

**Q1.** Basic concepts of probability.

Roll a set of two fair 6-sided dice, one colored red the other white. Assume that the two dice roll independently. Record the two numbers facing up as  $X_1$  and  $X_2$ .

- i) Mathematically describe the format of outcome. Describe the sample space.
- ii) Let  $X$  be the absolute difference between  $X_1$  and  $X_2$ . Is  $X$  a random variable?
- iii) Find the PMF of  $X$ . Find  $E(X)$ .
- iv) Let  $A = \{X_2^2 \geq 2X_1\}$ . Find  $\Pr(A|X_1 = n)$  for  $n = 1, 2, \dots, 6$ . Use the law of total probability to find  $\Pr(A)$ .

**Q2.** Properties of  $\text{Var}(\cdot)$ .

- i) Use the definition,  $\text{Var}(Y) \equiv E(Y^2) - E(Y)^2$ , to derive  $\text{Var}(\alpha X) = \alpha^2 \text{Var}(X)$ .
- ii) Finish the proof that if  $X$  and  $Y$  are independent, then we have  $\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$ .
- iii) Find an example demonstrating that  $\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$  does not imply  $X$  and  $Y$  are independent.

**Q3.** Properties of  $E(\cdot)$  and  $\text{Var}(\cdot)$ .

Let  $X \sim \text{Bino}(n, p)$  and  $Y \sim N(\mu, \sigma^2)$ . We know that

$$E(X) = np, \quad \text{Var}(X) = np(1-p), \quad E(Y) = \mu, \quad \text{Var}(Y) = \sigma^2$$

Note that  $X$  and  $Y$  are not given as independent. Calculate  $E(X^2 + Y^2)$ .

**Q4.** Probability of a normal random variable falling in a given interval.

Let  $X \sim N(\mu, \sigma^2)$ . Finish the derivation of the inequality below.

$$\Pr\left(\mu - \eta\sigma \leq X \leq \mu + \eta\sigma\right) = F_X(\mu + \eta\sigma) - F_X(\mu - \eta\sigma) = \dots = \text{erf}\left(\frac{\eta}{\sqrt{2}}\right)$$

The results above gives a numerical formula for calculating the probability on the left since the error function  $\text{erf}(\cdot)$  is available in all packages.

**Q5.** A simple convoluted experiment.

Consider a coin with  $\Pr(\text{"head"}) = p$ . In the first part of an experiment, we first flip the coin  $n$  times where  $n$  is a prescribed fixed number. Let

$$Y = \text{number of heads in the sequence of } n \text{ flips in the first part}$$

Then in the second part of the experiment, we flip the coin  $Y$  times. Let

$$X = \text{number of heads in the sequence of } Y \text{ flips in the second part}$$

Use the law of total expectation to calculate

- i)  $E(X)$ , and
- ii)  $E(XY)$

Hint:  $Y \sim \text{Bino}(n, p)$ ,  $(X|Y = m) \sim \text{Bino}(m, p)$ .

**Q6.** Verify the concept of confidence interval.

Write a code to draw numerically a data set of  $n = 10$  independent samples of  $X \sim N(\mu, \sigma^2)$  with  $\mu = 0.6$ ,  $\sigma = 1.3$ .

$$(\text{A data set}) = \{X_j, j = 1, 2, \dots, n\}$$

In real applications, one data set is measured while the true values of  $\mu$  and  $\sigma$  are unknown. We estimate the mean  $\mu$  as the sample mean of the data set. To illustrate the statistical behavior of the process, we repeat this experiment  $M = 500,000$  times.

- i) Consider the idealized case where  $\sigma = 1.3$  is given. For each data set, calculate the exact 95% confidence interval for estimating  $\mu$ . Out of  $M = 500,000$  repeats, calculate the fraction in which the confidence interval actually contains the true  $\mu$ . Report the fraction.
- ii) Consider the realistic case where  $\sigma$  is unknown. For each data set, calculate the approximate 95% confidence interval for estimating  $\mu$ .

$$\hat{\sigma} = \sqrt{\frac{1}{(n-1)} \sum_{j=1}^n (X_j - \hat{\mu})^2}, \quad \hat{\mu} = \frac{1}{n} \sum_{j=1}^n X_j$$

Out of  $M = 500,000$  repeats, calculate the fraction in which the confidence interval actually contains the true  $\mu$ . Report the fraction.

**Q7.** Analytical utility of characteristic functions.

A normal distribution is defined by its probability density

$$\rho_X(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$$

The corresponding characteristic function is

$$\phi_X(\xi) = \exp\left(i\mu\xi - \frac{\sigma^2\xi^2}{2}\right)$$

Use the properties of characteristic function  $\phi_X(\xi)$  to show

- i)  $E(X) = \mu$ , and
- ii)  $\text{Var}(X) = \sigma^2$

Hint: It is more convenient to differentiate

$$E\left(\exp(i\xi(X-\mu))\right) = \phi_X(\xi) \exp(-i\xi\mu) = \exp\left(-\frac{\sigma^2\xi^2}{2}\right)$$

The results above justify that we name parameters  $\mu$  and  $\sigma^2$  in the probability density, respectively, as the mean and the variance. This approach does not involve integration by parts. This approach will be very useful in the discussion of multivariate Gaussian.

- Q8.** An example illustrating that the well-posedness of a probability problem depends on whether or not we have enough information to repeat the random experiments.  
Consider a bin containing two types of coins in unknown proportions.

Coin type 1:  $\Pr(\text{“head”}) = 0.5$

Coin type 2:  $\Pr(\text{“head”}) = 0.7$

Take a coin randomly from the bin and flip it. We study the probability that the coin is type 2 when the flip outcome is “head”. Mathematically, we study the conditional probability

$$\Pr(\text{“type 2”} \mid \text{“head”})$$

- i) Is this probability problem completely specified (i.e, we have enough information to repeat the experiment in a lab setting)?
- ii) If not, What information we might impose to make it well-posed?