

# Adapting Word Prediction to Subject Matter without Topic-labeled Data

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## Problem

- alternative communication, slow communication rate



- word prediction speeds up communication rate
- evaluation: keystroke savings

$$KS = \frac{keys_{normal} - keys_{prediction}}{keys_{normal}} \times 100\%$$

**How can we improve keystroke savings?**

## Why topic adaptation?

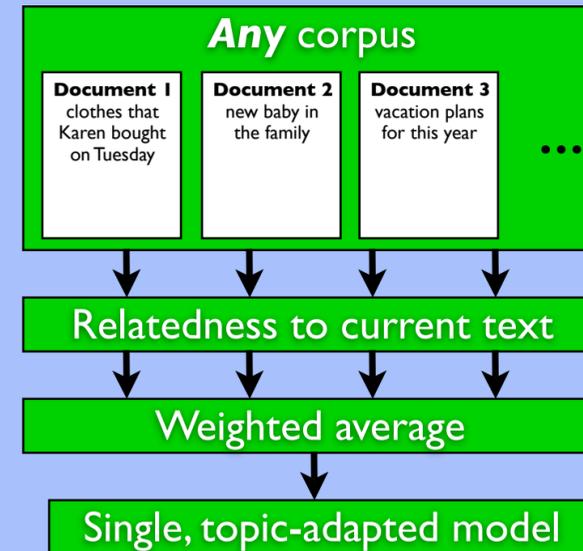
- topic can substantially affect word choice:

how long would it be in the

world  
 winter  
 city  
 area  
 United  
 future  
 mountains

- existing devices tend to use word prediction more for **fringe words**, which tend to be topical words

## Current work - unlabeled topics



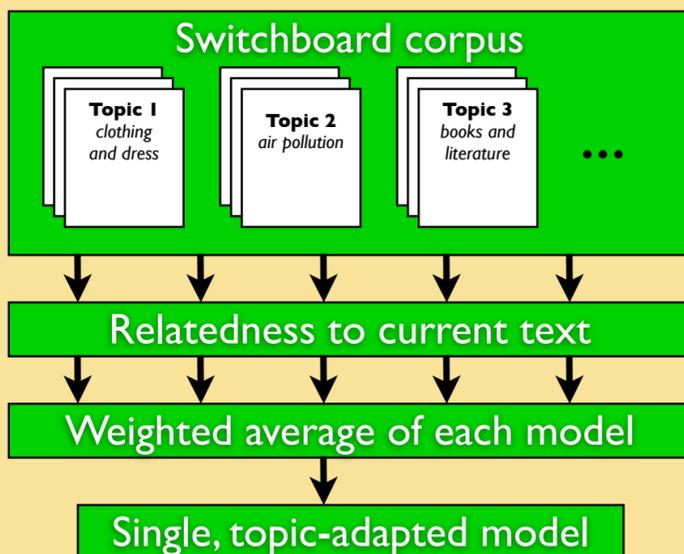
- compare each **document** to current conversation
- weight based on similarity of keyword usage
- take a weighted average of all the document models
- each document is treated as if it were a very specific topic
- tested using a mixture of corpora for training data (below)

Testing corpus	Switchb. topic	Mixed trigram	Mixed topic
AAC Email	43.53%	52.18%	<b>53.14%</b>
Callhome	49.52%	52.14%	<b>53.39%</b>
Charlotte	50.07%	53.50%	<b>53.92%</b>
SBCSAE	43.90%	47.78%	<b>48.84%</b>
Micase	46.99%	51.46%	<b>53.13%</b>
Switchboard	<b>61.48%</b>	59.80%	61.17%

### Advantages

- works with any kind of corpus
- can add more and more data (even general-purpose data) with little negative effect

## Previous work - labeled topic modeling



Testing corpus	Switchboard trigram	Switchboard topic
AAC Email	43.25%	<b>43.53%</b>
Callhome	49.33%	<b>49.52%</b>
Charlotte	49.64%	<b>50.07%</b>
SBCSAE	43.49%	<b>43.90%</b>
Micase	46.52%	<b>46.99%</b>
Switchboard	60.35%	<b>61.48%</b>

### Advantages compared to basic ngrams

- higher keystroke savings (even when tested on other corpora)
- predictions are appropriate for the overall topic (e.g., cooking)

### Disadvantages compared to basic ngrams

- need training corpus with topic labels (limited to training on Switchboard)

- compare each topic to current conversation
- weight based on similarity of keyword usage
- take a weighted average of all the topic models

## Future work

- can use this method to integrate user text into the language model
  - difficult to perform controlled evaluations
- use the same methodology with a different similarity score
  - e.g., syntactic similarity
- improving the similarity score to provide guarantees that there will be no loss in keystroke savings by adding more training texts