

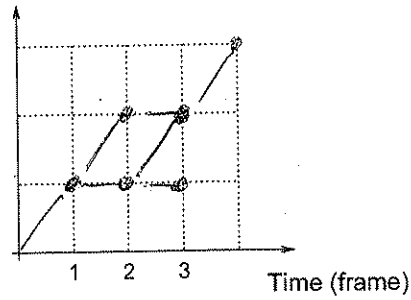
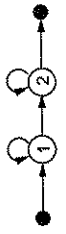
## Worked example of the forward procedure

$$\pi = [1 \quad 0]$$

$$A = \begin{bmatrix} 0.8 & 0.2 \\ 0 & 0.6 \end{bmatrix} \quad \eta = \begin{bmatrix} 0 \\ 0.4 \end{bmatrix}$$

$$B = \begin{array}{c} R \quad G \quad B \quad Y \\ \begin{bmatrix} 0.5 & 0.2 & 0.3 & 0 \\ 0 & 0.8 & 0.1 & 0.1 \end{bmatrix} \end{array}$$

State



$$O = \{R, G, G\}$$

$$t=1: \alpha_1(1) = 1 \times 0.5 = 0.5$$

$$\alpha_1(2) = 0 \times 0 = 0$$

$$t=2: \alpha_2(1) = [\alpha_1(1) \times 0.8] \times 0.2 = 0.08$$

$$\alpha_2(2) = [\alpha_1(1) \times 0.2] \times 0.8 = 0.08$$

$$t=3: \alpha_3(1) = [\alpha_2(1) \times 0.8] \times 0.2 = \del{0.0128} 0.0128$$

$$\alpha_3(2) = [\alpha_2(1) \times 0.2 + \alpha_2(2) \times 0.6] \times 0.8 = \del{0.0512} 0.0512$$

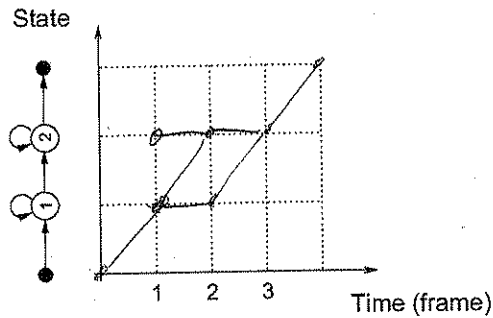
$$P(O|\lambda) = \alpha_3(2) \times 0.4 = 0.02048$$

## Worked example of the backward procedure

$$\pi = [1 \quad 0]$$

$$A = \begin{bmatrix} 0.8 & 0.2 \\ 0 & 0.6 \end{bmatrix} \quad \eta = \begin{bmatrix} 0 \\ 0.4 \end{bmatrix}$$

$$B = \begin{matrix} R & G & B & Y \\ \begin{bmatrix} 0.5 & 0.2 & 0.3 & 0 \\ 0 & 0.8 & 0.1 & 0.1 \end{bmatrix} \end{matrix}$$



$$O = \{R, G, G\}$$

$$\beta_3(1) = 0$$

$$\beta_2(1) = 0.2 \times 0.8 \times 0.4 = 0.064$$

$$\beta_3(2) = 0.4$$

$$\beta_2(2) = 0.6 \times 0.8 \times 0.4 = 0.192$$

$$\beta_1(1) = [a_{11} b_1(G) \beta_2(1) + a_{12} b_2(G) \beta_2(2)]$$

$$= 0.8 \times 0.2 \times 0.064 + 0.2 \times 0.8 \times 0.192$$

$$= 0.01024 + 0.03072$$

$$= 0.04096$$

$$\beta_1(2) = [a_{22} b_2(G) \beta_2(2)]$$

$$= 0.6 \times 0.8 \times 0.192 = 0.09216$$

$$P(O|\lambda) = [\pi_1 b_1(R) \beta_1(1) + \pi_2 b_2(R) \beta_1(2)]$$

$$= 1 \times 0.5 \times 0.04096$$

$$= 0.02048$$

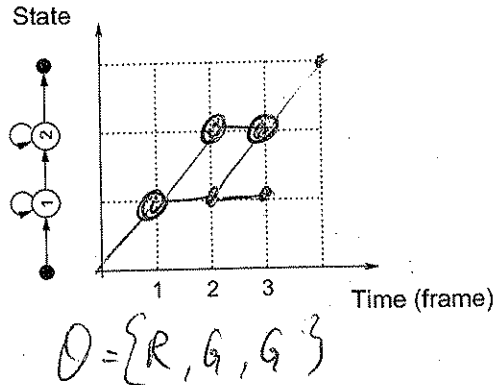
## Worked example of Viterbi algorithm

$$\bar{\pi} = [1 \ 0]$$

$$A = \begin{bmatrix} 0.5 & 0.2 \\ 0 & 0.6 \end{bmatrix} \quad \gamma = \begin{bmatrix} 0 \\ 0.4 \end{bmatrix}$$

$$B = \begin{bmatrix} 0.5 & 0.2 & 0.3 & 0 \\ 0 & 0.8 & 0.1 & 0.1 \end{bmatrix}$$

$R$ 
 $G$ 
 $B$ 
 $Y$



$t=1:$

$$\delta_1(1) = 1 \times 0.5 = 0.5$$

$$\delta_1(2) = 0 \times 0 = 0$$

$t=2:$

$$\delta_2(1) = [\delta_1(1) \times 0.8] \times 0.2 = 0.08$$

$$\delta_2(2) = [\delta_1(1) \times 0.2] \times 0.8 = 0.08$$

$t=3:$

$$\delta_3(1) = [\delta_2(1) \times 0.8] \times 0.2 = 0.0128$$

$$\delta_3(2) = \max[\delta_2(1) \times 0.2, \delta_2(2) \times 0.6] \times 0.8 = 0.0384$$

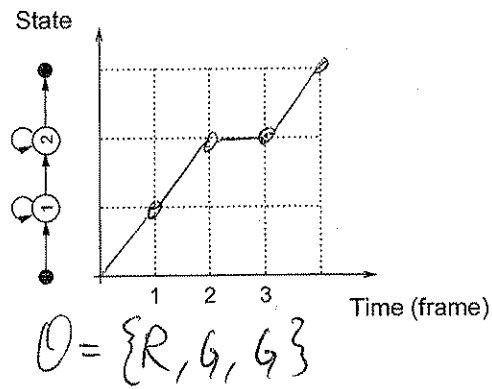
Finally,

$$\Delta_T^* = P(O, X^* | \lambda) = \delta_3(2) \times 0.4 = 0.01536$$

Thus, the best path  $X^* = \{1, 2, 2\}$ , which happens to match the actual state sequence that we used to generate the observations in the first place.

# Worked example of Viterbi re-estimation

$$\begin{aligned}
 q_1(1) &= 1 & q_1(2) &= 0 \\
 q_2(1) &= 0 & q_2(2) &= 1 \\
 q_3(1) &= 0 & q_3(2) &= 1
 \end{aligned}$$



$$\hat{\pi} = [1 \ 0]$$

$$\hat{A} = \begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix}$$

$$\hat{B} = \begin{bmatrix} R & G & B & Y \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix}$$

# Worked example of Baum-Welch re-estimation

$$\alpha_1(1) = 0.5$$

$$\alpha_1(2) = 0$$

$$\alpha_2(1) = 0.08$$

$$\alpha_2(2) = 0.08$$

$$\alpha_3(1) = 0.0128$$

$$\alpha_3(2) = 0.0512$$

$$\beta_1(1) = 0.04096$$

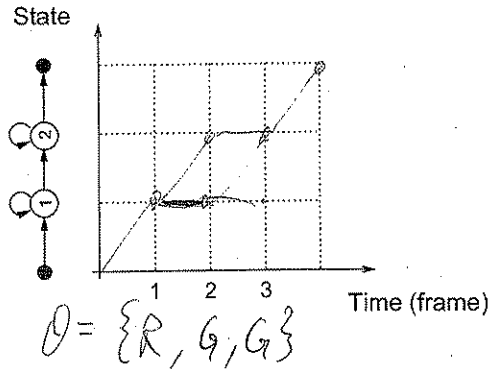
$$\beta_1(2) = 0.09216$$

$$\beta_2(1) = 0.064$$

$$\beta_2(2) = 0.192$$

$$\beta_3(1) = 0$$

$$\beta_3(2) = 0.4$$



$$P(\theta|\lambda) = 0.02048$$

$$\gamma_1(1) = \frac{0.5 \times 0.04096}{0.02048} = 1$$

$$\gamma_1(2) = \frac{0 \times 0.09216}{0.02048} = 0$$

$$\gamma_2(1) = \frac{0.08 \times 0.064}{0.02048} = 0.25$$

$$\gamma_2(2) = \frac{0.08 \times 0.192}{0.02048} = 0.75$$

$$\gamma_3(1) = \frac{0.0128 \times 0}{0.02048} = 0$$

$$\gamma_3(2) = \frac{0.0512 \times 0.4}{0.02048} = 1$$

$$\hat{\pi} = [1 \quad 0]$$

$$\hat{B} = \begin{bmatrix} & R & G & B & Y \\ \begin{matrix} R \\ G \\ B \\ Y \end{matrix} & \begin{bmatrix} 0.8 & 0.2 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix} \end{bmatrix}$$

$$\gamma_t(i) = \frac{\alpha_t(i) \beta_t(i)}{P(\theta|\lambda)}$$

$$\xi_t(i,j) = \frac{\alpha_{t+1}(j) a_{ij} b_j(o_{t+1}) \beta_t(j)}{P(\theta|\lambda)}$$

Considering only the non-zero prob. transitions:

$$\xi_2(1,1) = \frac{0.5 \times 0.8 \times 0.2 \times 0.064}{0.02048} = 0.25$$

$$\xi_2(1,2) = \frac{0.5 \times 0.2 \times 0.8 \times 0.192}{0.02048} = 0.75$$

$$\xi_3(1,2) = \frac{0.08 \times 0.2 \times 0.8 \times 0.4}{0.02048} = 0.25$$

$$\xi_3(2,2) = \frac{0.08 \times 0.6 \times 0.8 \times 0.4}{0.02048} = 0.75$$

$$\hat{A} = \begin{bmatrix} \frac{0.25}{1.25} & \frac{0.75+0.25}{1.25} \\ 0 & \frac{0.75}{0.75} \end{bmatrix} = \begin{bmatrix} 0.2 & 0.8 \\ 0 & 1 \end{bmatrix}$$