

Expectation / Variance

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1 Notations

The following document summarizes rules for computing the expected value, variance, and covariance of random variables and random vectors. x_i denotes a random variable with expectation value $m_i = E[x_i]$, variance $s_i = \text{Var}[x_i]$ and covariance $s_{ij} = \text{Cov}[x_i, x_j]$. a_i and b_i are constant scalars. \mathbf{x}_i denotes a random vector with expectation vector $\mathbf{m}_i = E[\mathbf{x}_i]$, variance matrix $S_i = \text{Var}[\mathbf{x}_i]$ and covariance matrix $S_{ij} = \text{Cov}[\mathbf{x}_i, \mathbf{x}_j]$. A_i and \mathbf{b}_i is a constant matrix and vector.

2 Expectation

2.1 Random variable

$$\begin{aligned} \mathbb{E}[x] &= \int p(x)x \, dx \\ \mathbb{E}[ax + b] &= am + b \\ \mathbb{E}[a_1x_1 + a_2x_2 + b] &= a_1m_1 + a_2m_2 + b \\ \mathbb{E}\left[\prod_{i=1}^N x_i\right] &= \prod_{1 \leq i \leq N} m_i + \sum_{1 \leq i < j \leq N} s_{ij} \\ \mathbb{E}[x_1x_2] &= m_1m_2 + s_{12} \\ \mathbb{E}[x^2] &= m^2 + s \end{aligned}$$

2.2 Random vector

$$\begin{aligned} \mathbb{E}[\mathbf{x}] &= \int p(\mathbf{x})\mathbf{x} \, d\mathbf{x} \\ \mathbb{E}[A\mathbf{x} + \mathbf{b}] &= A\mathbf{m} + \mathbf{b} \\ \mathbb{E}[A_1\mathbf{x}_1 + A_2\mathbf{x}_2 + \mathbf{b}] &= A_1\mathbf{m}_1 + A_2\mathbf{m}_2 + \mathbf{b} \\ \mathbb{E}[(\mathbf{x}_1 - \mathbf{b}_1)^T A(\mathbf{x}_2 - \mathbf{b}_2)] &= (\mathbf{m}_1 - \mathbf{b}_1)^T A(\mathbf{m}_2 - \mathbf{b}_2) + \text{Tr}[A^T S_{12}] \\ \mathbb{E}[\mathbf{x}_1^T A\mathbf{x}_2] &= \mathbf{m}_1^T A\mathbf{m}_2 + \text{Tr}[A^T S_{12}] \\ \mathbb{E}[\mathbf{x}_1^T \mathbf{x}_2] &= \mathbf{m}_1^T \mathbf{m}_2 + \text{Tr}[S_{12}] \\ \mathbb{E}[X_1^T A X_2]_{ij} &= \mathbb{E}[X_{1:i}]^T A \mathbb{E}[X_{2:j}] + \text{Tr}[A^T \text{Cov}[X_{1:i}, X_{2:j}]] \\ \mathbb{E}[(\mathbf{x}_1 - \mathbf{b}_1)(\mathbf{x}_2 - \mathbf{b}_2)^T] &= (\mathbf{m}_1 - \mathbf{b}_1)(\mathbf{m}_2 - \mathbf{b}_2)^T + S_{12} \\ \mathbb{E}[\mathbf{x}_1 \mathbf{x}_2^T] &= \mathbf{m}_1 \mathbf{m}_2^T + S_{12} \end{aligned}$$

3 Variance

3.1 Random variable

$$\begin{aligned}\text{Var}[x] &= \int p(x)(x - \mathbb{E}[x])^2 dx = \mathbb{E}[(x - \mathbb{E}[x])^2] \\ \text{Var}[ax + b] &= a^2 s \\ \text{Var}[a_1 x_1 + a_2 x_2 + b] &= a_1^2 s_1 + a_2^2 s_2 + 2s_{12} \\ \text{Var}[a_1 x_1 - a_2 x_2 + b] &= a_1^2 s_1 + a_2^2 s_2 - 2s_{12}\end{aligned}$$

If x_1 and x_2 are independent:

$$\text{Var}[x_1 x_2] = \mathbb{E}[x_1^2] \mathbb{E}[x_2^2] - \mathbb{E}[x_1]^2 \mathbb{E}[x_2]^2$$

3.2 Random vector

$$\begin{aligned}\text{Var}[\mathbf{x}] &= \int p(\mathbf{x})(\mathbf{x} - \mathbb{E}[\mathbf{x}])^2 d\mathbf{x} \\ &= \mathbb{E}[(\mathbf{x} - \mathbb{E}[\mathbf{x}])^2] \\ \text{Var}[A\mathbf{x} + \mathbf{b}] &= A \text{Var}[\mathbf{x}] A^T \\ \text{Var}[A_1 \mathbf{x}_1 + A_2 \mathbf{x}_2 + \mathbf{b}] &= A_1 \text{Var}[\mathbf{x}_1] A_1^T + A_2 \text{Var}[\mathbf{x}_2] A_2^T + 2A_1 S_{12} A_2^T\end{aligned}$$

Variance of quadratic form:

$$\begin{aligned}\text{Var}[\mathbf{x}^T A \mathbf{x}] &= 4\mathbf{m}^T A' S A' \mathbf{m} + 2 \text{Tr}[A' S A' S] \\ A' &= \frac{1}{2}(A + A^T) \\ \text{Var}[\mathbf{x}^T \mathbf{x}] &= 4\mathbf{m}^T S \mathbf{m} + 2 \text{Tr}[SS]\end{aligned}$$

If \mathbf{x}_1 and \mathbf{x}_2 are independent:

$$\begin{aligned}\text{Var}[\mathbf{x}_1^T \mathbf{x}_2] &= \sum_i \mathbb{E}[x_{1i}^2] \mathbb{E}[x_{2i}^2] + \mathbb{E}[x_{1i}]^2 \mathbb{E}[x_{2i}]^2 \\ &= \sum_i \mathbb{E}[x_{1i}]^2 \text{Var}[x_{2i}] + \mathbb{E}[x_{2i}]^2 \text{Var}[x_{1i}] + \text{Var}[x_{1i}] \text{Var}[x_{2i}]\end{aligned}$$

4 Covariance

4.1 Random variable

$$\begin{aligned}\text{Cov}[x_1, x_2] &= \int p(x_1, x_2)(x_1 - \mathbb{E}[x_1])(x_2 - \mathbb{E}[x_2]) \, dx \\ &= \mathbb{E}[(x_1 - \mathbb{E}[x_1])(x_2 - \mathbb{E}[x_2])] \\ \text{Cov}[a_1x_1 + b_1, a_2x_2 + b_2] &= a_1a_2s_{12}\end{aligned}$$

4.2 Random vector

$$\begin{aligned}\text{Cov}[\mathbf{x}_1, \mathbf{x}_2] &= \int p(\mathbf{x}_1, \mathbf{x}_2)(\mathbf{x}_1 - \mathbb{E}[\mathbf{x}_1])(\mathbf{x}_2 - \mathbb{E}[\mathbf{x}_2])^T \, d\mathbf{x} \\ &= \mathbb{E}[(\mathbf{x}_1 - \mathbb{E}[\mathbf{x}_1])(\mathbf{x}_2 - \mathbb{E}[\mathbf{x}_2])^T] \\ \text{Cov}[A_1\mathbf{x}_1 + \mathbf{b}_1, A_2\mathbf{x}_2 + \mathbf{b}_2] &= A_1s_{12}A_2^T\end{aligned}$$