## Note for Probability Theory and Statistics

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#### 1 Distribution

Multivariate normal distribution

$$f_{x}(x_{1},\dots,x_{k}) = \frac{e^{-\frac{(x-\mu)^{T}C^{-1}(x-\mu)}{2}}}{\sqrt{(2\pi)^{k}|C|}}$$

Where,

 $oldsymbol{C}$  The symmetric covariance matrix.

 $\boldsymbol{x}$  A real k-dimensional column vector.

#### 2 Distance

### 2.1 Distance of Random Variables (Same Distribution, Covariance Matrix=C)

Mahalanobis Distance

(1) Distance of vector  $\boldsymbol{x}$  and set:

$$D_M(\boldsymbol{x}) = \sqrt{(\boldsymbol{x} - \boldsymbol{\mu})^T \boldsymbol{C}^{-1} (\boldsymbol{x} - \boldsymbol{\mu})}$$

If x is a k-dimensional Gaussian random vector with mean vector  $\mu$  and rank k covariance matrix C, then the Mahalanobis distance  $D_M^2(x)$  follows a  $\chi^2$ -distribution with d degrees of freedom.

(2) Distance of two vectors  $\boldsymbol{x}$  and  $\boldsymbol{y}$ :

$$d(\boldsymbol{x}, \boldsymbol{y}) = \sqrt{(\boldsymbol{x} - \boldsymbol{\mu})^T \boldsymbol{C}^{-1} (\boldsymbol{x} - \boldsymbol{\mu})}$$

#### 3 Error Function

#### 3.1 Approximation

$$erf(x) \approx 1 - \frac{1}{(1 + a_1x + a_2x^2 + a_3x^3 + a_4x^4)^4} \quad x \ge 0$$

Where  $a_1 = 0.278393, a_2 = 0.230389, a_3 = 0.000972, a_4 = 0.078108, and its maximum error: <math>5 \times 10^{-4}$ 

# References