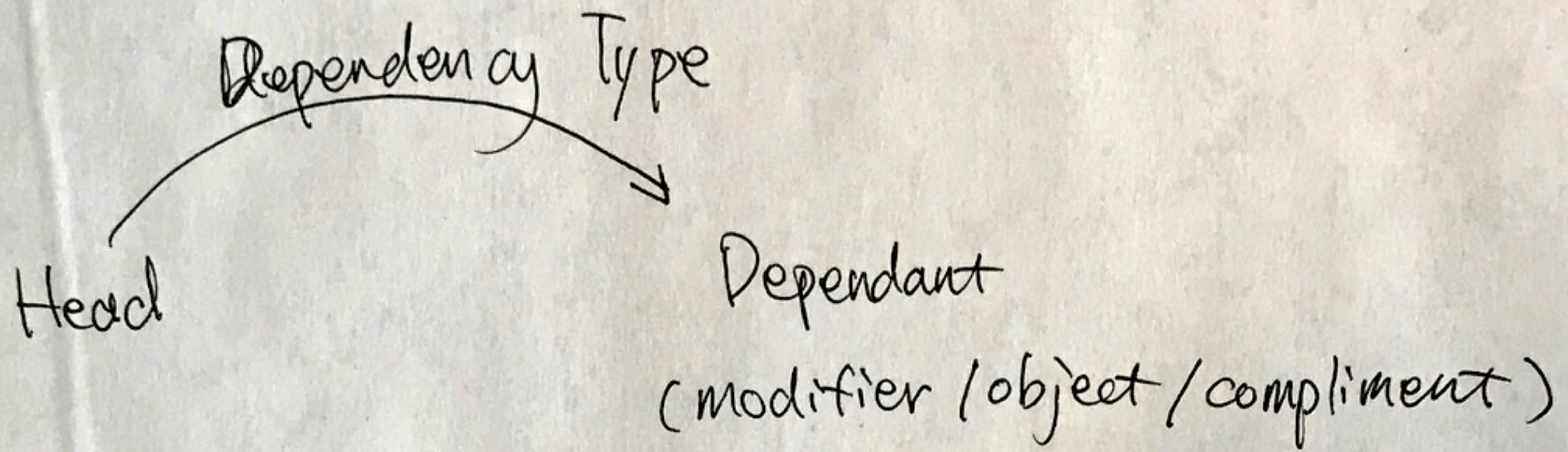


Dependency Grammars

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

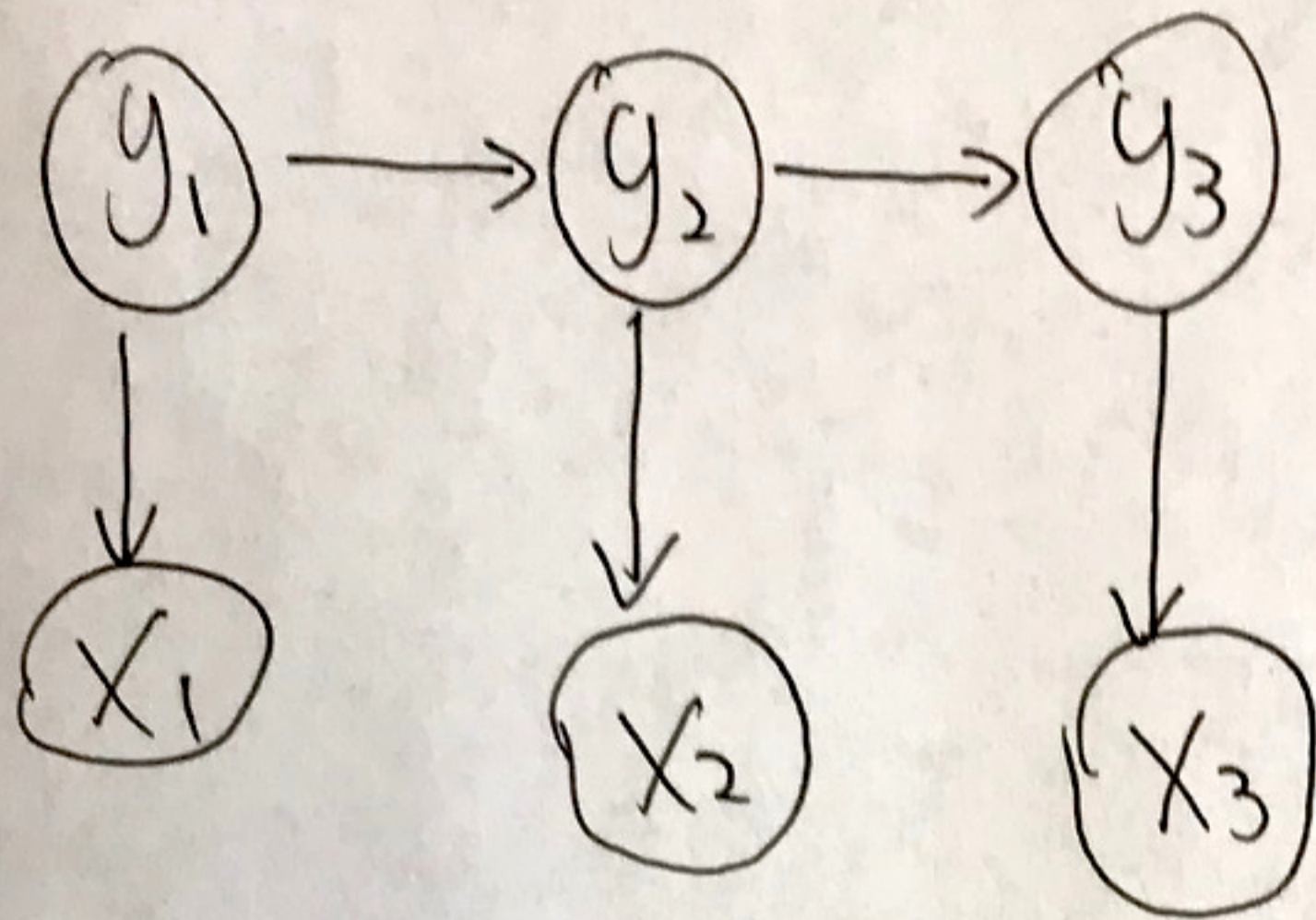


$$\underbrace{P(Y|X)}_{\text{posterior}} = \frac{P(Y, X)}{P(X)} = \frac{\overbrace{P(X|Y)}^{\text{likelihood}} \overbrace{P(Y)}^{\text{prior}}}{P(X)}$$

$$P(x_i | y_i) \propto \exp \theta^T \mathbf{f}(x_i, y_i)$$

suffix
hyphen
capital letters
numbers
:

First-Order Markov Assumption



Assume that

$$y_2 \perp\!\!\!\perp x_1 \mid y_1$$

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

$$y_3 \perp\!\!\!\perp y_1, x_1, x_2 \perp y_2$$

$$x_3 \perp\!\!\!\perp y_1, x_1, y_2, x_2 \perp y_3$$

$$P(\mathbf{x}, \mathbf{y}) = P(x_1, x_2, x_3, y_1, y_2, y_3)$$

~~$$= P(x_1) P(y_1 \mid x_1)$$~~

Chain Rule

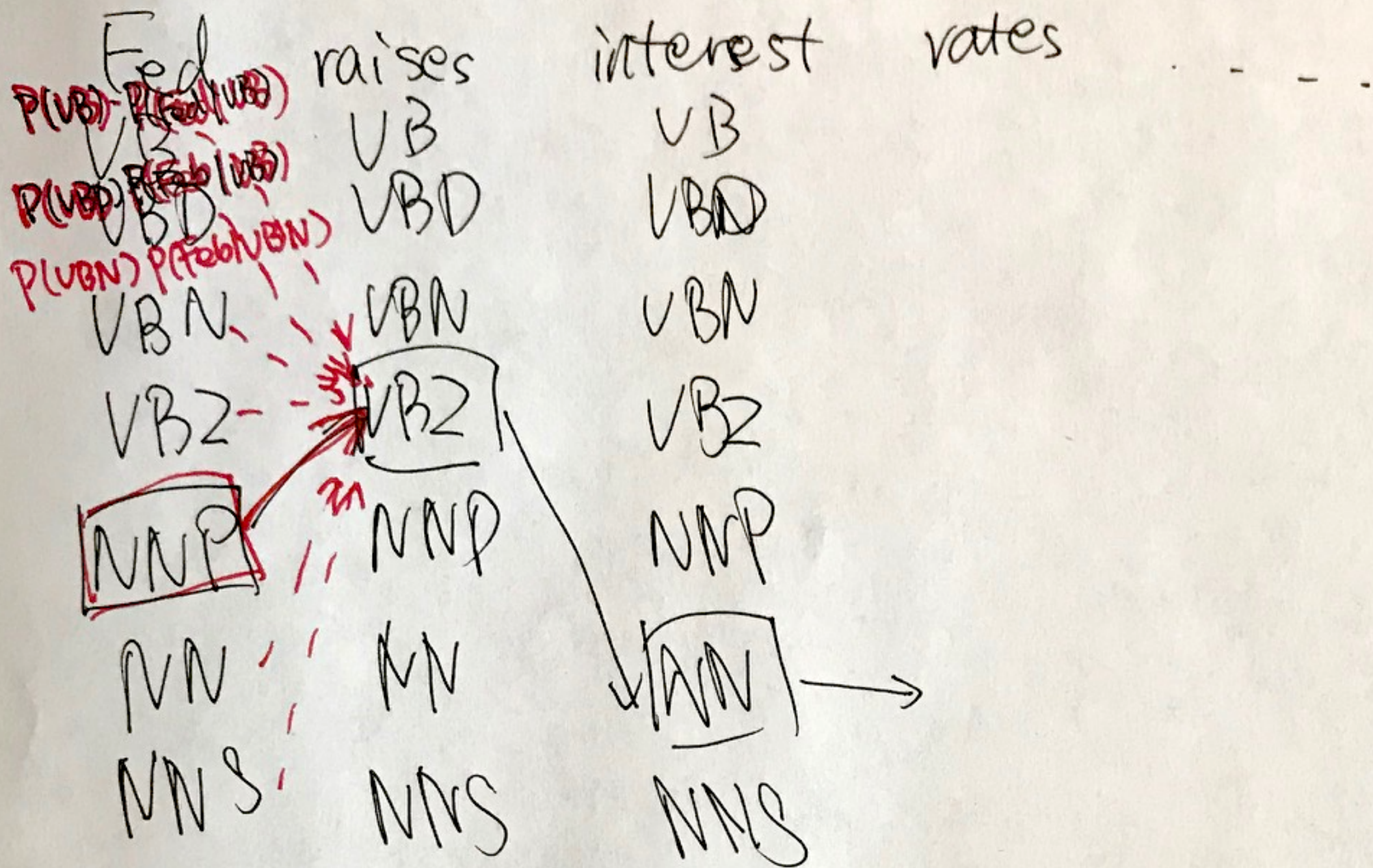
$$= P(y_1) P(x_1 \mid y_1) P(y_2 \mid y_1, x_1) P(x_2 \mid y_1, x_1, y_2)$$

$$P(y_3 \mid y_1, x_1, y_2, x_2) P(x_3 \mid y_1, x_1, y_2, x_2, y_3)$$

Markov Assumption

$$= P(y_1) P(x_1 \mid y_1) P(y_2 \mid y_1) P(x_2 \mid y_2)$$

$$P(y_3 \mid y_2) P(x_3 \mid y_3)$$



$$\text{score}_1(y_1) = P(y_1) P(x_1|y_1)$$

$$\text{score}_2(y_2) = \max_{y_1} P(y_2|y_1) P(x_2|y_2) \cdot \text{score}_1(y_1)$$

$$\text{score}_i(y_i) = \max_{y_{i-1}} P(y_i|y_{i-1}) P(x_i|y_i) \cdot \underline{\underline{\text{score}_{i-1}(y_{i-1})}}$$

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)

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

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

$$\geq 2T(n-2) + \underline{O(1)} \quad c.$$

~~$$\geq 2T(n-4) +$$~~

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

⋮

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

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

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

Fibonacci

Dynamic Programming
Memo = {}

fib(n):

if n in memo: return memo[n]

else: if n ≤ 2: f = 1

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

free

return f.

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

