

Standard MVN

$$\vec{z} \sim N_n(\vec{0}_n, I_n)$$

$$E[\vec{z}] = \vec{0}_n$$

$$\text{Var}[\vec{z}] = I_n$$

$$\text{If } \vec{x} = \vec{z} + \vec{\mu} \sim N(\vec{\mu}, I_n)$$

$$\text{If } \vec{x} = A\vec{z}, \quad E[\vec{x}] = AE[\vec{z}] = A\vec{0}_n = \vec{0}_m$$

$$\Sigma = \text{Var}[\vec{x}] = AA^T \in \mathbb{R}^{m \times n}.$$

Is  $\Sigma$  symmetric?

$$\Sigma \stackrel{?}{=} \Sigma^T$$

$$\Sigma^T = (AA^T)^T = \bar{A}^T \cdot A^T = AA^T \checkmark$$

We want  $f_{\vec{x}}(\vec{x})$ ?  
(jdf).

$$\vec{x} = g(\vec{z}) = A\vec{z} \quad \text{Assume } A \in \mathbb{R}^{n \times n} \text{ for now.}$$

we can say  $\vec{z} = h(x)$  where  $h$  is the inverse function.

$$\vec{z} = h(x) = A^{-1} \vec{x} \quad g \text{ is only 1:1 if } A \text{ is full rank.}$$

We need  $m=n$  i.e.  $A$  is square and  $A$  is full rank

let's do multivariate change of variable like before.

$$A^{-1} \vec{x} = \begin{bmatrix} h_1(\vec{x}) \\ h_2(\vec{x}) \\ \vdots \\ h_n(\vec{x}) \end{bmatrix}$$

$$f_{\vec{x}}(\vec{x}) = f_{\vec{z}}(h(\vec{x})) |J_n(\vec{x})|$$

generalization  
of univariate

$$\text{Let } B = A^{-1}$$

$$B = A^{-1} = \begin{bmatrix} \leftarrow b_{1.} \rightarrow \\ \leftarrow b_{2.} \rightarrow \\ \vdots \\ \leftarrow b_{n.} \rightarrow \end{bmatrix} = \begin{bmatrix} \uparrow & \uparrow & & \uparrow \\ b_{.1} & b_{.2} & \dots & b_{.n} \\ \downarrow & \downarrow & & \downarrow \end{bmatrix} \stackrel{\text{canonical notation}}{=} \begin{bmatrix} b_{11} & \dots & b_{1n} \\ \vdots & & \vdots \\ b_{n1} & \dots & b_{nn} \end{bmatrix}$$

$$\vec{z} = \vec{B} \vec{x} \stackrel{\parallel h(\vec{x})}{=} \begin{cases} h_1(\vec{x}) = \vec{b}_{1.} \vec{x} = b_{11}x_1 + \dots + b_{1n}x_n \\ h_2(\vec{x}) = \vec{b}_{2.} \vec{x} = b_{21}x_1 + \dots + b_{2n}x_n \\ \vdots \\ h_n(\vec{x}) = \vec{b}_{n.} \vec{x} = b_{n1}x_1 + \dots + b_{nn}x_n \end{cases}$$

$$J_n(\vec{x}) = \det \begin{pmatrix} \frac{\partial}{\partial x_1} [h_1(\vec{x})] & \dots & \frac{\partial}{\partial x_n} [h_1(\vec{x})] \\ \vdots & & \vdots \\ \frac{\partial}{\partial x_1} [h_n(\vec{x})] & \dots & \frac{\partial}{\partial x_n} [h_n(\vec{x})] \end{pmatrix}$$

$$= \det \begin{pmatrix} b_{11} & \dots & b_{1n} \\ \vdots & & \vdots \\ b_{n1} & \dots & b_{nn} \end{pmatrix} = \det(B) = \det(A^{-1})$$

$$(*) f_{\vec{x}}(\vec{x}) = f_{\vec{z}}(h(\vec{x})) |J_n(\vec{x})| = f_{\vec{z}}(\vec{A}^{-1} \vec{x}) | \det(A^{-1}) | \quad \text{Side note: } \frac{\partial}{\partial \vec{x}} [c \vec{x}] = c$$



Recall  $\det(A^{-1}) = \frac{1}{\det(A)}$

Proof: Recall  $AA^{-1} = I$   
 $\det(AA^{-1}) = \det(I)$

= 1  
 (product of the diagonal)

Recall  $\det(AB) = \det(A) \cdot \det(B)$

So,  $\det(A) \cdot \det(A^{-1}) = 1$

$\therefore \det(A^{-1}) = \frac{1}{\det(A)}$

$(A^T)^{-1} = (A^{-1})^T$

$AA^{-1} = I$

$(AA^{-1})^T = I^T = I$

$(A^{-1})^T \cdot A^T = I$

$A^T(A^T)^{-1} = I$

$(A^T)^{-1}A^T = I$

equal

So  $(A^T)^{-1} = (A^{-1})^T$

$\Sigma = AA^T$

$\Sigma^{-1} = (AA^T)^{-1} = (A^T)^{-1}A^{-1}$

$(AB)^{-1}(AB) = I$

It has to be

$\underbrace{B^{-1}A^{-1}}_I \cdot AB = I$

So  $(AB)^{-1} = B^{-1}A^{-1}$

$\det(\Sigma) = \det(A) \cdot \det(A^T)$   
 $= \det(A)^2$

$\Rightarrow \det A = \sqrt{\det \Sigma}$

$\det A = \det(A^T)$

Now  $f_{\vec{x}}(\vec{x}) = \frac{1}{(2\pi)^{n/2}} e^{-\frac{1}{2} (A^{-1}\vec{x})^T (A^{-1}\vec{x})} \cdot \frac{1}{|\det(A)|}$

$= \frac{1}{(2\pi)^{n/2}} e^{-\frac{1}{2} \vec{x}^T (A^{-1})^T \cdot A^{-1} \vec{x}} \cdot \frac{1}{|\det(A)|}$

$= \frac{1}{(2\pi)^{n/2}} e^{-\frac{1}{2} \vec{x}^T \underbrace{(A^T)^{-1} A^{-1}}_I \vec{x}} \cdot \frac{1}{\det(A)}$

$= \frac{1}{(2\pi)^{n/2}} \cdot e^{-\frac{1}{2} \vec{x}^T \Sigma^{-1} \vec{x}} \cdot \frac{1}{\det(A)}$

$= \frac{1}{\sqrt{(2\pi)^n \det(\Sigma)}} e^{-\frac{1}{2} \vec{x}^T \Sigma^{-1} \vec{x}}$

$f_{\vec{x}}(\vec{x}) = N_n(\vec{0}_n, \Sigma)$

centered at zero

$\Sigma = AA^T$

$\vec{x} = A\vec{z} + \mu \sim N_n(\vec{\mu}, \Sigma)$   
 general MVN  
 $\cdot Y = \vec{x} - \mu = A\vec{z}$

$N_n(\vec{\mu}, \Sigma) = \frac{1}{\sqrt{(2\pi)^n \det(\Sigma)}} e^{-\frac{1}{2} (\vec{x} - \vec{\mu})^T \Sigma^{-1} (\vec{x} - \vec{\mu})}$

characteristic function:

Recall:  $\phi_X(t) := E[e^{itX}]$  generalizes to  $\phi_{\vec{X}}(\vec{t}) = E[e^{i\vec{t}^T \vec{X}}$

Rules: Uni dimensional rules are valid for multidim.

$$\begin{aligned}\phi_{\vec{X}_1 + \vec{X}_2}(\vec{t}) &= E[e^{i\vec{t}^T (\vec{X}_1 + \vec{X}_2)}] = E[e^{i\vec{t}^T \vec{X}_1} \cdot e^{i\vec{t}^T \vec{X}_2}] \\ &= E[e^{i\vec{t}^T \vec{X}_1}] E[e^{i\vec{t}^T \vec{X}_2}] = \phi_{\vec{X}_1}(\vec{t}) \cdot \phi_{\vec{X}_2}(\vec{t})\end{aligned}$$

characteristic function

Recall:  $\phi_X(t) := E[e^{it^T X}]$  generalizes to

$$\phi_{\vec{X}}(\vec{t}) = E[e^{i\vec{t}^T \vec{X}}]$$

Rules: uni dimensional rules are valid for multi dime

$$\phi_{\vec{X}_1 + \vec{X}_2}(\vec{t}) = E[e^{i\vec{t}^T (\vec{X}_1 + \vec{X}_2)}] = E[e^{i\vec{t}^T \vec{X}_1} \cdot e^{i\vec{t}^T \vec{X}_2}]$$

$$= E[e^{i\vec{t}^T \vec{X}_1}] E[e^{i\vec{t}^T \vec{X}_2}] = \phi_{\vec{X}_1}(\vec{t}) \cdot \phi_{\vec{X}_2}(\vec{t})$$

$$\text{Let } \vec{Y} = A\vec{X} + \vec{c}$$

$$A \in \mathbb{R}^{m \times n}, \vec{c} \in \mathbb{R}^m$$

$$\vec{X} \in \mathbb{R}^n, \vec{Y} \in \mathbb{R}^m$$

$$\begin{aligned} \phi_{\vec{Y}}(\vec{t}) &= E[e^{i\vec{t}^T \vec{Y}}] = E[e^{i\vec{t}^T (A\vec{X} + \vec{c})}] \\ &= E[e^{i\vec{t}^T A\vec{X}} \cdot e^{i\vec{t}^T \vec{c}}] = e^{i\vec{t}^T \vec{c}} E[e^{i\vec{t}^T A\vec{X}}] \end{aligned}$$

$$\text{Let } \vec{t}'^T = \vec{t}^T A, \vec{t}' = (\vec{t}^T A)^T = A^T \vec{t}$$

$$= e^{i\vec{t}^T \vec{c}} E[e^{i\vec{t}'^T \vec{X}}] = e^{i\vec{t}^T \vec{c}} \phi_{\vec{X}}(\vec{t}')$$

$$= e^{i\vec{t}^T \vec{c}} \phi_{\vec{X}}(A^T \vec{t})$$

Need to get characteristic function of  $\vec{Z}$ .

$$\vec{Z} \sim N_n(\vec{0}_n, I_n) \Rightarrow \phi_{\vec{Z}}(\vec{t}) = E[e^{i\vec{t}^T \vec{Z}}]$$

$$= \int_{\mathbb{R}} \dots \int_{\mathbb{R}} e^{i\vec{t}^T \vec{z}} f_{\vec{Z}}(\vec{z}) d\vec{z}$$

$$= \int_{\mathbb{R}} \dots \int_{\mathbb{R}} e^{i(t_1 z_1 + \dots + t_n z_n)} \cdot \frac{1}{(\sqrt{2\pi})^n} e^{-\frac{1}{2}(z_1^2 + \dots + z_n^2)} dz_1 \dots dz_n$$



$$\Sigma = A A^T$$

$$\Sigma^{-1} = (A A^T)^{-1} \\ = (A^T)^{-1} A^{-1}$$

$$(AB)^{-1} (AB) = I$$

$$\text{It has to be } \underbrace{B^{-1} A^{-1} A B}_{I} = I$$

$$\text{So } (AB)^{-1} = B^{-1} A^{-1}$$

$$\text{So } \Rightarrow = \frac{1}{(2\pi)^{n/2}} e^{-\frac{1}{2} \vec{x}^T \Sigma^{-1} \vec{x}} \frac{1}{\det(A)}$$

$$\det(\Sigma) = \det(A) \det(A^T) = \det(A)^2$$

$$\Rightarrow \det(A) = \sqrt{\det(\Sigma)}$$

$$\det(A) = \det(A^T)$$

$$= \frac{1}{\sqrt{(2\pi)^n \det(\Sigma)}} e^{-\frac{1}{2} \vec{x}^T \Sigma^{-1} \vec{x}}$$

$$= N_n(\vec{0}_n, \Sigma)$$

↑  
centered  
at zero

$$\Sigma = A A^T$$

$$\textcircled{*} \vec{x} = A \vec{z} + \mu \sim N_n(\vec{\mu}, \Sigma) = \frac{1}{\sqrt{(2\pi)^n \det(\Sigma)}} e^{-\frac{1}{2} (\vec{x} - \vec{\mu})^T \Sigma^{-1} (\vec{x} - \vec{\mu})}$$

↑  
general  
MVN

$$\bullet \vec{y} = \vec{x} - \mu = A \vec{z}$$