntactic role of entity $e_i$ in sentence $s_j$ : S: subject; O: object; X: others; -: not present
$s_2$	S	O	-	-	
$s_3$	X	X	X	X	

Each row represents a sentence.

Each column represents an entity  $e_i$  in  $d$ .

Sequence of roles gives local transitions for each entity.

- Entity extraction options:** Coreference resolution or not.
- Represent document as vector  $\Phi(d) = (p_1(d), p_2(d), \dots, p_m(d))$ .  
 $p_t$ : proportions in text of each possible sequence  $t$ .

## 3. Experimental Setup

### In the original model

#### Sentence ordering task

- Scramble sentences of text to produce random permutations.
- Permutations are considered to be less coherent than their source document.
- Training and testing on the **pairwise** preferences between an original document and its permutations.

#### Summary coherence rating task

- System-generated and human-composed summaries, rated by human judges for coherence.
- Training and testing on the **pairwise** preferences between summaries generated from the same input cluster.

### In our extension

#### Sentence ordering task

- Assign **multiple ranks** to permutations, indicating the **dissimilarity** between their sentence orders and the original.
- Also train on the pairwise preferences among the permutations.
- Experiment with two sets of permutations:  $PS_{BL}$  (evenly distributed) and  $PS_M$  (favoring swapping near sentences).

#### Summary coherence rating task

- Automatically assign** scores to system-generated summaries, by computing the **dissimilarity** between their (rough) sentence orders and the one in the reference summary.

## 4. Multiple Ranks Assignment

### Dissimilarity metrics

Reference ordering:  $\sigma = (1, 2, \dots, N)$ ; test ordering:  $\pi = (o_1, o_2, \dots, o_N)$ .

- Kendall's  $\tau$**  (Lapata, 2006): measures the disagreement between  $\pi$  and  $\sigma$  in terms of  $m$ , the number of swaps of adjacent sentences to convert  $\pi$  into  $\sigma$ .
- Average continuity (AC)** (Bollegala et al., 2006): estimates the quality of  $\sigma$  by the number of correctly arranged continuous sentences, compared to  $\pi$ .
- Edit distance (ED)**: the minimum number of edits (insertions, deletions, and substitutions) needed to convert  $\pi$  into  $\sigma$ .

### Rank assignment

Two options for assigning ranks to the permutations:

- Raw:** rank the permutations by their dissimilarity scores.
- Stratified:**  $C$  (3 to 6) ranks are assigned to the permutations according to their raw dissimilarity scores.

## 5. Data

### Sentence ordering

- Two datasets:**  
Earthquakes: pronominal realization of entities.  
Accidents: string repetition of entities.
- Training and testing:** each with 100 texts and up to 20 permutations.

### Summary coherence rating

- Dataset:** MUC 2003 summaries (16 clusters, 5 systems).
- Training:** 144 pairwise rankings.
- Testing:**  
Same: 80 pairwise rankings among summaries within the same cluster.  
Full: 1520 pairwise rankings.

## 6. Results

### Sentence ordering

**Results:** We show the model configurations with the best accuracies.

Perms	Earthquakes				Accidents			
	Metric	C	F&H	B&L	Metric	C	F&H	B&L
Condition: full coreference resolution with oracular information								
$PS_{BL}$	ED	3	86.8	85.3	AC	3	83.3	83.2
$PS_M$	ED	$N$	87.9*	85.3	ED	4	86.3*	81.7
Condition: full coreference resolution without oracular information								
$PS_{BL}$	ED	4	77.4*	71.7	AC	3	74.5	73.8
$PS_M$	$\tau$	3	55.9	49.2	ED	5	52.3	53.2
Condition: no coreference resolution								
$PS_{BL}$	$\tau$	4	82.8	83.7	AC	3	84.2**	80.1
$PS_M$	ED	5	86.7**	82.6	AC	$N$	86.6**	77.5

Significantly better than B&L: \* ( $p < .05$ ), \*\* ( $p < .01$ ).

$C=N$ : using raw option for rank assignment.

➤ Multiple ranking is effective in improving accuracies, especially when trained on the more realistic permutations  $PS_M$ .

➤ Different influence on two datasets when trained on  $PS_{BL}$ .

➤ This condition is not a good option when trained on  $PS_M$ .

➤ Coreference resolution is crucial to *Earthquakes*.

➤ Consistently outperforms B&L's model by a large margin.

### Summary coherence rating

**Rough sentence orders:** via simple sentence alignment.

Entities	Metric	Same	Full
Coreference resolution	AC	82.5	72.6*
	ED	81.3	73.0**
	B&L	78.8	70.9
No coreference resolution	AC	76.3	72.0
	ED	78.8	71.7
	B&L	80.8	72.3

Unsupervised score assignment is competitive with B&L's model, which requires human annotations.

### References

- Regina Barzilay and Mirella Lapata. 2008. Modeling local coherence: an entity-based approach. *Computational Linguistics*, 34(1):1-34.
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