

$$P(y_1|y_2)P(x_1|y_1) \text{ and } P(y_3|y_2) \cdot \text{scores}_{(PRP)}$$

$$\left\{ \begin{array}{l} P(VB|DT)P(\text{and}|VB) \cdot \text{scores}_{(PRP)} \\ P(VB|NNS)P(\text{and}|VB) \cdot \text{scores}_{(NNS)} \\ P(VB|VB) \dots \\ P(VB|IN) \dots \\ P(PRP|NNS)P(\text{and}|VB) \cdot \text{scores}_{(PRP)} \end{array} \right.$$

$$\text{score}_4(VB) = \max_{y_3} \left\{ \begin{array}{l} \dots \text{scores}_{(DT)} \\ \dots \text{scores}_{(NNS)} \\ \dots \\ \dots \text{scores}_{(PRP)} \end{array} \right.$$

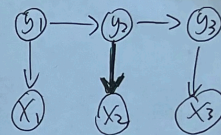
"cats" NNS
 $P(y_3 = \text{NNS} | x)$

$y = \text{PRP NNS NNS NNS DT}$
 $x = \text{I love cats and dogs}$

HMM

$n=3$

x_1	x_2	x_3
I	love	cats
PRP	VBP	NNS
y_1	y_2	y_3



edges convey conditional independence
 i.e. x_2 is conditionally independent of everything else given y_2

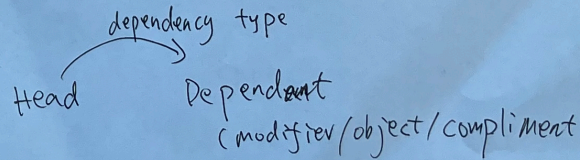
$$x_2 \perp\!\!\!\perp y_1, x_1 | y_2$$

$$P(x, y) = P(x_1, x_2, x_3, y_1, y_2, y_3) = P(y_2|y_1) \dots P(x_2|y_2) \dots$$

$$= P(y_1) \cdot P(x_1|y_1) \cdot \frac{P(y_2|y_1, x_1)}{P(y_2|y_1)} \cdot \underbrace{P(x_2|y_1, x_1, y_2)}_{P(x_2|y_2)} \cdot \underbrace{P(y_3|y_1, x_1, y_2, x_2)}_{P(y_3|y_2)} \cdot P(x_3|y_3)$$

Dependency Grammars

syntactic structure = lexical items linked by binary asymmetrical relations called dependencies



Dynamic Programming

1, 1, 2, 3, 5, 8, 13, ...

Fibonacci Numbers: $F_1 = F_2 = 1; F_n = F_{n-1} + F_{n-2}$

naive algorithm

fib(n):

if $n \leq 2$, return 1
 else return fib(n-1) + fib(n-2).

$\Rightarrow T(0) = T(1) = 1$

$T(n) = T(n-1) + T(n-2) + O(1) = c$ (constant)

$\geq 2T(n-2) + O(1) = c$

$\geq 2(2T(n-4) + c) + c = 4T(n-4) + 3c$

⋮

$\geq 2^k T(n-2k) + (2^k - 1)c$

when $n-2k=0 \Rightarrow k=n/2$

$T(n) \sim 2^{n/2} \cdot T(0)$

dynamic programming

memo = {}

fib(n):

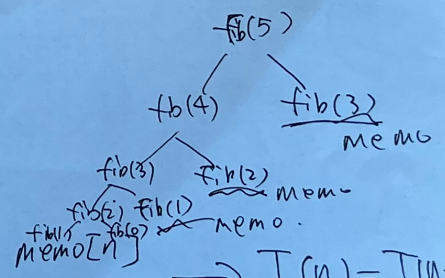
if n in memo, return

else: if $n \leq 2$: f = 1

else: f = fib(n-1) + fib(n-2)

memo[n] = f

return f.



$\Rightarrow T(n) = T(n-1) + O(1) = O(n)$

free lookup.