

Viterbi Algorithm

CS 181 Fall 2021



Definition: Inputs and Outputs

- Inputs $\langle \Theta, \lambda = (n, m, A, B, \pi) \rangle$
 - Θ = a sequence of observations
 - λ = a Hidden Markov Model
 - n = the number of hidden states, $S = \{s_1, s_2, \dots, s_n\}$
 - m = the number of observation symbols, $V = \{v_1, v_2, \dots, v_m\}$
 - A = the transition probability distribution $A = \{a_{ij} = \mathbb{P}(q_{t+1} = s_j | q_t = s_i)\}$
 - The (i, j) -th entry is the probability that the HMM is in state j at time $t+1$, given that it was in state i at time t
 - B = the emission probability distribution $B = \{b_j(k) = \mathbb{P}(v_k \text{ at time } t | q_t = s_j)\}$
 - The (j, k) -th entry is the probability that the HMM emits symbol k at time t , given that it was in state j at time t
 - π = the initial state distribution $\pi = \{\pi_i = \mathbb{P}(q_1 = s_i)\}$
- Outputs: Q for which $\mathbb{P}(Q|\Theta)$ is maximal
 - The most likely sequence of states (and the probability of observing it)

Definition: Auxiliary Data Structures

- δ = Scoring matrix $\delta_t(i) = \underset{q_1 \dots q_{t-1}}{\text{MAX}} \{ \mathbb{P}(\theta_1 \dots \theta_t, q_t = s_i) \}$
 - The (i, t) -th entry gives the maximum probability of observing $\theta_1, \dots, \theta_t$ along any sequence of states and ending in state i at time t
- ψ = Backtracking matrix
 - The (i, t) -th entry gives the state at time $t-1$ which produces the maximum probability path ending in state i at time t

The Algorithm

- 1) Initialize the matrices
- 2) Apply the recurrence relations to fill each matrix
- 3) Compute the maximum probability
- 4) Initialize the backtracking process
- 5) Complete the backtracking
- 6) Output p^* and q_1^*, \dots, q_T^*

```
/* Initialization
for i ← 1 to n do
    δ1(i) ← πibi(θ1)
    ψ1(i) ← 0
end

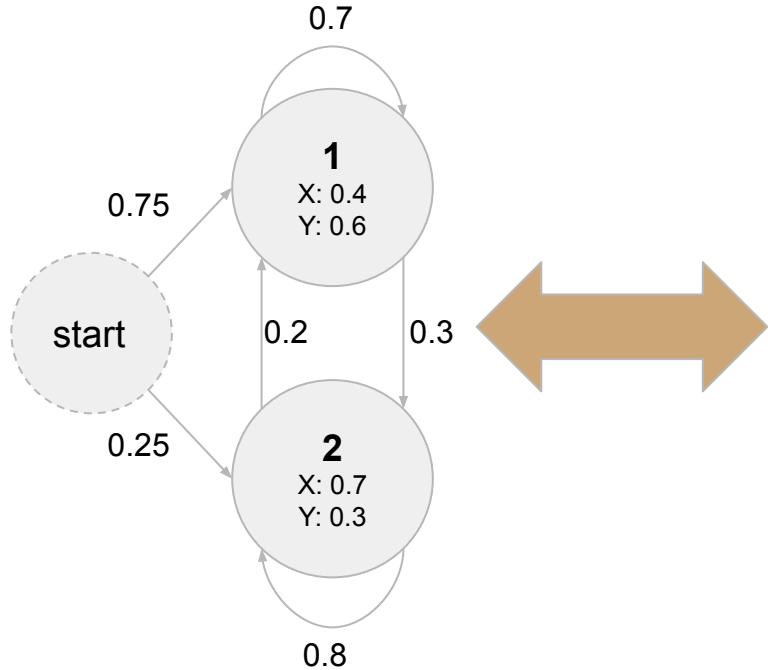
/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
        δt(j) ← MAX1≤i≤n[δt-1(i) · aij] · bj(θt)
        ψt(j) ← ARGMAX1≤i≤n[δt-1(i) · aij]
    end
end

/* Termination
p* ← MAX1≤i≤n[δT(i)]
qT* ← ARGMAX1≤i≤n[δT(i)]

/* Backtracking
for t ← T - 1 to 1 do
    | qt* ← ψt+1(qt+1*)
end
```

An Example

Observation sequence, θ :
XYXX



π	1	2
$P(q_1)$	0.75	0.25

A	1	2
1	0.7	0.3
2	0.2	0.8

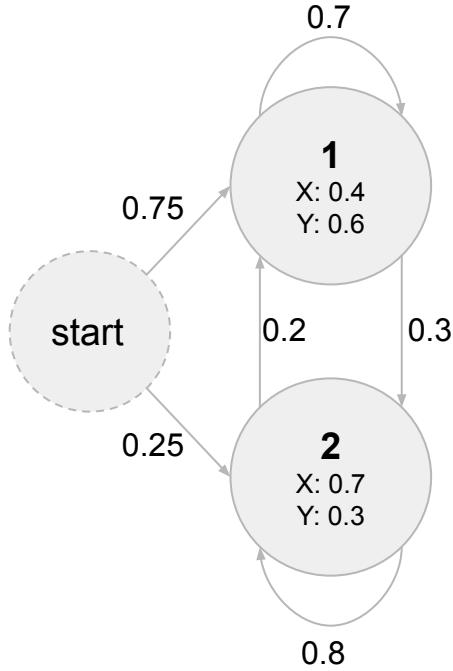
B	X	Y
1	0.4	0.6
2	0.7	0.3

XYYXX

δ	1	2
t = 1		
2		
3		
4		
5		

Ψ	1	2
t = 1		
2		
3		
4		
5		

```
/* Initialization
for i ← 1 to n do
     $\delta_1(i) \leftarrow \pi_i b_i(\theta_1)$ 
     $\psi_1(i) \leftarrow 0$ 
end
```



XYYXX

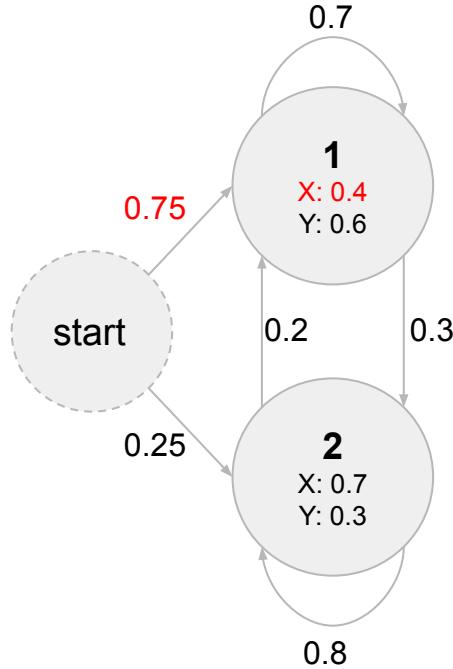
δ	1	2
$t = 1$	0.3	
2		
3		
4		
5		

Ψ	1	2
$t = 1$		
2		
3		
4		
5		

```

/* Initialization
for  $i \leftarrow 1$  to  $n$  do
     $\delta_1(i) \leftarrow \pi_i b_i(\theta_1)$ 
     $\psi_1(i) \leftarrow 0$ 
end

```



XYYXX

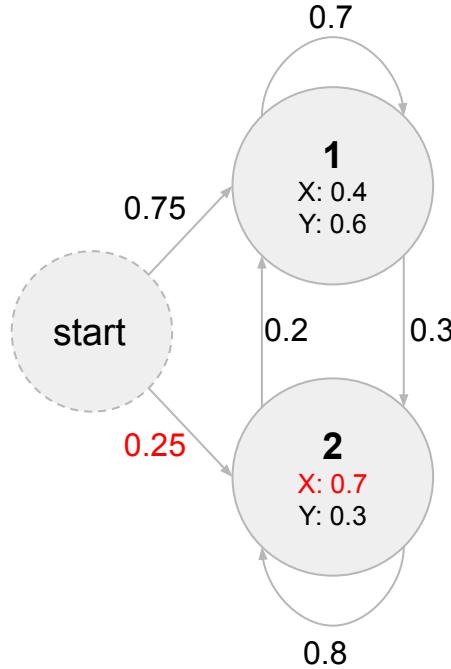
δ	1	2
$t = 1$	0.3	0.175
2		
3		
4		
5		

Ψ	1	2
$t = 1$		
2		
3		
4		
5		

```

/* Initialization
for  $i \leftarrow 1$  to  $n$  do
     $\delta_1(i) \leftarrow \pi_i b_i(\theta_1)$ 
     $\psi_1(i) \leftarrow 0$ 
end

```



XYYXX

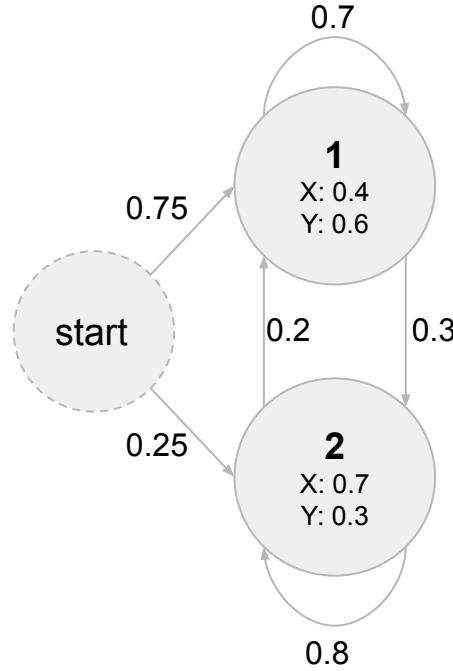
δ	1	2
$t = 1$	0.3	0.175
2		
3		
4		
5		

Ψ	1	2
$t = 1$	0	0
2		
3		
4		
5		

```

/* Initialization
for  $i \leftarrow 1$  to  $n$  do
     $\delta_1(i) \leftarrow \pi_i b_i(\theta_1)$ 
     $\psi_1(i) \leftarrow 0$ 
end

```



XYYXX

j=1: max(0.21, 0.035)

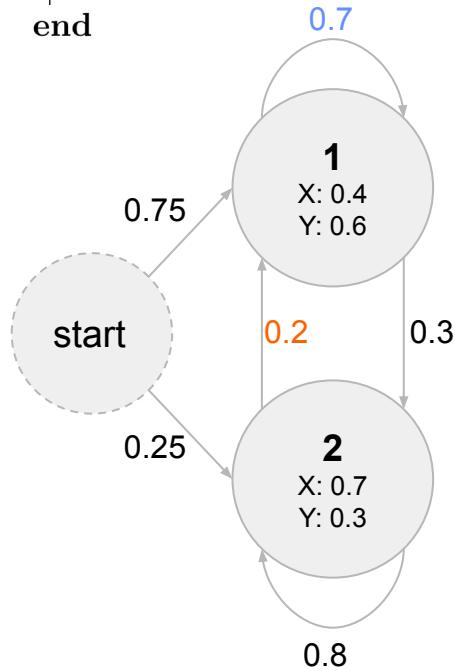
δ	1	2
t = 1	0.3	0.175
2		
3		
4		
5		

Ψ	1	2
t = 1	0	0
2		
3		
4		
5		

```

/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
         $\delta_t(j) \leftarrow \max_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}] \cdot b_j(\theta_t)$ 
         $\psi_t(j) \leftarrow \operatorname{ARGMAX}_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}]$ 
    end
end

```



XYYXX

j=1: max(0.21, 0.035)

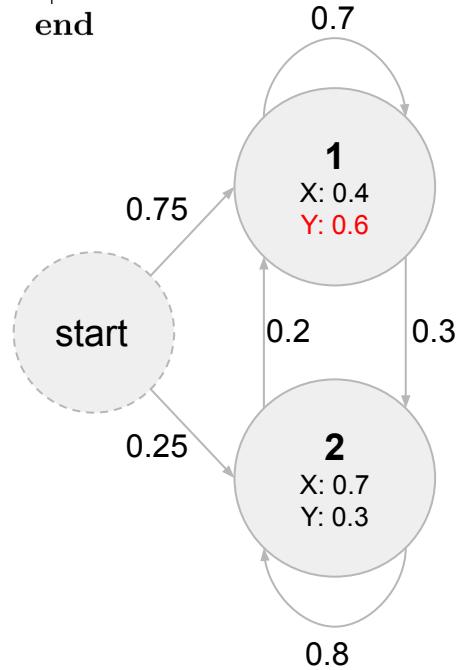
δ	1	2
t = 1	0.3	0.175
2	0.126	
3		
4		
5		

Ψ	1	2
t = 1	0	0
2	1	
3		
4		
5		

```

/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
         $\delta_t(j) \leftarrow \max_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}] \cdot b_j(\theta_t)$ 
         $\psi_t(j) \leftarrow \operatorname{ARGMAX}_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}]$ 
    end
end

```



XYYXX

j=1: max(0.09, 0.14)

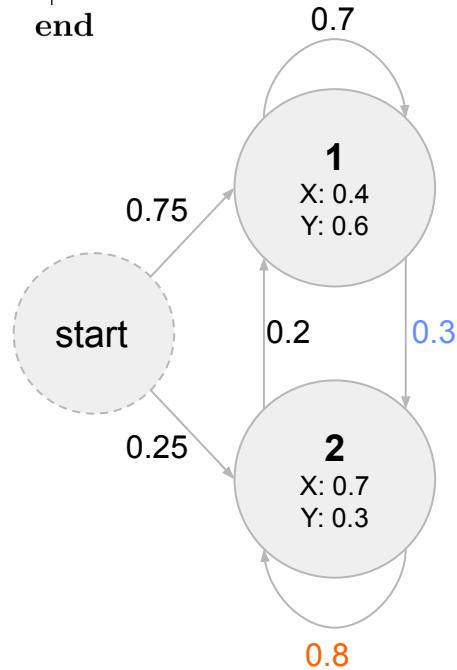
δ	1	2
t = 1	0.3	0.175
2	0.126	
3		
4		
5		

Ψ	1	2
t = 1	0	0
2	1	
3		
4		
5		

```

/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
         $\delta_t(j) \leftarrow \max_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}] \cdot b_j(\theta_t)$ 
         $\psi_t(j) \leftarrow \operatorname{ARGMAX}_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}]$ 
    end
end

```



XYYXX

j=1: max(0.09, **0.14**)

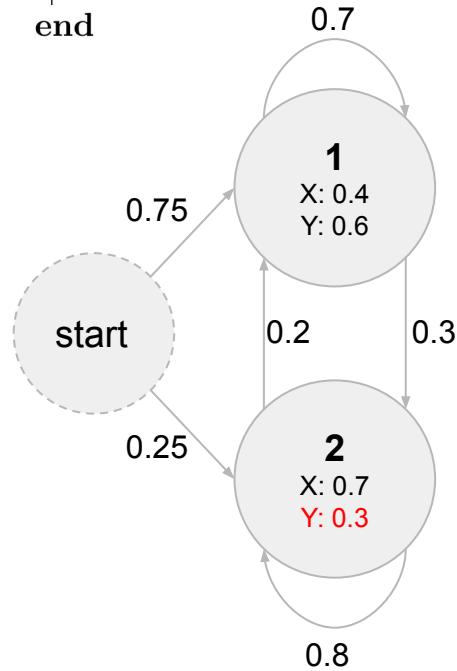
δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3		
4		
5		

Ψ	1	2
t = 1	0	0
2	1	2
3		
4		
5		

```

/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
         $\delta_t(j) \leftarrow \max_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}] \cdot b_j(\theta_t)$ 
         $\psi_t(j) \leftarrow \operatorname{ARGMAX}_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}]$ 
    end
end

```



XYYXX

j=1: max(0.0882, 0.0084)

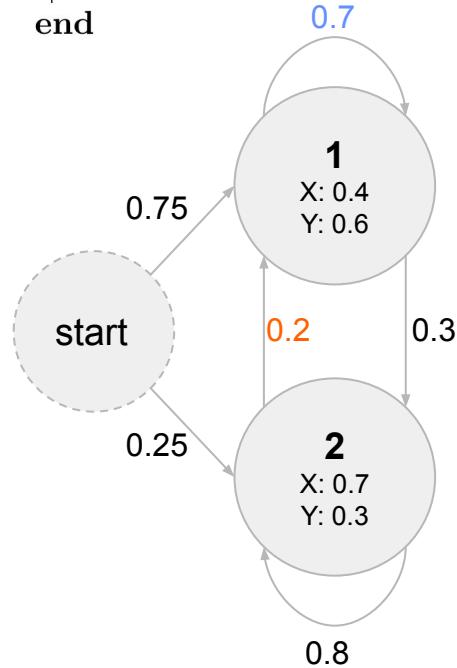
δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3		
4		
5		

Ψ	1	2
t = 1	0	0
2	1	2
3		
4		
5		

```

/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
         $\delta_t(j) \leftarrow \max_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}] \cdot b_j(\theta_t)$ 
         $\psi_t(j) \leftarrow \operatorname{ARGMAX}_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}]$ 
    end
end

```



XY_YXX

j=1: max(0.0882, 0.0084)

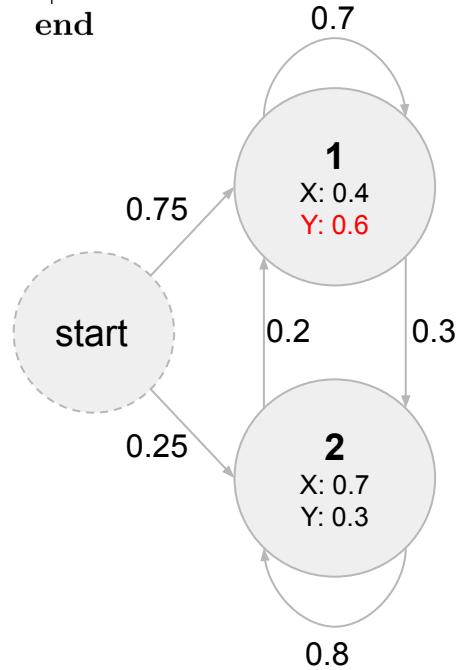
δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	
4		
5		

Ψ	1	2
t = 1	0	0
2	1	2
3	1	
4		
5		

```

/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
         $\delta_t(j) \leftarrow \max_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}] \cdot b_j(\theta_t)$ 
         $\psi_t(j) \leftarrow \operatorname{ARGMAX}_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}]$ 
    end
end

```



XYYXX

j=1: max(0.0378, 0.0336)

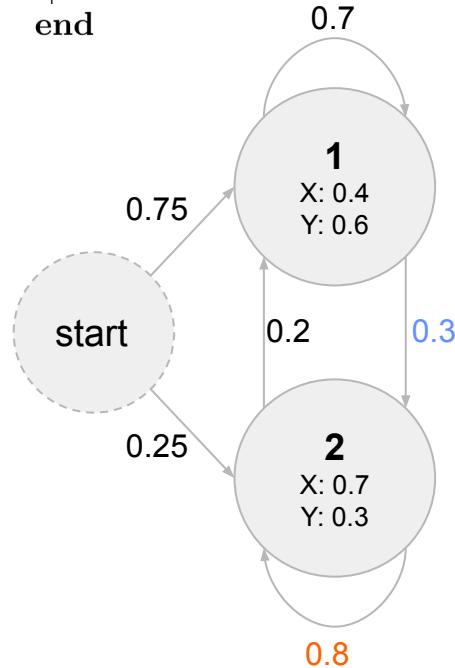
δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	
4		
5		

Ψ	1	2
t = 1	0	0
2	1	2
3	1	
4		
5		

```

/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
         $\delta_t(j) \leftarrow \max_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}] \cdot b_j(\theta_t)$ 
         $\psi_t(j) \leftarrow \operatorname{ARGMAX}_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}]$ 
    end
end

```



XY_YXX

j=1: max(0.0378, 0.0336)

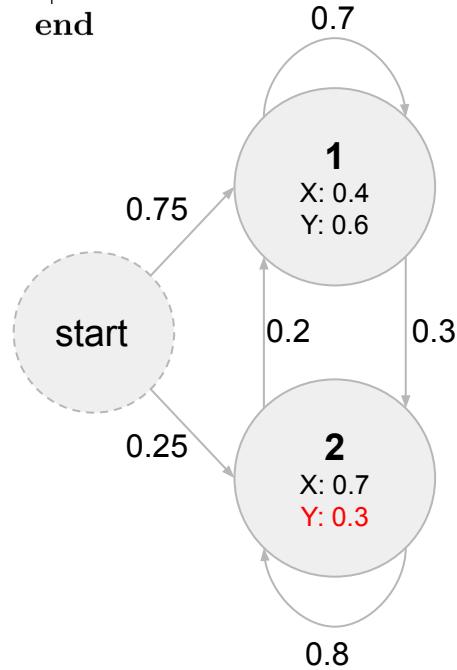
δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4		
5		

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4		
5		

```

/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
         $\delta_t(j) \leftarrow \max_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}] \cdot b_j(\theta_t)$ 
         $\psi_t(j) \leftarrow \operatorname{ARGMAX}_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}]$ 
    end
end

```



XYYXX

j=1: max(0.0370, 0.0027)

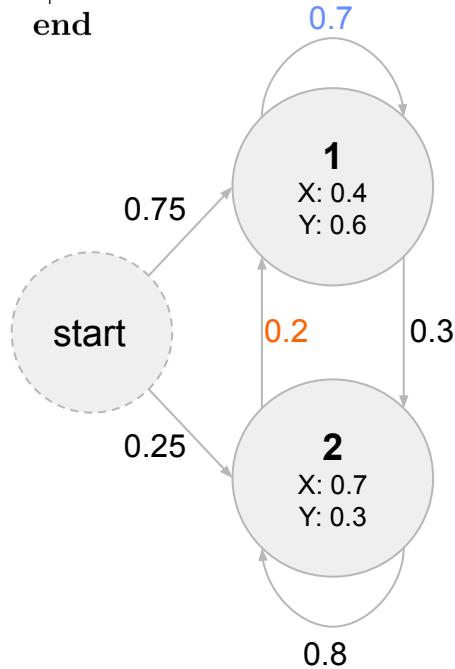
δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4		
5		

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4		
5		

```

/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
         $\delta_t(j) \leftarrow \max_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}] \cdot b_j(\theta_t)$ 
         $\psi_t(j) \leftarrow \operatorname{ARGMAX}_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}]$ 
    end
end

```



XYYXX

j=1: max(0.0370, 0.0027)

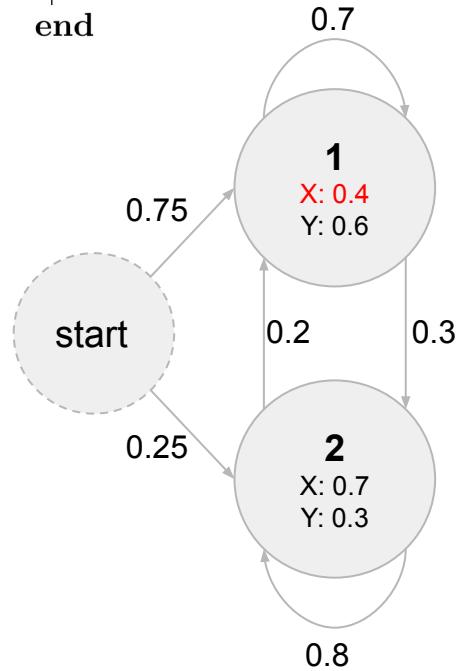
δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4	0.0148	
5		

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4	1	
5		

```

/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
         $\delta_t(j) \leftarrow \max_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}] \cdot b_j(\theta_t)$ 
         $\psi_t(j) \leftarrow \operatorname{ARGMAX}_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}]$ 
    end
end

```



XYYXX

j=1: max(0.0159, 0.0091)

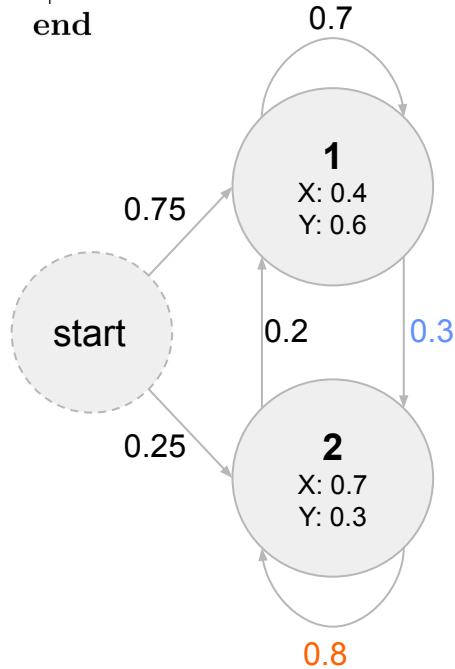
δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4	0.0148	
5		

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4	1	
5		

```

/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
         $\delta_t(j) \leftarrow \max_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}] \cdot b_j(\theta_t)$ 
         $\psi_t(j) \leftarrow \operatorname{ARGMAX}_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}]$ 
    end
end

```



XYYXX

j=1: max(0.0159, 0.0091)

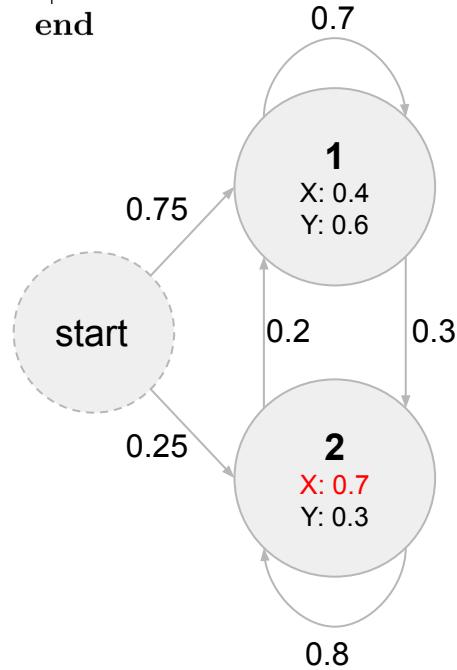
δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4	0.0148	0.0111
5		

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4	1	1
5		

```

/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
         $\delta_t(j) \leftarrow \max_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}] \cdot b_j(\theta_t)$ 
         $\psi_t(j) \leftarrow \operatorname{ARGMAX}_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}]$ 
    end
end

```



XYYXX

j=1: max(0.0104, 0.0022)

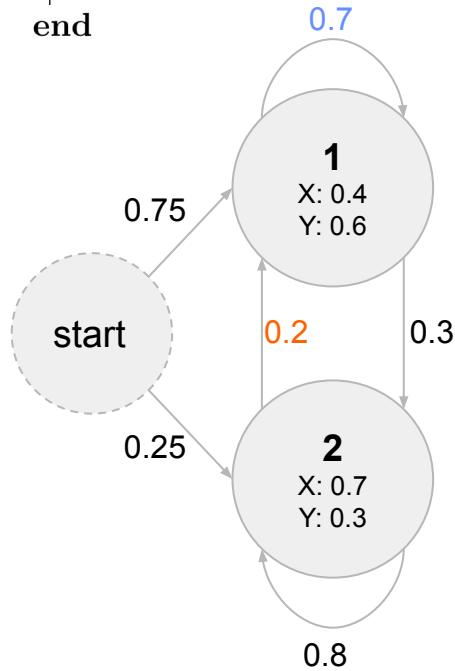
δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4	0.0148	0.0111
5		

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4	1	1
5		

```

/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
         $\delta_t(j) \leftarrow \max_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}] \cdot b_j(\theta_t)$ 
         $\psi_t(j) \leftarrow \operatorname{ARGMAX}_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}]$ 
    end
end

```



XYYXX

j=1: max(0.0104, 0.0022)

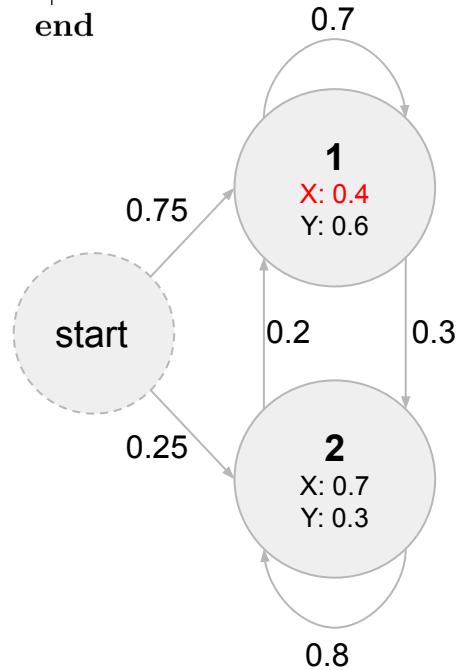
δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4	0.0148	0.0111
5	0.0041	

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4	1	1
5	1	

```

/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
         $\delta_t(j) \leftarrow \max_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}] \cdot b_j(\theta_t)$ 
         $\psi_t(j) \leftarrow \operatorname{ARGMAX}_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}]$ 
    end
end

```



XYYXX

j=1: max(0.0044, 0.0089)

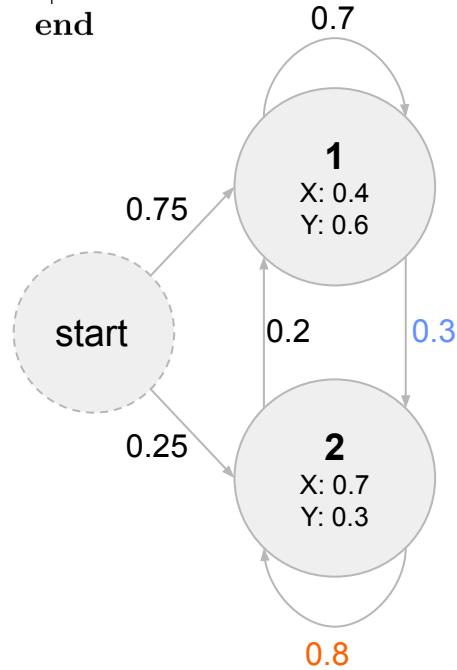
δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4	0.0148	0.0111
5	0.0041	

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4	1	1
5	1	

```

/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
         $\delta_t(j) \leftarrow \max_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}] \cdot b_j(\theta_t)$ 
         $\psi_t(j) \leftarrow \operatorname{ARGMAX}_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}]$ 
    end
end

```



XYYXX

j=1: max(0.0044, **0.0089**)

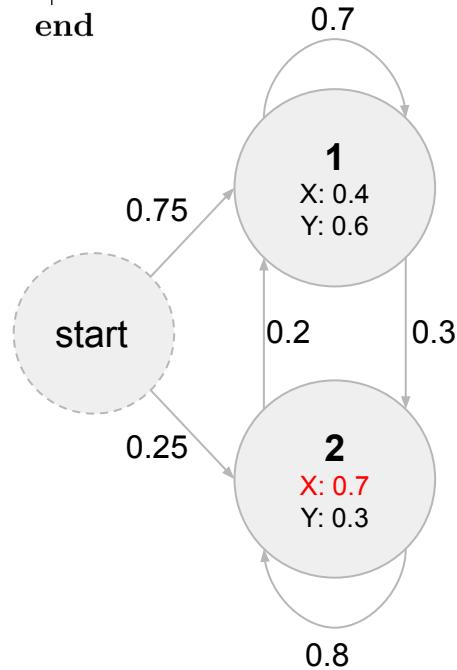
δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4	0.0148	0.0111
5	0.0041	0.0062

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4	1	1
5	1	2

```

/* Recurrence
for t ← 2 to T do
    for j ← 1 to n do
         $\delta_t(j) \leftarrow \max_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}] \cdot b_j(\theta_t)$ 
         $\psi_t(j) \leftarrow \operatorname{ARGMAX}_{1 \leq i \leq n} [\delta_{t-1}(i) \cdot a_{ij}]$ 
    end
end

```



XYYXX

δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4	0.0148	0.0111
5	0.0041	0.0062

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4	1	1
5	1	2

XYYXX

$$p^* = 0.006234$$

δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4	0.0148	0.0111
5	0.0041	0.0062

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4	1	1
5	1	2

/* Termination

$$p^* \leftarrow \underset{1 \leq i \leq n}{\text{MAX}}[\delta_T(i)]$$

$$q_T^* \leftarrow \underset{1 \leq i \leq n}{\text{ARGMAX}}[\delta_T(i)]$$

	q^*
t=1	
2	
3	
4	
5	2

XYYXX

$$p^* = 0.006234$$

δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4	0.0148	0.0111
5	0.0041	0.0062

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4	1	1
5	1	2

```

/* Backtracking
for  $t \leftarrow T - 1$  to 1 do
|  $q_t^* \leftarrow \psi_{t+1}(q_{t+1}^*)$ 
end

```

	q^*
t=1	
2	
3	
4	
5	2

XYYXX

$$p^* = 0.006234$$

δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4	0.0148	0.0111
5	0.0041	0.0062

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4	1	1
5	1	2

```

/* Backtracking
for t ← T - 1 to 1 do
|    $q_t^* \leftarrow \psi_{t+1}(q_{t+1}^*)$ 
end

```

	q^*
t=1	
2	
3	
4	2
5	2

XYYXX

$$p^* = 0.006234$$

δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4	0.0148	0.0111
5	0.0041	0.0062

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4	1	1
5	1	2

```

/* Backtracking
for t ← T - 1 to 1 do
|    $q_t^* \leftarrow \psi_{t+1}(q_{t+1}^*)$ 
end

```

	q^*
t=1	
2	
3	1
4	2
5	2

XYYXX

$$p^* = 0.006234$$

δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4	0.0148	0.0111
5	0.0041	0.0062

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4	1	1
5	1	2

```

/* Backtracking
for t ← T - 1 to 1 do
|    $q_t^* \leftarrow \psi_{t+1}(q_{t+1}^*)$ 
end

```

	q^*
t=1	
2	1
3	1
4	2
5	2

XYYXX

$$p^* = 0.006234$$

δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4	0.0148	0.0111
5	0.0041	0.0062

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4	1	1
5	1	2

```

/* Backtracking
for t ← T - 1 to 1 do
|    $q_t^* \leftarrow \psi_{t+1}(q_{t+1}^*)$ 
end

```

	q^*
t=1	1
2	1
3	1
4	2
5	2

XYYXX

$$p^* = 0.006234$$

δ	1	2
t = 1	0.3	0.175
2	0.126	0.042
3	0.0529	0.0113
4	0.0148	0.0111
5	0.0041	0.0062

Ψ	1	2
t = 1	0	0
2	1	2
3	1	1
4	1	1
5	1	2

	q^*
t=1	1
2	1
3	1
4	2
5	2

$\theta:$ XYYXX

State sequence:

1 1 1 2 2

Probability of
observing θ , given
the state sequence:

0.0062234

$$p^* = 0.006234$$

	q^*
t=1	1
2	1
3	1
4	2
5	2