

18.600: Lecture 16

Normal random variables

Scott Sheffield

MIT

Outline

Tossing coins

Normal random variables

Special case of central limit theorem

Tossing coins

Normal random variables

Special case of central limit theorem

Tossing coins

- ▶ Suppose we toss a million fair coins. How many heads will we get?
- ▶ About half a million, yes, but how close to that? Will we be off by 10 or 1000 or 100,000?
- ▶ How can we describe the error?
- ▶ Let's try this out.

Tossing coins

- ▶ Toss n coins. What is probability to see k heads?
- ▶ Answer: $2^{-k} \binom{n}{k}$.
- ▶ Let's plot this for a few values of n .
- ▶ Seems to look like it's converging to a curve.
- ▶ If we replace fair coin with p coin, what's probability to see k heads.
- ▶ Answer: $p^k (1 - p)^{n-k} \binom{n}{k}$.
- ▶ Let's plot this for $p = 2/3$ and some values of n .
- ▶ What does limit shape seem to be?

Outline

Tossing coins

Normal random variables

Special case of central limit theorem

Outline

Tossing coins

Normal random variables

Special case of central limit theorem

Standard normal random variable

- ▶ Say X is a (standard) **normal random variable** if
$$f_X(x) = f(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}.$$
- ▶ Clearly f is always non-negative for real values of x , but how do we show that $\int_{-\infty}^{\infty} f(x) dx = 1$?
- ▶ Looks kind of tricky.
- ▶ Happens to be a nice trick. Write $I = \int_{-\infty}^{\infty} e^{-x^2/2} dx$. Then try to compute I^2 as a two dimensional integral.
- ▶ That is, write

$$I^2 = \int_{-\infty}^{\infty} e^{-x^2/2} dx \int_{-\infty}^{\infty} e^{-y^2/2} dy = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-x^2/2} dx e^{-y^2/2} dy.$$

- ▶ Then switch to polar coordinates.

$$I^2 = \int_0^{\infty} \int_0^{2\pi} e^{-r^2/2} r d\theta dr = 2\pi \int_0^{\infty} r e^{-r^2/2} dr = -2\pi e^{-r^2/2} \Big|_0^{\infty},$$

so $I = \sqrt{2\pi}$.

Standard normal random variable mean and variance

- ▶ Say X is a (standard) **normal random variable** if $f(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$.
- ▶ Question: what are mean and variance of X ?
- ▶ $E[X] = \int_{-\infty}^{\infty} xf(x)dx$. Can see by symmetry that this zero.
- ▶ Or can compute directly:

$$E[X] = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} x dx = \frac{1}{\sqrt{2\pi}} e^{-x^2/2} \Big|_{-\infty}^{\infty} = 0.$$

- ▶ How would we compute $\text{Var}[X] = \int f(x)x^2 dx = \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} x^2 dx$?
- ▶ Try integration by parts with $u = x$ and $dv = xe^{-x^2/2} dx$. Find that $\text{Var}[X] = \frac{1}{\sqrt{2\pi}} (-xe^{-x^2/2} \Big|_{-\infty}^{\infty} + \int_{-\infty}^{\infty} e^{-x^2/2} dx) = 1$.

General normal random variables

- ▶ Again, X is a (standard) **normal random variable** if $f(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$.
- ▶ What about $Y = \sigma X + \mu$? Can we “stretch out” and “translate” the normal distribution (as we did last lecture for the uniform distribution)?
- ▶ Say Y is normal with parameters μ and σ^2 if $f(x) = \frac{1}{\sqrt{2\pi\sigma}} e^{-(x-\mu)^2/2\sigma^2}$.
- ▶ What are the mean and variance of Y ?
- ▶ $E[Y] = E[X] + \mu = \mu$ and $\text{Var}[Y] = \sigma^2 \text{Var}[X] = \sigma^2$.

Cumulative distribution function

- ▶ Again, X is a standard normal random variable if $f(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$.
- ▶ What is the cumulative distribution function?
- ▶ Write this as $F_X(a) = P\{X \leq a\} = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^a e^{-x^2/2} dx$.
- ▶ How can we compute this integral explicitly?
- ▶ Can't. Let's just give it a name. Write $\Phi(a) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^a e^{-x^2/2} dx$.
- ▶ Values: $\Phi(-3) \approx .0013$, $\Phi(-2) \approx .023$ and $\Phi(-1) \approx .159$.
- ▶ Rough rule of thumb: "two thirds of time within one SD of mean, 95 percent of time within 2 SDs of mean."

Outline

Tossing coins

Normal random variables

Special case of central limit theorem

Outline

Tossing coins

Normal random variables

Special case of central limit theorem

DeMoivre-Laplace Limit Theorem

- ▶ Let S_n be number of heads in n tosses of a p coin.
- ▶ What's the standard deviation of S_n ?
- ▶ Answer: \sqrt{npq} (where $q = 1 - p$).
- ▶ The special quantity $\frac{S_n - np}{\sqrt{npq}}$ describes the number of standard deviations that S_n is above or below its mean.
- ▶ What's the mean and variance of this special quantity? Is it roughly normal?
- ▶ **DeMoivre-Laplace limit theorem (special case of central limit theorem):**

$$\lim_{n \rightarrow \infty} P\left\{a \leq \frac{S_n - np}{\sqrt{npq}} \leq b\right\} \rightarrow \Phi(b) - \Phi(a).$$

- ▶ This is $\Phi(b) - \Phi(a) = P\{a \leq X \leq b\}$ when X is a standard normal random variable.

Problems

- ▶ Toss a million fair coins. Approximate the probability that I get more than 501,000 heads.
- ▶ Answer: well, $\sqrt{npq} = \sqrt{10^6 \times .5 \times .5} = 500$. So we're asking for probability to be over two SDs above mean. This is approximately $1 - \Phi(2) = \Phi(-2) \approx .159$.
- ▶ Roll 60000 dice. Expect to see 10000 sixes. What's the probability to see more than 9800?
- ▶ Here $\sqrt{npq} = \sqrt{60000 \times \frac{1}{6} \times \frac{5}{6}} \approx 91.28$.
- ▶ And $200/91.28 \approx 2.19$. Answer is about $1 - \Phi(-2.19)$.