

## Correction Detection

Compare a sentence with its revision, to detect corrections of individual mistakes

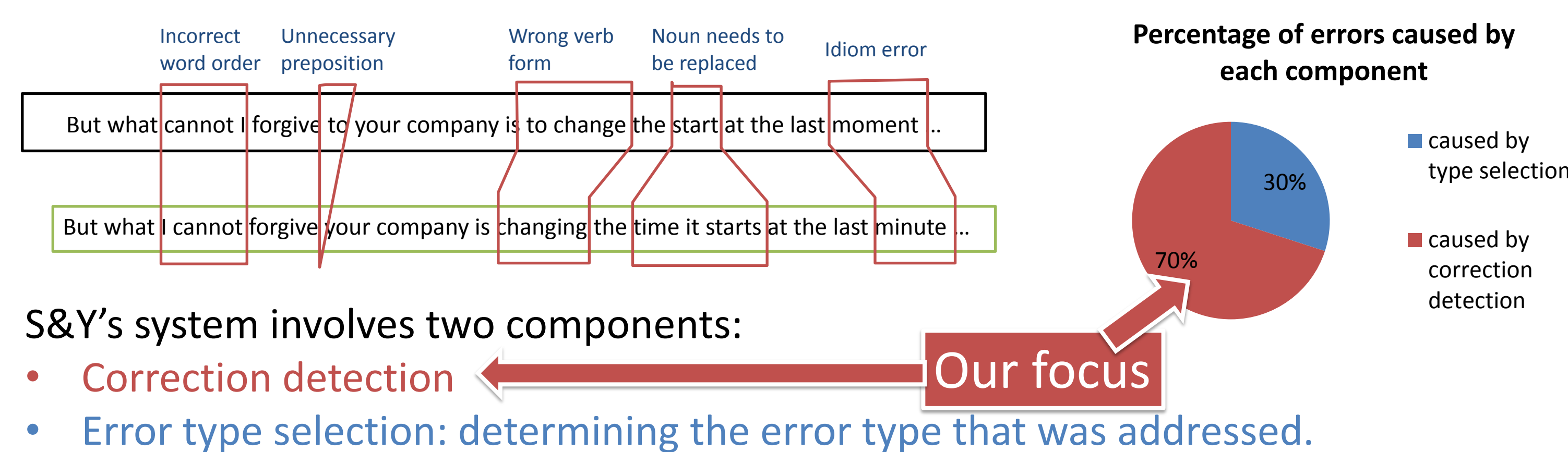
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| Quote | +

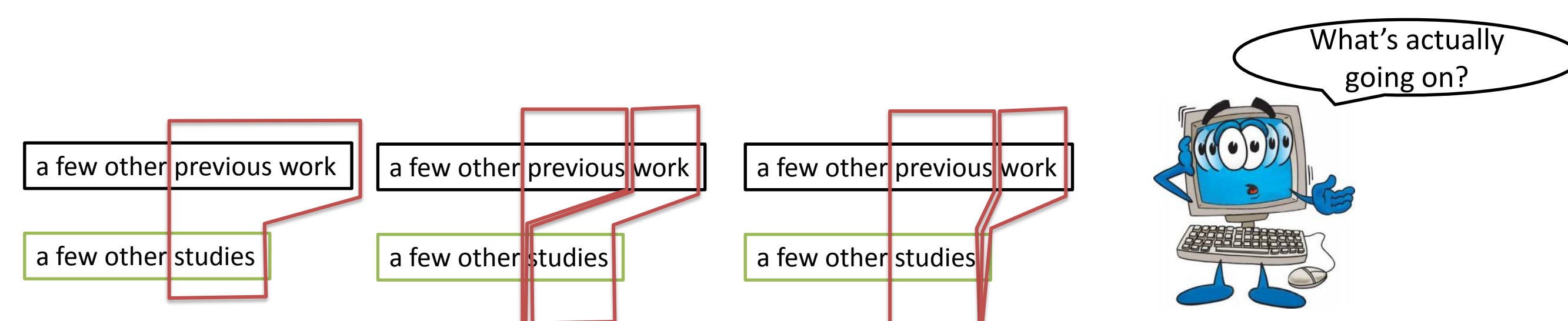
As a result, when the earthquake ~~that~~ happened in Lushan, ~~Jiangxi province~~ this year, some measures, such as temporal traffic management and emergency stocks mobilization, became more efficient, and more effective.

Swanson and Yamagil (2012): comparing two versions automatically

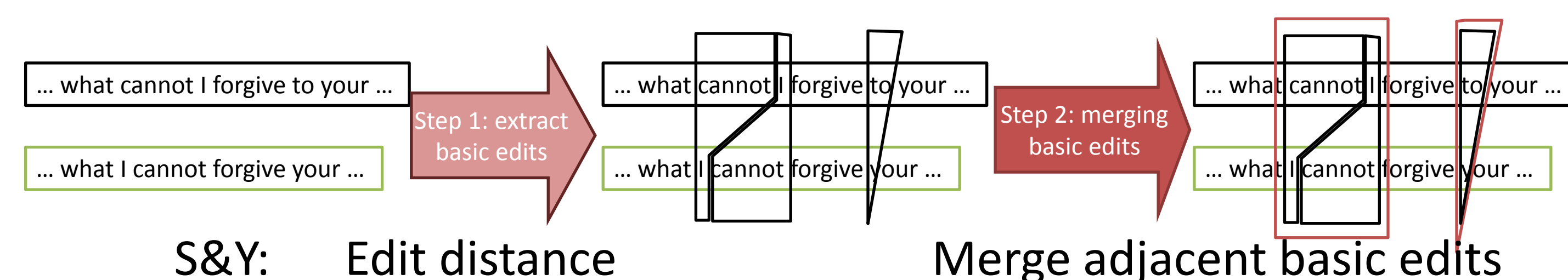


The challenge of correction detection – Ambiguity

There are often multiple ways to interpret one revision



S&Y's uses a deterministic algorithm for correction detection



The heuristics were developed from one single data collection

- Lacks the flexibility to adapt to a specific context
- May not work as well for other guidelines

S&Y's algorithm makes mistakes when there are ambiguities

**First step's mistakes**

The extracted basic edits might not match our linguistic intuition

Extracted by edit-distance: ... closed because of repairs, ... closed for repairs

Edits that makes more linguistic sense: It was closed because of repairs, It was closed for repairs

**Second step's mistakes**

Non-adjacent but should merge: ... existed always..., ... take it away..., ... have not..., ... always existed..., ... remove it..., ... don't have...

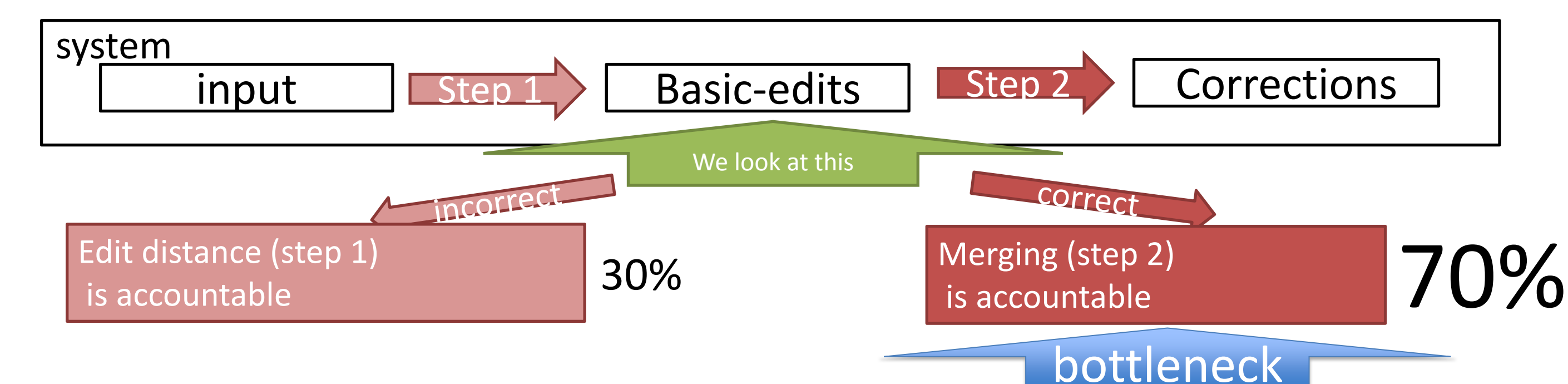
Adjacent but should not merge: He pick on a book., You can notice every students., He picks up a book., You may inform all students.

## Our contributions

- We show empirically that a major challenge in correction detection is to determine if two edits address the same error – do we need to merge them
- We have developed a merging model that reduces mis-detections by 1/3
- We show the merging model is generalizable – we have conducted experiments across multiple corpora

## How to improve correction detection? – An error analysis of S&Y's system

Which step is accountable for more errors?



This highlights the difference between correction detection and previous work

- Previous work focus on alignment -- step 1
  - Previous tasks also involve identify corresponding phrases between two sentences (Koehn et al., 2003, Cohn et al. 2008)
  - A bigger concern in previous work is to guarantee the extracted phrase pairs are indeed translations or paraphrases (Snover et al., 2009; Heilman and Smith, 2010)
- Correction detection concerns the granularity of the extracted phrase pairs – focuses on step 2

## We propose: A classifier to determine if consecutive basic edits address the same mistake

Intuition: certain patterns indicate whether two edits address the same writing mistake

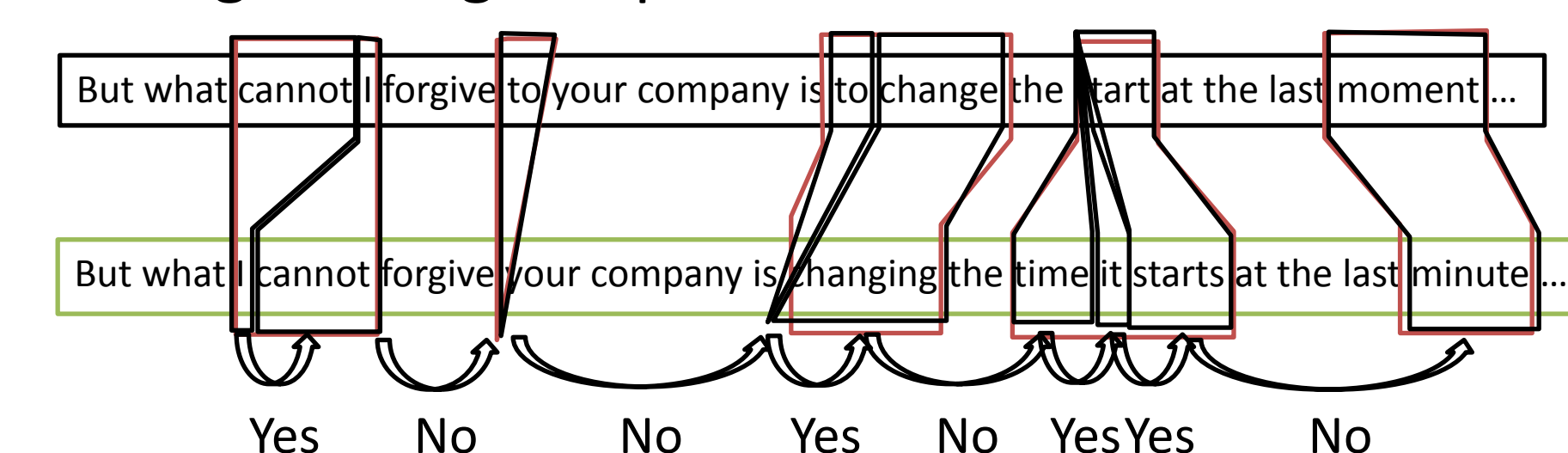
We encode these patterns into a classifier

- Input: features extracted from two consecutive basic edits
- Output: whether we should merge them

**Classifier with Contextual Features**

Does merging the two basic edits match the pattern for a single correction?	gap-between-edits	Yes
	Tense-change	
	Word-order-error	
	Same-word-set	
	Revised-to	
Does one basic-edit address one single error?	Editdistance=1	no
	Orig-word-not-in-dict	
	Same-POS	
	Both-prepositions	

Constructing training samples



## Experiments

Task: detect corrections from revisions, on different corpora

Evaluation Metric: Precision/Recall/ $F_1$ -measure

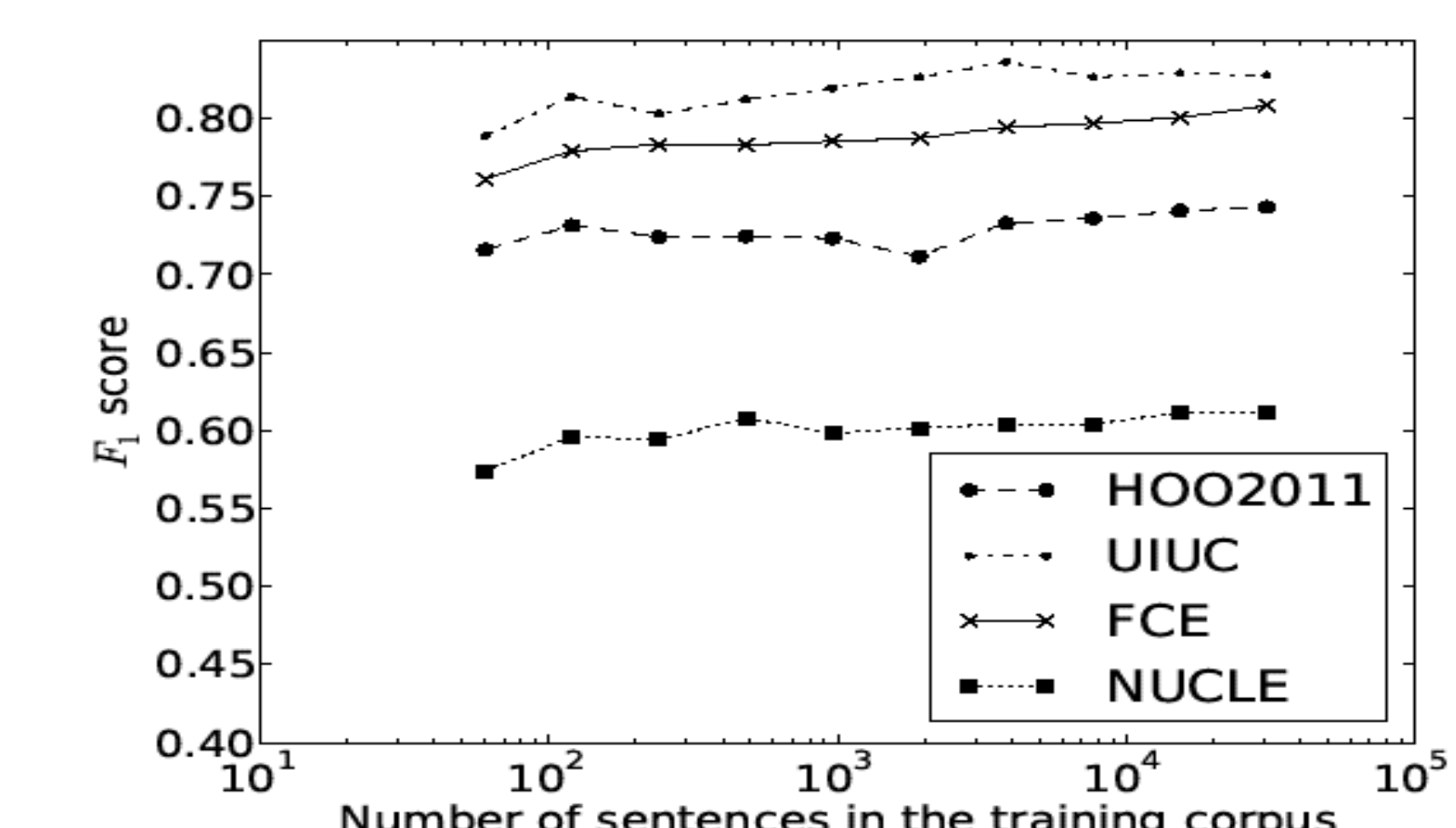
Did additional contextual information help?

Method	Corpus	Correction Detection F1	Overall F-score
S&Y	FCE	70.40%	57.10%
MaxEntMerger	FCE	<b>80.96%</b>	<b>66.36%</b>
S&Y	NUCLE	61.18%	39.32%
MaxEntMerger	NUCLE	<b>63.88%</b>	<b>41.00%</b>
S&Y	UIUC	76.57%	65.08%
MaxEntMerger	UIUC	<b>82.81%</b>	<b>70.55%</b>
S&Y	HOO2011	68.73%	50.95%
MaxEntMerger	HOO2011	<b>75.71%</b>	<b>56.14%</b>

How did our method generalize over revisions from different sources?

Testing \ Training	FCE	NUCLE	UIUC	HOO2011
S&Y	70.44%	61.18%	76.57%	68.73%
FCE	<b>80.96%</b>	61.26%	<b>83.07%</b>	75.43%
NUCLE	74.53%	<b>63.88%</b>	78.57%	74.73%
UIUC	77.25%	58.21%	<b>82.81%</b>	70.83%
HOO2011	71.94%	54.99%	71.19%	<b>75.71%</b>

- FCE corpus is a comparably good resource for training correction detection models.
- Big data size benefits model training



## Conclusions

The merging step accounts for 70% errors in correction detection

We proposed a merging model:

- Reduces 1/3 errors in correction
- Leads to significant overall system performance
- Generalizes over corpora