

$$N(x|\mu, \Sigma) = \frac{1}{(2\pi)^{\frac{D}{2}} |\Sigma|^{\frac{1}{2}}} e^{-\frac{1}{2} (x-\mu)^T \Sigma^{-1} (x-\mu)}$$

$$(x-\mu)^T \Sigma^{-1} (x-\mu)$$

①  $\Sigma$  - diagonal matrix

$$\Sigma = \begin{bmatrix} \Sigma_1 & & \\ & \Sigma_2 & \\ & & \ddots \\ & & & \Sigma_d \end{bmatrix} \Rightarrow \Sigma^{-1} = \begin{bmatrix} \frac{1}{\Sigma_1} & & \\ & \frac{1}{\Sigma_2} & \\ & & \ddots \\ & & & \frac{1}{\Sigma_d} \end{bmatrix}$$

$$(x-\mu)^T \Sigma^{-1} (x-\mu) = [(x_1-\mu_1), (x_2-\mu_2), \dots, (x_D-\mu_D)] \begin{bmatrix} \frac{1}{\Sigma_1} & & \\ & \frac{1}{\Sigma_2} & \\ & & \ddots \\ & & & \frac{1}{\Sigma_D} \end{bmatrix} \begin{bmatrix} x_1-\mu_1 \\ x_2-\mu_2 \\ \vdots \\ x_D-\mu_D \end{bmatrix} =$$

$$= (x_1-\mu_1) \cdot \frac{1}{\Sigma_1} \cdot (x_1-\mu_1) + \dots + (x_D-\mu_D) \cdot \frac{1}{\Sigma_D} (x_D-\mu_D) =$$

$$= \sum_{i=1}^D \underbrace{\frac{1}{\Sigma_i} (x_i-\mu_i)^2}_A$$

A - always positive

- small value  $(x_i-\mu_i) \rightarrow$  close to mean

-  $\Sigma_i$  - scales this value

$\Rightarrow \Sigma_i$  - big  $\Rightarrow$  uncertain about this dimension  $\Rightarrow$  large deviation from mean doesn't matter

$\Sigma_i$  - small  $\Rightarrow$  certain about this dimension  $\Rightarrow$  large deviation from mean matters a lot