

Introduction to Stochastic Process

Time: 18:30-21:05

Problem 1

[Sheldon M. Ross: Introduction to Probability Models Exercise 24 (p169) & Exercise 51(p92)]

A coin, having probability p of landing on head, is continually flipped until at least one head and one tail have been obtained. The flips of the coin are made independently.

(1) Find the expected number of flips needed. [10 marks]

- Lemma:

Let X be a random variable that follows a geometric distribution with success probability p , denoted as $X \sim \text{Geo}(p)$.

The probability mass function (pmf) of X is given by:

$$\text{pmf}(i) = (1 - p)^{i-1} p \quad (i = 1, 2, \dots)$$

Therefore the expectation of X is:

$$\begin{aligned}\mathbb{E}(X) &= \sum_{i=1}^{\infty} i \cdot \text{pmf}(i) \\ &= \sum_{i=1}^{\infty} i \cdot (1 - p)^{i-1} p \\ &= p \sum_{i=1}^{\infty} i \cdot q^{i-1} \quad (q := 1 - p) \\ &= p \sum_{i=1}^{\infty} \frac{d}{dq} (q^i) \\ &= p \frac{d}{dq} \left(\sum_{i=1}^{\infty} q^i \right) \\ &= p \frac{d}{dq} \left(\frac{q}{1-q} \right) \\ &= \frac{p}{(1-q)^2} \\ &= \frac{1}{p}\end{aligned}$$

Solution: ([Reference](#))

Consider the two scenarios based on the first coin toss result:

- First toss is a head (probability p): We then wait for the first tail.
- First toss is a tail (probability $1 - p$): We then wait for the first head.

Denote N as the total number of flips required.

- The conditional distribution $(N - 1 | \text{first toss is a head})$ follows a Geometric distribution with success probability $1 - p$, thus $\mathbb{E}[N - 1 | \text{first toss is a head}] = \frac{1}{1-p}$.

- The conditional distribution ($N - 1$ | first toss is a tail) follows a Geometric distribution with success probability p , thus $\mathbb{E}[N - 1 | \text{first toss is a tail}] = \frac{1}{p}$.

Therefore, the expected total number of flips can be calculated as follows:

$$\begin{aligned}\mathbb{E}[N] &= 1 + \mathbb{E}[N - 1] \\ &= 1 + p \cdot \mathbb{E}[N - 1 | \text{first toss is a head}] + (1 - p) \cdot \mathbb{E}[N - 1 | \text{first toss is a tail}] \\ &= 1 + \frac{p}{1-p} + \frac{1-p}{p} \\ &= \frac{1}{p(1-p)} - 1\end{aligned}$$

(2) Find the expected number of flips that land on heads. [10 marks]

Solution: [\(Reference\)](#)

Consider the two scenarios based on the first coin toss result:

- First toss is a head (probability p):** the outcome must be $HH \cdots HT$
- First toss is a tail (probability $1 - p$):** the outcome must be $TT \cdots TH$

Denote N' as the number of flips that land on heads after the first toss.

- The conditional distribution ($N' + 1 | \text{first toss is a head}$) follows a Geometric distribution with success probability $1 - p$, thus $\mathbb{E}[N' | \text{first toss is a head}] = \frac{1}{1-p} - 1$.
- The conditional distribution ($N' | \text{first toss is a tail}$) $\equiv 1$ follows a one-point distribution at 1 thus $\mathbb{E}[N' | \text{first toss is a tail}] = 1$

Therefore, the expected number of flips that land on heads can be calculated as follows:

$$\begin{aligned}\mathbb{E}[N] &= p \cdot [1 + \mathbb{E}[N' | \text{first toss is a head}]] + (1 - p) \cdot [0 + \mathbb{E}[N' | \text{first toss is a tail}]] \\ &= p \cdot (1 + \frac{1}{1-p} - 1) + (1 - p) \cdot 1 \\ &= \frac{p}{1-p} + (1 - p)\end{aligned}$$

Problem 2

Suppose $X = \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} \sim N \left(\begin{bmatrix} \mu_1 \\ \mu_2 \end{bmatrix}, \begin{bmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{bmatrix} \right)$, find the distribution of $X_1 | X_2 = x_2$.

Solution:

[\(Deriving the conditional distributions of a multivariate normal distribution\)](#)

Denote $Z := X_1 + AX_2$ (assume that Σ_{22} is nonsingular).

- First, we choose A so that Z and X_2 are independent:

$$\begin{aligned}\text{Cov}(Z, X_2) &= \text{Cov}(X_1 + AX_2, X_2) \\ &= \text{Cov}(X_1, X_2) + A\text{Cov}(X_2, X_2) \\ &= \Sigma_{12} + A\Sigma_{22} \\ &= 0\end{aligned}$$

which yields $A = -\Sigma_{12}\Sigma_{22}^{-1}$ (this is actually block Gaussian elimination).

Therefore Z and X_2 are uncorrelated.

Since they are jointly Gaussian, uncorrelatedness implies independence.

- Secondly, determine the expectation of $X_1|_{X_2=x_2}$:

$$\begin{aligned}\mathbb{E}[X_1|X_2=x_2] &= \mathbb{E}[Z-AX_2|X_2=x_2] \\ &= \mathbb{E}[Z|X_2=x_2]-Ax_2 \\ &= \mathbb{E}[Z]-Ax_2 \\ &= (\mu_1+A\mu_2)-Ax_2 \\ &= \mu_1+A(\mu_2-x_2) \\ &= \mu_1-\Sigma_{12}\Sigma_{22}^{-1}(\mu_2-x_2)\end{aligned}$$

- Thirdly, determine the variance of $X_1|_{X_2=x_2}$:

$$\begin{aligned}\text{Var}(X_1|X_2=x_2) &= \text{Var}(Z-AX_2|X_2=x_2) \\ &= \text{Var}(Z|X_2=x_2) \\ &= \text{Var}(Z) \\ &= \text{Var}(X_1+AX_2) \\ &= \text{Var}(X_1)+\text{Var}(AX_2)+\text{Cov}(X_1,AX_2)+\text{Cov}(AX_2,X_1) \\ &= \Sigma_{11}+A\Sigma_{22}A^T+\Sigma_{12}A^T+A\Sigma_{21} \\ &= \Sigma_{11}+\Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21}-\Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21}-\Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21} \\ &= \Sigma_{11}-\Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21}\end{aligned}$$

which is the **Schur complement** of $(1, 1)$ block of $\Sigma = \begin{bmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{bmatrix}$

Therefore, $X_1|_{X_2=x_2} \sim N(\mu_1-\Sigma_{12}\Sigma_{22}^{-1}(\mu_2-x_2), \Sigma_{11}-\Sigma_{12}\Sigma_{22}^{-1}\Sigma_{21})$

Problem 3

[Sheldon M. Ross: Introduction to Probability Models Example 5.18 (P319)]

Assume that the number of cars that enters into a parking lot in the interval $[0, t]$ follows a Poisson process with rate λ .

Moreover, assume the parking time of each car follows the distribution with CDF $G(x)$ and is independent of its arrival time.

Find the distribution of the number of cars that drive out of the parking lot in the interval $[0, t]$.

- **Lemma: (Poisson 过程的拆分, 原书命题 5.4)**

考虑一个参数为 $\lambda > 0$ 的 Poisson 过程 $\mathbf{N} = \{N(t) : t \geq 0\}$:

假设有 k 种可能类型的事件,

而一个事件被分类为类型 i ($i = 1, \dots, k$) 事件的概率依赖于事件发生的时间.

特别地, 我们假设若一个事件在时刻 t 发生,

则它将以概率 $p_i(t)$ 被分类到类型 i ($i = 1, \dots, k$) (独立于以前发生的任何事件),

其中 $\{p_i(t)\}_{i=1}^k$ 满足 $\sum_{i=1}^k p_i(t) = 1$

记 $N_i(t)$ ($i = 1, \dots, k$) 为到时刻 t 为止类型 i 事件发生的个数,

则 $N_i(t)$ ($i = 1, \dots, k$) 是具有均值 $\mathbb{E}[N_i(t)] = \lambda \int_0^t p_i(s)ds$ 的独立 Poisson 随机变量,
即 $N_i(t) \sim \text{Poisson}(\lambda \int_0^t p_i(s)ds)$ 且它们相互独立.

这样我们就拆分得到了一系列非齐次 Poisson 过程 $\{N_i(t) : t \geq 0\}$.

Solution:

对于任意给定的 t , 我们将进入停车场的车辆分为两类:

- 第一类: 在时刻 t 之前 (包含时刻 t) 离开了停车场;

- 第二类: 在时刻 t 之前 (包含时刻 t) 没有离开停车场;

现在考虑一个在时刻 $s \leq t$ 进入停车场的车辆.

如果它的停车时间小于等于 $t - s$, 则它属于第一类, 其概率为 $G(t - s)$;

否则, 它属于第二类, 其概率为 $1 - G(t - s)$.

根据引理, 我们知道:

- $[0, t]$ 时间内驶离停车场的车辆数 (即第一类车辆数) $X(t)$ 的分布是
均值为 $\mathbb{E}[X(t)] = \lambda \int_0^t G(t-s)ds = -\lambda \int_t^0 G(x)dx = \lambda \int_0^t G(x)dx$ 的 Poisson 分布.
- 类似地, 在时刻 t 仍在停车场的车辆数 (即第二类车辆数) $Y(t)$ 的分布是
均值为 $\mathbb{E}[Y(t)] = \lambda \int_0^t [1 - G(t-s)]ds = \lambda t - \lambda \int_0^t G(y)dy$ 的 Poisson 分布.
- 而且 $X(t) \perp Y(t)$

Problem 4

Denote by $\mathbf{X} = \{X(t) : t \geq 0\}$ a compound Poisson process.

Compute $\text{Cov}(X(s), X(t))$.

- **Lemma:**

设 X_1, X_2, \dots 是一列独立同分布的随机变量, 具有均值 $\mathbb{E}[X] = \mu$ 和方差 $\text{Var}(X) = \sigma^2$
假设它们与取非负整数值的随机变量 N 独立.

我们称 $S = \sum_{i=1}^N X_i$ 为**复合随机变量**, 有 $\begin{cases} \mathbb{E}[S] = \mu \mathbb{E}[N] \\ \text{Var}[S] = \sigma^2 \mathbb{E}[N] + \mu^2 \text{Var}(N) \end{cases}$ 成立.

Solution:

设 $X(t)$ 具有形式 $X(t) = \sum_{i=1}^{N(t)} \xi_i$

其中 $\mathbf{N} = \{N(t) : t \geq 0\}$ 是均值为 $\lambda > 0$ 的 Poisson 过程,

而 $\{\xi_i\}$ 是一列均值为 μ , 方差为 σ^2 的独立同分布的随机变量, 且与 \mathbf{N} 独立.

不失一般性地, 假设 $0 < s < t$.

$$\begin{aligned} \text{Cov}(X(s), X(t)) &= \text{Cov}(X(s), X(t) - X(s) + X(s)) \\ &= \text{Cov}(X(s), X(t) - X(s)) + \text{Cov}(X(s), X(s)) \quad (\text{note that } X(s) \perp X(t) - X(s)) \\ &= 0 + \text{Var}(X(s)) \\ &= \mathbb{E} \left[\text{Var} \left(\sum_{i=1}^{N(s)} \xi_i \mid \mathbf{N} \right) \right] + \text{Var} \left[\mathbb{E} \left[\sum_{i=1}^{N(s)} \xi_i \mid \mathbf{N} \right] \right] \\ &= \mathbb{E}[N(s)\sigma^2] + \text{Var}[N(s)\mu] \\ &= \lambda s \cdot \sigma^2 + \lambda s \cdot \mu^2 \\ &= (\mu^2 + \sigma^2)\lambda s \end{aligned}$$

Problem 5

Consider independent tosses of a fair die.

Let X_n be the maximum of numbers appearing in the first n throws, $n = 1, 2, \dots$

(a) Verify that $\{X_n : n \geq 1\}$ is a Markov chain.

Describe the state space and the transition probability matrix of this homogeneous Markov Chain.

Solution:

记第 n 次投掷骰子的结果为 ξ_n , 其分布为 $P\{\xi_n = i\} = \frac{1}{6}$ ($i = 1, 2, 3, 4, 5, 6$)

则对于任意 $n \in \mathbb{N}$ 都有 $X_{n+1} = \max\{\xi_1, \dots, \xi_n, \xi_{n+1}\} = \max\{X_n, \xi_{n+1}\}$

显然有 $P\{X_{n+1} = x_{n+1} | X_n = x_n, \dots, X_1 = x_1\} = P\{X_{n+1} = x_{n+1} | X_n = x_n\}$ 成立

说明 $\{X_n : n \geq 1\}$ 具有 Markov 性, 因此是一个 Markov 链.

其状态空间为 $\{1, 2, 3, 4, 5, 6\}$, 考虑转移概率:

- 当前 n 次投掷处于最大值 m 时, 下一步仍停在 m 的概率为:

$$P_{m,m} = P\{\xi_{n+1} \leq m\} = \frac{m}{6}$$

- 当前 n 次投掷处于最大值 m 时, 下一步变为 $j > m$ 的概率为:

$$P_{m,m} = P\{\xi_{n+1} = j\} = \frac{1}{6}$$

因此状态转移矩阵为:

$$P = \begin{bmatrix} \frac{1}{6} & \frac{1}{6} & \frac{1}{6} & \frac{1}{6} & \frac{1}{6} & \frac{1}{6} \\ \frac{2}{6} & \frac{1}{6} & \frac{1}{6} & \frac{1}{6} & \frac{1}{6} & \frac{1}{6} \\ & \frac{3}{6} & \frac{1}{6} & \frac{1}{6} & \frac{1}{6} & \frac{1}{6} \\ & & \frac{4}{6} & \frac{1}{6} & \frac{1}{6} & \frac{1}{6} \\ & & & \frac{5}{6} & \frac{1}{6} & \\ & & & & & 1 \end{bmatrix}$$

(b) Find out the classes of all states.

Is there any state closed (a state is closed if, once entered, it cannot be left)?

Is there any state recurrent/transient?

Is there any state positive recurrent?

Write your reasons briefly.

Solution:

- 状态 6 是闭的 (即吸收态), 因为 $P\{X_{n+1} = 6 | X_n = 6\} = P_{6,6} = 1$

- 状态 6 是常返态 (具体来说, 是正常返态), 因为它是一个吸收态.

状态 1, 2, 3, 4, 5 是瞬时态,

因为它们都可到达状态 6, 可是一旦过程进入状态 6, 它就无法返回到其他状态.

Problem 6

For all integer $n \geq 3$, is the origin point for the n -dimensional symmetric random walk recurrent or transient?

Prove your answer using Stirling's formula: $k! \sim \sqrt{2\pi k} \cdot k^k e^{-k}$

Solution:

对于维度 $n \geq 3$ 的情况, 即使是对称的随机游动也是瞬时的.

考察状态 0, 尝试确定 $\sum_{m=0}^{\infty} P_{00}^{(m)}$ 是有限还是无穷大.

- 奇数次行走后肯定不能回到原点 0, 因此 $P_{00}^{(2k-1)} = 0$ ($\forall k = 1, 2, \dots$)

- 偶数次行走最终回到原点 0 当且仅当沿着每个维度，向正方向的步数等于向负方向的步数.

设对于第 j 个维度，正向步数和负向步数均为 r_j , 满足 $2 \sum_{j=1}^n r_j = 2k$.

对应的概率为:

$$\begin{aligned} P_{00}^{(2k)} &= \frac{1}{(2n)^{2k}} \sum_{r_1+\dots+r_n=k} \frac{(2k)!}{\prod_{j=1}^n (r_j! r_j!)} \\ &= \frac{(2k)!}{(2n)^{2k}} \sum_{r_1+\dots+r_n=k} \frac{1}{(\prod_{j=1}^n r_j!)^2} \end{aligned}$$

注意到存在某个正常数 $C_1 > 0$ 使得:

$$\max_{r_1+\dots+r_n=k} \frac{1}{\prod_{j=1}^n r_j!} \leq \frac{C_1}{(\lfloor k/n \rfloor!)^n}$$

于是我们有:

$$\begin{aligned} P_{00}^{(2k)} &= \frac{(2k)!}{(2n)^{2k}} \sum_{r_1+\dots+r_n=k} \frac{1}{(\prod_{j=1}^n r_j!)^2} \\ &= \frac{(2k)!}{(2n)^{2k}} \max_{r_1+\dots+r_n=k} \frac{1}{\prod_{j=1}^n r_j!} \cdot \frac{1}{k!} \sum_{r_1+\dots+r_n=k} \frac{k!}{\prod_{j=1}^n r_j!} \\ &\leq \frac{(2k)!}{(2n)^{2k}} \cdot \frac{C_1}{(\lfloor k/n \rfloor!)^n} \cdot \frac{1}{k!} \quad (\text{note that } k! \sim \sqrt{2\pi k} \cdot k^k e^{-k}) \\ &\lesssim C_1 \frac{(2k)^{2k}}{(2n)^{2k}} \cdot \frac{n^k}{k^k} \cdot \frac{1}{(k/n)^{k+n/2}} \cdot \frac{\sqrt{2}}{(2\pi)^{-n/2}} \\ &\leq \frac{C_2}{k^{n/2}} \end{aligned}$$

其中 $C_2 > 0$ 为某个正常数.

因此我们有:

$$\sum_{m=0}^{\infty} P_{00}^{(m)} = \sum_{k=1}^{\infty} P_{00}^{(2k)} \leq C_2 \sum_{k=1}^{\infty} k^{-n/2} < \infty$$

故状态 0 是瞬时态.

The End