



Encontrar μ_{ab} y Σ_{ab} completando cuadrados Para $P(x_a|x_b)$

Apartir de la distribución del termino ~~x_b~~ ~~x_a~~

$$-\frac{1}{2}(x_a - \mu_a, x_b - \mu_b)^T \Sigma^{-1} (x_a - \mu_a, x_b - \mu_b) \quad \text{gaussiana condicional multivariante}$$

Distancia "norme yard"

$$-\frac{1}{2}(x - \mu)^T \Sigma^{-1} (x - \mu), \text{ sin descomponer } x$$

Abra, $P(x) = P([x_a, x_b])$; $P(x_a|x_b) = N(x_a | \mu_{ab}, \Sigma_{ab})$

Se toma x_b como una constante

Reescribiendo: distributiva

$$-\frac{1}{2}(x - \mu)^T \Sigma^{-1} (x - \mu) = -\frac{1}{2} [(x_a - \mu_a)^T \Sigma_{aa}^{-1} (x_a - \mu_a) + (x_b - \mu_b)^T \Sigma_{ba}^{-1} (x_a - \mu_a) + (x_a - \mu_a)^T \Sigma_{ab}^{-1} (x_b - \mu_b) + (x_b - \mu_b)^T \Sigma_{bb}^{-1} (x_b - \mu_b)]$$

$$= -\frac{1}{2} [x_a^T \Sigma_{aa}^{-1} x_a + x_a^T \Sigma_{aa}^{-1} \mu_a \mu_a^T - \mu_a^T \Sigma_{aa}^{-1} x_a + \mu_a^T \Sigma_{aa}^{-1} \mu_a + x_b^T \Sigma_{ba}^{-1} x_a - x_b^T \Sigma_{ba}^{-1} \mu_a - \mu_b^T \Sigma_{ba}^{-1} x_a + \mu_b^T \Sigma_{ba}^{-1} \mu_a + x_a^T \Sigma_{ab}^{-1} x_b - x_a^T \Sigma_{ab}^{-1} \mu_b - \mu_a^T \Sigma_{ab}^{-1} x_b + \mu_a^T \Sigma_{ab}^{-1} \mu_b + x_b^T \Sigma_{bb}^{-1} x_b - x_b^T \Sigma_{bb}^{-1} \mu_b - \mu_b^T \Sigma_{bb}^{-1} x_b + \mu_b^T \Sigma_{bb}^{-1} \mu_b]$$

Se toman los valores que dependen de x_a y x_b en x_a

El termino cuadrático en x_a : $-\frac{1}{2} x_a^T \Sigma_{aa}^{-1} x_a$ tiene la información de la matriz de covarianza

del termino cuadrático $\Sigma_{ab} = \Sigma_{aa}^{-1}$

Ahora buscamos los términos lineales en x_a

$$\frac{1}{2} \left[x_a^T \Delta_{aa} u_a - \underline{u_a^T \Delta_{aa} x_a} + x_b^T \Delta_{ba} x_a - u_b^T \Delta_{ba} x_a + x_a^T \Delta_{ab} x_b - x_a^T \Delta_{ab} u_b \right]$$

se tiene, se pueden factorizar (factor común en x_a^T)

$$x_a^T \Delta_{aa} u_a - x_a^T \Delta_{ab} x_b + x_a^T \Delta_{ab} u_b = x_a^T (\Delta_{aa} u_a - \Delta_{ab} x_b + \Delta_{ab} u_b)$$

$$\text{se factoriza } \Delta_{ab} = x_a^T (\Delta_{aa} u_a - \Delta_{ab} (x_b - u_b))$$

Se busca despejar el término lineal en x donde $x^T \tilde{A}^{-1} u$

$$\underline{x_a^T \tilde{A}^{-1} \Delta_{ab} u_b} = x_a^T (\Delta_{aa} u_a - \Delta_{ab} (u_b - x_b))$$

Como $\tilde{A}^{-1} \Delta_{ab} = \Delta_{aa}$, se multiplica a ambos lados por \tilde{A}_{ab}

x_a^T se cancela (ambos lados) (No se cancela, se plantea $A^{-1} A = I$)

$$\tilde{A}_{ab} \tilde{A}^{-1} \Delta_{ab} u_b = \tilde{A}_{ab} \Delta_{aa} u_a - \tilde{A}_{ab} \Delta_{ab} (u_b - x_b) \quad \tilde{A}_{ab}^{-1} = \Delta_{aa}$$

$$u_b = u_a - \Delta_{aa}^{-1} \Delta_{ab} (u_b - x_b)$$

se deja todo en términos de Δ
 $\Delta_{aa}^{-1} = \tilde{A}_{ab}$

luego se reemplaza Δ por \tilde{A}



$$MM^T = I - M^T M$$

$$\Sigma^{-1} = \begin{bmatrix} \Sigma_{aa} & \Sigma_{ab} \\ \Sigma_{ba} & \Sigma_{bb} \end{bmatrix}^{-1} = \Delta = \begin{bmatrix} \Delta_{aa} & \Delta_{ab} \\ \Delta_{ba} & \Delta_{bb} \end{bmatrix} \quad \begin{bmatrix} A & B \\ C & D \end{bmatrix}^{-1} =$$

$$\Delta = \Sigma^{-1} = \begin{bmatrix} (\Sigma_{aa} - \Sigma_{ab} \Sigma_{bb}^{-1} \Sigma_{ba})^{-1} & -(\Sigma_{aa} - \Sigma_{ab} \Sigma_{bb}^{-1} \Sigma_{ba})^{-1} \Sigma_{ab} \Sigma_{bb}^{-1} \\ -\Sigma_{bb}^{-1} \Sigma_{ba} (\Sigma_{aa} - \Sigma_{ab} \Sigma_{bb}^{-1} \Sigma_{ba})^{-1} & \Sigma_{bb}^{-1} + \Sigma_{bb}^{-1} \Sigma_{ba} (\Sigma_{aa} - \Sigma_{ab} \Sigma_{bb}^{-1} \Sigma_{ba})^{-1} \Sigma_{ab} \Sigma_{bb}^{-1} \end{bmatrix} = \begin{bmatrix} M & -MBD^{-1} \\ -D^{-1}CM & D^{-1} + D^{-1}CMBD^{-1} \end{bmatrix}$$

$$\mu_{ab} = \mu_a - \Delta_{aa}^{-1} \Delta_{ab} (\mu_b - x_b)$$

$$= \mu_a - \underbrace{(\Sigma_{aa} - \Sigma_{ab} \Sigma_{bb}^{-1} \Sigma_{ba})^{-1}}_{M^{-1}} \underbrace{(\Sigma_{aa} - \Sigma_{ab} \Sigma_{bb}^{-1} \Sigma_{ba})^{-1} \Sigma_{ab} \Sigma_{bb}^{-1}}_{M} (\mu_b - x_b)$$

$$\boxed{\mu_{bb} = \mu_a + \Sigma_{ab} \Sigma_{bb}^{-1} (\mu_b - x_b)}$$

$$\Sigma_{ab} = \Delta_{aa}^{-1}$$

$$\Sigma_{abb}^{-1} = (\Sigma_{aa} - \Sigma_{ab} \Sigma_{bb}^{-1} \Sigma_{ba})^{-1}$$

$$\boxed{\Sigma_{ab} = \Sigma_{aa} - \Sigma_{ab} \Sigma_{bb}^{-1} \Sigma_{ba}} = \Delta_{aa}^{-1}$$

$$M = (A - BD^{-1}C)^{-1}$$

M^{-1} = complementary Schur

$$P(x_b | x_a) = N(x_b | \mu_{b|a}, \Sigma_{b|a}) \quad ; \quad P(x) = P([x_a, x_b])$$

Se distribuye la distancia de Mahalanobis (euclídea elíptica) $\frac{1}{2}(x-\mu)^T \Sigma^{-1}(x-\mu)$
 y se sustituye $\Sigma^{-1} = \Delta \rightarrow$ (precisión)
 \rightarrow (covarianza)

$$= \frac{1}{2} [(x_a - \mu_a)^T \Delta_{aa} (x_a - \mu_a) + (x_b - \mu_b)^T \Delta_{ba} (x_a - \mu_a) + (x_a - \mu_a)^T \Delta_{ab} (x_b - \mu_b) + (x_b - \mu_b)^T \Delta_{bb} (x_b - \mu_b)]$$

$$= \frac{1}{2} [x_a^T \Delta_{aa} x_a - x_a^T \Delta_{aa} \mu_a - \mu_a^T \Delta_{aa} x_a + \mu_a^T \Delta_{aa} \mu_a + x_b^T \Delta_{ba} x_a + x_b^T \Delta_{ba} \mu_a - \mu_b^T \Delta_{ba} x_a + \mu_b^T \Delta_{ba} \mu_a + x_a^T \Delta_{ab} x_b - x_a^T \Delta_{ab} \mu_b - \mu_a^T \Delta_{ab} x_b + \mu_a^T \Delta_{ab} \mu_b + x_b^T \Delta_{bb} x_b - x_b^T \Delta_{bb} \mu_b - \mu_b^T \Delta_{bb} x_b + \mu_b^T \Delta_{bb} \mu_b]$$

Se toma el término cuadrático de x_b

$$\frac{1}{2} [x_b^T \Delta_{bb} x_b], \Rightarrow \Delta_{bb} \text{ tiene la información de la covarianza } \Sigma_{b|a}$$

$$\boxed{\Sigma_{b|a} = \Delta_{bb}^{-1}}$$

Para los términos lineales de x_b

$$\frac{1}{2} [-x_b^T \Delta_{ba} x_a + x_b^T \Delta_{ba} \mu_a - x_b^T \Delta_{bb} \mu_b - \mu_b^T \Delta_{bb} x_b + x_a^T \Delta_{ab} x_b - \mu_a^T \Delta_{ab} x_b]$$

$$- x_a^T \Delta_{aa} \mu_a - \mu_a^T \Delta_{aa} x_a$$



$x_b, x_a = \text{Vector}$



Se agrupan con x_b^T

$$\begin{aligned} -x_b^T \Delta_{ba} x_a + x_b^T \Delta_{ba} \mu_a + x_b^T \Delta_{bb} \mu_b &= x_b^T (-\Delta_{ba} x_a + \Delta_{ba} \mu_a + \Delta_{bb} \mu_b) \\ &= x_b^T (\Delta_{bb} \mu_b + \Delta_{ba} (\mu_a - x_a)) \end{aligned}$$

con $x^T \Sigma^{-1} \mu$; $\Sigma_{bb}^{-1} = \Delta_{bb}$

$$x_b^T \Sigma_{b|a}^{-1} \mu_{b|a} = x_b^T (\Delta_{bb} \mu_b + \Delta_{ba} (\mu_a - x_a)) \quad \beta = \alpha$$

Se multiplica a ambos lados por $\Sigma_{b|a}$ y se "cancela" x_b^T

$$\underbrace{\Sigma_{b|a} \Sigma_{b|a}^{-1}}_I \mu_{b|a} = \Sigma_{b|a} (\Delta_{bb} \mu_b + \Delta_{ba} (\mu_a - x_a))$$

$$\mu_{b|a} = \underbrace{\Sigma_{b|a} \Delta_{bb}}_I \mu_b + \Sigma_{b|a} \Delta_{ba} (\mu_a - x_a) \quad (\Delta_{ba} = \Sigma_{ba}^{-1})$$

$$\mu_{b|a} = \mu_b + \Sigma_{b|a} \Delta_{ba} (\mu_a - x_a)$$

$$\mu_{b|a} = \mu_b + \Sigma_{b|a} \Sigma_{ba}^{-1} (\mu_a - x_a)$$

$$\boxed{\mu_{b|a} = \mu_b + \Delta_{bb}^{-1} \Delta_{ba} (\mu_a - x_a)}$$

$$\Delta_{ba} = -\Sigma_{bb}^{-1} \Sigma_{ba} (\Sigma_{aa} - \Sigma_{ab} \Sigma_{bb}^{-1} \Sigma_{ba})^{-1}$$

$$\Delta b/a = \Delta b b^{-1}$$

$$\Delta b b = \underbrace{\Sigma b b^{-1}}_{D^{-1}} + \underbrace{\Sigma b b^{-1}}_{D^{-1}} \underbrace{\Sigma b a}_{C} (\underbrace{\Sigma a a - \Sigma a b \Sigma b b^{-1} \Sigma b a}_{M})^{-1} \underbrace{\Sigma a b}_{B} \underbrace{\Sigma b b^{-1}}_{D^{-1}}$$

$$\Delta b/a = (\Sigma b b^{-1} + \Sigma b b^{-1} \Sigma b a (\Sigma a a - \Sigma a b \Sigma b b^{-1} \Sigma b a)^{-1} \Sigma a b \Sigma b b^{-1})^{-1}$$

$$M|A \supset D - C A^{-1} B$$

$$\mu_{b/a} = \mu_b + \Delta b b^{-1} \Delta b a (\mu_a - x_a)$$

$$(A-B)^{-1} = A^{-1} + A^{-1} B (A-B)^{-1}$$

$$\mu_{b/a} = \mu_b + (\Sigma b b^{-1} + \Sigma b b^{-1} \Sigma b a (\Sigma a a - \Sigma a b \Sigma b b^{-1} \Sigma b a)^{-1} \Sigma a b \Sigma b b^{-1})^{-1} \\ (- \Sigma b b^{-1} \Sigma b a (\Sigma a a - \Sigma a b \Sigma b b^{-1} \Sigma b a)^{-1}) (\mu_a - x_a)$$

