

Math 170E

Lecture Notes Section 5.6 ^{*†}

Central limit theorem

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NOTE: Materials that appear in the textbook but do not appear in the lecture notes might still be tested. Please send me an email if you find typos.

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[†]This notes is based on March Boedihardjo's and Jamie Haddock's notes from the past quarters, and I would like to thank them for their generosity. “*Nanos gigantum humeris insidentes* (I am but a dwarf standing on the shoulders of giants)”.

1 Super weak law of large numbers

Recall from Section 5.3 the following theorem.

Theorem 1. *Let X_1, X_2, \dots be independent random variables with mean μ and variance σ^2 . Let \bar{X} be the sampled mean*

$$\bar{X} := \frac{X_1 + \dots + X_n}{n}.$$

Then,

$$E[\bar{X}] = \mu; \qquad \text{Var}[\bar{X}] = \frac{\sigma^2}{n}.$$

So, as $n \rightarrow \infty$, the average \bar{X} gets closer and closer to the mean μ .

2 Law of large numbers: Intuition, stock

One can think of X_i as the profit of a stock investment at the i -th year.

When investing, we need to pay attention to both **profit** and **risk**.

Then $E[\bar{X}] = \mu$ is the **expected profit** if you invest for n years on average.

Furthermore, $\text{Var}[\bar{X}]$ measures the **expected risk** of your investment for n years.

Typical question when investing:

“**How much money will I make** if I invest in n years?”

This question cannot be answered with certainty, since no one can predict the future. A better question is

“**What is the probability** that I will make x amount of money if I invest in n years?”

This question can be answered by the **central limit theorem**.

3 Central limit theorem: Example

Let \bar{X} be the average of the profit of your company for $n = 25$ years. Suppose that X_1, \dots, X_n has mean 15 (millions USD) and variance 4 (millions USD).

Approximate the probability

$$P(14.4 < \bar{X} < 15.6).$$

4 Central limit theorem

Theorem 2. *Let X_1, X_2, \dots be independent random variables with mean μ and variance σ^2 . Let W be the random variable*

$$W = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} = \frac{(X_1 + \dots + X_n) - n\mu}{\sigma\sqrt{n}}.$$

Then the cdf of W converges to the cdf of standard normal random variable as $n \rightarrow \infty$.

That is to say, when n is large, we have for all real w ,

$$P(W \leq w) \approx \int_{-\infty}^w \frac{1}{\sqrt{2\pi}} e^{-z^2/2} dz = \Phi(w).$$

where $\Phi(\cdot)$ is the cdf of standard normal RV from Table Va. This is the reason why normal RVs and bell curves are so important.

5 Central limit theorem: Answer

Answer: Let $W = \frac{\bar{X} - \mu}{\sigma/\sqrt{n}}$ be as in the central limit theorem. Then we have

$$\begin{aligned}\bar{X} = 14.4 &\quad \implies \quad W = \frac{(14.4) - (15)}{(2)/\sqrt{(25)}} = \frac{-0.6}{0.4} = -1.5; \\ \bar{X} = 15.6 &\quad \implies \quad W = \frac{(15.6) - (15)}{(2)/\sqrt{(25)}} = \frac{0.6}{0.4} = 1.5.\end{aligned}$$

By the central limit theorem, we then have

$$\begin{aligned}P(14.4 < \bar{X} < 15.6) &= P(-1.5 < W < 1.5) \\ &\approx \Phi(1.5) - \Phi(-1.5) = 2\Phi(1.5) - 1 \\ &= 0.8664.\end{aligned}$$