

Please provide complete and well-written solutions to the following exercises.

Due January 24, at the beginning of class.

Homework 1

Exercise 1. Let A, B be events in a sample space. Let C_1, \dots, C_n be events such that $C_i \cap C_j = \emptyset$ for any $i, j \in \{1, \dots, n\}$ with $i \neq j$, and such that $\cup_{i=1}^n C_i$ is the whole sample space. Show:

$$\mathbf{P}(A|B) = \sum_{i=1}^n \mathbf{P}(A|B, C_i) \mathbf{P}(C_i|B).$$

(Hint: consider using the Total Probability Theorem and that $\mathbf{P}(\cdot|B)$ is a probability law.)

Exercise 2. By definition, a random vector $Z = (Z_1, \dots, Z_d) \in \mathbf{R}^d$ is **Gaussian** if, for any $v_1, \dots, v_d \in \mathbf{R}$, the random variable $\sum_{i=1}^d v_i Z_i$ is a Gaussian random variable. Equivalently, for any $v \in \mathbf{R}^d$, the random variable $\langle v, Z \rangle$ is a Gaussian random variable. The covariance matrix $(a_{ij})_{1 \leq i, j \leq d}$ of Z is defined by

$$a_{ij} := \mathbf{E}((Z_i - \mathbf{E}Z_i)(Z_j - \mathbf{E}Z_j)).$$

Let $Z = (Z_1, \dots, Z_d) \in \mathbf{R}^d$ be a Gaussian random vector.

- Show that the covariance matrix $(a_{ij})_{1 \leq i, j \leq d}$ of Z is symmetric, positive semidefinite. That is, for any $v \in \mathbf{R}^d$, we have

$$v^T a v = \sum_{i, j=1}^d v_i v_j a_{ij} \geq 0.$$

- Given any symmetric positive semidefinite matrix $(b_{ij})_{1 \leq i, j \leq d}$, show that there exists a Gaussian random vector Z such that the covariance matrix of Z is $(b_{ij})_{1 \leq i, j \leq d}$. (Hint: write the matrix b in its Cholesky decomposition $b = r r^*$, where r is a $d \times d$ real matrix. Let $e^{(1)}, \dots, e^{(d)}$ be the rows of r . Let X_1, \dots, X_d be independent standard Gaussian random variables. Let $X := (X_1, \dots, X_d)$. Define $Z_i := \langle X, e^{(i)} \rangle$ for any $1 \leq i \leq d$.)

Exercise 3. Let Y_0, Y_1, \dots be independent standard Gaussian random variables (so that they each have mean zero and variance one). Let $a, b \in \mathbf{R}$ be unknown (deterministic) parameters. For any $n \geq 0$, define

$$X_n := a + bY_n.$$

Suppose the data X_0, X_1, \dots, X_{30} are given by the following

0.7118 0.7587 0.9143 - 0.3666 2.2630 1.5951 0.9470 2.2222 0.3731 1.3387 1.2551 1.5915
1.5877 0.8811 1.3820 1.3011 1.7766 2.0560 2.0656 0.8818 1.4464 0.6715 0.7319 1.3959

2.3196 0.9382 1.6228 1.2646 2.0704 0.7466 1.4195

To the best of your ability, estimate a and b .

Exercise 4. Let μ be a probability measure on \mathbf{R}^n , where \mathbf{R}^n has the Borel σ -algebra. Define the **distribution function** $F: \mathbf{R}^n \rightarrow [0, 1]$ associated to μ by

$$\begin{aligned} F(t_1, \dots, t_n) &:= \mu((-\infty, t_1] \times \dots \times (-\infty, t_n]) \\ &= \mu(\{(x_1, \dots, x_n) \in \mathbf{R}^n: -\infty < x_i \leq t_i, \forall 1 \leq i \leq n\}), \quad \forall t_1, \dots, t_n \in \mathbf{R}. \end{aligned}$$

Show the following properties of F :

- F is nondecreasing. ($F(t_1, \dots, t_n) \leq F(t'_1, \dots, t'_n)$ whenever $t_i \leq t'_i \forall 1 \leq i \leq n$.)
- $\lim_{t_1, \dots, t_n \rightarrow -\infty} F(t_1, \dots, t_n) = 0$ and $\lim_{t_1, \dots, t_n \rightarrow \infty} F(t_1, \dots, t_n) = 1$.
- F is right continuous, i.e. $F(t_1, \dots, t_n) = \lim_{(s_1, \dots, s_n) \rightarrow (t_1, \dots, t_n)^+} F(s_1, \dots, s_n)$ for all $t_1, \dots, t_n \in \mathbf{R}$, where the limit restricts that $s_i \geq t_i \forall 1 \leq i \leq n$.
- If $t_{i,0} \leq t_{i,1} \forall 1 \leq i \leq n$, then

$$\sum_{(\omega_1, \dots, \omega_n) \in \{0,1\}^n} (-1)^{\omega_1 + \dots + \omega_n + n} F(t_{1,\omega_1}, \dots, t_{n,\omega_n}) \geq 0.$$

Exercise 5. A **finite Markov Chain** is a stochastic process (X_0, X_1, X_2, \dots) together with a finite set Ω , which is called the **state space** of the Markov Chain, and an $|\Omega| \times |\Omega|$ real matrix P . The random variables X_0, X_1, \dots take values in the finite set Ω . The matrix P is **stochastic**, that is all of its entries are nonnegative and

$$\sum_{y \in \Omega} P(x, y) = 1, \quad \forall x \in \Omega.$$

And the stochastic process satisfies the following **Markov property**: for all $x, y \in \Omega$, for any $n \geq 1$, and for all events H_{n-1} of the form $H_{n-1} = \bigcap_{k=0}^{n-1} \{X_k = x_k\}$, where $x_k \in \Omega$ for all $0 \leq k \leq n-1$, such that $\mathbf{P}(H_{n-1} \cap \{X_n = x\}) > 0$, we have

$$\mathbf{P}(X_{n+1} = y | H_{n-1} \cap \{X_n = x\}) = \mathbf{P}(X_{n+1} = y | X_n = x) = P(x, y).$$

That is, the next location of the Markov chain only depends on its current location. And the transition probability is defined by $P(x, y)$.

Suppose we have a Markov Chain with state space Ω . Let $n \geq 0$, $\ell \geq 1$, let $x_0, \dots, x_n \in \Omega$ and let $A \subseteq \Omega^\ell$. Using the (usual) Markov property, show that

$$\begin{aligned} \mathbf{P}((X_{n+1}, \dots, X_{n+\ell}) \in A | (X_0, \dots, X_n) = (x_0, \dots, x_n)) \\ = \mathbf{P}((X_{n+1}, \dots, X_{n+\ell}) \in A | X_n = x_n). \end{aligned}$$

Then, show that

$$\mathbf{P}((X_{n+1}, \dots, X_{n+\ell}) \in A | X_n = x_n) = \mathbf{P}((X_1, \dots, X_\ell) \in A | X_0 = x_n).$$

(Hint: it may be helpful to use the Multiplication Rule for conditional probabilities.)

Exercise 6. Let P, Q be stochastic matrices of the same size. Show that PQ is a stochastic matrix. Conclude that, if r is a positive integer, then P^r is a stochastic matrix.

Exercise 7. Let $Y_1, Y_2, \dots: \Omega \rightarrow \mathbf{R}$ be random variables that converge almost surely to a random variable $Y: \Omega \rightarrow \mathbf{R}$. Show that Y_1, Y_2, \dots converges in probability to Y in the following way.

- For any $\varepsilon > 0$ and for any positive integer n , let

$$A_{n,\varepsilon} := \bigcup_{m=n}^{\infty} \{\omega \in \Omega: |Y_m(\omega) - Y(\omega)| > \varepsilon\}.$$

Show that $A_{n,\varepsilon} \supseteq A_{n+1,\varepsilon} \supseteq A_{n+2,\varepsilon} \supseteq \dots$.

- Show that $\mathbf{P}(\bigcap_{n=1}^{\infty} A_{n,\varepsilon}) = 0$.
- Using Continuity of the Probability Law, deduce that $\lim_{n \rightarrow \infty} \mathbf{P}(A_{n,\varepsilon}) = 0$.

Now, show that the converse is false. That is, find random variables Y_1, Y_2, \dots that converge in probability to Y , but where Y_1, Y_2, \dots do not converge to Y almost surely.

Exercise 8. Suppose random variables $Y_1, Y_2, \dots : \Omega \rightarrow \mathbf{R}$ converge in probability to a random variable $Y : \Omega \rightarrow \mathbf{R}$. Prove that Y_1, Y_2, \dots converge in distribution to Y .

Then, show that the converse is false.