

Lecture reviews — Week 07

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Week(s ~~6~~ & ~~7~~) keypoints

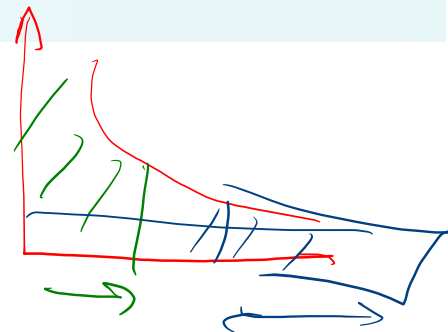
5 6

Week ~~6~~:

- ▶ (related to week ~~5~~ as well) what "lemmatization" is
- ▶ what "part-of-speech tagging" is
- ▶ two hypothesis to transform PoS tagging into "the second problem" of HMMs
- ▶ order of magnitude of performances

Week ~~7~~:

- ▶ what an HMM is
- ▶ the 3 problems and how it relates to PoS tagging
- ▶ Viterbi algorithm
- ▶ properties of Baum-Welch algorithm



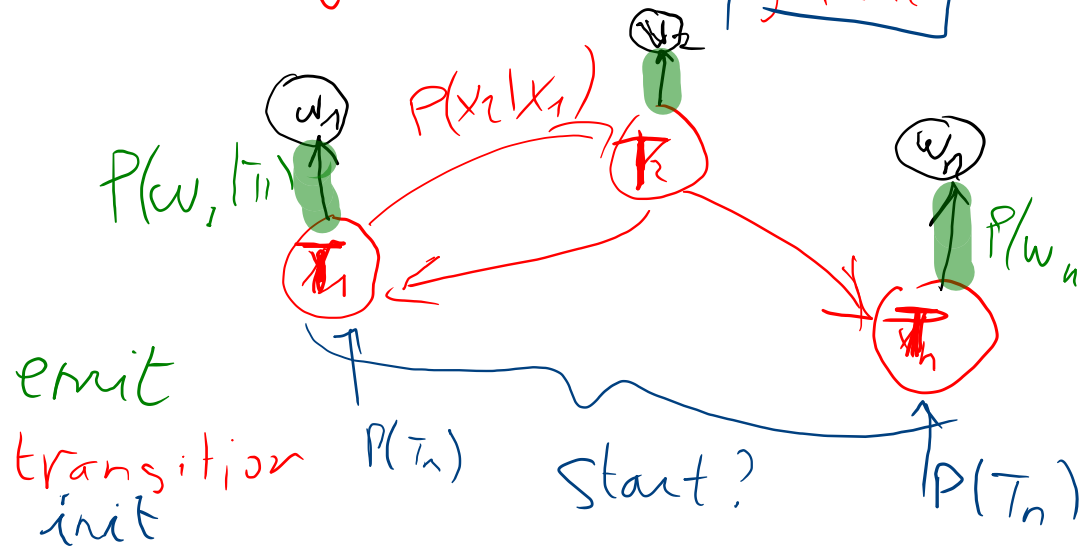
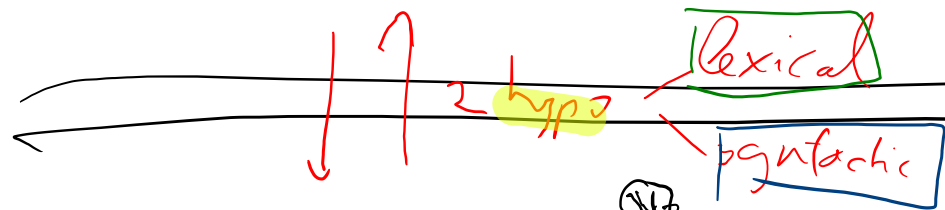
Input: $w_1 \dots w_n$

Output: $T_1 \dots T_n$

$$P(w_i | T_1 \dots T_n, w_1 \dots w_{i-1}, w_{i+1} \dots w_n)$$

$$= P(w_i | T_i)$$

$$\frac{P(T_i | T_1 \dots T_{i-1})}{P(T_i | T_{i-1})}$$



$$P(x_i | x_1 \dots x_{i-1})$$

$$P(w_n | T_n) = P(x_i | x_{i-1})$$

$$\theta = \left(\begin{array}{l} P(T_1) \dots P(T_n) \\ P(T_1 | T_2) \dots \\ P(w_i | T_i) \dots \end{array} \right)$$

Input: $w_1 w_2 \dots w_n$

Output: $T_1 T_2 \dots T_n$

Lexicon:
 $w^{(1)} w^{(2)} \dots w^{(n)}$
Tag set:
 $T^{(1)} T^{(2)} \dots T^{(n)}$

Hypothesis:

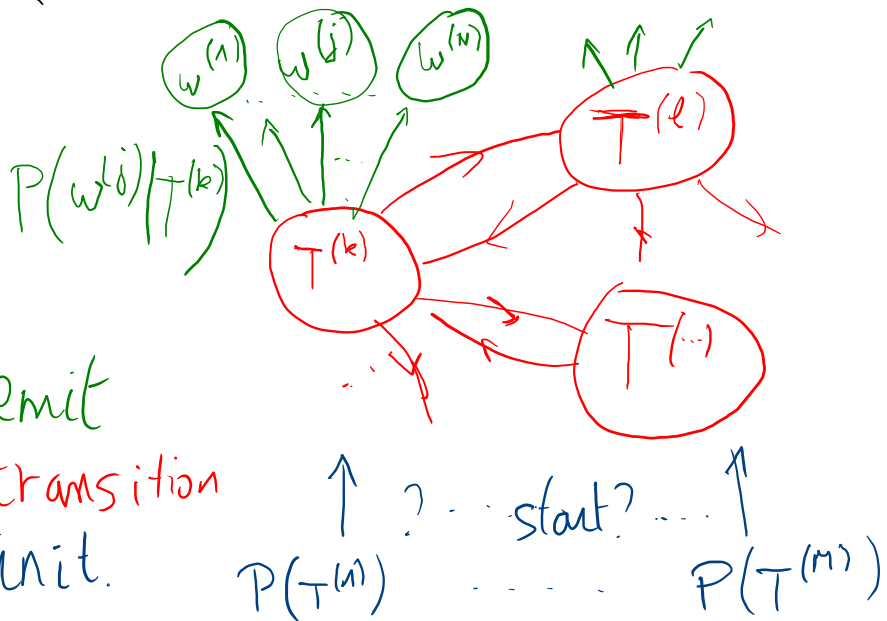
lexical:

$$P(w_i | T_1 \dots T_n, w_1 \dots w_{i-1}, w_{i+1} \dots w_n)$$

$$= P(w_i | T_i)$$

syntactic:

$$P(T_i | T_1 \dots T_{i-1}) = P(T_i | T_{i+1})$$



$$H = \left(P(T^{(1)}), P(T^{(2)}), \dots, P(T^{(m)}), \right. \\ \left. P(T^{(1)} | T^{(2)}) P(T^{(2)} | T^{(1)}) \dots \right. \\ \left. \dots P(T^{(i)} | T^{(h)}) \dots P(T^{(m)} | T^{(m)}), \right. \\ \left. P(w^{(1)} | T^{(1)}) \dots P(w^{(i)} | T^{(k)}) \dots P(w^{(m)} | T^{(m)}) \right)$$

$$P(X_{pos} = a \mid \text{Universe})$$

$$\approx P(X_{pos} = a \mid \underbrace{\mathcal{N}(X_{pos})}_{\text{neighborhood}})$$

Markov \uparrow

Week 7 practice example (1/3)

- ① Consider an order-1 HMM PoS tagger using a lexicon with N entries, and a tag set with T tags. Furthermore, assume that the entries of the lexicon are associated, on the average, with t distinct tags.

Provide the total number P of (not necessarily free) parameters to be estimated to exploit the order-1 HMM model, assuming that no guesser has been implemented.

② several

Justify your answer.

$$P = T + T \times T + N \times \left\{ \begin{matrix} T \\ (T-1) \end{matrix} \right. \quad \left| \begin{matrix} \text{free} \\ (T-1) + T \times (T-1) + T \\ (N-1) \end{matrix} \right.$$

- ② Consider the following lexicon excerpt, where D, N, P, and V are the tags associated with the entries

(D stands for determiner, N for noun, P for pronoun, and V for verb):

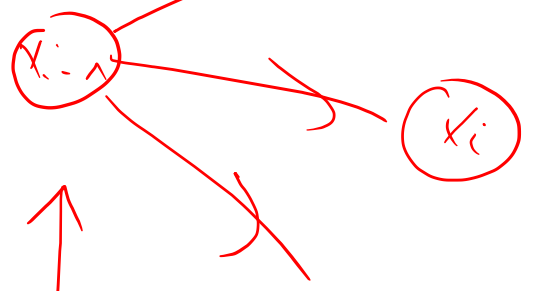
cat: N, V	saw: N, V
run: N, V	the: D
running: N, V	you: P

Provide and justify the number M of potential PoS taggings that have to be considered for the following sentence:

the cat you saw running
 1 · 2 · 1 · 2 · 2

$M = 8$

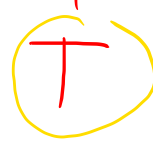
$P(w^{(k)} | x_{i-1})$
 $P(x_i | x_{i-1})$



T

$\sum_{x_i} P(x_i | x_{i-1}) = 1$
 $\rightarrow T-1$

$\Sigma = 1$



$w^{(1)}$ ate
 $w^{(2)}$ cat
 $w^{(3)}$ the
 $w^{(4)}$ zulu
 } lexicon

the cat ate
 w_1 w_2 w_3

$T \times (N-1)$

w_1 \dots w_n
 $\underset{\text{Pos}}{=}$

$P(w_i | T_i) = P(w_i = w^{(j)} | T_i = \boxed{T^{(j)}})$

Week 7 practice example (2/3)

Tag set $\mathcal{T} = \{N, V, D, P\}$

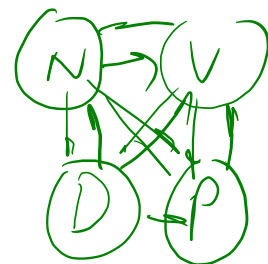
$\uparrow^{(1)} \quad \uparrow^{(2)}$

$T = 4$

cat: $\overset{\leftarrow 2}{N}, \vec{V}$
 run: N, \vec{V}
 running: $\overset{\leftarrow}{N}, \vec{V}$

$$t = \frac{10}{6}$$

saw: N, V
 the: D
 you: P
1



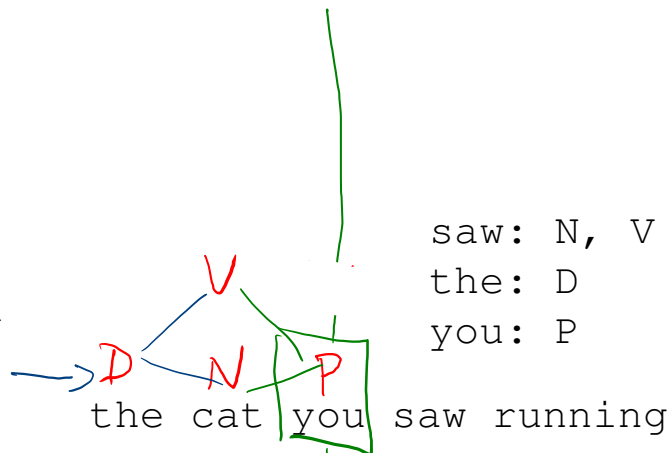
the cat you saw running

- ③ What is the condition to be verified by the parameters of the order-1 HMM model (using the provided lexicon excerpt) for the word “cat” to be tagged as a noun in the above sentence?
Justify your answer.

Week 7 practice example (2/3)

cat: N, V
 run: N, V
 running: N, V

saw: N, V
 the: D
 you: P



- ③ What is the condition to be verified by the parameters of the order-1 HMM model (using the provided lexicon excerpt) for the word “cat” to be tagged as a noun in the above sentence?

Justify your answer.

$$P(V|D) \cdot P(\text{cat}|V) \cdot P(P|V) < P(N|D) P(\text{cat}|N) \cdot P(P|N)$$

global
max \rightarrow $P(x_1 \dots x_n)$

$$= \prod \dots P(x_i | x_{i-1})$$

1-order Markov:
how the global proba is written
(NOT how is it maximized!)

Week 7 practice example (3/3)

T_1	T_2	T_3	T_4	T_n) Task
D	N	P	V	V	
the w_1	cat w_2	you	saw	running w_n	

④ What is the most probable tagging (using data provided below)?

lexicon

$w^{(1)}$
 cat: N ($1e-4$), V ($2e-6$)
 run: N ($3e-6$), V ($4e-4$)
 running: N ($5e-6$), V ($6e-4$)

saw: N ($7e-4$), V ($8e-5$)
 the: D
 you: P
 $w^{(m)}$

) data Model

$P_i(D) = 0.35$ $P_i(N) = 0.25$ $P_i(V) = 0.15$ $P_i(P) = 0.1$

$P(D|D) = 0$ $P(N|D) = 0.8$ $P(V|D) = 0$ $P(P|D) = 0$

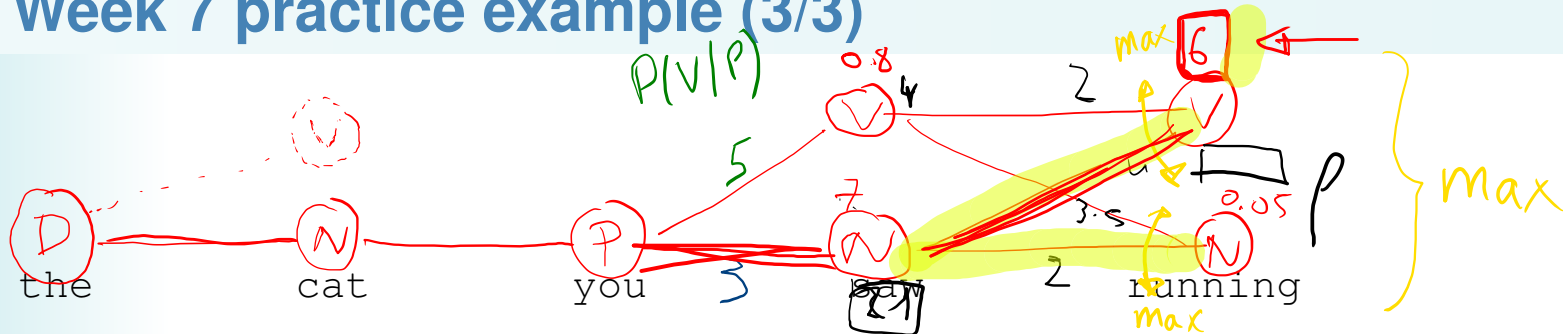
$P(D|N) = 0.1$ $P(N|N) = 0.2$ $P(V|N) = 0.4$ $P(P|N) = 0.3$

$P(D|V) = 0.15$ $P(N|V) = 0.35$ $P(V|V) = 0.2$ $P(P|V) = 0.25$

$P(D|P) = 0.1$ $P(N|P) = 0.3$ $P(V|P) = 0.5$ $P(P|P) = 0$

④

Week 7 practice example (3/3)



④ What is the most probable tagging (using data provided below)?

emit

cat: N ($1e-4$), ~~V ($2e-6$)~~
 run: N ($3e-6$), V ($4e-4$)
 running: N ($5e-6$), V ($6e-4$)

saw: N ($7e-4$), V ($8e-5$)
 the: D
 you: P

$P_i(D) = 0.35$ $P_i(N) = 0.25$ $P_i(V) = 0.15$ $P_i(P) = 0.1$

D N P N V

$P(D|D) = 0$ $P(N|D) = 0.8$ $P(V|D) = 0$ $P(P|D) = 0$
 $P(D|N) = 0.1$ $P(N|N) = 0.2$ $P(V|N) = 0.4$ $P(P|N) = 0.3$
 $P(D|V) = 0.15$ $P(N|V) = 0.35$ $P(V|V) = 0.2$ $P(P|V) = 0.25$
 $P(D|P) = 0.1$ $P(N|P) = 0.3$ $P(V|P) = 0.5$ $P(P|P) = 0$