

AND

(3) (11)

$$p(\underline{x}|\underline{y}) = N(\underline{x} | \underline{\Sigma}_N \{ \underline{A}^T \underline{L} (\underline{y} - \underline{b}) + \underline{\Lambda} \underline{\mu} \}, \underline{\Sigma}_N)$$

$$= N(\underline{x} | \underline{m}_N, \underline{\Sigma}_N)$$

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

EXAMPLE 7:

$$y(n) = \theta_0 + e(n) = y_n(n) + e(n)$$

N = 1 ONE MEASUREMENT

$$\underline{L} \underline{\Phi}^T \underline{\theta}$$

$$[1] [\theta_0]$$

$$p(\theta) = \frac{1}{\sqrt{2\pi}\sigma_\theta} \exp\left[-\frac{1}{2\sigma_\theta^2}(\theta - \theta_0)^2\right] \leftarrow \text{prior}$$

$$p(y|\theta) = \frac{1}{\sqrt{2\pi}\sigma_e} \exp\left[-\frac{1}{2\sigma_e^2}(y - \theta)^2\right] \leftarrow \text{MEASUREMENT}$$

$$\ell(\theta) \Rightarrow \frac{\partial}{\partial \theta} \left[-\frac{1}{2\sigma_e^2}(y - \theta)^2 - \frac{1}{2\sigma_\theta^2}(\theta - \theta_0)^2 \right]$$

$$\frac{(y - \theta)}{\sigma_e^2} - \frac{(\theta - \theta_0)}{\sigma_\theta^2} = 0$$

$$\hat{\theta}_{MAP} = \frac{\frac{1}{\sigma_e^2} \theta_0 + \frac{1}{\sigma_e^2} y}{\frac{1}{\sigma_e^2} + \frac{1}{\sigma_\theta^2}}$$

EXAMPLE 8: MORE MEASUREMENTS

$$p(y(n)|\underline{\theta}) \sim N(\underline{\Phi} \underline{\theta}, \sigma_e^2) = \frac{1}{\sqrt{2\pi}\sigma_e} \exp\left[-\frac{1}{2\sigma_e^2}(y(n) - \theta)^2\right]$$

$$[1] \cdot \theta$$

$$p(\underline{y}|\underline{\theta}) \sim N(\underline{\Phi} \underline{\theta}, \sigma_e^2 \underline{I}) = \left(\frac{1}{\sqrt{2\pi}}\right)^N \frac{1}{\sigma_e^N} \exp\left[-\frac{1}{2\sigma_e^2} \sum (y(n) - \theta)^2\right]$$

$$\begin{bmatrix} 1 \\ 1 \\ \vdots \end{bmatrix} \theta$$

$$|\sigma_e^2 \underline{I}|^{1/2} = \sigma_e^N$$

$$-\frac{1}{2}(\underline{y} - \underline{\Phi} \underline{\theta})^T (\sigma_e^2 \underline{I})^{-1} (\underline{y} - \underline{\Phi} \underline{\theta})$$

$$\frac{\partial}{\partial \theta} \rightarrow \frac{1}{\sigma_e^2} \left(\sum_n (y(n) - \theta) \right) - \frac{1}{\sigma_\theta^2} (\theta - \theta_0) = 0$$

$$\frac{1}{\sigma_e^2} = \beta \quad (\text{NOISY MEASUREMENT PRECISION})$$

$$\frac{1}{\sigma_\theta^2} = \alpha \quad (\text{A PRIORI KNOWLEDGE PRECISION})$$

$$\hat{\theta}_{MAP} = \frac{\alpha \theta_0 + N\beta \left(\frac{1}{N} \sum y(n) \right)}{\alpha + N\beta}$$

$$\hat{\theta}_{ML}$$