

Machine Learning Assignment 53

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Problem 1

(a) Given that $X \sim U[0, 1]$, compute $\mathbf{Cov}[X, X^2]$.

$$\begin{aligned} E[X] &= \frac{1}{b-a} = 1 \\ E[X^2] &= \frac{a^2 + ab + b^2}{3} = \frac{1}{3} \\ \mathbf{Cov}[X, X^2] &= (x_1 - \bar{x}_1)(x_2 - \bar{x}_2)dx \\ &= \int_0^1 (x - 1)(x^2 - \frac{1}{3})dx \\ &= \int_0^1 x^3 - x^2 - x + 1dx \\ &= \left. \frac{x^4}{4} - \frac{x^3}{3} - \frac{x^2}{6} + \frac{x}{3} \right|_0^1 \\ &= \frac{1}{12} \end{aligned}$$

(b) Given that $X_1, X_2 \sim U[0, 1]$, compute $\mathbf{Cov}[X_1, X_2]$.

$$\begin{aligned}
E[X_1] &= \frac{a+b}{2} = \frac{1}{2} \\
E[X_2] &= \frac{a+b}{2} = \frac{1}{2} \\
\text{Cov}[X, X^2] &= \iint_{[0,1] \times [0,1]} (x_1 - \bar{x}_1)(x_2 - \bar{x}_2) dx \\
&= \iint_{[0,1] \times [0,1]} (x_1 - \frac{1}{2})(x_2 - \frac{1}{2}) dx \\
&= \iint_{[0,1] \times [0,1]} x_1 x_2 - 0.5x_1 - 0.5x_2 + 1 dx \\
&= \left. \frac{(x_1^2 - x)(x_2^2 - x_2)}{4} \right|_0^1 \\
&= 0
\end{aligned}$$

(c) Prove that $\text{Var}[X_1 + X_2] = \text{Var}[X_1] + \text{Var}[X_2] + 2\text{Cov}[X_1, X_2]$.

$$\begin{aligned}
\text{Var}[X_1 + X_2] &= E[(X_1 + X_2) - E[(X_1 + X_2)]]^2 \\
&= E[(X_1 + X_2 - E[X_1] - E[X_2])]^2 \\
&= E[(X_1 - E[X_1]) + (X_2 - E[X_2])]^2 \\
&= E[(X_1 - E[X_1])^2 + 2(X_1 - E[X_1])(X_2 - E[X_2]) + (X_2 - E[X_2])^2] \\
&= E[(X_1 - E[X_1])^2] + E[2(X_1 - E[X_1])(X_2 - E[X_2])] + E[(X_2 - E[X_2])^2] \\
&= \text{Var}[X_1] + \text{Var}[X_2] + 2\text{Cov}[X_1, X_2]
\end{aligned}$$

(d) Prove that $\text{Cov}[X_1, X_2] = E[X_1 X_2] - E[X_1]E[X_2]$.

$$\begin{aligned}
\text{Cov}[X_1, X_2] &= E[(X_1 - E[X_1])(X_2 - E[X_2])] \\
&= E[(X_1 X_2) - X_1 E[X_2] - X_2 E[X_1] + E[X_1]E[X_2]] \\
&= E[X_1 X_2] - E[X_1]E[X_2] - E[X_2]E[X_1] + E[X_1]E[X_2] \\
&= E[X_1 X_2] - E[X_1]E[X_2]
\end{aligned}$$