

## Moment Generating Function

**Def:** The Moment Generating Function (MGF) of a r.v.  $X$  is

$$M(t) = M_X(t) = \mathbb{E} e^{tX}, \quad t \in \mathbb{R}.$$

Basic properties of MGF:

- $\mathbb{E} X^n = M^{(n)}(0)$  and  $\text{Var}(X) = M''(0) - (M'(0))^2$ .
- If  $X$  and  $Y$  are ind. with MGF  $M_X(t)$  and  $M_Y(t)$ , respectively, then

$$M_{X+Y}(t) = M_X(t) \cdot M_Y(t).$$

**Ex 11.1:** Determine  $M_X(t)$  and  $\mathbb{E}(X^n)$  for Bernoulli r.v with parameter  $p$ .

**HW:** S11.1: 1, 4, 5; S11.5: 2, 3, 7, R11: 1, 11, 18.

## Law of Large Numbers (LLN)

**Thm. 11.10:** (Weak Law of Large Numbers). Let  $X_1, X_2, \dots$  be a sequence of independent and identically distributed r.v.'s with  $\mathbb{E} X_i = \mu$ . Then for any  $\varepsilon > 0$ ,

$$\lim_{n \rightarrow \infty} \mathbb{P} \left( \left| \frac{X_1 + \dots + X_n}{n} - \mu \right| \geq \varepsilon \right) = 0$$

**Pf:** Use Chebyshev's Inequality.

**Thm. 11.11:** (Strong Law of Large Numbers). Let  $X_1, X_2, \dots$  be a sequence of independent and identically distributed r.v.'s with  $\mathbb{E} X_i = \mu$ . Then

$$\mathbb{P} \left( \lim_{n \rightarrow \infty} \frac{X_1 + \dots + X_n}{n} = \mu \right) = 1.$$

# The Central Limit Theorem (CLT)

**Thm. 3.1:** (Central Limit Theorem). Let  $X_1, X_2, \dots$  be a sequence of independent and identically distributed r.v.'s with  $\mathbb{E} X_i = \mu$  and  $\sigma^2 = \text{Var}(X_i)$ . Then, as  $n \rightarrow \infty$ ,

$$\begin{aligned} \mathbb{P} \left( \frac{X_1 + \dots + X_n - n\mu}{\sigma\sqrt{n}} \leq x \right) &\rightarrow \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-t^2/2} dt \\ &= \Phi(x) = \mathbb{P}(Z \leq x) \end{aligned}$$

where  $Z \sim N(0, 1)$ .

**Idea of Pf:** Show  $M_{Z_n}(t) \rightarrow M_Z(t)$  for

$$Z_n = \frac{X_1 + \dots + X_n - n\mu}{\sigma\sqrt{n}} = \frac{S_n - \mathbb{E} S_n}{\sqrt{\text{Var}(S_n)}}.$$

where  $S_n = X_1 + \dots + X_n$ .

• Let  $\bar{X}_n = (X_1 + \dots + X_n)/n$  denote the sample mean. Then  $\frac{\bar{X}_n - \mu}{\sigma/\sqrt{n}}$  is approximately standard normal, i.e

$$\frac{\bar{X}_n - \mu}{\sigma/\sqrt{n}} \approx Z.$$

**Ex 11.27:** The lifetime of a TV tube (in years) is an exponential random variable with mean 10. What is the probability that the average lifetime of a random sample of 36 TV tubes is at least 10.5?

**Ex 11.29:** If 20 random numbers are selected independently from the interval  $(0, 1)$ , what is the approximate probability that the sum of these numbers is at least eight?

●Additional questions: At most 11? Between 8 and 13? More than 12 or less than 8?

## Simulation

All of simulation starts with the question “how do I choose a random number uniformly between 0 and 1?” This is an intricate question, and most time we use pseudorandom numbers. These days, any self-respecting programming language or environment has a routine for this task (typically something like `random[]`, `rand`, `rnd`, or some other variant thereof). In this class, we will use such random number generators to generate a few other random variables of interest.

- Generating a Bernolli Random Variable  $X$ . If  $U$  is uniformly distributed on the interval  $[0, 1]$ , then it follows that  $\mathbb{P}(U \leq p) = p$ . So, if we defined

$$X := \begin{cases} 1 & \text{if } U \leq p \\ 0 & \text{if } U > p \end{cases}$$

then  $\mathbb{P}(X = 1) = \mathbb{P}(U \leq p) = p$  and  $\mathbb{P}(X = 0) = \mathbb{P}(U > p) = 1 - p$ .

- Generating Discrete Random Variables. Assume  $\mathbb{P}(X = x_i) = p_i$ ,  $i = 0, 1, \dots$ , with  $p_0 + p_1 + \dots = 1$ . Let  $U$  be uniform on  $[0, 1]$ . Define

$$X := \begin{cases} x_0 & \text{if } U \leq p_0 \\ x_1 & \text{if } p_0 < U \leq p_0 + p_1 \\ x_2 & \text{if } p_0 + p_1 < U \leq p_0 + p_1 + p_2 \\ \dots & \dots \end{cases}$$

- The Inverse Transform Method.
- Monte Carlo Method
- Rejection Method