

# Ma 3b Practical – Recitation 5

February 27, 2025

Recall Poisson distribution, Bayer's theorem, Normal distribution, and central limit theorem.

**Exercise 1.** (Poisson process) Suppose that an office receives telephone calls as a Poisson process with  $\lambda = 0.5$  per min. What is the probability of exactly one call in first 5 mininutes?

**Exercise 2.** (Random variables) The Rayleigh distribution has CDF

$$F(x) = 1 - e^{-x^2/2}, \quad x > 0$$

Compute the following

1.  $f(X)$
2.  $E(X)$ ,  $\text{Var}(X)$
3.  $P(X > 2)$

**Intuition** Let  $F$  be the CDF and  $f$  be the PDF of a continuous r.v.  $X$ . It's important that  $f(x)$  is not a probability; for example, we could have  $f(3) > 1$ , and we know  $P(X = 3) = 0$ . But thinking about the probability of  $X$  being very close to 3 gives us a way to interpret  $f(3)$ . Specifically, the probability of  $X$  being in a tiny interval of length  $\epsilon$ , centered at 3, will essentially be  $f(3)\epsilon$ . This is because

$$P(3 - \epsilon/2 < X < 3 + \epsilon/2) = \int_{3-\epsilon/2}^{3+\epsilon/2} f(x) dx \approx f(3)\epsilon,$$

This also implies that  $P(X \leq 3) = P(X < 3)$ .

**Exercise 3.** (normal distribution) We estimate that the distribution of a birthdate, in days from conception, has a normal distribution with mean 270 (in days) and a variance of 100. What is the probability that a child was born after the 300-th day or before the 240-th day after conception?

**Exercise 4.** (central limit theorem) We try to measure the intensity of a signal, but we can only make approximate measurements. These measurements are i.i.d. random variables  $X_i$  with mean  $\mu$  (this is the true intensity, which is unknown) and variance 10. How many measurements must be taken to find the intensity with an error of at most 1 unit, and 95% certainty?

**Exercise 5.** (central limit theorem) Using the CLT and a continuity correction<sup>1</sup>, estimate the probability that a Poisson r.v. with parameter  $\lambda = 100$  is larger than 120.

**Exercise 6.** (central limit theorem) We roll 100 unbiased dice and we want to estimate the probability that the sum of the results is between 300 and 400 (both included).

---

<sup>1</sup> *Let's forget about continuity correction if not covered yet. It's some minor improvement on estimation for discrete i.i.d random variables.*

**Solution.** The number of calls in a 5 -min. interval follows a Poisson distribution with parameter  $\omega = 5\lambda = 2.5$ . Thus, the probability of no calls in a 5 -min. interval is  $e^{-2.5} = .082$ . The probability of exactly one call is  $2.5e^{-2.5} = .205$ .

**Notes:** Section 5.6 of Introduction to Probability-Joseph K. Blitzstein, Jessica Hwang.

Recall that the Poisson distribution  $\text{Pois}(\lambda)$  has pmf  $P(X = k) = \frac{\lambda^k}{k!} e^{-\lambda}$  where  $\lambda = 2$ .

A Poisson process is a sequence of arrivals occurring at different points on a time-line, such that the number of arrivals in a particular interval of time has a Poisson distribution. A process of arrivals in continuous time is called a Poisson process with rate  $\lambda$  if the following two conditions hold.

1. The number of arrivals that occur in an interval of length  $t$  is a  $\text{Pois}(\lambda t)$  random variable, i.e. pmf  $P_t(X = k) = \frac{(\lambda t)^k}{k!} e^{-\lambda t}$ .

2. The numbers of arrivals that occur in disjoint intervals are independent of each other. For example, the numbers of arrivals in the intervals  $(0, 10)$ ,  $[10, 12)$ , and  $[15, 1)$  are independent.

**Exercise 7.** (normal distribution) We estimate that the distribution of a birthdate, in days from conception, has a normal distribution with mean 270 (in days) and a variance of 100 . What is the probability that a child was born after the 300-th day or before the 240-th day after conception?

**Solution.** Let  $X$  be the birthdate, in days from conception, then

$$X \sim \mathcal{N}(270, 100)$$

We compute

$$\begin{aligned} &= 1 - P(240 \leq X \leq 300) \\ &= 1 - P\left(\frac{240 - 270}{10} \leq \frac{X - 270}{10} \leq \frac{300 - 270}{10}\right) \\ &= 1 - P(-3 \leq Z \leq 3) \quad \text{where } Z \stackrel{\circ}{=} \frac{X - \mu}{\sigma} \sim \mathcal{N}(0, 1) \\ &= 1 - [P(Z \leq 3) - P(Z \leq -3)] = 1 - P(Z \leq 3) + 1 - P(Z \leq 3) \\ &= 2 - 2F_Z(3) \\ &= 2 - 2(0.9987) \quad (\text{ see the } \mathcal{N}(0, 1) \text{ c.d.f. table } ) \\ &= 0.0026. \end{aligned}$$

**Exercise 8.** (central limit theorem) We try to measure the intensity of a signal, but we can only make approximate measurements. These measurements are i.i.d. random variables  $X_i$  with mean  $\mu$  (this is the true intensity, which is unknown) and variance 10 . How many measurements must be taken to find the intensity with an error of at most 1 unit, and 95% certainty?

**Solution.** We want to choose  $n$  so that  $P(|\bar{X}_n - \mu| \leq 1) = 0.95$ .

Equivalently,  $P\left(\left|\frac{\bar{X}_n - \mu}{\sqrt{10/n}}\right| \leq \frac{1}{\sqrt{10/n}}\right) = 0.95$ .

According to the CLT, this is a normal distribution

$$P\left(\left|\frac{\bar{X}_n - \mu}{\sqrt{10/n}}\right| \leq \alpha\right) = 0.95. \iff \alpha = 1.96$$

So  $\frac{1}{\sqrt{10/n}} = 1.96$  and  $n = 10(1.96)^2 = 38.416$ .

**Solution.** Let  $X_i \stackrel{\text{i.i.d.}}{\sim} \text{Pois}(1)$ , then we know that  $X \stackrel{\circ}{=} \sum_{i=1}^{100} X_i \sim \text{Pois}(100)$ . Note that  $E[X_i] = 1$  and  $\text{Var}(X_i) = 1$ . Therefore, by the CLT (also applying a continuity correction), we have

$$\begin{aligned} P(X > 120) &= P\left(\frac{X - 100 \cdot 1}{\sqrt{100 \cdot 1}} > \frac{120 - 100 \cdot 1}{\sqrt{100 \cdot 1}}\right) \\ &\approx 1 - F_Z\left(\frac{120 + 1/2 - 100}{10}\right) \quad \text{where } Z \sim \mathcal{N}(0, 1) \\ &= 1 - F_Z(2.05) \\ &= 1 - 0.9798 \\ &= 0.0202 \end{aligned}$$

**Solution.** Let  $X_i, i \in \{1, 2, \dots, 100\}$ , be i.i.d. r.v.s such that

$$P(X_i = k) = 1/6 \quad \text{for all } k \in \{1, 2, \dots, 6\}.$$

We have

$$E[X_i] = \sum_{k=1}^6 k \frac{1}{6} = \frac{6 \cdot 7}{2} \frac{1}{6} = 7/2 = 3.5$$

$$E[X_i^2] = \sum_{k=1}^6 k^2 \frac{1}{6} = \frac{6 \cdot 7 \cdot 13}{6} \frac{1}{6} = 91/6$$

$$\text{Var}(X_i) = E[X_i^2] - (E[X_i])^2 = 91/6 - (7/2)^2 = \frac{182 - 147}{12} = 2.91\bar{6}$$

Let  $X \stackrel{\circ}{=} \sum_{i=1}^{100} X_i$ , and note that

$$E[X] = 100 \cdot 3.5 = 350, \quad \text{and} \quad \text{Var}(X) = 100 \cdot 2.91\bar{6} = 291.\bar{6}$$

By the CLT (also applying a continuity correction), we have

$$\begin{aligned} P(X \in [300, 400]) &= P(|X - 350| \leq 50) \\ &= P\left(\left|\frac{X - 350}{\sqrt{291.6}}\right| \leq \frac{50}{\sqrt{291.6}}\right) \\ &= 2F_Z\left(\frac{50 + 1/2}{\sqrt{291.6}}\right) - 1 \\ &= 2F_Z(2.96) - 1 = 2 \cdot 0.9985 - 1 = 0.997 \end{aligned}$$

## 1 Remark on continuity correction(optional)

Since the cumulative distribution function of a binomial r.v. jumps at discrete values and the one for the normal is continuous, the approximation can be improved if we apply what is called a continuity correction. When we look at the probability that  $S_n$  is inside some interval of the form  $(c, d)$ ,  $[c, d)$ ,  $(c, d]$ ,  $[c, d]$ ,

- we add  $+1/2$  to the LEFT endpoint if it is strict (it doesn't include the endpoint);
- we add  $-1/2$  to the LEFT endpoint if it is not strict (it includes the endpoint);
- we add  $-1/2$  to the RIGHT endpoint if it is strict (it doesn't include the endpoint);
- we add  $+1/2$  to the RIGHT endpoint if it is not strict (it includes the endpoint);

## 2 Theorem (Central limit theorem (CLT))

Let  $X_1, X_2, X_3, \dots$  be a sequence of i.i.d. r.v.s with expectation  $\mu$  and (finite) variance  $\sigma^2$ , and let  $\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$ . Then, we have the following convergence in law :

$$\lim_{n \rightarrow \infty} P \left( \frac{\bar{X}_n - \mu}{\sqrt{\sigma^2/n}} \leq a \right) = F_Z(a) \stackrel{\circ}{=} \int_{-\infty}^a \frac{1}{\sqrt{2\pi}} \exp \left( -\frac{z^2}{2} \right) dz, \quad \forall a \in \mathbb{R},$$

where  $Z \sim \mathcal{N}(0, 1)$ . We can also write:

$$\lim_{n \rightarrow \infty} P \left( \frac{\sum_{i=1}^n X_i - n\mu}{\sqrt{n\sigma^2}} \leq a \right) = F_Z(a), \quad \forall a \in \mathbb{R}.$$