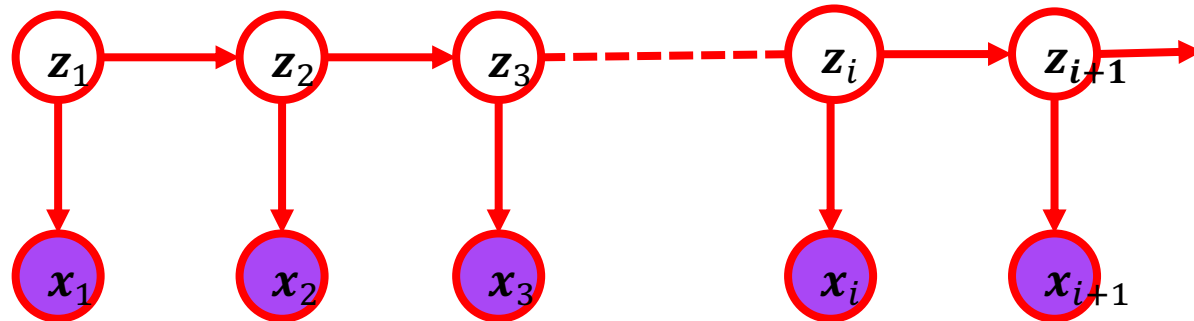


$$p(\mathbf{x}_1, \cdots, \mathbf{x}_N, \mathbf{z}_1, \cdots, \mathbf{z}_N) = p(\mathbf{z}_1) \left[ \prod_{i=2}^N p(\mathbf{z}_i | \mathbf{z}_{i-1}) \right] \left[ \prod_{i=1}^N p(\mathbf{x}_i | \mathbf{z}_i) \right]$$

$$p(\mathbf{X}, \mathbf{Z}) = p(\mathbf{z}_1) \left[ \prod_{i=2}^N p(\mathbf{z}_i | \mathbf{z}_{i-1}) \right] \left[ \prod_{j=1}^N p(\mathbf{x}_j | \mathbf{z}_j) \right]$$

$$p(\mathbf{X}, \mathbf{Z} | \boldsymbol{\theta}) = p(\mathbf{z}_1 | \boldsymbol{\pi}) \left[ \prod_{i=2}^N p(\mathbf{z}_i | \mathbf{z}_{i-1}, \mathbf{A}) \right] \left[ \prod_{j=1}^N p(\mathbf{x}_j | \mathbf{z}_j, \boldsymbol{\phi}) \right] \quad \boldsymbol{\theta} = \{\boldsymbol{\pi}, \mathbf{A}, \boldsymbol{\phi}\}$$

# HMM EM 알고리즘



$$\ln p(\mathbf{X}|\boldsymbol{\theta}) = \sum_{n=1}^N \ln p(\mathbf{x}_n|\boldsymbol{\theta}) = \sum_{n=1}^N \ln \sum_{\mathbf{z}_n} p(\mathbf{x}_n, \mathbf{z}_n|\boldsymbol{\theta})$$

$$= \sum_{n=1}^N \sum_{\mathbf{z}_n} q(\mathbf{z}_n) \ln \left\{ \frac{p(\mathbf{x}_n, \mathbf{z}_n|\boldsymbol{\theta})}{q(\mathbf{z}_n)} \right\} - \sum_{\mathbf{z}_n} \sum_{n=1}^N q(\mathbf{z}_n) \ln \left\{ \frac{p(\mathbf{z}_n|\mathbf{x}_n, \boldsymbol{\theta})}{q(\mathbf{z}_n)} \right\}$$

E-step :  $KL(q||p) = 0 \quad \longrightarrow \quad q(\mathbf{z}_n) = p(\mathbf{z}_n|\mathbf{x}_n, \boldsymbol{\theta}^{old}) \quad q(\mathbf{z}_n) \text{를 정의?}$

M-step :  $\boldsymbol{\theta}^{new} = \arg \max_{\boldsymbol{\theta}} \mathcal{L}(q, \boldsymbol{\theta})$

$$\mathcal{L}(q, \boldsymbol{\theta}) = \sum_{n=1}^N \sum_{\mathbf{z}_n} p(\mathbf{z}_n|\mathbf{x}_n, \boldsymbol{\theta}^{old}) \ln p(\mathbf{x}_n, \mathbf{z}_n|\boldsymbol{\theta}) + const$$

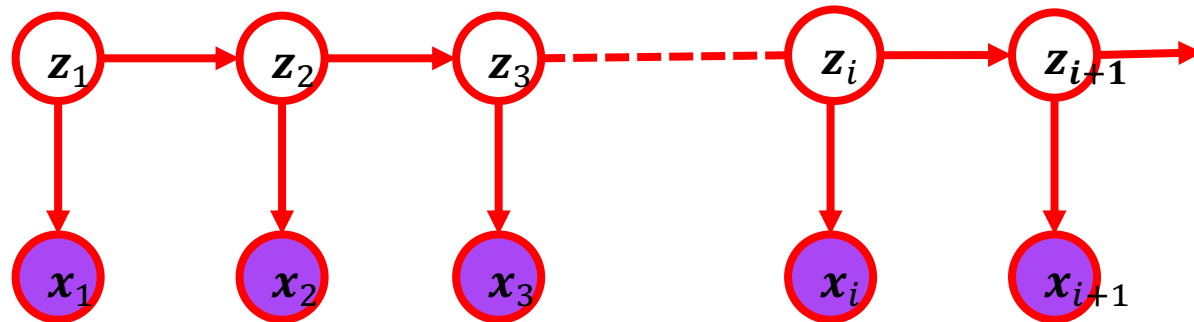
## HMM EM 알고리즘

$$\ln p(\mathbf{X}|\boldsymbol{\theta}) = \sum_{n=1}^N \ln p(\mathbf{x}_n|\boldsymbol{\theta}) = \sum_{n=1}^N \ln \sum_{\mathbf{z}_n} p(\mathbf{x}_n, \mathbf{z}_n|\boldsymbol{\theta})$$

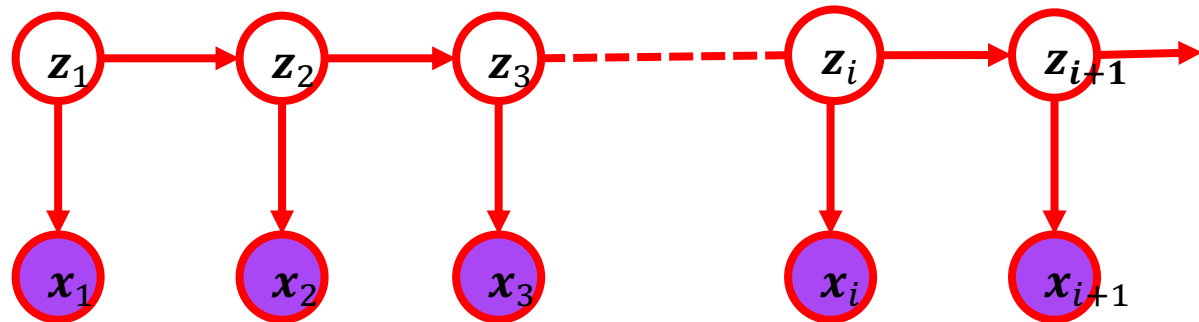
$$= \sum_{n=1}^N \sum_{\mathbf{z}_n} q(\mathbf{z}_n) \ln \left\{ \frac{p(\mathbf{x}_n, \mathbf{z}_n|\boldsymbol{\theta})}{q(\mathbf{z}_n)} \right\} - \sum_{\mathbf{z}_n} \sum_{n=1}^N q(\mathbf{z}_n) \ln \left\{ \frac{p(\mathbf{z}_n|\mathbf{x}_n, \boldsymbol{\theta})}{q(\mathbf{z}_n)} \right\}$$

$$p(\mathbf{X}, \mathbf{Z}) = p(\mathbf{z}_1) \left[ \prod_{i=2}^N p(\mathbf{z}_i|\mathbf{z}_{i-1}) \right] \left[ \prod_{j=1}^N p(\mathbf{x}_j|\mathbf{z}_j) \right]$$

$$\ln p(\mathbf{X}, \mathbf{Z}) = \ln p(\mathbf{z}_1) + \sum_{i=2}^N \ln p(\mathbf{z}_i|\mathbf{z}_{i-1}) + \sum_{j=1}^N \ln p(\mathbf{x}_j|\mathbf{z}_j)$$



# HMM EM 알고리즘



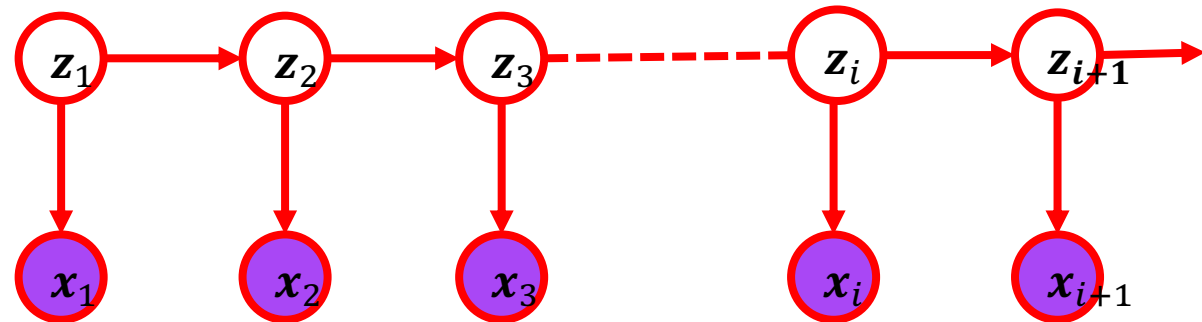
$$p(\mathbf{X}, \mathbf{Z}) = p(\mathbf{z}_1) \left[ \prod_{i=2}^N p(\mathbf{z}_i | \mathbf{z}_{i-1}) \right] \left[ \prod_{j=1}^N p(\mathbf{x}_j | \mathbf{z}_j) \right]$$

$$\ln \left\{ \sum_{\mathbf{Z}} p(\mathbf{X}, \mathbf{Z} | \boldsymbol{\theta}) \right\} = \ln \left\{ \sum_{\mathbf{Z}} p(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_N, \mathbf{z}_1, \mathbf{z}_2, \dots, \mathbf{z}_N | \boldsymbol{\theta}) \right\} = \ln \left\{ \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \dots \sum_{\mathbf{z}_N} p(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_N, \mathbf{z}_1, \mathbf{z}_2, \dots, \mathbf{z}_N | \boldsymbol{\theta}) \right\}$$

$$Q(\boldsymbol{\theta}, \boldsymbol{\theta}^{old}) = \sum_{\mathbf{Z}} p(\mathbf{Z} | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{X}, \mathbf{Z} | \boldsymbol{\theta})$$

$$Q(\boldsymbol{\theta}, \boldsymbol{\theta}^{old}) = \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \dots \sum_{\mathbf{z}_N} p(\mathbf{Z} | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln \left\{ p(\mathbf{z}_1 | \boldsymbol{\theta}) \left[ \prod_{i=2}^N p(\mathbf{z}_i | \mathbf{z}_{i-1}, \boldsymbol{\theta}) \right] \left[ \prod_{j=1}^N p(\mathbf{x}_j | \mathbf{z}_j, \boldsymbol{\theta}) \right] \right\}$$

# HMM EM 알고리즘



$$\gamma(\mathbf{z}_n) = p(\mathbf{z}_n | \mathbf{X}, \boldsymbol{\theta}^{old})$$

$$\xi(\mathbf{z}_{n-1}, \mathbf{z}_n) = p(\mathbf{z}_{n-1}, \mathbf{z}_n | \mathbf{X}, \boldsymbol{\theta}^{old})$$

$$\begin{aligned} Q(\boldsymbol{\theta}, \boldsymbol{\theta}^{old}) &= \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \cdots \sum_{\mathbf{z}_N} p(\mathbf{z}_1, \mathbf{z}_2, \cdots, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln \left\{ p(\mathbf{z}_1 | \boldsymbol{\theta}) \left[ \prod_{i=2}^N p(\mathbf{z}_i | \mathbf{z}_{i-1}, \boldsymbol{\theta}) \right] \left[ \prod_{j=1}^N p(\mathbf{x}_j | \mathbf{z}_j, \boldsymbol{\theta}) \right] \right\} \\ &= \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \cdots \sum_{\mathbf{z}_N} p(\mathbf{z}_1, \mathbf{z}_2, \cdots, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \left\{ \ln p(\mathbf{z}_1 | \boldsymbol{\theta}) + \sum_{i=2}^N \ln p(\mathbf{z}_i | \mathbf{z}_{i-1}, \boldsymbol{\theta}) + \sum_{j=1}^N \ln p(\mathbf{x}_j | \mathbf{z}_j, \boldsymbol{\theta}) \right\} \\ &= \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \cdots \sum_{\mathbf{z}_N} p(\mathbf{z}_1, \mathbf{z}_2, \cdots, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{z}_1 | \boldsymbol{\theta}) + \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \cdots \sum_{\mathbf{z}_N} p(\mathbf{z}_1, \mathbf{z}_2, \cdots, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \sum_{i=2}^N \ln p(\mathbf{z}_i | \mathbf{z}_{i-1}, \boldsymbol{\theta}) \\ &\quad + \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \cdots \sum_{\mathbf{z}_N} p(\mathbf{z}_1, \mathbf{z}_2, \cdots, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \sum_{j=1}^N \ln p(\mathbf{x}_j | \mathbf{z}_j, \boldsymbol{\theta}) \end{aligned}$$

$$\sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \cdots \sum_{\mathbf{z}_N} p(\mathbf{z}_1, \mathbf{z}_2, \cdots, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{z}_1 | \boldsymbol{\theta})$$

$$= \sum_{\mathbf{z}_1} \ln p(\mathbf{z}_1 | \boldsymbol{\theta}) \sum_{\mathbf{z}_2} \cdots \sum_{\mathbf{z}_N} p(\mathbf{z}_1, \mathbf{z}_2, \cdots, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old})$$

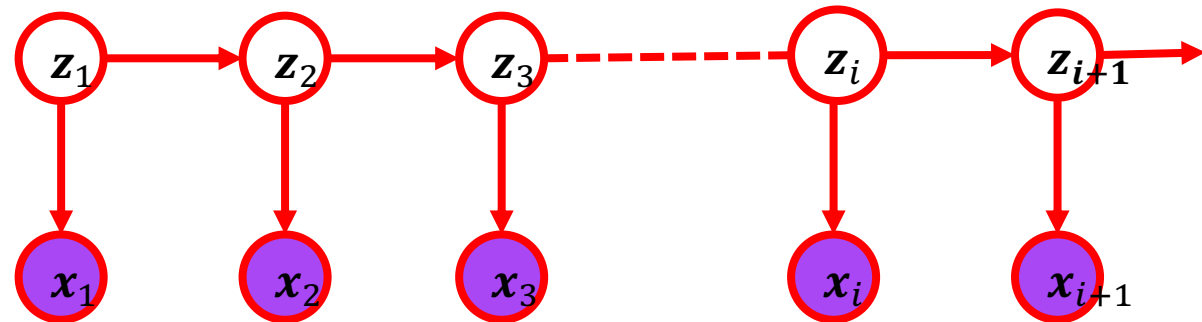
$$= \sum_{\mathbf{z}_1} p(\mathbf{z}_1 | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{z}_1 | \boldsymbol{\theta})$$



$$\begin{aligned}
& \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \cdots \sum_{\mathbf{z}_N} p(\mathbf{z}_1, \mathbf{z}_2, \cdots, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \sum_{i=2}^N \ln p(\mathbf{z}_i | \mathbf{z}_{i-1}, \boldsymbol{\theta}) \\
&= \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \cdots \sum_{\mathbf{z}_N} p(\mathbf{z}_1, \mathbf{z}_2, \cdots, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \{ \ln p(\mathbf{z}_2 | \mathbf{z}_1, \boldsymbol{\theta}) + \ln p(\mathbf{z}_3 | \mathbf{z}_2, \boldsymbol{\theta}) + \cdots + \ln p(\mathbf{z}_N | \mathbf{z}_{N-1}, \boldsymbol{\theta}) \} \\
&= \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \cdots \sum_{\mathbf{z}_N} p(\mathbf{z}_1, \mathbf{z}_2, \cdots, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{z}_2 | \mathbf{z}_1, \boldsymbol{\theta}) + \cdots + \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \cdots \sum_{\mathbf{z}_N} p(\mathbf{z}_1, \mathbf{z}_2, \cdots, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{z}_N | \mathbf{z}_{N-1}, \boldsymbol{\theta}) \\
&= \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} p(\mathbf{z}_1, \mathbf{z}_2 | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{z}_2 | \mathbf{z}_1, \boldsymbol{\theta}) + \sum_{\mathbf{z}_2} \sum_{\mathbf{z}_3} p(\mathbf{z}_2, \mathbf{z}_3 | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{z}_3 | \mathbf{z}_2, \boldsymbol{\theta}) + \\
&\quad \cdots + \sum_{\mathbf{z}_{N-1}} \sum_{\mathbf{z}_N} p(\mathbf{z}_{N-1}, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{z}_N | \mathbf{z}_{N-1}, \boldsymbol{\theta}) \\
&= \sum_{n=2}^N \sum_{\mathbf{z}_{n-1}} \sum_{\mathbf{z}_n} p(\mathbf{z}_{n-1}, \mathbf{z}_n | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{z}_n | \mathbf{z}_{n-1}, \boldsymbol{\theta})
\end{aligned}$$

$$\begin{aligned}
& \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \cdots \sum_{\mathbf{z}_N} p(\mathbf{z}_1, \mathbf{z}_2, \cdots, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \sum_{j=1}^N \ln p(\mathbf{x}_j | \mathbf{z}_j, \boldsymbol{\theta}) \\
&= \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \cdots \sum_{\mathbf{z}_N} p(\mathbf{z}_1, \mathbf{z}_2, \cdots, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \{ \ln p(\mathbf{x}_1 | \mathbf{z}_1, \boldsymbol{\theta}) + \ln p(\mathbf{x}_2 | \mathbf{z}_2, \boldsymbol{\theta}) + \cdots + \ln p(\mathbf{x}_N | \mathbf{z}_N, \boldsymbol{\theta}) \} \\
&= \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \cdots \sum_{\mathbf{z}_N} p(\mathbf{z}_1, \mathbf{z}_2, \cdots, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{x}_1 | \mathbf{z}_1, \boldsymbol{\theta}) + \cdots + \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \cdots \sum_{\mathbf{z}_N} p(\mathbf{z}_1, \mathbf{z}_2, \cdots, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{x}_N | \mathbf{z}_N, \boldsymbol{\theta}) \\
&= \sum_{\mathbf{z}_1} p(\mathbf{z}_1 | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{x}_1 | \mathbf{z}_1, \boldsymbol{\theta}) + \sum_{\mathbf{z}_2} p(\mathbf{z}_2 | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{x}_2 | \mathbf{z}_2, \boldsymbol{\theta}) + \cdots + \sum_{\mathbf{z}_N} p(\mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{x}_N | \mathbf{z}_N, \boldsymbol{\theta}) \\
&= \sum_{n=1}^N \sum_{\mathbf{z}_n} p(\mathbf{z}_n | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{x}_n | \mathbf{z}_n, \boldsymbol{\theta})
\end{aligned}$$

# HMM EM 알고리즘



$$\gamma(\mathbf{z}_n) = p(\mathbf{z}_n | \mathbf{X}, \boldsymbol{\theta}^{old})$$

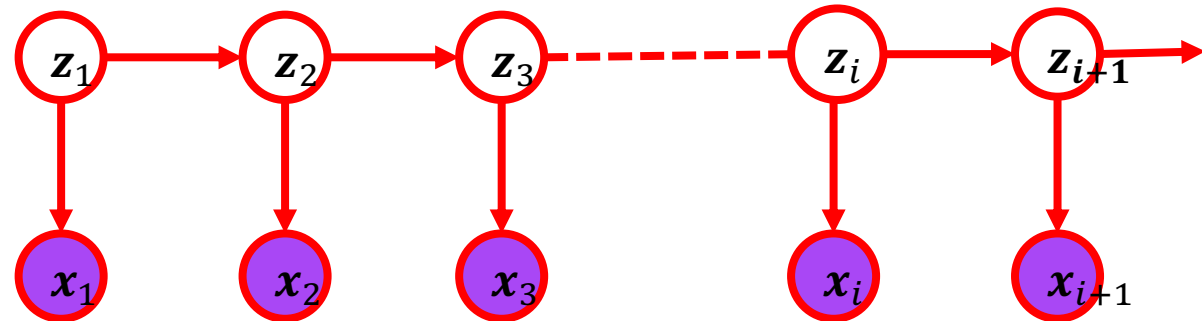
$$\xi(\mathbf{z}_{n-1}, \mathbf{z}_n) = p(\mathbf{z}_{n-1}, \mathbf{z}_n | \mathbf{X}, \boldsymbol{\theta}^{old})$$

$$Q(\boldsymbol{\theta}, \boldsymbol{\theta}^{old}) = \sum_{\mathbf{z}_1} \sum_{\mathbf{z}_2} \dots \sum_{\mathbf{z}_N} p(\mathbf{z}_1, \mathbf{z}_2, \dots, \mathbf{z}_N | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln \left\{ p(\mathbf{z}_1 | \boldsymbol{\theta}) \left[ \prod_{i=2}^N p(\mathbf{z}_i | \mathbf{z}_{i-1}, \boldsymbol{\theta}) \right] \left[ \prod_{j=1}^N p(\mathbf{x}_j | \mathbf{z}_j, \boldsymbol{\theta}) \right] \right\}$$

$$= \sum_{\mathbf{z}_1} p(\mathbf{z}_1 | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{z}_1 | \boldsymbol{\theta}) + \sum_{n=2}^N \sum_{\mathbf{z}_{n-1}} \sum_{\mathbf{z}_n} p(\mathbf{z}_{n-1}, \mathbf{z}_n | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{z}_n | \mathbf{z}_{n-1}, \boldsymbol{\theta}) + \sum_{n=1}^N \sum_{\mathbf{z}_n} p(\mathbf{z}_n | \mathbf{X}, \boldsymbol{\theta}^{old}) \ln p(\mathbf{x}_n | \mathbf{z}_n, \boldsymbol{\theta})$$

$$= \sum_{\mathbf{z}_1} \gamma(\mathbf{z}_1) \ln p(\mathbf{z}_1 | \boldsymbol{\theta}) + \sum_{n=2}^N \sum_{\mathbf{z}_{n-1}} \sum_{\mathbf{z}_n} \xi(\mathbf{z}_{n-1}, \mathbf{z}_n) \ln p(\mathbf{z}_n | \mathbf{z}_{n-1}, \boldsymbol{\theta}) + \sum_{n=1}^N \sum_{\mathbf{z}_n} \gamma(\mathbf{z}_n) \ln p(\mathbf{x}_n | \mathbf{z}_n, \boldsymbol{\theta})$$

# HMM EM 알고리즘



$$\gamma(\mathbf{z}_n) = p(\mathbf{z}_n | \mathbf{X}, \boldsymbol{\theta}^{old})$$

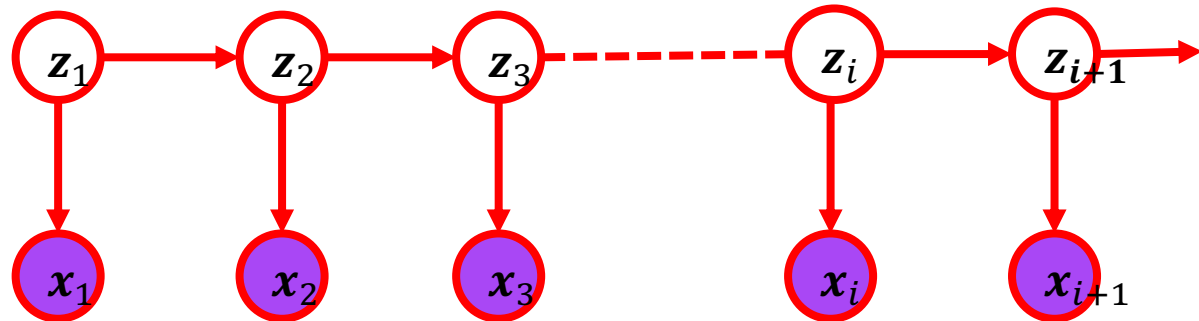
$$\xi(\mathbf{z}_{n-1}, \mathbf{z}_n) = p(\mathbf{z}_{n-1}, \mathbf{z}_n | \mathbf{X}, \boldsymbol{\theta}^{old})$$

$$p(\mathbf{z}_1 | \boldsymbol{\pi}) = \prod_{i=1}^K \pi_i^{z_{1i}} \quad p(\mathbf{z}_n | \mathbf{z}_{n-1}, \mathbf{A}) = \prod_{i=1}^K \prod_{j=1}^K A_{j,i}^{z_{n-1,j} z_{n,i}} \quad p(\mathbf{x}_n | \mathbf{z}_n, \boldsymbol{\phi}) = \prod_{i=1}^K p(\mathbf{x}_n | \phi_i)^{z_{n,i}}$$

$$Q(\boldsymbol{\theta}, \boldsymbol{\theta}^{old}) = \sum_{\mathbf{z}_1} \gamma(\mathbf{z}_1) \ln p(\mathbf{z}_1 | \boldsymbol{\theta}) + \sum_{n=2}^N \sum_{\mathbf{z}_{n-1}} \sum_{\mathbf{z}_n} \xi(\mathbf{z}_{n-1}, \mathbf{z}_n) \ln p(\mathbf{z}_n | \mathbf{z}_{n-1}, \boldsymbol{\theta}) + \sum_{n=1}^N \sum_{\mathbf{z}_n} \gamma(\mathbf{z}_n) \ln p(\mathbf{x}_n | \mathbf{z}_n, \boldsymbol{\theta})$$

$$= \sum_{i=1}^K \left\{ \sum_{\mathbf{z}_1} \gamma(\mathbf{z}_1) z_{1i} \right\} \ln \pi_i + \sum_{n=2}^N \sum_{i=1}^K \sum_{j=1}^K \left\{ \sum_{\mathbf{z}_{n-1}} \sum_{\mathbf{z}_n} \xi(\mathbf{z}_{n-1}, \mathbf{z}_n) z_{n-1,j} z_{n,i} \right\} \ln A_{j,i} + \sum_{n=1}^N \sum_{i=1}^K \left\{ \sum_{\mathbf{z}_n} \gamma(\mathbf{z}_n) z_{n,i} \right\} \ln p(\mathbf{x}_n | \phi_i)$$

## HMM EM 알고리즘



$$\gamma(z_{n,i}) = \mathbb{E}[z_{n,i}] = \sum_{z_n} \gamma(z_n) z_{n,i}$$

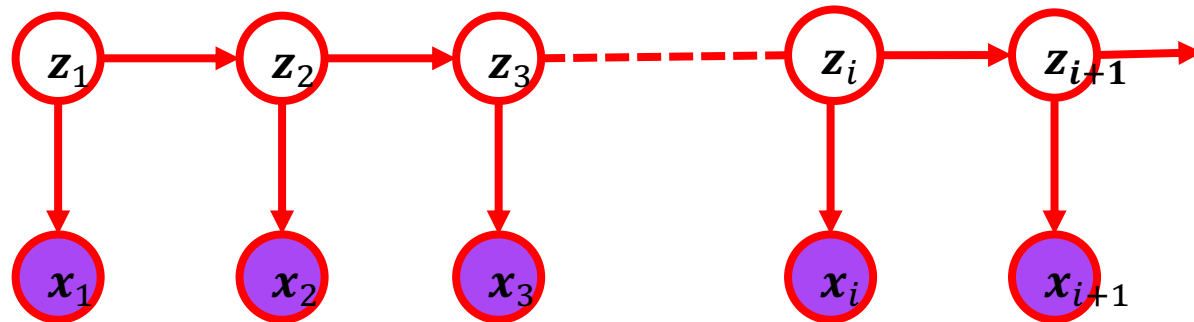
$$\xi(z_{n-1,j} z_{n,i}) = \mathbb{E}[z_{n-1,j} z_{n,i}] = \sum_{z_{n-1}} \sum_{z_n} \xi(z_{n-1}, z_n) z_{n-1,j} z_{n,i}$$

$$\begin{aligned} Q(\theta, \theta^{old}) &= \sum_{i=1}^K \left\{ \sum_{z_1} \gamma(z_1) z_{1,i} \right\} \ln \pi_i + \sum_{n=2}^N \sum_{i=1}^K \sum_{j=1}^K \left\{ \sum_{z_{n-1}} \sum_{z_n} \xi(z_{n-1}, z_n) z_{n-1,j} z_{n,i} \right\} \ln A_{j,i} + \sum_{n=1}^N \sum_{i=1}^K \left\{ \sum_{z_n} \gamma(z_n) z_{n,i} \right\} \ln p(x_n | \phi_i) \\ &= \sum_{i=1}^K \gamma(z_{1,i}) \ln \pi_i + \sum_{n=2}^N \sum_{i=1}^K \sum_{j=1}^K \xi(z_{n-1,j} z_{n,i}) \ln A_{j,i} + \sum_{n=1}^N \sum_{i=1}^K \gamma(z_{n,i}) \ln p(x_n | \phi_i) \end{aligned}$$

# HMM EM 알고리즘

$$p(\mathbf{X}, \mathbf{Z} | \boldsymbol{\theta}) = p(\mathbf{z}_1 | \boldsymbol{\pi}) \left[ \prod_{i=2}^N p(\mathbf{z}_i | \mathbf{z}_{i-1}, \mathbf{A}) \right] \left[ \prod_{j=1}^N p(\mathbf{x}_j | \mathbf{z}_j, \boldsymbol{\phi}) \right]$$

$$\boldsymbol{\theta} = \{\boldsymbol{\pi}, \mathbf{A}, \boldsymbol{\phi}\}$$



E-step :

$$\gamma(z_{n,i}) = \mathbb{E}[z_{n,i}] = \sum_{\mathbf{z}_n} \gamma(\mathbf{z}_n) z_{n,i} = p(z_{n,i} | \mathbf{X}, \boldsymbol{\theta}^{old})$$

$$\xi(z_{n-1,j}, z_{n,i}) = \mathbb{E}[z_{n-1,j}, z_{n,i}] = \sum_{\mathbf{z}_{n-1}} \sum_{\mathbf{z}_n} \xi(\mathbf{z}_{n-1}, \mathbf{z}_n) z_{n-1,j} z_{n,i} = p(z_{n-1,j}, z_{n,i} | \mathbf{X}, \boldsymbol{\theta}^{old})$$

M-step :

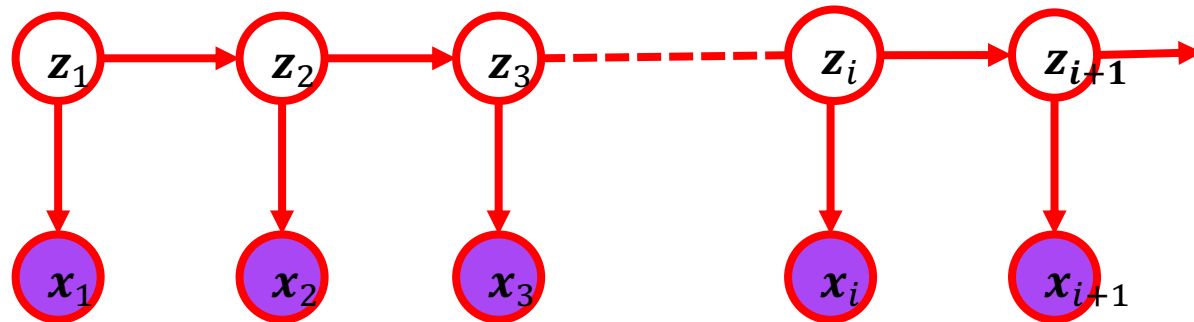
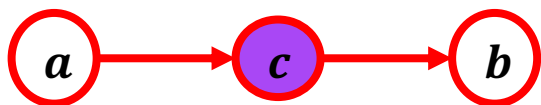
$$Q(\boldsymbol{\theta}, \boldsymbol{\theta}^{old}) = \sum_{i=1}^K \gamma(z_{1,i}) \ln \pi_i + \sum_{n=2}^N \sum_{i=1}^K \sum_{j=1}^K \xi(z_{n-1,j}, z_{n,i}) \ln A_{j,i} + \sum_{n=1}^N \sum_{i=1}^K \gamma(z_{n,i}) \ln p(\mathbf{x}_n | \phi_i)$$

알파 베타 알고리즘

$c$ 는 머리 대 꼬리(*head-to-tail*) 노드

$$p(a, b|c) = \frac{p(a, b, c)}{p(c)} = \frac{p(b|c)p(c|a)p(a)}{p(c)}$$

$$= p(a|c)p(b|c)$$



$$p(x_1, x_2, \dots, x_n, x_{n+1} \dots, x_N | z_n) = p(x_1, x_2, \dots, x_n | z_n) p(x_{n+1} \dots, x_N | z_n)$$

$$\gamma(z_n) = p(z_n | X) = \frac{p(X | z_n) p(z_n)}{p(X)} = \frac{p(x_1, x_2, \dots, x_n | z_n) p(x_{n+1} \dots, x_N | z_n) p(z_n)}{p(X)}$$

$$= \frac{p(x_1, x_2, \dots, x_n, z_n) p(x_{n+1} \dots, x_N | z_n)}{p(X)} = \frac{\alpha(z_n) \beta(z_n)}{p(X)}$$

$$\alpha(z_n) \equiv p(x_1, x_2, \dots, x_n, z_n)$$

$$\beta(z_n) \equiv p(x_{n+1} \dots, x_N | z_n)$$

## 알파 베타 알고리즘

$$\alpha(z_n) \equiv p(x_1, x_2, \dots, x_n, z_n)$$

$$\beta(z_n) \equiv p(x_{n+1} \dots, x_N | z_n)$$

$$\gamma(z_n) = p(z_n | X) = \frac{p(X | z_n) p(z_n)}{p(X)} = \frac{\alpha(z_n) \beta(z_n)}{p(X)}$$

$$\alpha(z_n) = p(x_1, x_2, \dots, x_n, z_n)$$

$$= p(x_1, x_2, \dots, x_n | z_n) p(z_n)$$

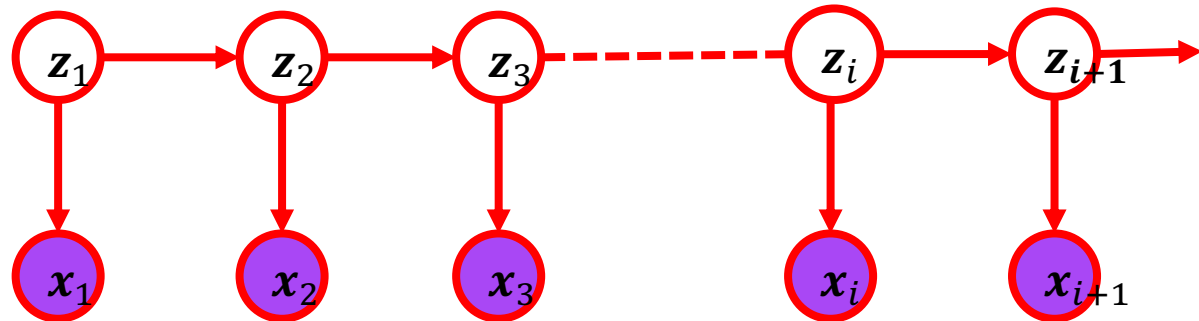
$$= p(x_n | z_n) p(x_1, x_2, \dots, x_{n-1} | z_n) p(z_n) \rightarrow d\text{분리이용}$$

$$= p(x_n | z_n) p(x_1, x_2, \dots, x_{n-1}, z_n)$$

$$= p(x_n | z_n) \sum_{z_{n-1}} p(x_1, x_2, \dots, x_{n-1}, z_{n-1}, z_n)$$

$$= p(x_n | z_n) \sum_{z_{n-1}} p(x_1, x_2, \dots, x_{n-1}, z_n | z_{n-1}) p(z_{n-1})$$

$$= p(x_n | z_n) \sum_{z_{n-1}} p(x_1, x_2, \dots, x_{n-1} | z_{n-1}) p(z_n | z_{n-1}) p(z_{n-1}) \rightarrow d\text{분리이용}$$



$$= p(x_n | z_n) \sum_{z_{n-1}} p(x_1, x_2, \dots, x_{n-1}, z_{n-1}) p(z_n | z_{n-1})$$

$$\alpha(z_n) = p(x_n | z_n) \sum_{z_{n-1}} \alpha(z_{n-1}) p(z_n | z_{n-1})$$



## 알파 베타 알고리즘

$$\alpha(\mathbf{z}_n) \equiv p(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n, \mathbf{z}_n)$$

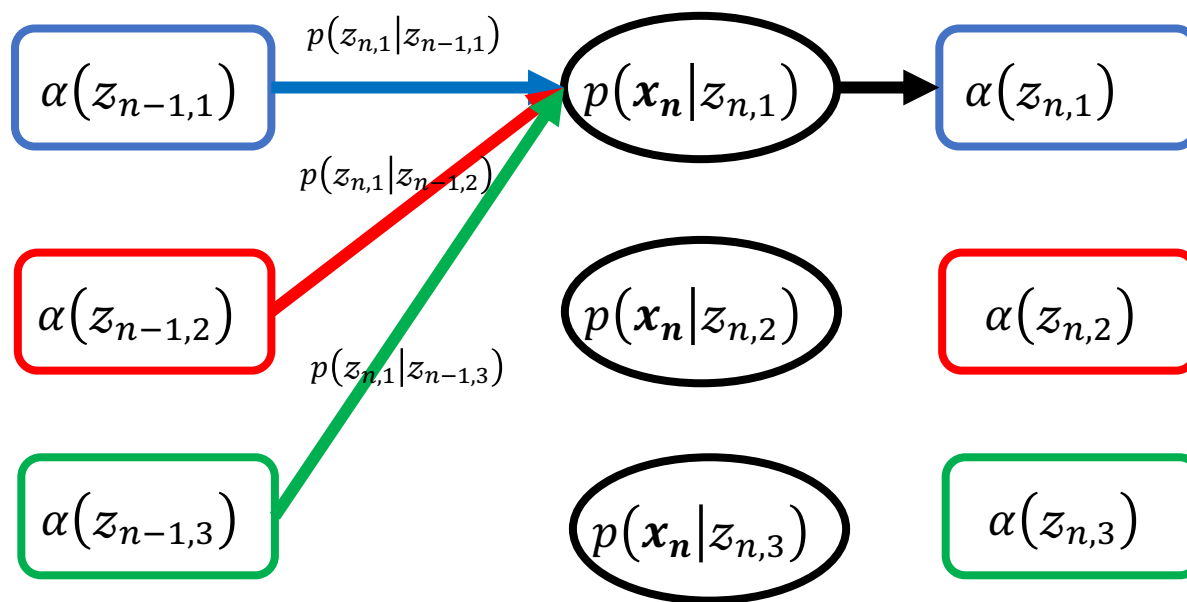
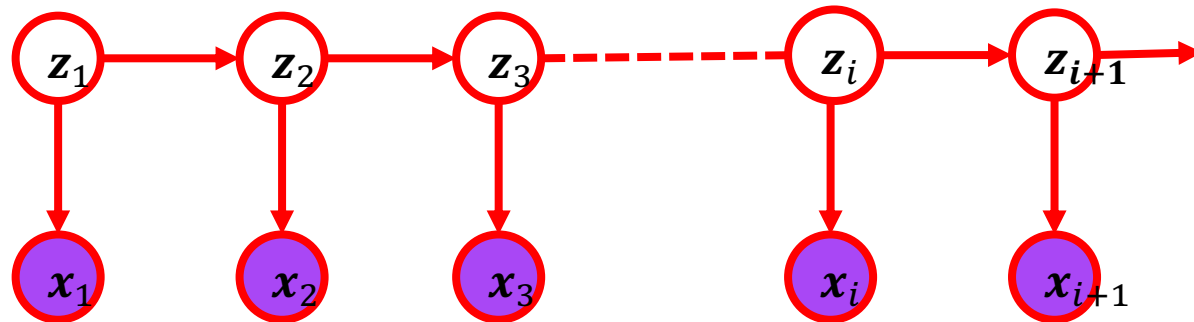
$$\beta(\mathbf{z}_n) \equiv p(\mathbf{x}_{n+1} \dots, \mathbf{x}_N | \mathbf{z}_n)$$

$$\gamma(\mathbf{z}_n) = p(\mathbf{z}_n | \mathbf{X}) = \frac{p(\mathbf{X} | \mathbf{z}_n) p(\mathbf{z}_n)}{p(\mathbf{X})} = \frac{\alpha(\mathbf{z}_n) \beta(\mathbf{z}_n)}{p(\mathbf{X})}$$

$$\alpha(\mathbf{z}_n) = p(\mathbf{x}_n | \mathbf{z}_n) \sum_{\mathbf{z}_{n-1}} \alpha(\mathbf{z}_{n-1}) p(\mathbf{z}_n | \mathbf{z}_{n-1})$$

$$\alpha(\mathbf{z}_1) = p(\mathbf{x}_1, \mathbf{z}_1)$$

$$\alpha(\mathbf{z}_{n,i}) = p(\mathbf{x}_n | \mathbf{z}_{n,i}) \sum_{j=1}^K \alpha(\mathbf{z}_{n-1,j}) p(\mathbf{z}_{n,i} | \mathbf{z}_{n-1,j})$$



## 알파 베타 알고리즘

$$\alpha(\mathbf{z}_n) \equiv p(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n, \mathbf{z}_n)$$

$$\beta(\mathbf{z}_n) \equiv p(\mathbf{x}_{n+1} \dots, \mathbf{x}_N | \mathbf{z}_n)$$

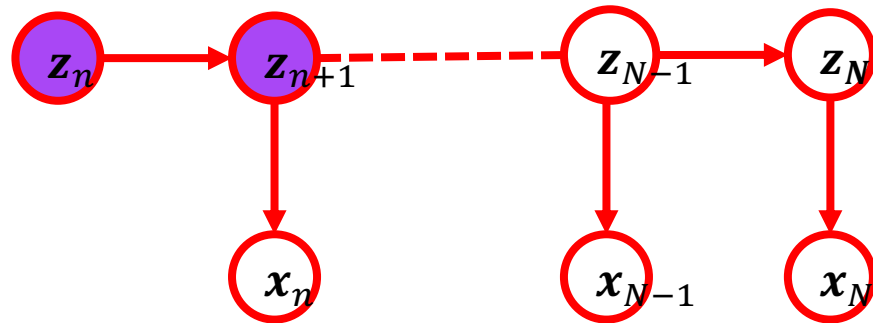
$$\gamma(\mathbf{z}_n) = p(\mathbf{z}_n | \mathbf{X}) = \frac{p(\mathbf{X} | \mathbf{z}_n) p(\mathbf{z}_n)}{p(\mathbf{X})} = \frac{\alpha(\mathbf{z}_n) \beta(\mathbf{z}_n)}{p(\mathbf{X})}$$

$$\beta(\mathbf{z}_n) = p(\mathbf{x}_{n+1} \dots, \mathbf{x}_N | \mathbf{z}_n) = \sum_{\mathbf{z}_{n+1}} p(\mathbf{x}_{n+1} \dots, \mathbf{x}_N, \mathbf{z}_{n+1} | \mathbf{z}_n)$$

$$= \sum_{\mathbf{z}_{n+1}} p(\mathbf{x}_{n+1} \dots, \mathbf{x}_N | \mathbf{z}_n, \mathbf{z}_{n+1}) p(\mathbf{z}_{n+1} | \mathbf{z}_n) \rightarrow \text{d분리이용}$$

$$= \sum_{\mathbf{z}_{n+1}} p(\mathbf{x}_{n+2} \dots, \mathbf{x}_N | \mathbf{z}_{n+1}) p(\mathbf{x}_{n+1} | \mathbf{z}_{n+1}) p(\mathbf{z}_{n+1} | \mathbf{z}_n)$$

$$\beta(\mathbf{z}_n) = \sum_{\mathbf{z}_{n+1}} \beta(\mathbf{z}_{n+1}) p(\mathbf{x}_{n+1} | \mathbf{z}_{n+1}) p(\mathbf{z}_{n+1} | \mathbf{z}_n)$$



$$p(\mathbf{x}_{n+1} \dots, \mathbf{x}_N | \mathbf{z}_n, \mathbf{z}_{n+1}) = \frac{p(\mathbf{x}_{n+1}, \dots, \mathbf{x}_N, \mathbf{z}_n, \mathbf{z}_{n+1})}{p(\mathbf{z}_n, \mathbf{z}_{n+1})}$$

$$= \frac{\frac{p(\mathbf{x}_{n+1}, \dots, \mathbf{x}_N, \mathbf{z}_n, \mathbf{z}_{n+1})}{p(\mathbf{z}_{n+1})}}{\frac{p(\mathbf{z}_n, \mathbf{z}_{n+1})}{p(\mathbf{z}_{n+1})}}$$

$$= \frac{p(\mathbf{x}_{n+1} \dots, \mathbf{x}_N, \mathbf{z}_n | \mathbf{z}_{n+1})}{p(\mathbf{z}_n | \mathbf{z}_{n+1})}$$

$$= \frac{p(\mathbf{x}_{n+1} \dots, \mathbf{x}_N | \mathbf{z}_{n+1}) p(\mathbf{z}_n | \mathbf{z}_{n+1})}{p(\mathbf{z}_n | \mathbf{z}_{n+1})}$$

## 알파 베타 알고리즘

$$\alpha(\mathbf{z}_n) \equiv p(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n, \mathbf{z}_n)$$

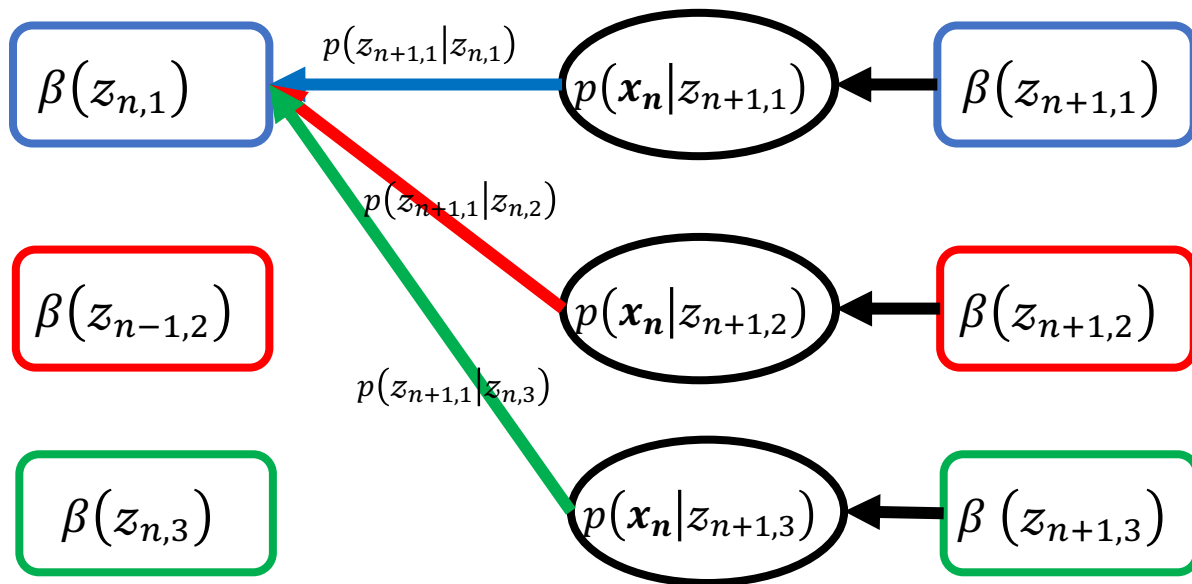
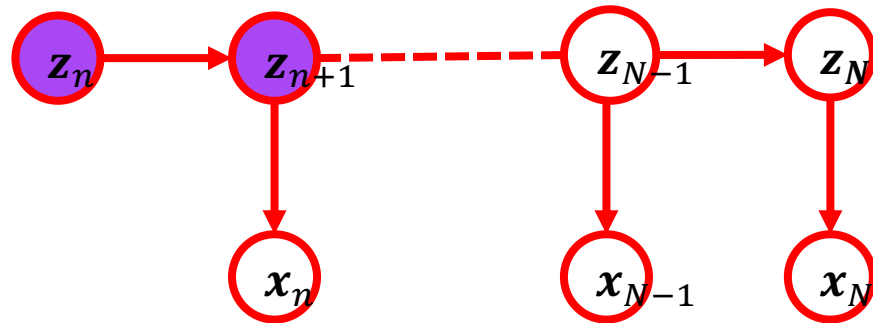
$$\beta(\mathbf{z}_n) \equiv p(\mathbf{x}_{n+1} \dots, \mathbf{x}_N | \mathbf{z}_n)$$

$$\gamma(\mathbf{z}_n) = p(\mathbf{z}_n | \mathbf{X}) = \frac{p(\mathbf{X} | \mathbf{z}_n) p(\mathbf{z}_n)}{p(\mathbf{X})} = \frac{\alpha(\mathbf{z}_n) \beta(\mathbf{z}_n)}{p(\mathbf{X})}$$

$$\beta(\mathbf{z}_n) = \sum_{\mathbf{z}_{n+1}} \beta(\mathbf{z}_{n+1}) p(\mathbf{x}_{n+1} | \mathbf{z}_{n+1}) p(\mathbf{z}_{n+1} | \mathbf{z}_n)$$

$$\beta(\mathbf{z}_N) = \mathbf{1}$$

$$\beta(\mathbf{z}_{n,i}) = \sum_{j=1}^K \beta(\mathbf{z}_{n+1,j}) p(\mathbf{x}_{n+1} | \mathbf{z}_{n+1,j}) p(\mathbf{z}_{n+1,j} | \mathbf{z}_{n,i})$$



## 알파 베타 알고리즘

$$\alpha(\mathbf{z}_n) \equiv p(\mathbf{x}_1, \mathbf{x}_2, \cdots, \mathbf{x}_n, \mathbf{z}_n)$$

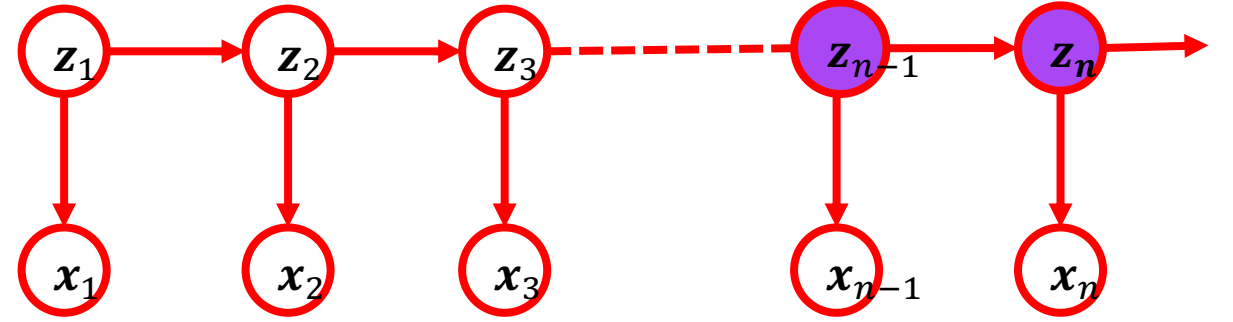
$$\beta(\mathbf{z}_n) \equiv p(\mathbf{x}_{n+1} \cdots, \mathbf{x}_N | \mathbf{z}_n)$$

$$\begin{aligned} \xi(\mathbf{z}_{n-1}, \mathbf{z}_n) &= p(\mathbf{z}_{n-1}, \mathbf{z}_n | \mathbf{X}) = \frac{p(\mathbf{X}, \mathbf{z}_{n-1}, \mathbf{z}_n)}{p(\mathbf{X})} = \frac{p(\mathbf{X} | \mathbf{z}_{n-1}, \mathbf{z}_n) p(\mathbf{z}_{n-1}, \mathbf{z}_n)}{p(\mathbf{X})} \\ &= \frac{p(\mathbf{X} | \mathbf{z}_{n-1}, \mathbf{z}_n) p(\mathbf{z}_n | \mathbf{z}_{n-1}) p(\mathbf{z}_{n-1})}{p(\mathbf{X})} \\ &= \frac{p(\mathbf{x}_1, \mathbf{x}_2, \cdots, \mathbf{x}_{n-1} | \mathbf{z}_{n-1}) p(\mathbf{x}_n | \mathbf{z}_n) p(\mathbf{x}_{n+1}, \mathbf{x}_{n+2}, \cdots, \mathbf{x}_N | \mathbf{z}_n) p(\mathbf{z}_n | \mathbf{z}_{n-1}) p(\mathbf{z}_{n-1})}{p(\mathbf{X})} \\ &= \frac{\alpha(\mathbf{z}_{n-1}) p(\mathbf{x}_n | \mathbf{z}_n) p(\mathbf{z}_n | \mathbf{z}_{n-1}) \beta(\mathbf{z}_n)}{p(\mathbf{X})} \end{aligned}$$

참고0

참고1

$$\rightarrow p(x_1, x_2, \dots, x_n | z_{n-1}, z_n)$$



$$= p(x_1, x_2, \dots, x_{n-1} | z_{n-1}, z_n) p(x_n | z_{n-1}, z_n) \rightarrow d\text{분리이용}$$

$$= \frac{p(x_1, x_2, \dots, x_{n-1}, z_{n-1}, z_n)}{p(z_{n-1}, z_n)} \frac{p(x_n, z_{n-1}, z_n)}{p(z_{n-1}, z_n)}$$

$$= \frac{\frac{p(x_1, x_2, \dots, x_{n-1}, z_{n-1}, z_n)}{p(z_{n-1})}}{\frac{p(z_{n-1}, z_n)}{p(z_{n-1})}} \frac{\frac{p(x_n, z_{n-1}, z_n)}{p(z_n)}}{\frac{p(z_{n-1}, z_n)}{p(z_n)}} = \frac{p(x_1, x_2, \dots, x_{n-1}, z_n | z_{n-1})}{p(z_n | z_{n-1})} \frac{p(x_n, z_{n-1} | z_n)}{p(z_{n-1} | z_n)}$$

$$= \frac{p(x_1, x_2, \dots, x_{n-1} | z_{n-1}) p(z_n | z_{n-1})}{p(z_n | z_{n-1})} \frac{p(x_n | z_n) p(z_{n-1} | z_n)}{p(z_{n-1} | z_n)}$$

$$= p(x_1, x_2, \dots, x_{n-1} | z_{n-1}) p(x_n | z_n)$$

참고2

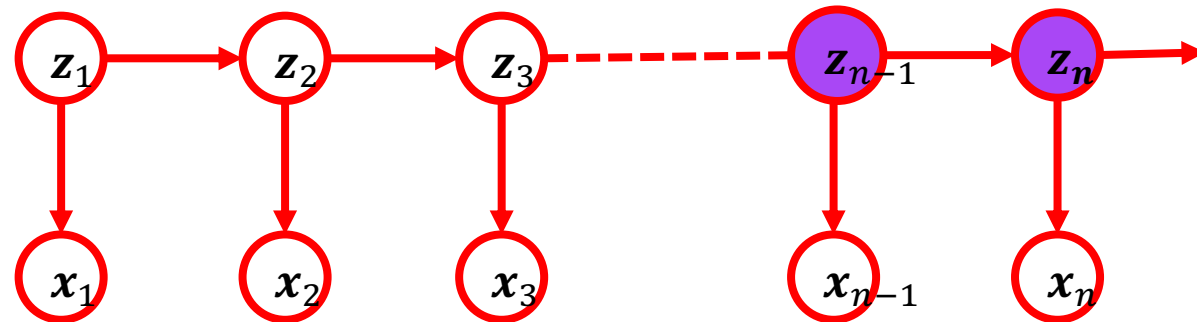
$$\rightarrow p(\mathbf{x}_{n+1}, \mathbf{x}_{n+2}, \dots, \mathbf{x}_N | \mathbf{z}_{n-1}, \mathbf{z}_n)$$

$$= \frac{p(\mathbf{x}_{n+1}, \mathbf{x}_{n+2}, \dots, \mathbf{x}_N, \mathbf{z}_{n-1}, \mathbf{z}_n)}{p(\mathbf{z}_{n-1}, \mathbf{z}_n)}$$

$$= \frac{\frac{p(\mathbf{x}_{n+1}, \mathbf{x}_{n+2}, \dots, \mathbf{x}_N, \mathbf{z}_{n-1}, \mathbf{z}_n)}{p(\mathbf{z}_n)}}{\frac{p(\mathbf{z}_{n-1}, \mathbf{z}_n)}{p(\mathbf{z}_n)}}$$

$$= \frac{p(\mathbf{x}_{n+1}, \mathbf{x}_{n+2}, \dots, \mathbf{x}_N, \mathbf{z}_{n-1} | \mathbf{z}_n)}{p(\mathbf{z}_{n+1} | \mathbf{z}_n)} = \frac{p(\mathbf{x}_{n+1}, \mathbf{x}_{n+2}, \dots, \mathbf{x}_N | \mathbf{z}_n) p(\mathbf{z}_{n-1} | \mathbf{z}_n)}{p(\mathbf{z}_{n+1} | \mathbf{z}_n)}$$

$$= p(\mathbf{x}_{n+1}, \mathbf{x}_{n+2}, \dots, \mathbf{x}_N | \mathbf{z}_n)$$



$\rightarrow d$ 분리이용

$$\alpha(\mathbf{z}_N) = p(\mathbf{X}_N, \mathbf{z}_N)$$

$$\begin{aligned}
p(\mathbf{x}_{N+1}|\mathbf{X}) &= \sum_{\mathbf{z}_{N+1}} p(\mathbf{x}_{N+1}, \mathbf{z}_{N+1}|\mathbf{X}) = \sum_{\mathbf{z}_{N+1}} \frac{p(\mathbf{X}, \mathbf{x}_{N+1}, \mathbf{z}_{N+1})}{p(\mathbf{X})} \frac{p(\mathbf{X}, \mathbf{z}_{N+1})}{p(\mathbf{X}, \mathbf{z}_{N+1})} = \sum_{\mathbf{z}_{N+1}} \frac{p(\mathbf{X}, \mathbf{x}_{N+1}, \mathbf{z}_{N+1})}{p(\mathbf{X}, \mathbf{z}_{N+1})} \frac{p(\mathbf{X}, \mathbf{z}_{N+1})}{p(\mathbf{X})} \\
&= \sum_{\mathbf{z}_{N+1}} \frac{\frac{p(\mathbf{X}, \mathbf{x}_{N+1}, \mathbf{z}_{N+1})}{p(\mathbf{z}_{N+1})}}{\frac{p(\mathbf{X}, \mathbf{z}_{N+1})}{p(\mathbf{z}_{N+1})}} \frac{p(\mathbf{X}, \mathbf{z}_{N+1})}{p(\mathbf{X})} = \sum_{\mathbf{z}_{N+1}} \frac{p(\mathbf{X}, \mathbf{x}_{N+1}|\mathbf{z}_{N+1})}{p(\mathbf{X}|\mathbf{z}_{N+1})} p(\mathbf{z}_{N+1}|\mathbf{X}) = \sum_{\mathbf{z}_{N+1}} \frac{p(\mathbf{X}|\mathbf{z}_{N+1})p(\mathbf{x}_{N+1}|\mathbf{z}_{N+1})}{p(\mathbf{X}|\mathbf{z}_{N+1})} p(\mathbf{z}_{N+1}|\mathbf{X}) \\
&= \sum_{\mathbf{z}_{N+1}} p(\mathbf{x}_{N+1}|\mathbf{z}_{N+1})p(\mathbf{z}_{N+1}|\mathbf{X}) = \sum_{\mathbf{z}_{N+1}} p(\mathbf{x}_{N+1}|\mathbf{z}_{N+1}) \sum_{\mathbf{z}_N} p(\mathbf{z}_N, \mathbf{z}_{N+1}|\mathbf{X}) = \sum_{\mathbf{z}_{N+1}} p(\mathbf{x}_{N+1}|\mathbf{z}_{N+1}) \sum_{\mathbf{z}_N} p(\mathbf{z}_{N+1}|\mathbf{z}_N)p(\mathbf{z}_N|\mathbf{X}) \\
&= \sum_{\mathbf{z}_{N+1}} p(\mathbf{x}_{N+1}|\mathbf{z}_{N+1}) \sum_{\mathbf{z}_N} p(\mathbf{z}_{N+1}|\mathbf{z}_N) \frac{p(\mathbf{X}, \mathbf{z}_N)}{p(\mathbf{X})} = \sum_{\mathbf{z}_{N+1}} p(\mathbf{x}_{N+1}|\mathbf{z}_{N+1}) \sum_{\mathbf{z}_N} p(\mathbf{z}_{N+1}|\mathbf{z}_N) \frac{\alpha(\mathbf{z}_N)}{p(\mathbf{X})} \\
c_{N+1} \equiv p(\mathbf{x}_{N+1}|\mathbf{X}) &= \frac{1}{p(\mathbf{X})} \sum_{\mathbf{z}_{N+1}} p(\mathbf{x}_{N+1}|\mathbf{z}_{N+1}) \sum_{\mathbf{z}_N} p(\mathbf{z}_{N+1}|\mathbf{z}_N) \alpha(\mathbf{z}_N)
\end{aligned}$$



$$\hat{\alpha}(\mathbf{z}_n) = p(\mathbf{z}_n | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n) = \frac{\alpha(\mathbf{z}_n)}{p(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n)}$$

$$\begin{aligned} \alpha(\mathbf{z}_n) &= p(\mathbf{x}_n | \mathbf{z}_n) \sum_{\mathbf{z}_{n-1}} \alpha(\mathbf{z}_{n-1}) p(\mathbf{z}_n | \mathbf{z}_{n-1}) \\ &= p(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n, \mathbf{z}_n) \end{aligned}$$

$$\frac{\alpha(\mathbf{z}_n)}{p(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1})} = p(\mathbf{x}_n | \mathbf{z}_n) \sum_{\mathbf{z}_{n-1}} \frac{\alpha(\mathbf{z}_{n-1})}{p(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1})} p(\mathbf{z}_n | \mathbf{z}_{n-1})$$

$$\frac{\frac{\alpha(\mathbf{z}_n)}{p(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n)}}{\frac{p(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1})}{p(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1}, \mathbf{x}_n)}} = p(\mathbf{x}_n | \mathbf{z}_n) \sum_{\mathbf{z}_{n-1}} \frac{\alpha(\mathbf{z}_{n-1})}{p(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1})} p(\mathbf{z}_n | \mathbf{z}_{n-1})$$

$$c_n \hat{\alpha}(\mathbf{z}_n) = p(\mathbf{x}_n | \mathbf{z}_n) \sum_{\mathbf{z}_{n-1}} \hat{\alpha}(\mathbf{z}_{n-1}) p(\mathbf{z}_n | \mathbf{z}_{n-1})$$

$$c_1 = p(\mathbf{x}_1)$$

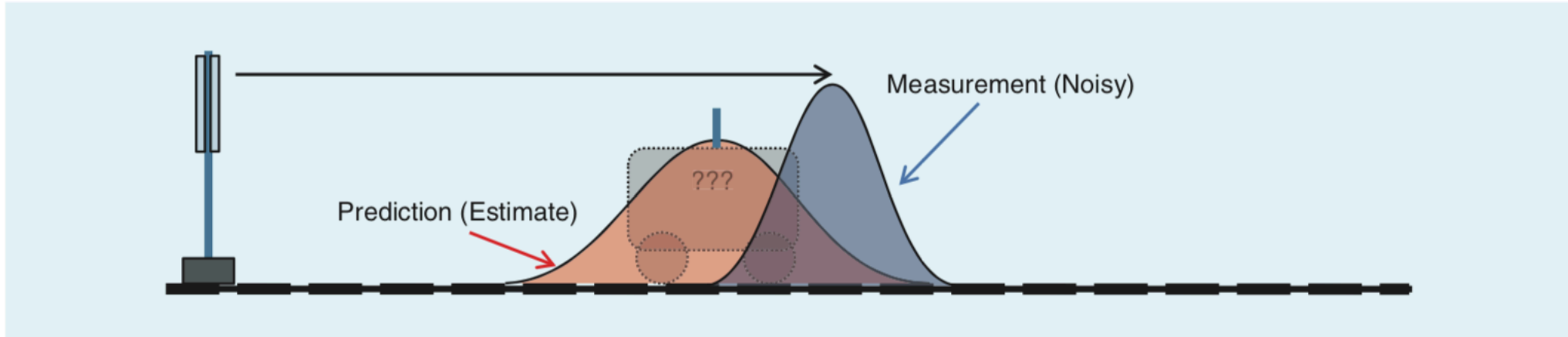
$$c_2 = p(\mathbf{x}_2 | \mathbf{x}_1)$$

$$c_3 = p(\mathbf{x}_3 | \mathbf{x}_1, \mathbf{x}_2)$$

$$c_4 = p(\mathbf{x}_4 | \mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3)$$

$$c_n = p(\mathbf{x}_n | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1})$$





**[FIG4]** Shows the measurement of the location of the train at time  $t = 1$  and the level of uncertainty in that noisy measurement, represented by the blue Gaussian pdf. The combined knowledge of this system is provided by multiplying these two pdfs together.

알려지지 않은 수량  $z_n$ 의 값을 노이즈가 포함되는 센서를 이용해서 측정해야 한다.

$$z_n = Az_{n-1} + \text{noise}$$

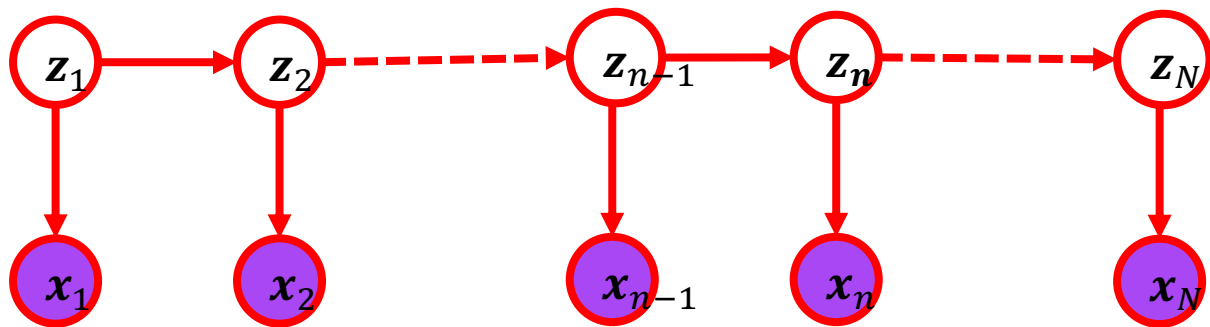
$z_n \rightarrow n$ 번째 추정값

$z_{n-1} \rightarrow n - 1$ 번째 추정값

센서로부터 주어진 관측값  $x_n$ 은  $z_n$ 의 값에 0평균 가우시안 노이즈를 더한 것이다.

$$x_n = Cz_n + \text{noise}$$

$x_n \rightarrow n$ 번째 측정값



알려지지 않은 수량  $z_n$ 의 값을 노이즈가 포함되는 센서를 이용해서 측정해야 한다.

$$z_n = Az_{n-1} + \text{noise}$$

$z_n \rightarrow n$ 번째 추정값

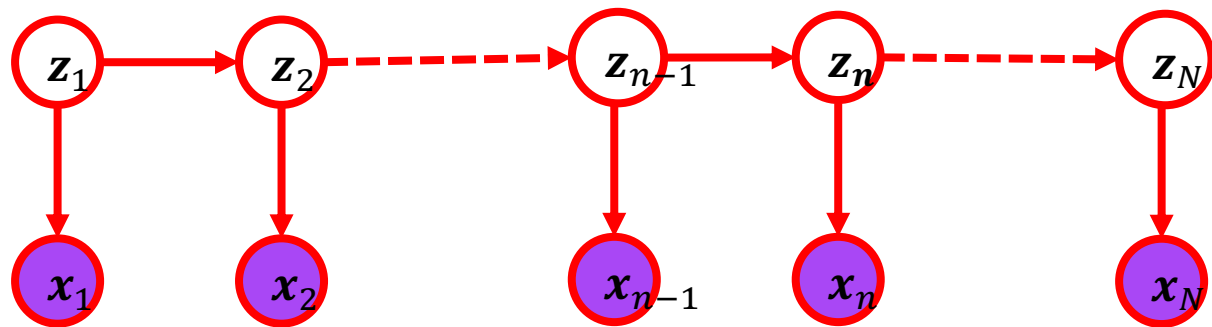
$z_{n-1} \rightarrow n-1$ 번째 추정값

센서로부터 주어진 관측값  $x_n$ 은  $z_n$ 의 값에 0평균 가우시안 노이즈를 더한 것이다.

$$x_n = Cz_n + \text{noise}$$

$x_n \rightarrow n$ 번째 측정값

# 선형 동적 시스템에서의 추론



$$p(\mathbf{z}_n | \mathbf{z}_{n-1}) = \mathcal{N}(\mathbf{z}_n | \mathbf{A}\mathbf{z}_{n-1}, \mathbf{\Gamma}) \rightarrow \text{전이 확률}$$

$$p(\mathbf{x}_n | \mathbf{z}_n) = \mathcal{N}(\mathbf{x}_n | \mathbf{C}\mathbf{z}_n, \mathbf{\Sigma}) \rightarrow \text{방사 확률}$$

$$p(\mathbf{z}_1) = \mathcal{N}(\mathbf{z}_1 | \boldsymbol{\mu}_0, \mathbf{P}_0) \rightarrow \text{초기값}$$

알려지지 않은 수량  $\mathbf{z}_n$



$$p(\mathbf{z}_n | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n)$$

$$\boldsymbol{\theta} = \{\mathbf{A}, \mathbf{\Gamma}, \mathbf{C}, \mathbf{\Sigma}, \boldsymbol{\mu}_0, \mathbf{P}_0\}$$

센서로부터 주어진 관측값  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$

## 선형 동적 시스템에서의 추론

$$\hat{\alpha}(\mathbf{z}_n) = p(\mathbf{z}_n | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n) = \mathcal{N}(\mathbf{z}_n | \boldsymbol{\mu}_n, \mathbf{V}_n)$$

$$c_n = p(\mathbf{x}_n | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1})$$

$$c_n \hat{\alpha}(\mathbf{z}_n) = p(\mathbf{x}_n | \mathbf{z}_n) \sum_{\mathbf{z}_{n-1}} \hat{\alpha}(\mathbf{z}_{n-1}) p(\mathbf{z}_n | \mathbf{z}_{n-1})$$

$$c_n \hat{\alpha}(\mathbf{z}_n) = p(\mathbf{x}_n | \mathbf{z}_n) \int \hat{\alpha}(\mathbf{z}_{n-1}) p(\mathbf{z}_n | \mathbf{z}_{n-1}) d\mathbf{z}_{n-1}$$

$$p(\mathbf{z}_n | \mathbf{z}_{n-1}) = \mathcal{N}(\mathbf{z}_n | \mathbf{A}\mathbf{z}_{n-1}, \boldsymbol{\Gamma}) \rightarrow \text{전이 확률}$$

$$p(\mathbf{x}_n | \mathbf{z}_n) = \mathcal{N}(\mathbf{x}_n | \mathbf{C}\mathbf{z}_n, \boldsymbol{\Sigma}) \rightarrow \text{방사 확률}$$

$$p(\mathbf{z}_1) = \mathcal{N}(\mathbf{z}_1 | \boldsymbol{\mu}_0, \mathbf{P}_0) \rightarrow \text{초기 값}$$

센서로부터 주어진 관측값  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$

알려지지 않은 수량  $\mathbf{z}_n$

## 선형 동적 시스템에서의 추론

$$\hat{\alpha}(\mathbf{z}_n) = p(\mathbf{z}_n | x_1, x_2, \dots, x_n) = \mathcal{N}(\mathbf{z}_n | \boldsymbol{\mu}_n, \mathbf{V}_n)$$

$$p(\mathbf{z}_n | \mathbf{z}_{n-1}) = \mathcal{N}(\mathbf{z}_n | \mathbf{A}\mathbf{z}_{n-1}, \boldsymbol{\Gamma}) \rightarrow \text{전이 확률}$$

$$p(x_n | \mathbf{z}_n) = \mathcal{N}(x_n | \mathbf{C}\mathbf{z}_n, \boldsymbol{\Sigma}) \rightarrow \text{방사 확률}$$

$$p(\mathbf{z}_1) = \mathcal{N}(\mathbf{z}_1 | \boldsymbol{\mu}_0, \mathbf{P}_0) \rightarrow \text{초기값}$$

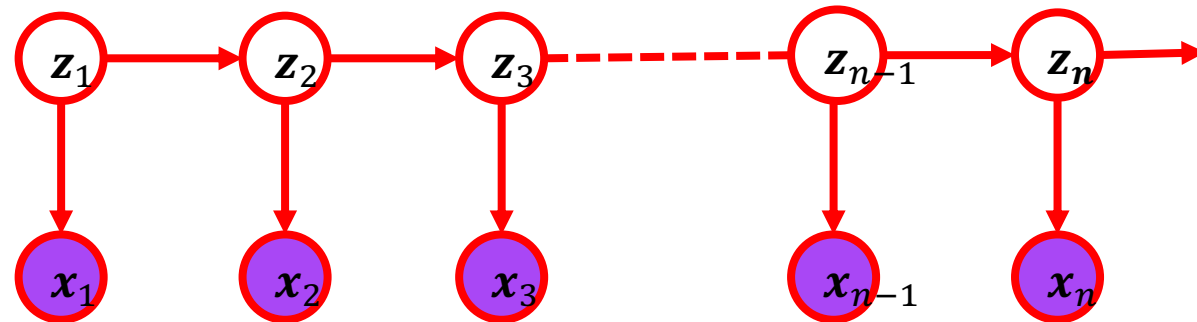
센서로부터 주어진 관측값  $x_1, x_2, \dots, x_n$

알려지지 않은 수량  $\mathbf{z}_n$

$$c_n \hat{\alpha}(\mathbf{z}_n) = p(x_n | \mathbf{z}_n) \int \hat{\alpha}(\mathbf{z}_{n-1}) p(\mathbf{z}_n | \mathbf{z}_{n-1}) d\mathbf{z}_{n-1}$$

$$c_n \mathcal{N}(\mathbf{z}_n | \boldsymbol{\mu}_n, \mathbf{V}_n) = \mathcal{N}(x_n | \mathbf{C}\mathbf{z}_n, \boldsymbol{\Sigma}) \int \mathcal{N}(\mathbf{z}_{n-1} | \boldsymbol{\mu}_{n-1}, \mathbf{V}_{n-1}) \mathcal{N}(\mathbf{z}_n | \mathbf{A}\mathbf{z}_{n-1}, \boldsymbol{\Gamma}) d\mathbf{z}_{n-1}$$

## 선형 동적 시스템에서의 추론



$$c_n \hat{\alpha}(\mathbf{z}_n) = p(\mathbf{x}_n | \mathbf{z}_n) \int \hat{\alpha}(\mathbf{z}_{n-1}) p(\mathbf{z}_n | \mathbf{z}_{n-1}) d\mathbf{z}_{n-1}$$

$$c_n p(\mathbf{z}_n | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n) = p(\mathbf{x}_n | \mathbf{z}_n) \int p(\mathbf{z}_n | \mathbf{z}_{n-1}) p(\mathbf{z}_{n-1} | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1}) d\mathbf{z}_{n-1}$$

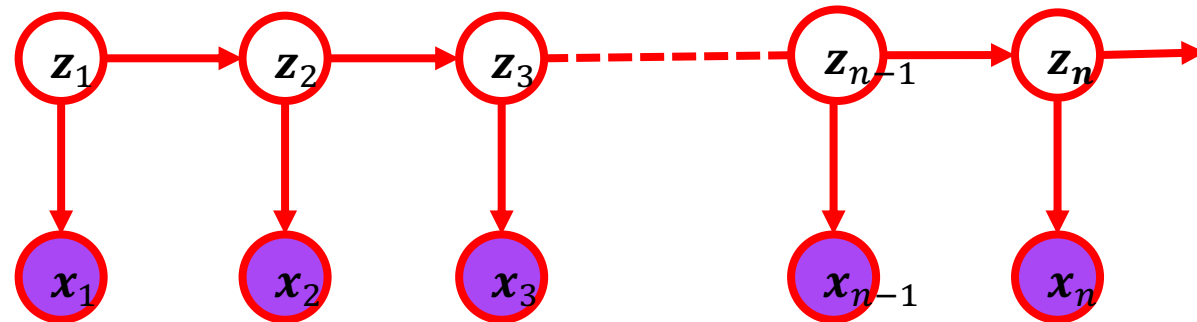
$$\rightarrow p(\mathbf{z}_{n-1}, \mathbf{z}_n | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1}) = p(\mathbf{z}_n | \mathbf{z}_{n-1}) p(\mathbf{z}_{n-1} | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1})$$

$$c_n p(\mathbf{z}_n | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n) = p(\mathbf{x}_n | \mathbf{z}_n) \int p(\mathbf{z}_{n-1}, \mathbf{z}_n | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1}) d\mathbf{z}_{n-1}$$

$$c_n p(\mathbf{z}_n | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n) = p(\mathbf{x}_n | \mathbf{z}_n) p(\mathbf{z}_n | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1})$$



## 선형 동적 시스템에서의 추론



$$p(z_{n-1}, z_n | x_1, x_2, \dots, x_{n-1}) = p(z_n | z_{n-1}, x_1, x_2, \dots, x_{n-1}) p(z_{n-1} | x_1, x_2, \dots, x_{n-1})$$

$$x_1, x_2, \dots, x_{n-1} \perp\!\!\!\perp z_n | z_{n-1}$$

$$\rightarrow p(z_{n-1}, z_n | x_1, x_2, \dots, x_{n-1}) = p(z_n | z_{n-1}) p(z_{n-1} | x_1, x_2, \dots, x_{n-1})$$

$$c_n p(z_n | x_1, x_2, \dots, x_n) = p(x_n | z_n, x_1, x_2, \dots, x_{n-1}) p(z_n | x_1, x_2, \dots, x_{n-1})$$

$$x_1, x_2, \dots, x_{n-1} \perp\!\!\!\perp x_n | z_n$$

$$c_n p(z_n | x_1, x_2, \dots, x_n) = p(x_n | z_n) p(z_n | x_1, x_2, \dots, x_{n-1})$$

## 선형 동적 시스템에서의 추론

$$\hat{\alpha}(\mathbf{z}_n) = p(\mathbf{z}_n | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n) = \mathcal{N}(\mathbf{z}_n | \boldsymbol{\mu}_n, \mathbf{V}_n)$$

$$c_n \hat{\alpha}(\mathbf{z}_n) = p(\mathbf{x}_n | \mathbf{z}_n) \int \hat{\alpha}(\mathbf{z}_{n-1}) p(\mathbf{z}_n | \mathbf{z}_{n-1}) d\mathbf{z}_{n-1}$$

$$c_n \mathcal{N}(\mathbf{z}_n | \boldsymbol{\mu}_n, \mathbf{V}_n) = \mathcal{N}(\mathbf{x}_n | \mathbf{C}\mathbf{z}_n, \boldsymbol{\Sigma}) \int \mathcal{N}(\mathbf{z}_{n-1} | \boldsymbol{\mu}_{n-1}, \mathbf{V}_{n-1}) \mathcal{N}(\mathbf{z}_n | \mathbf{A}\mathbf{z}_{n-1}, \boldsymbol{\Gamma}) d\mathbf{z}_{n-1}$$

$$p(\mathbf{z}_n | \mathbf{z}_{n-1}) = \mathcal{N}(\mathbf{z}_n | \mathbf{A}\mathbf{z}_{n-1}, \boldsymbol{\Gamma}) \rightarrow \text{전이 확률}$$

$$p(\mathbf{x}_n | \mathbf{z}_n) = \mathcal{N}(\mathbf{x}_n | \mathbf{C}\mathbf{z}_n, \boldsymbol{\Sigma}) \rightarrow \text{방사 확률}$$

$$p(\mathbf{z}_1) = \mathcal{N}(\mathbf{z}_1 | \boldsymbol{\mu}_0, \mathbf{P}_0) \rightarrow \text{초기 값}$$

센서로부터 주어진 관측값  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$

알려지지 않은 수량  $\mathbf{z}_n$

식 2.113~2.115 참고

## 선형 동적 시스템에서의 추론

식 2.113  $p(\mathbf{x}) = \mathcal{N}(\mathbf{x} | \boldsymbol{\mu}, \boldsymbol{\Lambda}^{-1})$

$$p(\mathbf{z}_{n-1} | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1}) = \mathcal{N}(\mathbf{z}_{n-1} | \boldsymbol{\mu}_{n-1}, \mathbf{V}_{n-1})$$

식 2.114  $p(\mathbf{y} | \mathbf{x}) = \mathcal{N}(\mathbf{y} | \mathbf{A}\mathbf{x}, \mathbf{L}^{-1})$

$$p(\mathbf{z}_n | \mathbf{z}_{n-1}) = \mathcal{N}(\mathbf{z}_n | \mathbf{A}\mathbf{z}_{n-1}, \boldsymbol{\Gamma})$$

식 2.115  $p(\mathbf{y}) = \mathcal{N}(\mathbf{y} | \mathbf{A}\boldsymbol{\mu}, \mathbf{L}^{-1} + \mathbf{A}\boldsymbol{\Lambda}^{-1}\mathbf{A}^T)$

$$p(\mathbf{z}_n | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1}) = \mathcal{N}(\mathbf{z}_n | \mathbf{A}\boldsymbol{\mu}_{n-1}, \mathbf{P}_{n-1})$$

$$\mathbf{P}_{n-1} = \boldsymbol{\Gamma} + \mathbf{A}\mathbf{V}_{n-1}\mathbf{A}^T$$

$$p(\mathbf{y}) = \int p(\mathbf{y} | \mathbf{x}) p(\mathbf{x}) d\mathbf{x}$$

$$p(\mathbf{z}_n | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1}) = \int p(\mathbf{z}_n | \mathbf{z}_{n-1}) p(\mathbf{z}_{n-1} | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_{n-1}) d\mathbf{z}_{n-1}$$

선형 동적 시스템에서의 추론

$$\hat{\alpha}(\mathbf{z}_n) = p(\mathbf{z}_n | \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n) = \mathcal{N}(\mathbf{z}_n | \boldsymbol{\mu}_n, \mathbf{V}_n)$$

$$c_n \hat{\alpha}(\mathbf{z}_n) = p(\mathbf{x}_n | \mathbf{z}_n) \int \hat{\alpha}(\mathbf{z}_{n-1}) p(\mathbf{z}_n | \mathbf{z}_{n-1}) d\mathbf{z}_{n-1}$$

$$c_n \mathcal{N}(\mathbf{z}_n | \boldsymbol{\mu}_n, \mathbf{V}_n) = \mathcal{N}(\mathbf{x}_n | \mathbf{C}\mathbf{z}_n, \boldsymbol{\Sigma}) \int \mathcal{N}(\mathbf{z}_{n-1} | \boldsymbol{\mu}_{n-1}, \mathbf{V}_{n-1}) \mathcal{N}(\mathbf{z}_n | \mathbf{A}\mathbf{z}_{n-1}, \boldsymbol{\Gamma}) d\mathbf{z}_{n-1}$$

식 2.113~2.115 참고

$$c_n \mathcal{N}(\mathbf{z}_n | \boldsymbol{\mu}_n, \mathbf{V}_n) = \mathcal{N}(\mathbf{x}_n | \mathbf{C}\mathbf{z}_n, \boldsymbol{\Sigma}) \mathcal{N}(\mathbf{z}_n | \mathbf{A}\boldsymbol{\mu}_{n-1}, \mathbf{P}_{n-1})$$

$$\mathbf{P}_{n-1} = \boldsymbol{\Gamma} + \mathbf{A}\mathbf{V}_{n-1}\mathbf{A}^T$$

$$p(\mathbf{z}_n | \mathbf{z}_{n-1}) = \mathcal{N}(\mathbf{z}_n | \mathbf{A}\mathbf{z}_{n-1}, \boldsymbol{\Gamma}) \rightarrow \text{전이 확률}$$

$$p(\mathbf{x}_n | \mathbf{z}_n) = \mathcal{N}(\mathbf{x}_n | \mathbf{C}\mathbf{z}_n, \boldsymbol{\Sigma}) \rightarrow \text{방사 확률}$$

$$p(\mathbf{z}_1) = \mathcal{N}(\mathbf{z}_1 | \boldsymbol{\mu}_n, \mathbf{P}_0) \rightarrow \text{초기 값}$$

센서로부터 주어진 관측값  $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$

알려지지 않은 수량  $\mathbf{z}_n$

## 선형 동적 시스템에서의 추론

$$\mathbf{P}_{n-1} = \mathbf{\Gamma} + \mathbf{A}\mathbf{V}_{n-1}\mathbf{A}^T$$

$$\text{식 2.113 } p(\mathbf{x}) = \mathcal{N}(\mathbf{x}|\boldsymbol{\mu}, \mathbf{\Lambda}^{-1})$$

$$p(\mathbf{z}_n|\mathbf{x}_1, \mathbf{x}_2, \cdots, \mathbf{x}_{n-1}) = \mathcal{N}(\mathbf{z}_n|\mathbf{A}\boldsymbol{\mu}_{n-1}, \mathbf{P}_{n-1})$$

$$\text{식 2.114 } p(\mathbf{y}|\mathbf{x}) = \mathcal{N}(\mathbf{y}|\mathbf{A}\mathbf{x}, \mathbf{L}^{-1})$$

$$p(\mathbf{x}_n|\mathbf{z}_n) = \mathcal{N}(\mathbf{x}_n|\mathbf{C}\mathbf{z}_n, \boldsymbol{\Sigma})$$

$$\text{식 2.115 } p(\mathbf{y}) = \mathcal{N}(\mathbf{y}|\mathbf{A}\boldsymbol{\mu}, \mathbf{L}^{-1} + \mathbf{A}\mathbf{\Lambda}^{-1}\mathbf{A}^T)$$

$$c_n = ?$$

$$\text{식 2.116 } p(\mathbf{x}|\mathbf{y}) = \mathcal{N}(\mathbf{x}|\boldsymbol{\Sigma}(\mathbf{A}^T\mathbf{L}\mathbf{y} + \mathbf{A}\boldsymbol{\mu}), \boldsymbol{\Sigma})$$

$$p(\mathbf{z}_n|\mathbf{x}_1, \mathbf{x}_2, \cdots, \mathbf{x}_n) = \mathcal{N}(\mathbf{z}_n|\boldsymbol{\mu}_n, \mathbf{V}_n) = ?$$

$$\boldsymbol{\Sigma} = (\mathbf{\Lambda} + \mathbf{A}^T\mathbf{L}\mathbf{A})^{-1}$$

$$p(\mathbf{y})p(\mathbf{x}|\mathbf{y}) = p(\mathbf{y}|\mathbf{x})p(\mathbf{x})$$

$$c_n p(\mathbf{z}_n|\mathbf{x}_1, \mathbf{x}_2, \cdots, \mathbf{x}_n) = p(\mathbf{x}_n|\mathbf{z}_n)p(\mathbf{z}_n|\mathbf{x}_1, \mathbf{x}_2, \cdots, \mathbf{x}_{n-1})$$

## 선형 동적 시스템에서의 추론

$$\mathbf{P}_{n-1} = \mathbf{\Gamma} + \mathbf{A}\mathbf{V}_{n-1}\mathbf{A}^T$$

식 2.113  $p(\mathbf{x}) = \mathcal{N}(\mathbf{x}|\boldsymbol{\mu}, \mathbf{\Lambda}^{-1})$

$$p(\mathbf{z}_n|\mathbf{x}_1, \mathbf{x}_2, \cdots, \mathbf{x}_{n-1}) = \mathcal{N}(\mathbf{z}_n|\mathbf{A}\boldsymbol{\mu}_{n-1}, \mathbf{P}_{n-1})$$

식 2.114  $p(\mathbf{y}|\mathbf{x}) = \mathcal{N}(\mathbf{y}|\mathbf{A}\mathbf{x}, \mathbf{L}^{-1})$

$$p(\mathbf{x}_n|\mathbf{z}_n) = \mathcal{N}(\mathbf{x}_n|\mathbf{C}\mathbf{z}_n, \boldsymbol{\Sigma})$$

식 2.115  $p(\mathbf{y}) = \mathcal{N}(\mathbf{y}|\mathbf{A}\boldsymbol{\mu}, \mathbf{L}^{-1} + \mathbf{A}\mathbf{\Lambda}^{-1}\mathbf{A}^T)$

$$c_n = \mathcal{N}(\mathbf{x}_n|\mathbf{C}\mathbf{A}\boldsymbol{\mu}_{n-1}, \boldsymbol{\Sigma} + \mathbf{C}\mathbf{P}_{n-1}\mathbf{C}^T)$$

식 2.116  $p(\mathbf{x}|\mathbf{y}) = \mathcal{N}(\mathbf{x}|\boldsymbol{\Sigma}(\mathbf{A}^T\mathbf{L}\mathbf{y} + \mathbf{\Lambda}\boldsymbol{\mu}), \boldsymbol{\Sigma})$

$$p(\mathbf{z}_n|\mathbf{x}_1, \mathbf{x}_2, \cdots, \mathbf{x}_n) \rightarrow$$

$$\boldsymbol{\Sigma} = (\mathbf{\Lambda} + \mathbf{A}^T\mathbf{L}\mathbf{A})^{-1}$$

$$\mathcal{N}(\mathbf{z}_n|\boldsymbol{\mu}_n, \mathbf{V}_n) = \mathcal{N}(\mathbf{z}_n|\mathbf{V}_n(\mathbf{C}^T\boldsymbol{\Sigma}^{-1}\mathbf{x}_n + \mathbf{P}_{n-1}^{-1}\mathbf{A}\boldsymbol{\mu}_{n-1}), \mathbf{V}_n)$$

$$\mathbf{V}_n = (\mathbf{P}_{n-1}^{-1} + \mathbf{C}^T\boldsymbol{\Sigma}^{-1}\mathbf{C})^{-1}$$

## 선형 동적 시스템에서의 추론

$$\mu_n = V_n (\mathbf{C}^T \Sigma^{-1} x_n + \mathbf{P}_{n-1}^{-1} A \mu_{n-1})$$

$$(P^{-1} + B^T R^{-1} B)^{-1} B^T R^{-1} = P B^T (B P B^T + R)^{-1} \quad (\text{식 C.5})$$

$$V_n = (\mathbf{P}_{n-1}^{-1} + \mathbf{C}^T \Sigma^{-1} \mathbf{C})^{-1}$$

$$(A + B D^{-1} C)^{-1} = A^{-1} - A^{-1} B (D + C A^{-1} B)^{-1} C A^{-1} \quad (\text{식 C.7})$$

$$c_n = \mathcal{N}(x_n | \mathbf{C} A \mu_{n-1}, \Sigma + \mathbf{C} \mathbf{P}_{n-1} \mathbf{C}^T)$$

$$V_n = (\mathbf{P}_{n-1}^{-1} + \mathbf{C}^T \Sigma^{-1} \mathbf{C})^{-1} = \mathbf{P}_{n-1} - \mathbf{P}_{n-1} \mathbf{C}^T (\Sigma + \mathbf{C} \mathbf{P}_{n-1} \mathbf{C}^T)^{-1} \mathbf{C} \mathbf{P}_{n-1} \quad (\text{식 C.7) 이용}$$

$$= \mathbf{P}_{n-1} - \mathbf{K}_n \mathbf{C} \mathbf{P}_{n-1}$$

$$V_n = (\mathbf{I} - \mathbf{K}_n \mathbf{C}) \mathbf{P}_{n-1}$$

$$\mathbf{K}_n = \mathbf{P}_{n-1} \mathbf{C}^T (\mathbf{C} \mathbf{P}_{n-1} \mathbf{C}^T + \Sigma)^{-1}$$

## 선형 동적 시스템에서의 추론

$$\mu_n = V_n(\mathbf{C}^T \Sigma^{-1} x_n + \mathbf{P}_{n-1}^{-1} \mathbf{A} \mu_{n-1})$$

$$\mathbf{V}_n = (\mathbf{I} - \mathbf{K}_n \mathbf{C}) \mathbf{P}_{n-1}$$

$$\mathbf{K}_n = \mathbf{P}_{n-1} \mathbf{C}^T (\mathbf{C} \mathbf{P}_{n-1} \mathbf{C}^T + \Sigma)^{-1}$$

$$c_n = \mathcal{N}(x_n | \mathbf{C} \mathbf{A} \mu_{n-1}, \Sigma + \mathbf{C} \mathbf{P}_{n-1} \mathbf{C}^T)$$

$$\mu_n = \mathbf{V}_n \mathbf{C}^T \Sigma^{-1} x_n + \mathbf{V}_n \mathbf{C} \mathbf{A} \mu_{n-1}$$

$$= \mathbf{K}_n x_n + \mathbf{A} \mu_{n-1} - \mathbf{K}_n \mathbf{C} \mathbf{A} \mu_{n-1}$$

$$= \mathbf{A} \mu_{n-1} + \mathbf{K}_n (x_n - \mathbf{K}_n \mathbf{C} \mathbf{A} \mu_{n-1})$$

$$(P^{-1} + B^T R^{-1} B)^{-1} B^T R^{-1} = P B^T (B P B^T + R)^{-1} \quad (\text{식 } C.5)$$

$$(A + B D^{-1} C)^{-1} = A^{-1} - A^{-1} B (D + C A^{-1} B)^{-1} C A^{-1} \quad (\text{식 } C.7)$$

$$\mathbf{V}_n \mathbf{C}^T \Sigma^{-1} = (\mathbf{P}_{n-1}^{-1} + \mathbf{C}^T \Sigma^{-1} \mathbf{C})^{-1} \mathbf{C}^T \Sigma^{-1} = \mathbf{P}_{n-1} \mathbf{C}^T (\mathbf{C} \mathbf{P}_{n-1} \mathbf{C}^T + \Sigma)^{-1}$$

$$\mathbf{V}_n \mathbf{P}_{n-1}^{-1} \mathbf{A} \mu_{n-1} = (\mathbf{I} - \mathbf{K}_n \mathbf{C}) \mathbf{P}_{n-1} \mathbf{P}_{n-1}^{-1} \mathbf{A} \mu_{n-1}$$



## 선형 동적 시스템에서의 추론

$$\mu_n = A\mu_{n-1} + K_n(x_n - CA\mu_{n-1})$$

$$V_n = (I - K_nC)P_{n-1}$$

$$K_n = P_{n-1} C^T (CP_{n-1}C^T + \Sigma)^{-1}$$

$$P_{n-1} = \Gamma + AV_{n-1}A^T$$

$$\hat{\alpha}(z_n) = p(z_n|x_1, x_2, \dots, x_n) = \mathcal{N}(z_n|\mu_n, V_n)$$

$$c_n \hat{\alpha}(z_n) = p(x_n|z_n) \int \hat{\alpha}(z_{n-1}) p(z_n|z_{n-1}) dz_{n-1}$$

$$p(z_n|z_{n-1}) = \mathcal{N}(z_n|Az_{n-1}, \Gamma) \rightarrow \text{전이 확률}$$

$$p(x_n|z_n) = \mathcal{N}(x_n|Cz_n, \Sigma) \rightarrow \text{방사 확률}$$

$$p(z_1) = \mathcal{N}(z_1|\mu_0, P_0) \rightarrow \text{초기 값}$$

센서로부터 주어진 관측값  $x_1, x_2, \dots, x_n$

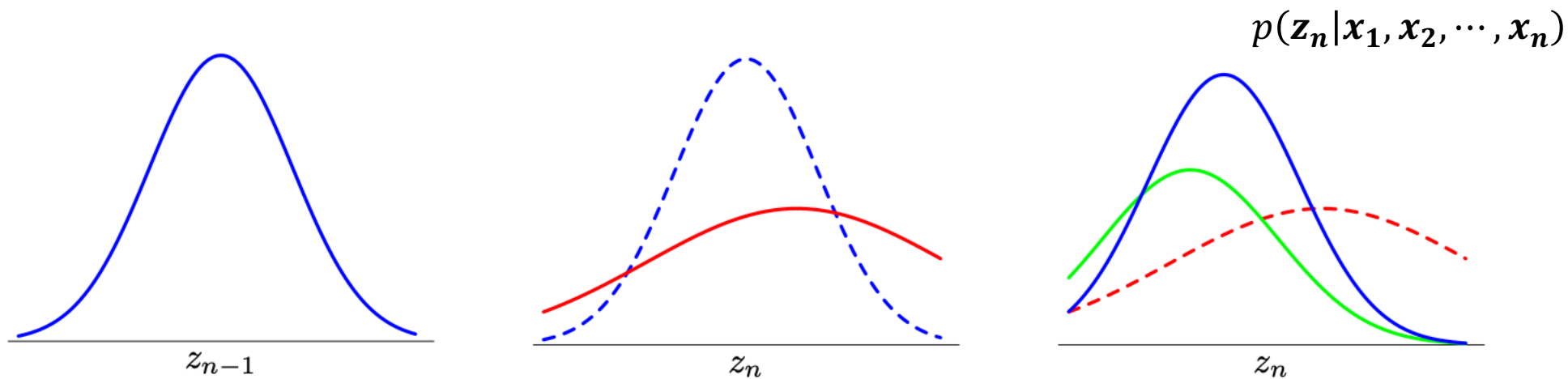
알려지지 않은 수량  $z_n$

$$c_n = p(x_n|x_1, x_2, \dots, x_{n-1})$$

$$\hat{\alpha}(z_n) = p(z_n|x_1, x_2, \dots, x_n)$$

$$c_n = \mathcal{N}(x_n|CA\mu_{n-1}, \Sigma + CP_{n-1}C^T)$$

## 선형 동적 시스템에서의 추론



$$p(z_n | x_1, x_2, \dots, x_{n-1}) = \int p(z_n | z_{n-1}) p(z_{n-1} | x_1, x_2, \dots, x_{n-1}) dz_{n-1}$$

$$p(x_n | z_n)$$

$$c_n p(z_n | x_1, x_2, \dots, x_n) = p(x_n | z_n) p(z_n | x_1, x_2, \dots, x_{n-1})$$