

Central Limit Theorem

of heads in n coin tosses is $\text{Bin}(n, \frac{1}{2})$

$$P(k) = \binom{n}{k} \left(\frac{1}{2}\right)^k \left(\frac{1}{2}\right)^{n-k} = \binom{n}{k} \left(\frac{1}{2}\right)^n$$

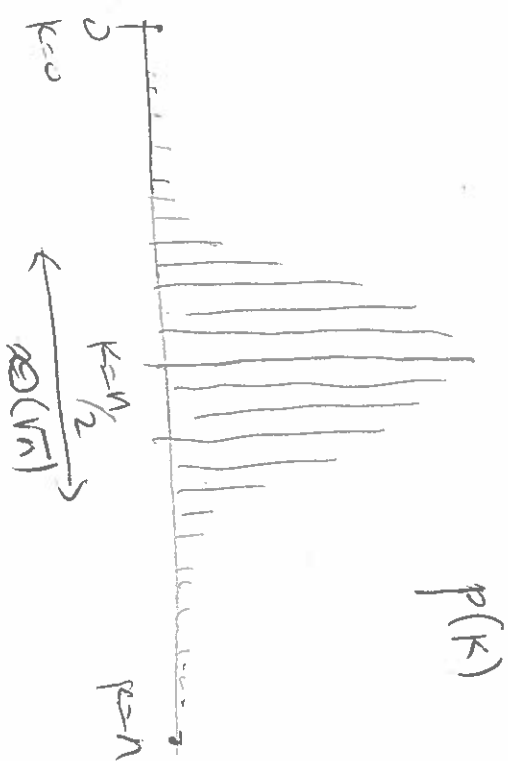
Stirling: $n! \sim \sqrt{2\pi n} \cdot \left(\frac{n}{e}\right)^n$

$$P(k) = \frac{n!}{k!(n-k)! 2^n}$$

$$P(k) \sim \frac{\sqrt{2\pi n} \left(\frac{n}{e}\right)^n}{\sqrt{2\pi k} \left(\frac{k}{e}\right)^k \sqrt{2\pi(n-k)} \left(\frac{n-k}{e}\right)^{n-k} 2^n}$$

If $k \approx \frac{n}{2}$; $k = \frac{n}{2} + t$

$$P\left(\frac{n}{2} + t\right) \sim \frac{1}{\sqrt{\pi n}} e^{-t^2/2n}$$



$$P\left(\frac{n}{2}\right) \sim \frac{\sqrt{2\pi n}}{(\sqrt{2\pi n/2})^2} = \frac{1}{\sqrt{\pi n}}$$

valid if $n \gg 1$ $t \ll n$

Similarly: Poi(λ): $f(\lambda+t) \approx \frac{1}{\sqrt{2\pi\lambda}} e^{-t/\lambda}$

Gamma RV, also has normal approx.

Central Limit Theorem (CLT):

Let X_1, X_2, \dots be iid R.V.s with $\mu = EX_i$ and

$\sigma^2 = \text{Var}(X_i)$. Let $S_n = \sum_{i=1}^n X_i$. Then

$$\underbrace{\frac{S_n - n\mu}{\sqrt{n}\sigma}}_{Z_n} \xrightarrow{D} N(0, 1).$$

$$\left[P(a \leq Z_n \leq b) \xrightarrow{n \rightarrow \infty} \int_a^b \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx \right] \quad \text{A} \quad [a, b]$$

$$Z_n \approx N(0, 1)$$

$$S_n = \mu n + \sigma \sqrt{n} Z_n$$

$$LLN: \frac{1}{n} S_n \longrightarrow \mu$$

Proof of CLT: by continuity thm, need to show that

$$\phi_{Z_n}(t) \xrightarrow{n \rightarrow \infty} e^{-t^2/2}$$

$$\text{Let } Y_n = X_n - \mu \quad \text{so } EY_n = 0 \quad \text{Var}(Y_n) = \sigma^2$$

$$Z_n = \frac{Y_1 + Y_2 + \dots + Y_n}{\sigma \sqrt{n}}$$

$$\phi_{Z_n}(t) = \phi_{Y_1 + Y_2 + \dots + Y_n} \left(\frac{t}{\sigma \sqrt{n}} \right)$$

$$= \left[\phi_Y \left(\frac{t}{\sigma \sqrt{n}} \right) \right]^n$$

$$\phi_Y(s) = 1 + is EY + \frac{i^2 s^2}{2!} EY^2 + O(s^3)$$

$$\phi_Y(s) = 1 + \frac{s^2}{2} \cdot \sigma^2 + \text{err}$$

$$\left[\phi_Y\left(\frac{t}{\sigma\sqrt{n}}\right) \right]^n = \left[1 - \frac{t^2 \sigma^2}{2 \cdot n^2} + \text{err} \right]^n \longrightarrow e^{-t^2/2}$$

as needed.

□

~~Note~~ If $EX = \infty$ or not defined then $\frac{S_n}{n}$ has no limit.

If $EX = \mu$ but $\text{Var}(X) = \infty$ then $\frac{S_n}{n} \longrightarrow \mu$

but $|S_n - n\mu| \gg \sqrt{n}$

Eq. want 50 new students.

Historically 40% of admitted students accept the offer.

If 100 students admitted, what is $P(\text{more than 50 come})$?

$X \sim \text{Bin}(100, 0.4)$ need $P(X > 50)$

X is a sum of 100 $\text{Bern}(0.4)$

$$E \text{Bern}(p) = p = 0.4$$

$$\text{Var}(\text{Bern}(p)) = p(1-p) = 0.2$$

$$\text{CLT} \quad \frac{X - 100p}{\sqrt{100 p(1-p)}} \approx N(0,1)$$

$$Z = \frac{X - 40}{\sqrt{24}} \approx N(0,1)$$

$$X > 50 \iff \frac{X-40}{\sqrt{24}} > \frac{50-40}{\sqrt{24}} = \frac{10}{\sqrt{24}} = 2.04 \text{ bit} = 2.04$$

$$P(Z > 2) = 1 - \Phi(2.04)$$

$$\Phi(t) = P(N(0,1) \leq t)$$