

# Modeling Local Coherence: An Entity-Based Approach



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# Abstract

This article proposes a **novel framework** for representing and measuring **local coherence**. Central to this approach is the **entity-grid representation of discourse**, which captures patterns of entity distribution in a text. The algorithm introduced in the article automatically abstracts a text into **a set of entity transition sequences** and records distributional, syntactic, and referential information about discourse entities. We re-conceptualize coherence assessment as a **learning task** and show that our entity-based representation is well-suited for ranking-based generation and text classification tasks. Using the proposed representation, we achieve **good performance on text ordering, summary coherence evaluation, and readability assessment**.

# Overview

- Introduction
- Previous work
- The Model
- Experiment 1, 2, and 3
- Wrapping it up

# Introduction

Local Coherence: text relatedness at the level of sentence-to-sentence transitions.

- Important for global coherence
- Important for automatically produced text
- Difficult to do automatically

# Introduction

“The distribution of entities in locally coherent texts exhibits certain regularities.”

The algorithm converts text into a set of entity transition sequences.

Incorporates information that is:

- Distributional
- Syntactic
- Referential

# Introduction

Entirely automatic; without manual annotation or a predefined knowledge base.

Tested in three ways:

- Text ordering
- Automatic evaluation of summary coherence
- Readability assessment

# Related Work

Entity-Based Approaches to Local Coherence:

- Linguistic Modeling (cf. Centering Theory)
- Computational Modeling

Ranking approaches to NLG

- Manual specifications



# Entity-Grids

“A local entity transition is a sequence  $\{S, O, X, -\}_n$  that represents entity occurrences and their syntactic roles in  $n$  adjacent sentences.”

Each transition has a probability: frequency of occurrence over total number of transitions of that length.

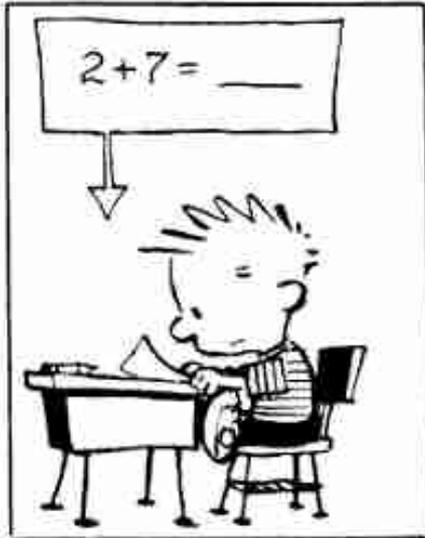
# Entity-Grids

	Department	Trial	Microsoft	Evidence	Competitors	Markets	Products	Brands	Case	Netscape	Software	Tactics	Government	Suit	Earnings	
1	s	O	s	x	O	-	-	-	-	-	-	-	-	-	-	1
2	-	-	O	-	-	x	s	O	-	-	-	-	-	-	-	2
3	-	-	s	O	-	-	-	-	s	O	O	-	-	-	-	3
4	-	-	s	-	-	-	-	-	-	-	-	s	-	-	-	4
5	-	-	-	-	-	-	-	-	-	-	-	-	s	O	-	5
6	-	x	s	-	-	-	-	-	-	-	-	-	-	-	O	6

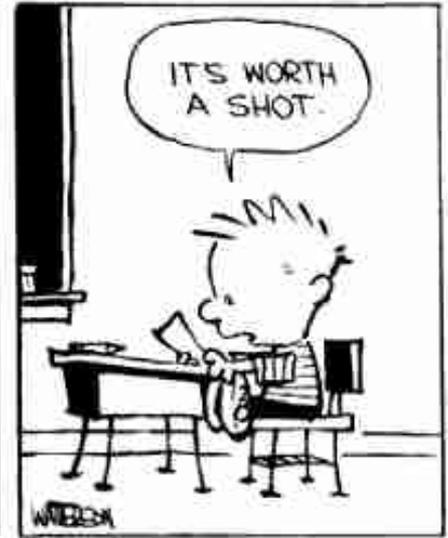
Example:

Probability of [O-]:  $7 / 75 = 0.093$

# Exercise!



I cannot answer this question, as it is against my religious principles.



# Exercise 1 Answers

	Curdie	Peter	Father	Mother	Cottage	Mountain	Thing	Me	Thing	terrors
1	S	X	-	-	-	-	-	-	-	-
2	S	-	X	X	X	X	-	-	-	-
3	S	-	X	-	-	X	-	-	-	-
4	-	-	-	-	-	S	-	-	O	-
5	-	-	-	-	-	S	-	X	-	-

What is the probability of [X-]? 0.1 ( $4/(10^4)$ )

What is the probability of [-xx]? 0.06 ( $2/(10^3)$ )

# Grid Construction

- What linguistic knowledge is useful?
- Considerations of parameters:
  - How important a parameter is
  - Accuracy of automatic computation
  - Size of the resulting feature space

# Grid Computation

- Entity Extraction – Coreference +/-
  - Use Ng & Cardie (2002) coreference resolution system
- Grammatical Function – Syntax +/-
  - S,O,X,-. Using Collins (1997) parser.
- Saliency – Salient +/-
  - Frequency

# Learning

- Given the feature vector representation, coherence assessment is an ML problem.
- Ranking alternative renderings works as a feature for learning

# Experiments

- Exp. 1: Information ordering
- Exp. 2: Summary coherence rating
- Exp. 3: Readability assessment

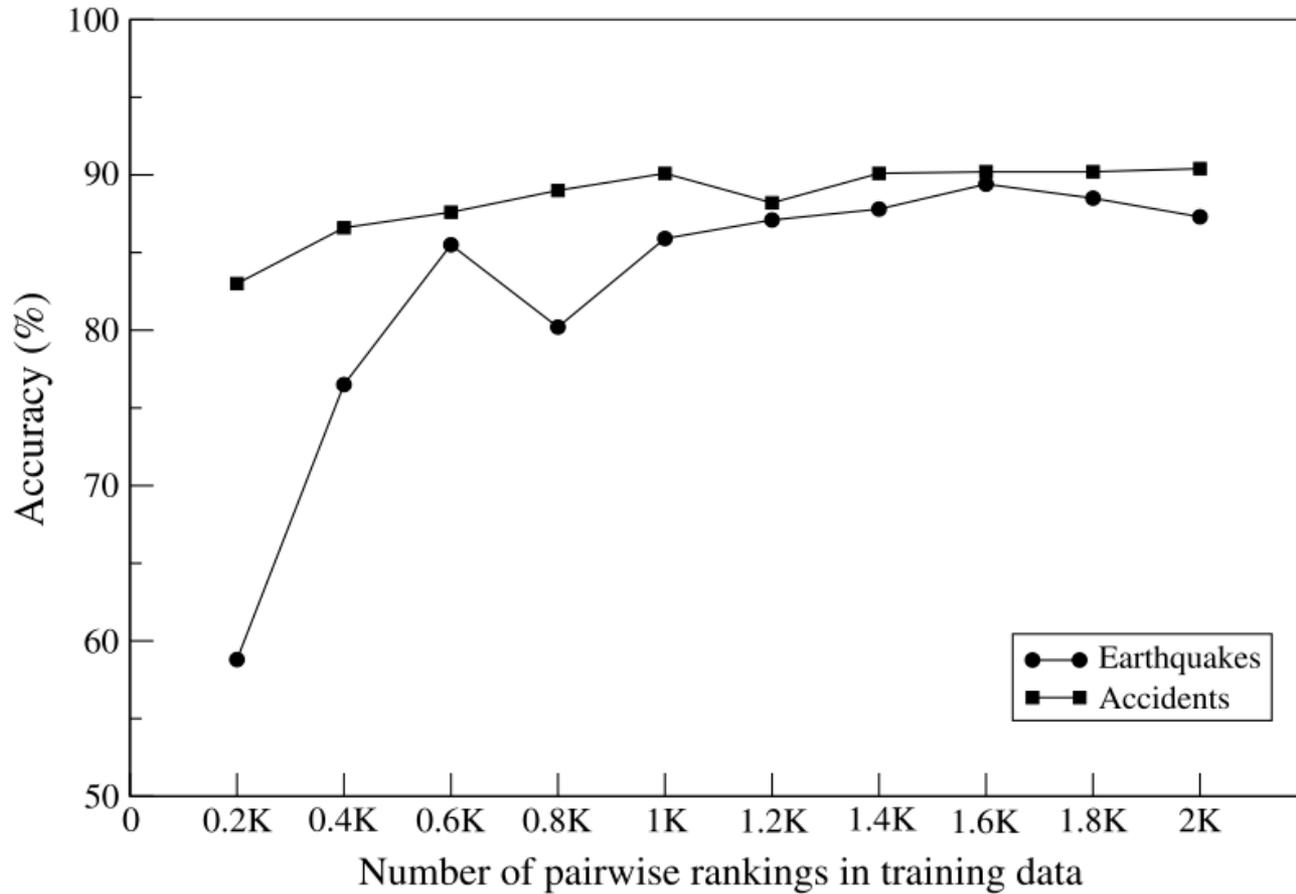
# Exp. 1: Sentence Ordering

- Task: Generate random permutations of a test document and measure how often a permutation is ranked higher than the original.
- Data: 100 source articles, up to 20 random permutations for training, same for testing
- Original document held to be coherent; corpus includes pairwise rankings that comprise the original document and one permutations.

# Exp. 1: Results

Model	Earthquakes	Accidents
<b>Coreference+Syntax+Saliency+</b>	<b>87.2</b>	<b>90.4</b>
Coreference+Syntax+Saliency-	88.3	90.1
Coreference+Syntax-Saliency+	86.6	88.4**
Coreference-Syntax+Saliency+	83.0**	89.9
Coreference+Syntax-Saliency-	86.1	89.2
Coreference-Syntax+Saliency-	82.3**	88.6*
Coreference-Syntax-Saliency+	83.0**	86.5**
Coreference-Syntax-Saliency-	81.4**	86.0**
HMM-based Content Models	88.0	75.8**
Latent Semantic Analysis	81.0**	87.3**

# Exp. 1: Learning Curve



# Exp. 1: Learning Domains

## Coreference+Syntax+Saliency

Train \ Test	Earthquakes	Accidents
Earthquakes	<b>87.3</b>	67.0**
Accidents	69.7**	<b>90.4</b>
EarthAccid	86.7	88.5*

## HMM-Based Content Models

Train \ Test	Earthquakes	Accidents
Earthquakes	<b>88.0</b>	31.7**
Accidents	60.3**	<b>75.8</b>

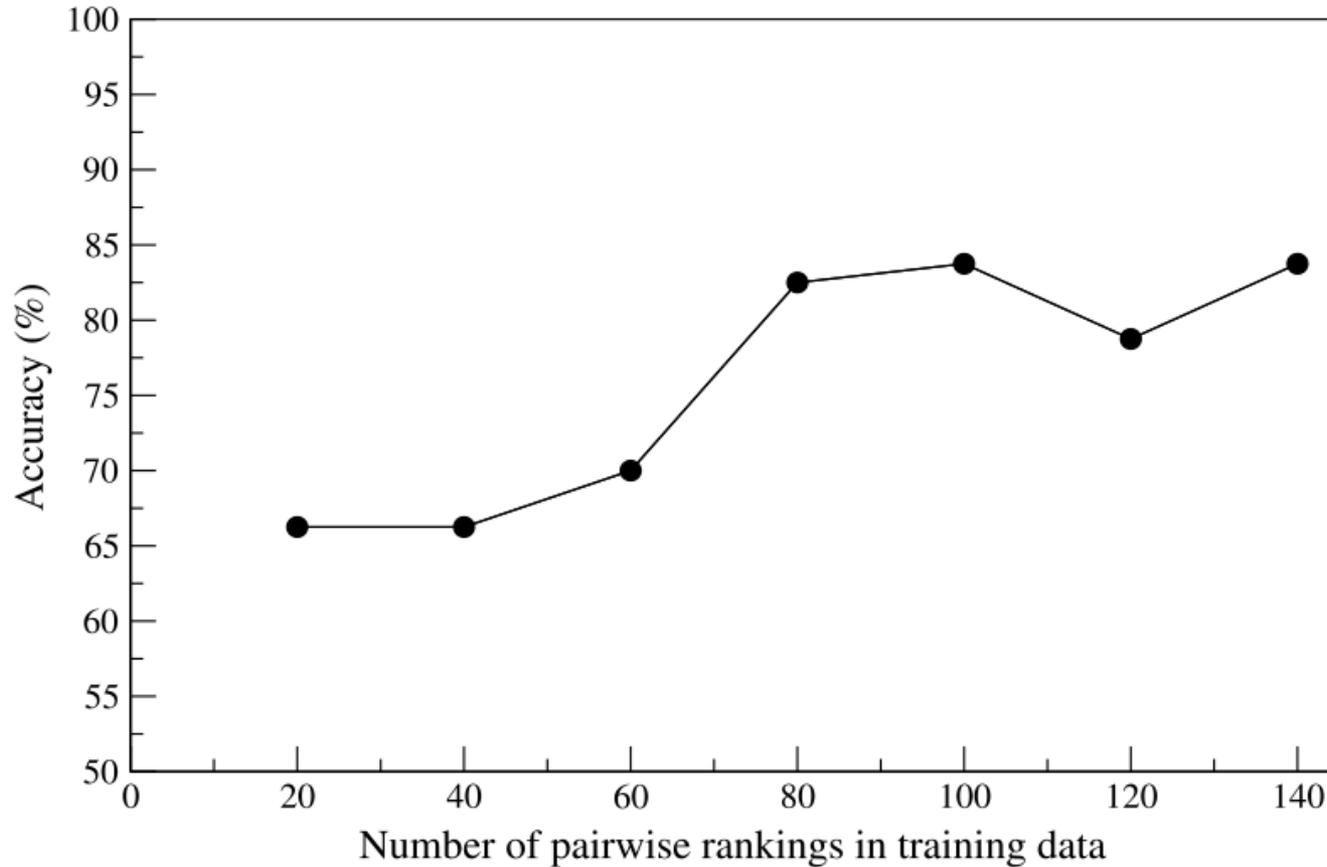
# Exp. 2: Summary Coherence Rating

- Test: compare model-induced rankings against rankings elicited by human judges
- Data: Training data was 144 student-annotated summaries. Test data was 80 pairwise rankings.

# Exp. 2: Results

Model	Accuracy
Coreference+Syntax+Saliency+	80.0
Coreference+Syntax+Saliency-	75.0
Coreference+Syntax-Saliency+	78.8
<b>Coreference-Syntax+Saliency+</b>	<b>83.8</b>
Coreference+Syntax-Saliency-	71.3*
Coreference-Syntax+Saliency-	78.8
Coreference-Syntax-Saliency+	77.5
Coreference-Syntax-Saliency-	73.8*
Latent Semantic Analysis	52.5**

# Exp. 2: Learning Curve



# Exp. 3: Readability Assessment

- Task: embed entity-grids into a system that assesses document readability.
- Readability as a classification task: a learner needs to predict if an article is easy or difficulty to read.
- Data: corpus of Encyclopedia articles

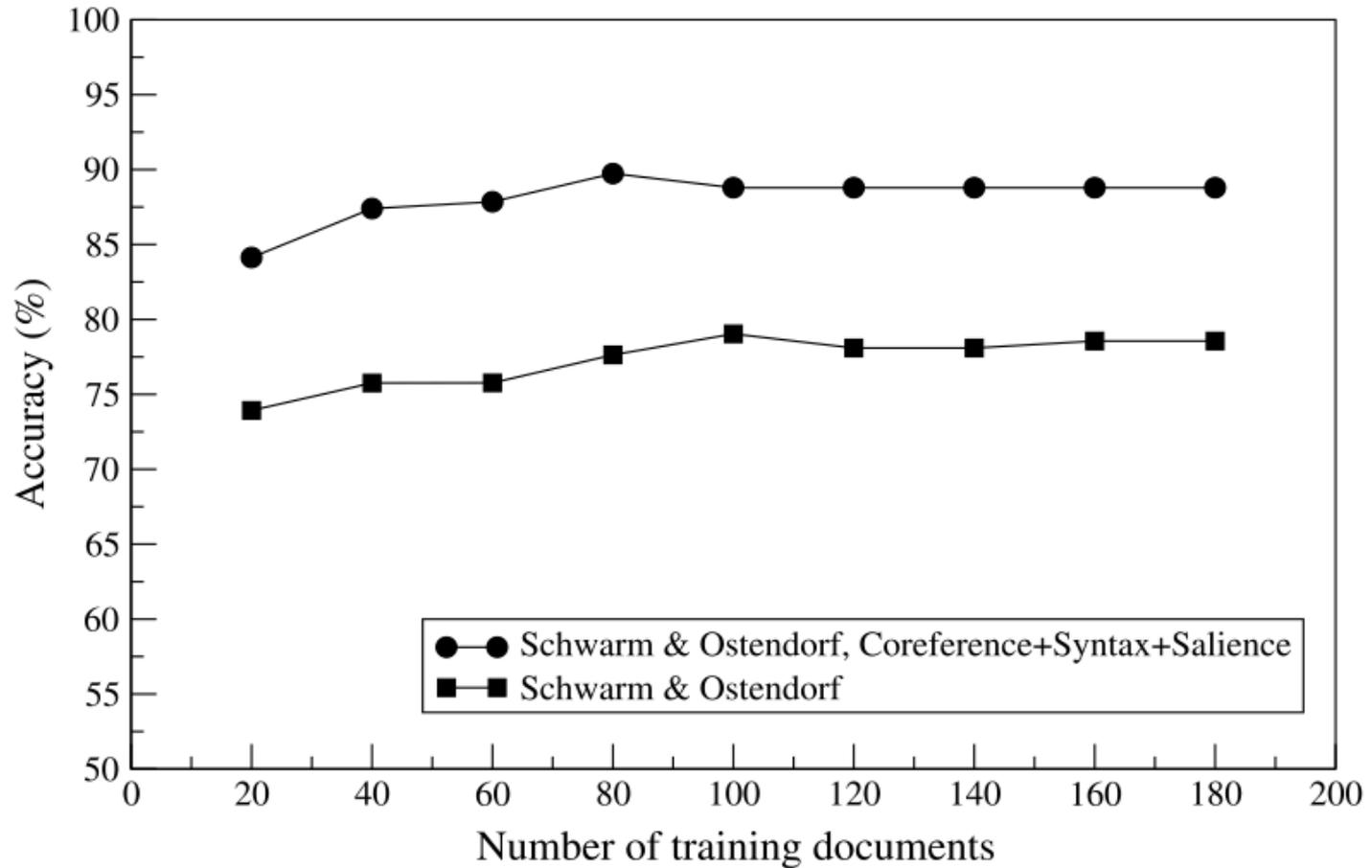
# Exp. 3: Versions

- 1. Use Schwarm and Ostendorf (2005), who use syntactic, semantic, and combination features.
- 2. Enrich Schwarm and Ostendorf with coherence-based features

# Exp. 3: Results

Model	Accuracy
Schwarm & Ostendorf	78.56
<b>Schwarm &amp; Ostendorf, Coreference+Syntax+Saliency+</b>	<b>88.79*</b>
Schwarm & Ostendorf, Coreference-Syntax+Saliency+	79.49
Schwarm & Ostendorf, Latent Semantic Analysis	78.56
Coreference+Syntax+Saliency+	50.90**
Coreference-Syntax+Saliency+	49.55**
Latent Semantic Analysis	48.58**

# Exp. 3: Learning



# Discussion

- Empirical validation of the importance of salience and syntactic information for coherence-based models
- Sacrificed linguistic faithfulness by being theory-agnostic
- ?: What about lexico-semantic knowledge?
- ?: Combined with global discourse analysis?
- ?: other languages?

# Reference

- Barzilay and Lapata (2008). Modeling Local Coherence: An Entity-Based Approach. ACL.

Thanks!

