



# Affine transformation applied to a multivariate Gaussian random variable - what is the mean vector and covariance matrix of the new variable?

Asked 9 years ago   Modified 4 years, 5 months ago   Viewed 34k times



40



29



Given a random vector  $\mathbf{x} \sim N(\bar{\mathbf{x}}, \mathbf{C}_x)$  with normal distribution.  $\bar{\mathbf{x}}$  is the mean value vector and  $\mathbf{C}_x$  is the covariance matrix of  $\mathbf{x}$ .

An affine transformation is applied to the  $\mathbf{x}$  vector to create a new random  $\mathbf{y}$  vector:

$$\mathbf{y} = \mathbf{A}\mathbf{x} + \mathbf{b}$$

Can we find mean value  $\bar{\mathbf{y}}$  and covariance matrix  $\mathbf{C}_y$  of this new vector  $\mathbf{y}$  in terms of already given parameters ( $\bar{\mathbf{x}}$ ,  $\mathbf{C}_x$ ,  $\mathbf{A}$  and  $\mathbf{b}$ )?

Can you please show the steps. Once I learn the method, I will use it on several other distributions myself.

linear-algebra

multivariable-calculus

normal-distribution

random-variables

Share   Cite   Follow

edited Apr 21, 2016 at 18:20



Cong Ma

115   6

asked Mar 17, 2013 at 2:19



hkBattousai

4,095   7   36   53

2   Hint:  $\bar{\mathbf{y}} = E[\mathbf{y}] = E[\mathbf{A}\mathbf{x} + \mathbf{b}]$ . Now apply *linearity of expectation*.

$\mathbf{C}_y = E[(\mathbf{y} - \bar{\mathbf{y}})(\mathbf{y} - \bar{\mathbf{y}})^T] = E[\mathbf{y}\mathbf{y}^T] - E[\bar{\mathbf{y}}\bar{\mathbf{y}}^T]$ . – Dilip Sarwate Mar 17, 2013 at 2:46

1 Answer

Sorted by:

Highest score (default)



We find the mean of  $\mathbf{y}$  by using the fact that  $\mathbb{E}\{\}$  is a linear operator.

49

$$\bar{\mathbf{y}} = \mathbb{E}\{\mathbf{y}\} = \mathbb{E}\{\mathbf{A}\mathbf{x} + \mathbf{b}\} = \mathbf{A}\mathbb{E}\{\mathbf{x}\} + \mathbf{b} = \mathbf{A}\bar{\mathbf{x}} + \mathbf{b}$$



Then we find covariance of





$$\begin{aligned}
 \mathbf{C}_y &\triangleq \mathbb{E}\{(\mathbf{y} - \bar{\mathbf{y}})(\mathbf{y} - \bar{\mathbf{y}})^\top\} \\
 &= \mathbb{E}\left\{\left[(\mathbf{A}\mathbf{x} + \mathbf{b}) - (\mathbf{A}\bar{\mathbf{x}} + \mathbf{b})\right]\left[(\mathbf{A}\mathbf{x} + \mathbf{b}) - (\mathbf{A}\bar{\mathbf{x}} + \mathbf{b})\right]^\top\right\} \\
 &= \mathbb{E}\left\{\left[\mathbf{A}(\mathbf{x} - \bar{\mathbf{x}})\right]\left[\mathbf{A}(\mathbf{x} - \bar{\mathbf{x}})\right]^\top\right\} \\
 &= \mathbb{E}\left\{\mathbf{A}(\mathbf{x} - \bar{\mathbf{x}})(\mathbf{x} - \bar{\mathbf{x}})^\top \mathbf{A}^\top\right\} \\
 &= \mathbf{A}\mathbb{E}\left\{(\mathbf{x} - \bar{\mathbf{x}})(\mathbf{x} - \bar{\mathbf{x}})^\top\right\}\mathbf{A}^\top \\
 &= \mathbf{A}\mathbf{C}_x\mathbf{A}^\top
 \end{aligned}$$

Then,  $\mathbf{y}$  is defined as,

$$\mathbf{y} \sim \mathcal{N}(\mathbf{A}\bar{\mathbf{x}} + \mathbf{b}, \mathbf{A}\mathbf{C}_x\mathbf{A}^\top)$$

That is,

$$f_Y(\mathbf{y}) = \frac{1}{\sqrt{|2\pi\mathbf{A}\mathbf{C}_x\mathbf{A}^\top|}} \exp\left(-\frac{1}{2}\left[\mathbf{y} - (\mathbf{A}\bar{\mathbf{x}} + \mathbf{b})\right]^\top (\mathbf{A}\mathbf{C}_x\mathbf{A}^\top)^{-1} \left[\mathbf{y} - (\mathbf{A}\bar{\mathbf{x}} + \mathbf{b})\right]\right)$$

Share Cite Follow

edited Oct 16, 2017 at 20:44



Atcold

213 2 10

answered Mar 17, 2013 at 12:07



hkBattousai

4,095 7 36 53

Are there restrictions on  $\mathbf{A}$ ? I'm thinking, for example, about  $\mathbf{A}=\{\{1,1\},\{0,0\}\}$  (Ones are top row, os are bottom row) and  $\sigma=\{\{1,0\},\{0,1\}\}$ . Then  $\text{rank } \mathbf{A} \cdot \text{sigma} \cdot \text{Transpose}(\mathbf{A})$  is 1 and not invertible any more.  
– BenB Jun 1, 2015 at 2:20

@BenB If the rank of  $\mathbf{A}$  is not full, then the transformation is not linear. And that is not the case mentioned in the question statement. – hkBattousai Feb 6, 2016 at 13:24

- 4 Not all linear transformations have full rank. If the rank isn't full, the formula  $y \sim N(\mathbf{A}\bar{\mathbf{x}} + \mathbf{b}, \mathbf{A}\mathbf{C}_x\mathbf{A}^\top)$  still works. The pdf also works if interpreted to use pseudoinverses and pseudodeterminants. – Geoffrey Irving May 8, 2016 at 4:58

To be even more complete this answer could include the cross covariance! nice answer btw – Marco Aguiar Feb 26, 2018 at 13:20