

week 4

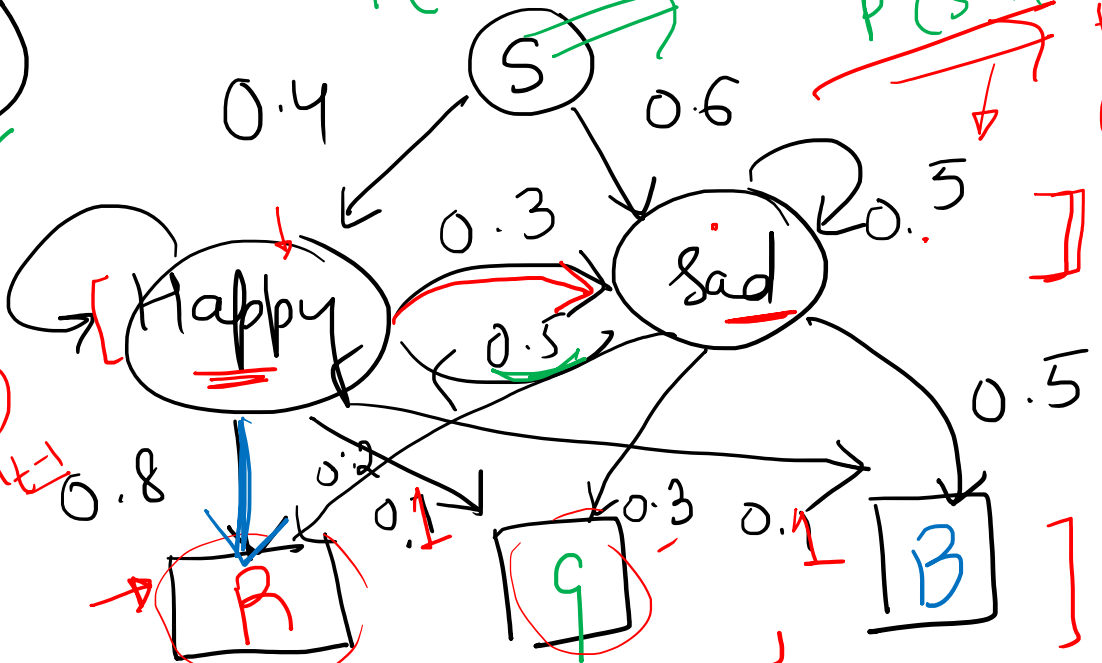
① ⇒ Hidden Markov Model

Viterbi Algorithm

}

HMM ✓

$P(H|start) = 0.4$
 $P(sad|start) = 0.6$
 hidden states



$P(H|H)$
 M_t, M_{t-1}

$P(s|H) = \frac{0.3}{0.5}$
 $P(H|s) =$

S: start state

t : current time
 $t-1$: previous time

$P(M_t | M_{t-1})$
 ** ~~two~~ ~~rows~~ transition = 1

States

Mood

observed

Transition

M_t


	H	s
H	$H H = 0.7$	$P(s H) = 0.3$
s	$H s = 0.5$	0.5

$(M_t | H)$

Emission

	R	G	B
H	$R H = 0.8$	$G H = 0.1$	0.1
s	$R s = 0.2$	$G s = 0.3$	$B s = 0.5$

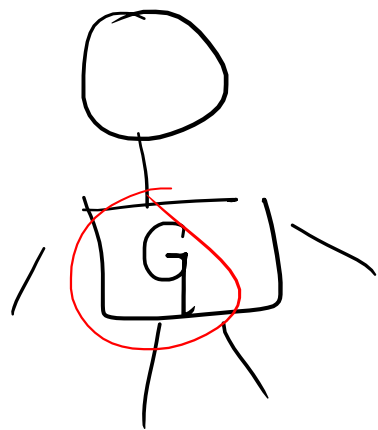
Transition prob \equiv Hidden states \equiv Mood
Emission prob \equiv Observed states } R, G, B



Hidden states directly affect the observed states
HMM

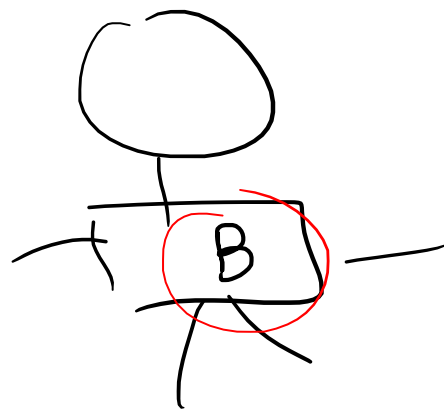
Basic Idea of HMM

There are some hidden states which we don't know about ~~those hidden~~ but they directly affect the observed states.



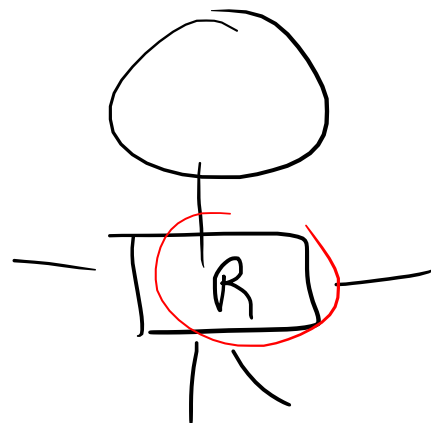
Day 1

$C_1 = G$



Day 2

$C_2 = B$



Day 3

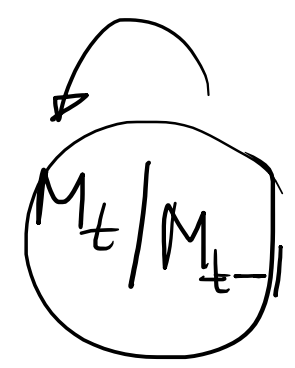
$C_3 = R$

$\begin{matrix} G & B & B \\ \hline G & R & R \end{matrix}$

$\{ M_1, M_2, M_3 \} = ?$

M : mood (hidden state)
 C : obs

$$\text{Max } P(C_1 = \underline{G}, C_2 = \underline{B}, C_3 = \underline{R})$$

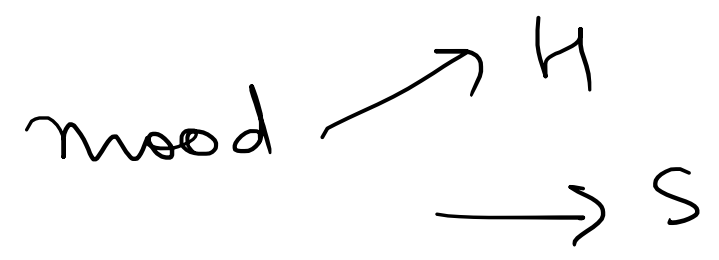


obs states
↑
hidden states

$$M_1 = m_1, M_2 = m_2, M_3 = m_3$$

realisation of random var $(m_1 \rightarrow M_1, m_2 \rightarrow M_2, m_3 \rightarrow M_3)$

Ques: what is the most likely sequence of moods of the prof on these 3 days?



H S	H S	H S
<hr/>	<hr/>	<hr/>
D ₁	D ₂	D ₃

How many possible seq. of moods are there for
3 days?

$$H = 2^3$$

$$2^3 = 8$$

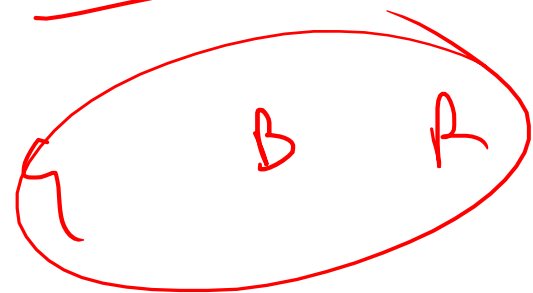
$$- - - = 8$$

$$2 * 2 * 2$$

$$2 * 2 * 2$$



HMM says the prob. of seeing the color seq.



should be maximised the prob. of
hidden states + obs. states

Maximise the prob \equiv Maximise the prob. of
observed states & hidden states & find which
~~combinational~~ sequence of hidden states (?) ✓

Assumptions

①

color shirt

→ ~~Transition prob = ?~~
→ Emission prob = ?
worn on any given day

only depends on the mood of the professor on that day & nothing else.

$$P(C_i | M_i)$$

②

Markov assumption

$$M_t | M_{t-1}$$

Mood on any given day depends directly on the mood on the very previous day

$$P(C_3 | C_2, C_1, M_3, M_2, M_1)$$

$$\times P(\underline{C_2} | C_1, M_3, M_2, M_1)$$

$$\times P(\underline{C_1} | M_3, M_2, M_1)$$

$$\times P(\textcircled{M_3} | M_2, M_1)$$

$$\times P(M_2 | M_1) \times P(M_1)$$

$$P(C_3 | M_3)$$

\times

$$P(C_2 | M_2)$$

$$\times P(C_1 | M_1)$$

$$\times P(M_3 | M_2)$$

$$\times P(M_2 | M_1)$$

$$\rightarrow P(M_1)$$

~~SSH~~

SHS

HHH

HSH

SSS

SHH

HHS

HSS

$C = \{R, G, B\}$; $M = \{H, S\}$

Day 1

C_1

M_1

M_4

R, G, B, R, G, B

Day 2

C_2

M_2

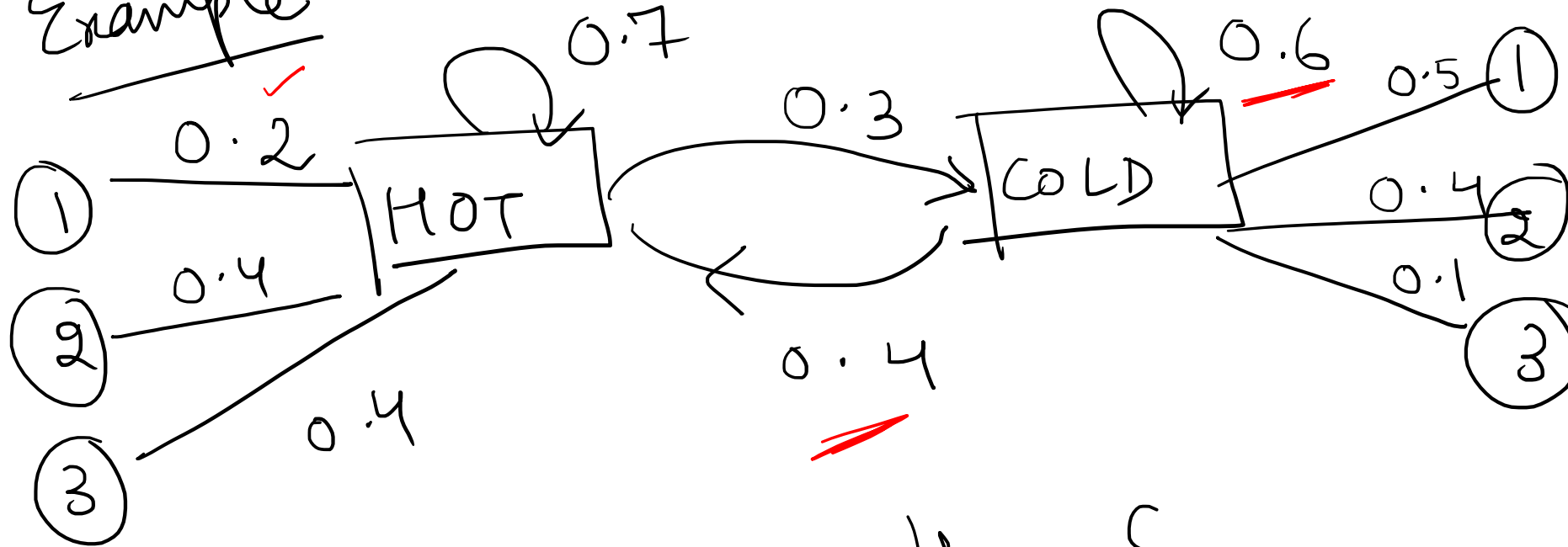
Day 3

C_3

M_3 $\rightarrow S$
 $\rightarrow H$

mood of pref
[on day 3?]

Example



$$\pi_{\text{start}} = \left\{ \begin{array}{cc} H & C \\ 0.8 & 0.2 \end{array} \right\}$$

Transition matrix

	H	C
H	0.7	0.3
C	0.4	0.6

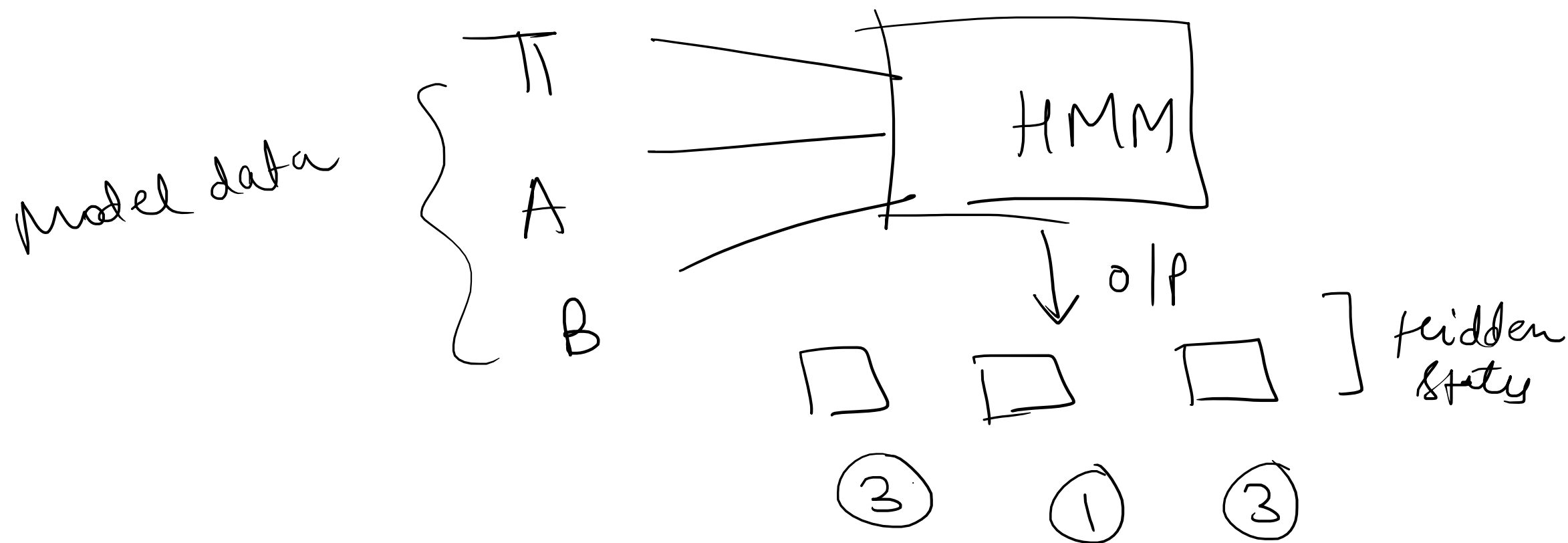
w_t / w_{t-1}

Emission matrix

	1	2	3
H	0.2	0.4	0.4
C	0.5	0.4	0.1

Observable states
(# means a person eats)

3 1 3 } i/p



weather : hidden states \equiv Transition

$w = \{H, C\}$

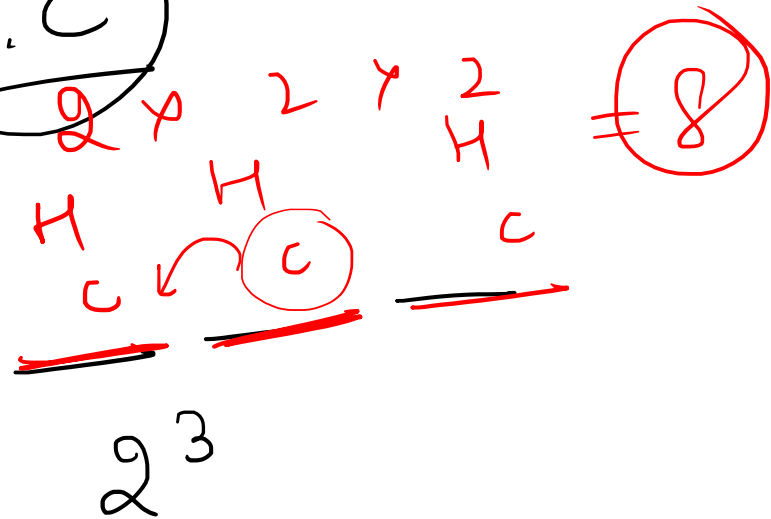
ice-cream : observed states \equiv emission

$\{1, 2, 3\}$

$N=2$;

$T_0 = 3$

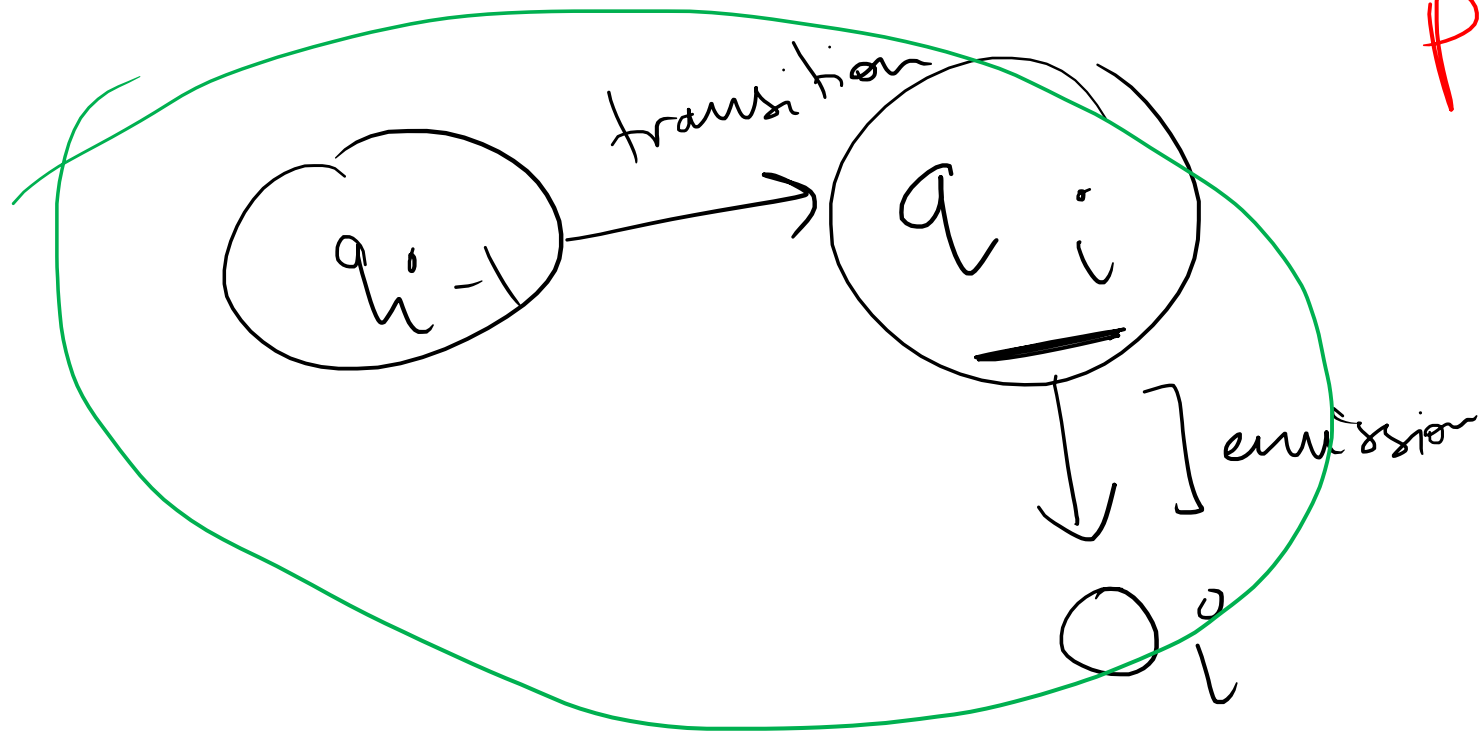
weather sequence?
→



$$P(\underline{O} | Q)$$

arg max

$$\left[\pi \overbrace{P(O_i | q_i)}^{\text{emission}}; \underbrace{P(q_i | q_{i-1})}_{\text{transition}} \right]$$



⑧ ✓

hidden states = 50

obs states = 10

~~Brute force~~

50¹⁰

≈

million/billion

⇒ Dynamic

prog

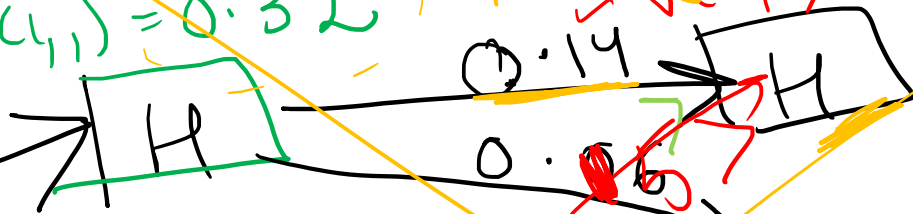
break down data into smaller chunks & store
all of intermediate steps to be used in
later steps.

$V(1)$
 $V(1,1) = 0.32$

0.0448
 $V(2)$

$V(3)$
 $V(3,1)$

$V(3,2)$



max $\{ 0.3 * 0.02, 0.15 * 0.32 \}$

3

1

3

$V(2,1) = \max$

$\{ 0.14 * 0.32, 0.08 * 0.02 \}$

Step 1 : (VI)

$$P(A, B) = P(A|B) \cdot P(B)$$

V(1,1)

$$\frac{P(\underline{3}, \underline{H})}{\underline{A}} = P(3|H) \cdot P(H)$$
$$= 0.4 \times 0.8 = \underline{0.32}$$

V(1,2)

$$P(3, C) = P(3|C) \cdot P(C) =$$
$$0.1 * 0.2 = \underline{0.02}$$

Step 2 (i) $P(1, H) = P(1|H) \cdot P(H|H)$

$$= 0.2 * 0.7 = 0.14$$

(ii) $P(1, C) = P(1|C) \cdot P(C|H)$

$$= 0.5 * 0.3 = 0.15$$

(iii) $P(1, C) = P(1|C) \cdot P(C|C)$

$$= 0.5 * 0.6 = 0.3$$

$$P(1|H) \cdot P(H|C)$$

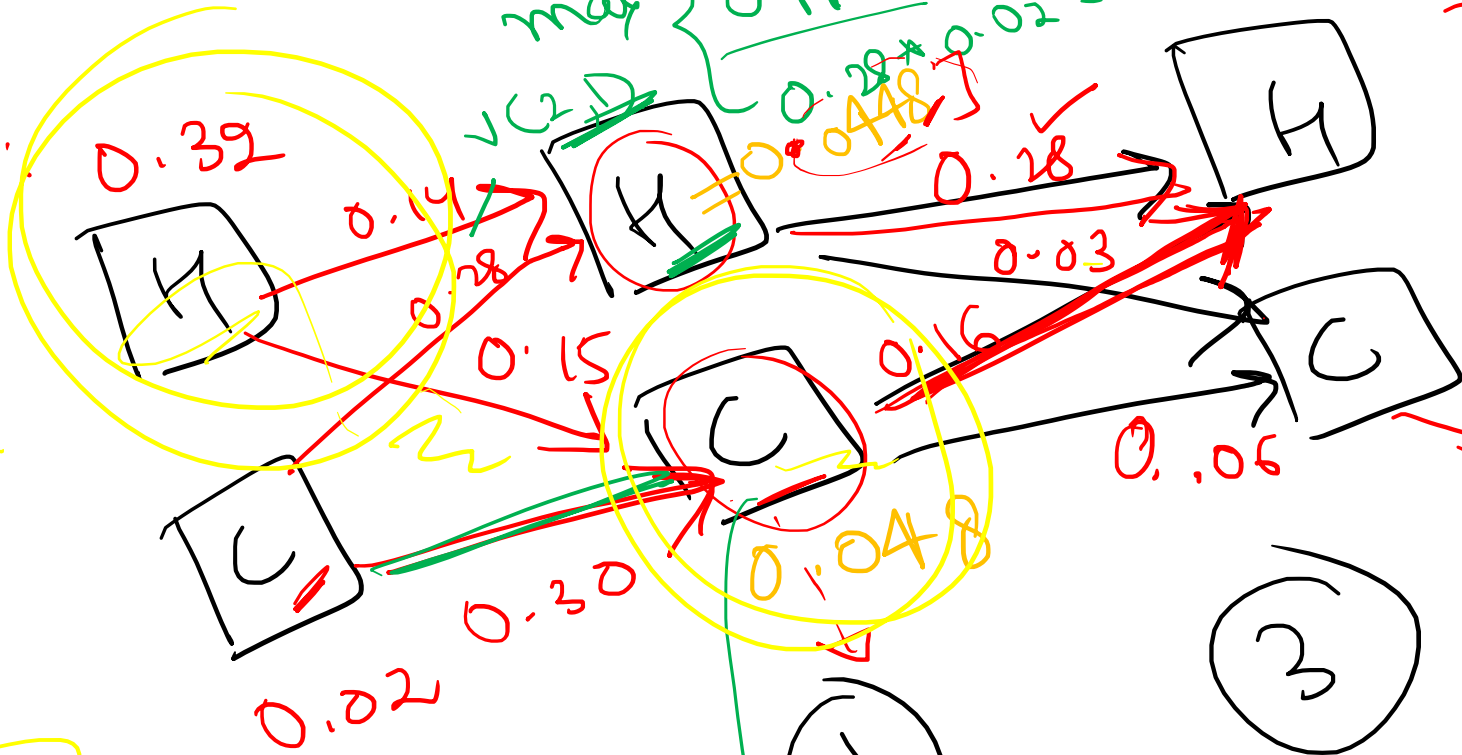
$$= 0.2 * 0.6 = 0.12$$

(iv) $P(1, H) =$

$$0.3$$

$$0.08$$

Step 2



HC H 3

$$\max \left\{ \frac{0.15 * 0.39}{0.30 * 0.02} \right\}$$

Step 2:

$$P(I, H) = P(I|H) \cdot P(H|H)$$

$0.2 \quad * \quad 0.7$

$$= \frac{0.14}{1}$$

$$P(I, C) = P(I|C) \cdot P(C|H)$$
$$= 0.5 * 0.3 = 0.15$$

$$P(1, H)$$

$$= P(1|H) \cdot P(H|C)$$

$$= 0.2 * 0.4 = 0.08$$

$$P(1, C) =$$

$$P(1|C) \cdot P(C|C)$$

$$= 0.5 * 0.6 = 0.30$$

step 3:

$$P(3, H) = P(3|H) \cdot P(\underline{H}|H).$$

$$= 0.4 * 0.7 = 0.28$$

$$P(3, \underline{H}) = P(3|H) \cdot P(H|C)$$

$$= 0.4 * 0.4 = 0.16$$

$$P(3, C) = P(3|C) \cdot P(C)$$

$$= 0.1 * 0.3 = 0.03$$

$$P(3, C) = P(3|C) \cdot P(C|C)$$

$$= 0.1 * 0.6 = \underline{0.06}$$