

Cerebellum: Character-based Language Model

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Introduction

The goal of the CBLM is:

- ▶ get rid of tokenization and any other rule-based processing
- ▶ limit language units not by length but by frequency
- ▶ do not work with word units, use bytes as they are universal for all languages represented in computers
- ▶ keep language modelling simple
- ▶ ...and simpler

Related work

- ▶ character-based models not common
- ▶ usually used as auxiliary models
- ▶ when Markov rule (n th order) is used, character-based models lack sufficient context width
- ▶ the principle of CBLM proposed in the past: suffix-array language model (SALM, 2007), but on word-level

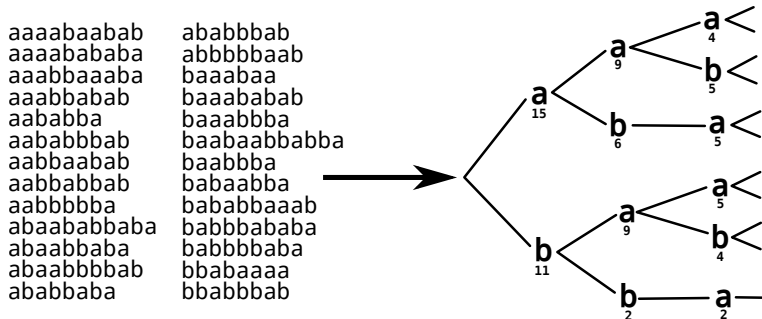
Suffix tree example

Input is any plain text data, one sentence per line.

i	suffix	SA	sorted suffix	LCP
0	popocatepetl	5	atepetl	0
1	opocatepetl	4	catepetl	0
2	pocatepetl	7	epetl	0
3	ocatepetl	9	etl	1
4	catepetl	11	l	0
5	atepetl	3	ocatepetl	0
6	tepetl	1	opocatepetl	1
7	epetl	8	petl	0
8	petl	2	pocatepetl	1
9	etl	0	popocatepetl	2
10	tl	6	tepetl	0
11	l	10	tl	1

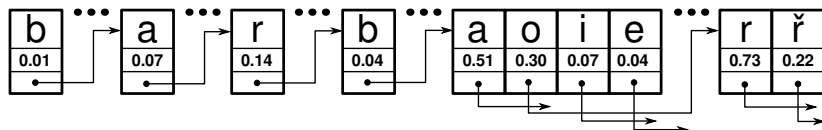
LCP up to 255 (longer sequences not stored).

From SA to trie



Parameter N : all sequences occurring $> N \times$ are put to trie.

Trie as stored on disk



Example of a Czech model, prefix *barb*

Language model

Assign probability to any byte sequence:

```
P = ROOT          // pointer to a node in trie
index = 0          // position in input sequence
probability = 1.0
current_path = [] // path from ROOT to P
while index < length(input):
    find input[index] among children(P)
    if not found:
        shorten from left and translate the current_path to ROOT
        until it can be prolonged with input[index] byte
        // current_path may be emptied
        P = current_path[-1] or ROOT
    else:
        P = found children position
    probability *= probability of P
    current_path.append(P)
    index++
return probability
```

Entropy & perplexity

$$H(m) = -\frac{1}{N} \sum_{i=1}^N \log p(b_i)$$

$$PP(m) = 2^{H(m)}$$

Model _N	Test	Size	H	PP
BNC ₂	Alice	330 M	2.286	4.879
BNC ₃	Alice	212 M	2.294	4.904
BNC ₂	1984	330 M	1.664	3.170
BNC ₃	1984	212 M	1.671	3.184
SYN2000 ₅	1984	362 M	1.837	3.574
csWiki ₃	1984	300 M	1.850	3.607
csWeb ₄	1984	312 M	1.571	2.972

Example random sentences

English First there is the fact that he was listening to the sound of the shot and killed in the end a precise answer to the control of the common ancestor of the modern city of Katherine Street, and when the final result may be the structure of conservative politics; and they were standing in the corner of the room.

Czech Pornoherečka Sharon Stone se nachází v blízkosti lesa. ¶ Máme malý byt, tak jsem tu zase. ¶ Změna je život a tak by nás nevolili. ¶ Petrovi se to začalo projevovat na veřejnosti. ¶ Vojáci byli po zásluze odměněni pohledem na tvorbu mléka. ¶ Graf znázorňuje utrpení Kristovo, jež mělo splňovat následující kritéria.

Hungarian Az egyesület székhelye: 100 m-es uszonyos gyorsúszásban a következő években is részt vettek a díjat az égre nézve szójaszármazékot. ¶ Az oldal az első lépés a tengeri akvarisztikával foglalkozó szakemberek számára is ideális szállás költsége a vevőt terhelik.

Future work

- ▶ design the *learning* process: using a fixed length window, put new data into the model (trie) on-the-fly
- ▶ compute relations with gaps (...atka → ...ala)
- ▶ use these relations to infer segmentation and grammar rules (be able to parse *Žmoulačka chřpíná u trpného hzunáka*).
- ▶ mimic short term memory by decaying contribution of nodes visited in the past
- ▶ CBLM for translation (...ed → ...aný)
- ▶ measure entropy and perplexity on standard datasets
- ▶ more extrinsic evaluation (NP segmentation, speech-to-text)