# Word Prediction With Suffix Trie HMM's

Bernhard Bauer < bauerb@in.tum.de > Jul 4, 2008

### The task

- Statistical word prediction
- Online learning

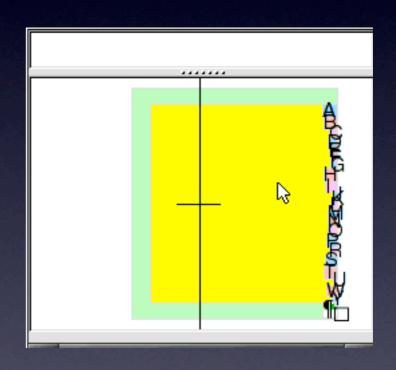
### Applications

- Mobile devices (T9)
- Dasher (<a href="http://">http://</a>

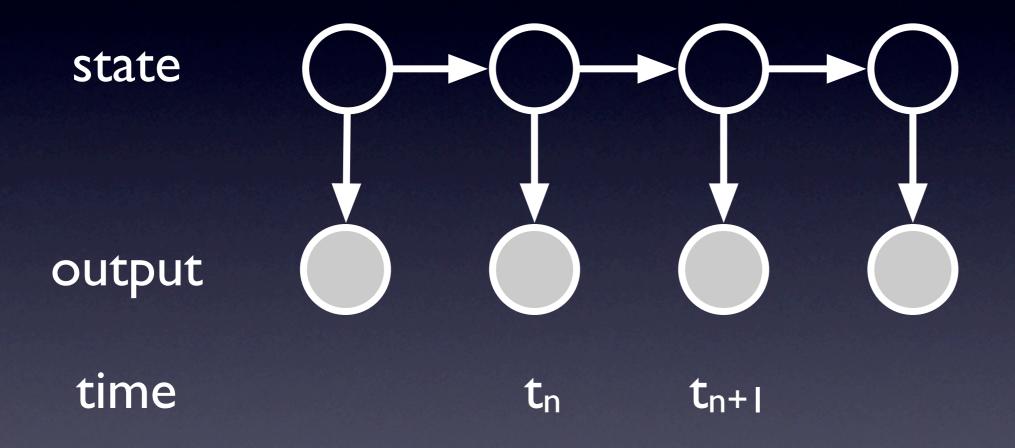
  www.inference.phy.cam.ac.uk/
  dasher/)

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- Dasher (<u>http://</u> <u>www.inference.phy.cam.ac.uk/</u> <u>dasher/</u>)

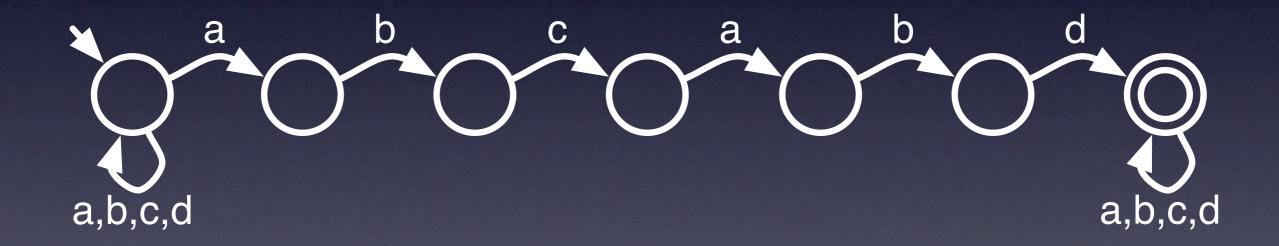


### HM's



### String searching

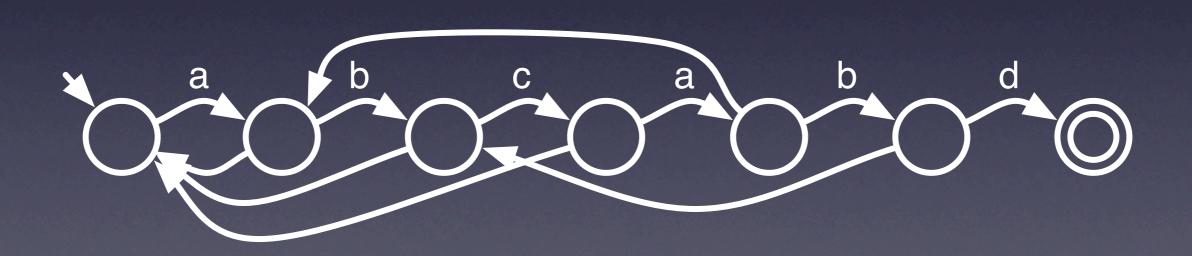
#### Finite automaton



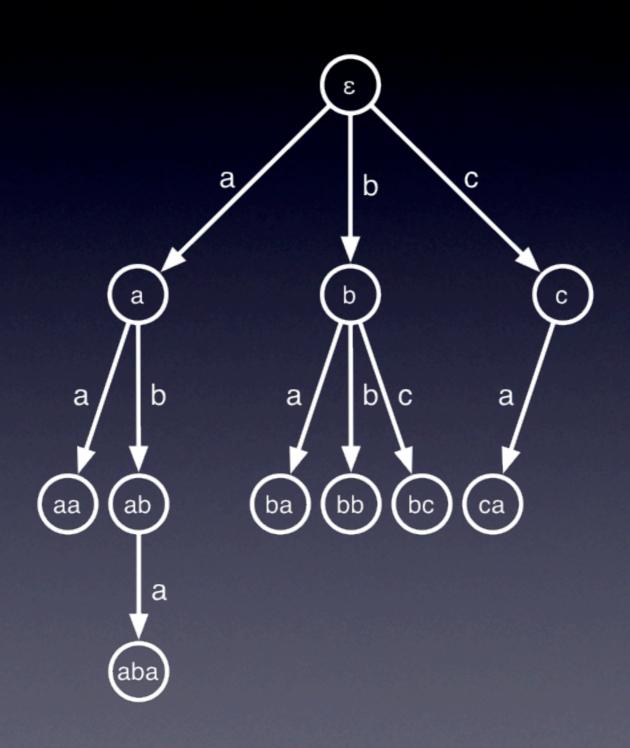
### String searching

Knuth-Morris-Pratt

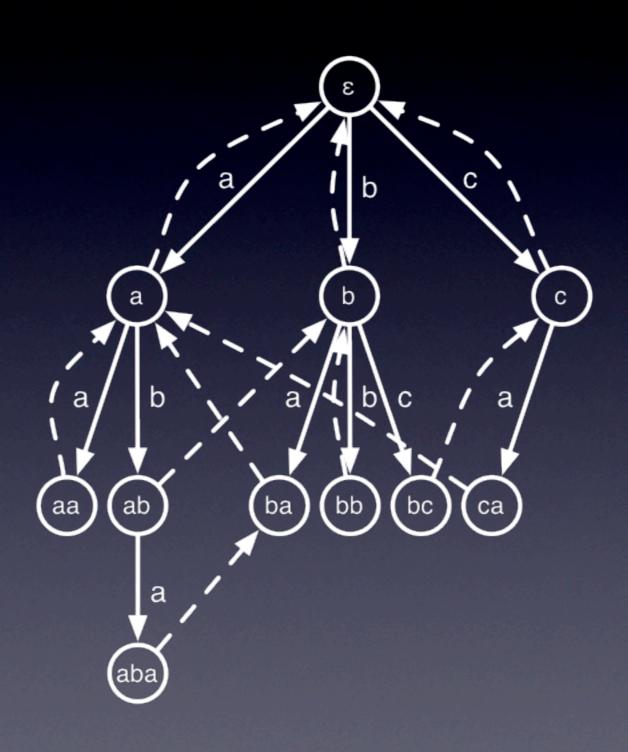
Α	В	С	Α	В	D
-1	0	0	0		2



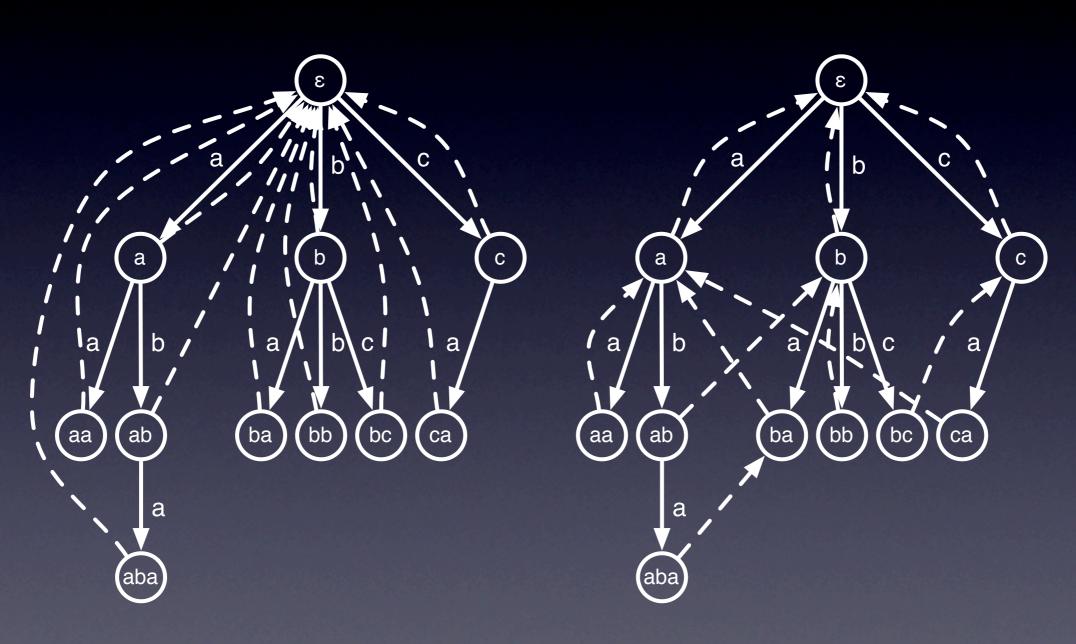
### Suffix trie



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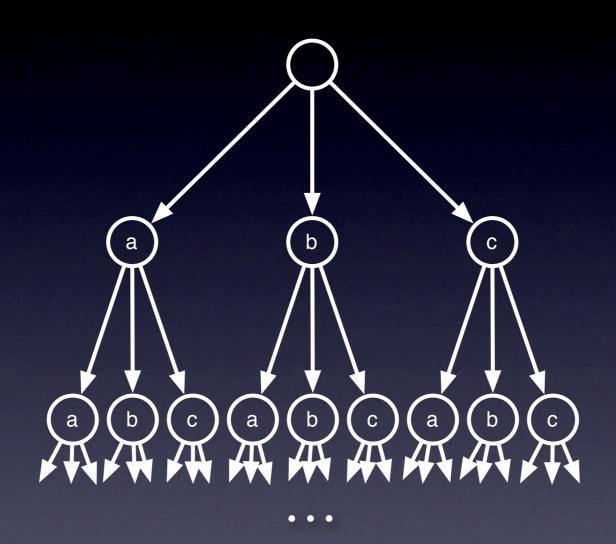
#### State model



Full backlinks

Partial backlinks

### Output model



### Space complexity

- $O(n^2 \cdot m)$ 
  - n: word length
  - *m*: number of words
- Less if words share common prefixes
- Suffix trees can achieve  $O(n \cdot m)$ , but are not applicable

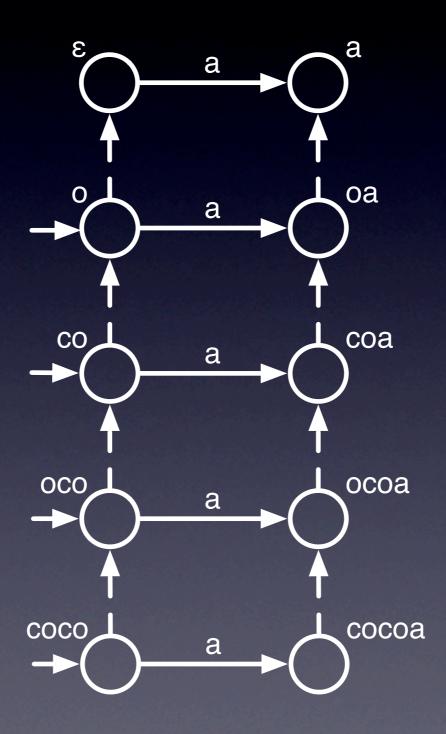


•  $O(n^2 \cdot m)$  potential states, but most of them have probability zero

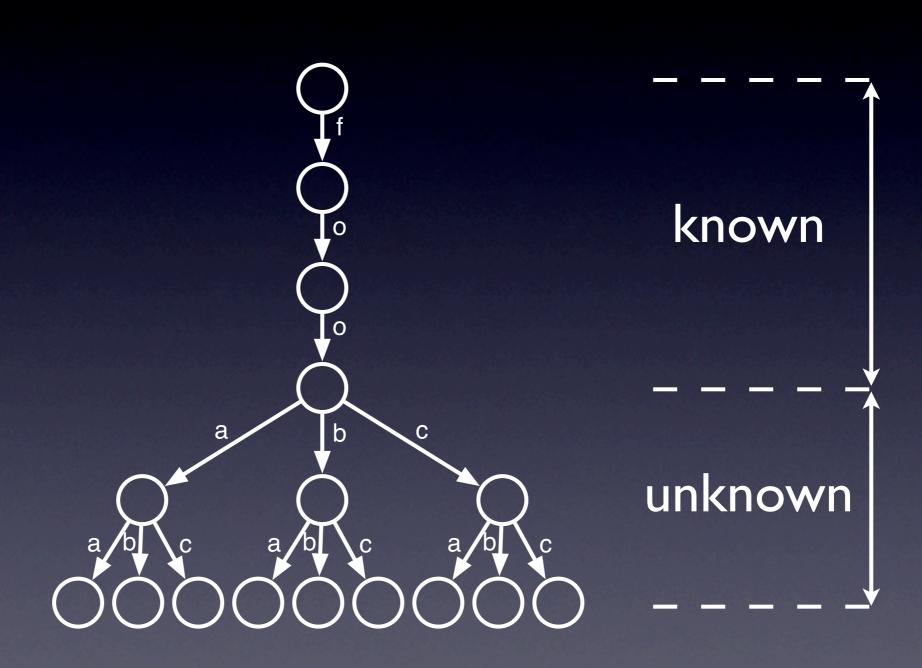
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- Store only longest suffix and list of probabilities



### Word completion

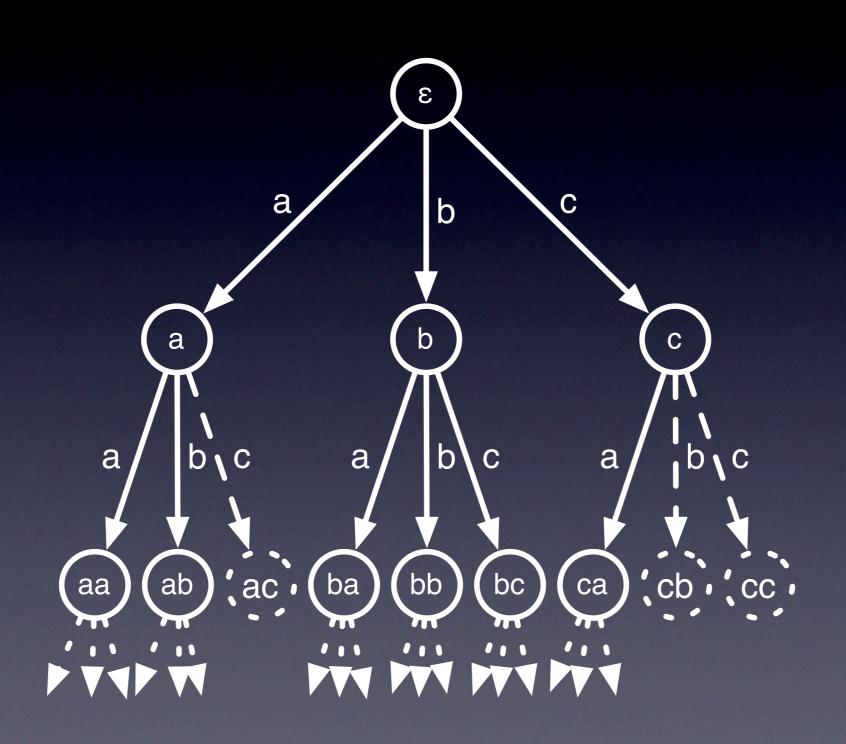


Online version of Baum-Welch algorithm:

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  - Use intermediate transition counts for inference

### Suffix trie construction



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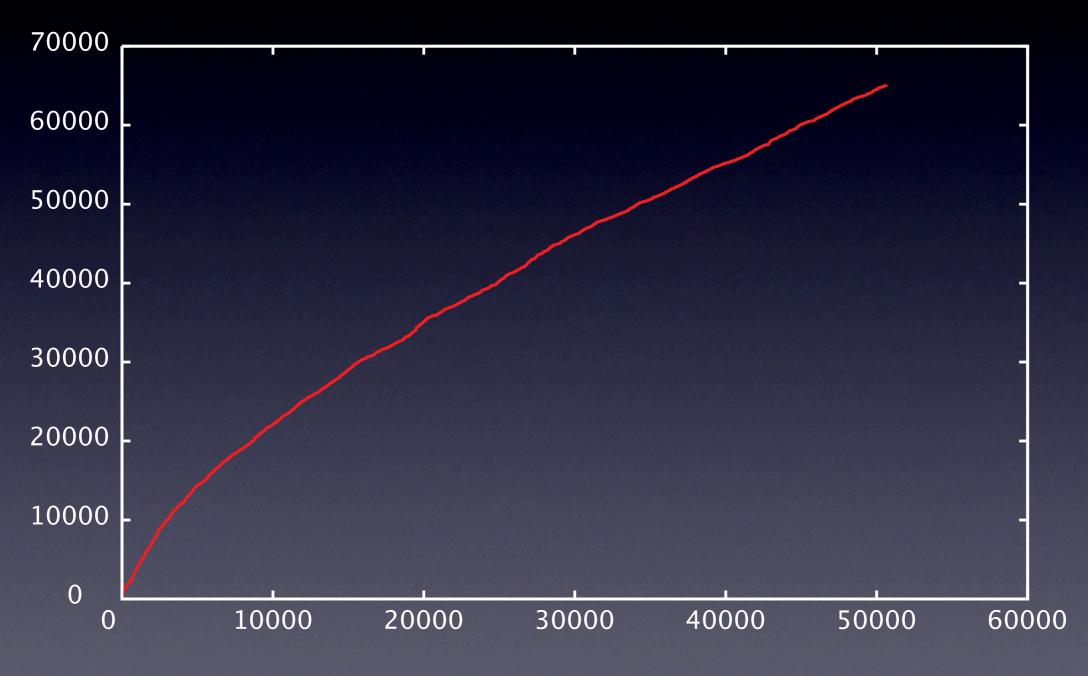
- Problem: probabilities stay zero
  ⇒ Smoothing
- Value: current state probability
- Total transition counts increase by I (like during update-step)
- Uses simple estimation of new values for inference, then updates estimation

#### Evaluation

- English text corpus from Dasher,
  ≈56,000 words, 300 kb
- 90% training, 10% testing
- Measuring number of states and average perplexity of test set

### Results

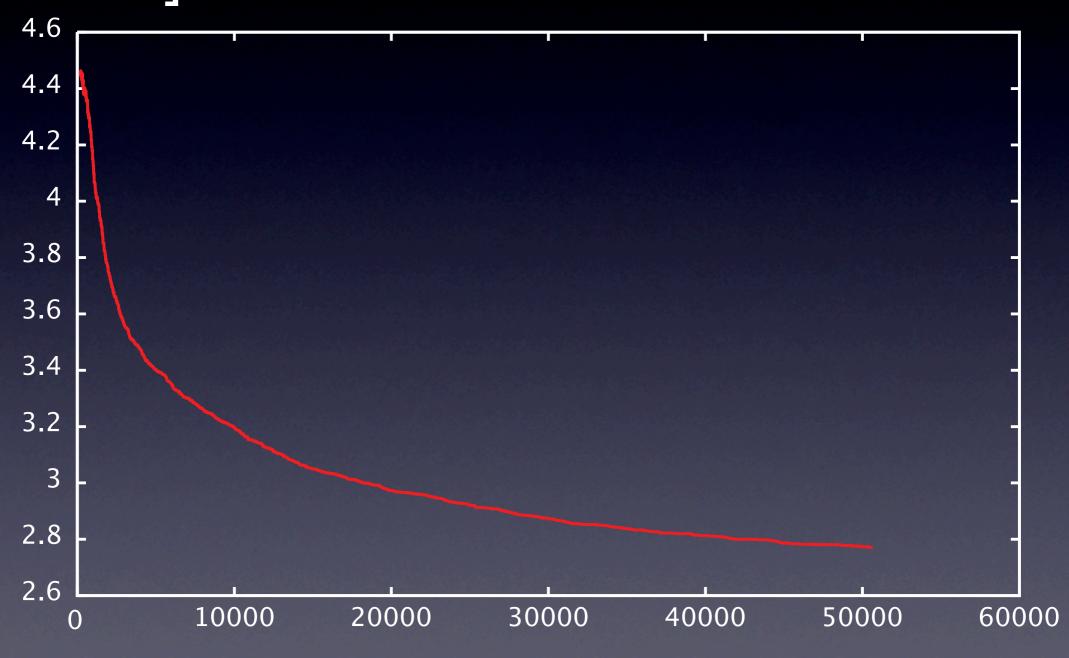
#### States



Words learned

## Avg. Perplexity [bits/char]

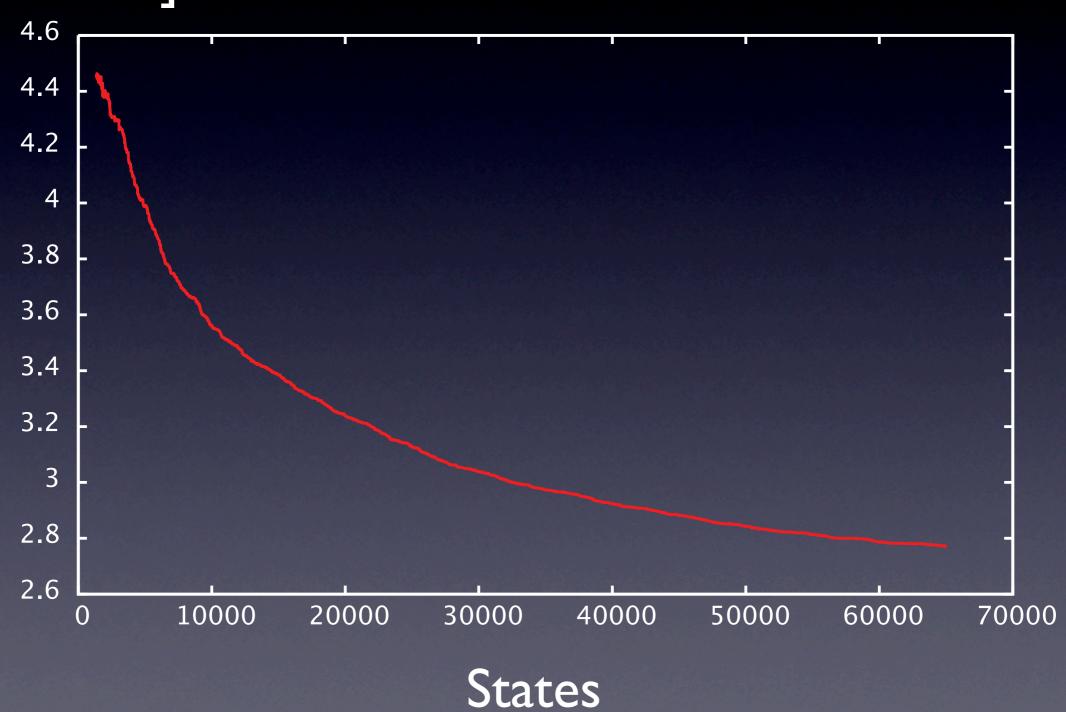
### Results



Words learned

# Avg. Perplexity [bits/char]

### Results



### An example

• T9-like text input:

http://home.in.tum.de/~bauerb/t9/