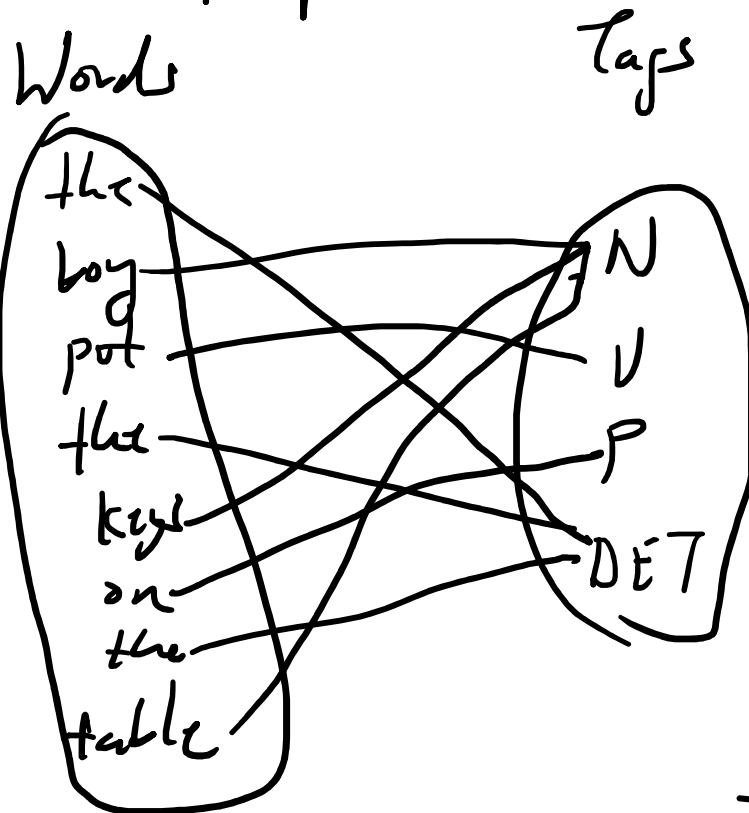


→ Disambiguation : Race ~ Semantic Analysis .

→ Parts-of-speech (POS)



→ Open Class Words (Content)

↳ Nouns, Verbs, Adjectives, Adverbs .

→ Closed Class Words

↳ Pronouns, Determiners, Prepositions, Connectives

→ N - Noun
P - Pronoun

PR - Pronoun

→ Noun

V - Verb

ADJ - Adjective

ADV - Adverb

P - Preposition

PRON - Pronoun

DET - Determiner.

→ The grand jury commented on a number of other topics.

→ The / DT → grand / JJ → jury / NN →
commented / VBD → on / IN → a / DT →
number / NN → of / IN → other / JJ →
topics / NNS.

→ Back

→ The back door. ~ JJ

→ On my back. → NN

→ Win the votes back. → RB

→ Promised to back the bill. → VB

→ the word itself:

→ arrows

→ Like, flies

→ Local Context: (the a) the bear
run/drink the grizzly bear

→ Rule Based Approach:

→ Statistical Tagging: (TBL Tagger)

→ Probabilistic: T / W

→ TBL Tagger:

→ the can was rusted.

The /DT → can / MD → was / VBD → rusted / VBD.

→ MD → NW: DT / _

→ VBP → VBN: VBD / _

-> Probabilistic Tagging:

Generative Model: (Joint)

Discriminative Model: (Conditional)

-> $P(d, c)$ in terms of $P(d|c)$

-> Naive Bayes, Hidden Markov Model. v 44

-> Maximum Entropy, CRF v DME.

-> Probabilistic Tagging:

$W = w_1 \dots w_n \sim \text{Observed}$

$T = t_1 \dots t_n \sim (\text{Unknown})$

$$\begin{aligned} \hat{T} &= \operatorname{argmax}_T P(T|W) \\ &= \operatorname{argmax}_T P(t_1 \dots t_n | w_1 \dots w_n) \\ &\rightarrow \operatorname{argmax}_T P(W|T) \times P(T) \end{aligned}$$

$$\hookrightarrow \operatorname{argmax}_T \frac{P(W|T) \times P(T)}{P(W)}$$

$$\operatorname{argmax}_T P(W|T) \times \underline{P(T)}$$

$$= \operatorname{argmax}_T \prod_i^n P(w_i | w_1 \dots w_{i-1}, t_1 \dots t_i) \times P(t_i | t_1 \dots t_{i-1})$$

Bigram Assumption:

$$P(t_i | t_1 \dots t_{i-1}) \approx P(t_i | t_{i-1})$$

\therefore Using Simplification

$$\frac{1}{T} = \operatorname{argmax}_T \prod_i^n P(w_i | t_i) \times P(t_i | t_{i-1})$$