

MATH 3235 Probability Theory

11/10/22

Flip a coin

H $+1$

T -1

$2N$ Times

N

large n

$$S_N = \sum_{i=1}^{2N} X_i$$

X_i : individual coin flip.

$$m \approx \delta \sqrt{N}$$

Chebyshev

$$\begin{aligned} \mathbb{P}(|S_N| > t) &\leq \frac{\mathbb{E}(S_N^2)}{t^2} \\ &= \frac{2N}{t^2} \end{aligned}$$

$$t \approx \sqrt{N}$$

$$t = 2\sqrt{N}$$

$$\mathbb{P}(|S_N| > t) \leq \frac{1}{2}$$

$$m = \delta \sqrt{N}$$

$$\mathbb{P}(2m - 1 \leq S_N \leq 2m + 1) = \binom{2N}{N-m} 2^{-2N}$$

$$N! \approx \sqrt{2\pi N} N^N e^{-N}$$

$$\binom{2N}{N-m} = \frac{(2N)!}{(N-m)! (N+m)!}$$

$$\frac{\sqrt{2\pi 2N}}{\sqrt{2\pi(N-m)} \sqrt{2\pi(N+m)}} \approx \frac{\sqrt{2}}{\sqrt{2\pi N}}$$

$$m = \delta \sqrt{N}$$

$$\frac{\binom{2N}{N-m} 2^{-2N} e^{-2N}}{(N-m)^{N-m} e^{-(N-m)} (N+m)^{N+m} e^{-(N+m)}} \approx$$

$$\frac{(N-m)^{-(N-m)} (N+m)^{-(N+m)}}{N^{-2N}} =$$

$$\left(1 - \frac{m}{N}\right)^{-(N-m)} \left(1 + \frac{m}{N}\right)^{-(N+m)} =$$

Remark:

$$\left(1 + \frac{1}{N}\right)^N \rightarrow e$$

$$= \exp\left(- (N-m) \log\left(1 - \frac{m}{N}\right) - (N+m) \log\left(1 + \frac{m}{N}\right)\right) =$$

$$= \exp\left(-N \left(\left(1 - \frac{m}{N}\right) \log\left(1 - \frac{m}{N}\right) + \left(1 + \frac{m}{N}\right) \log\left(1 + \frac{m}{N}\right) \right)\right)$$

$$\frac{m}{N} = \frac{\delta}{N} \quad \text{small}$$

$$\log(1+x) = x - \frac{x^2}{2} + \dots$$

$$\exp\left(-N\left(1 - \frac{\delta}{\sqrt{N}}\right)\left(-\frac{\delta}{\sqrt{N}} - \frac{\delta^2}{2N}\right) + \left(1 + \frac{\delta}{\sqrt{N}}\right)\left(\frac{\delta}{\sqrt{N}} - \frac{\delta^2}{2N}\right)\right) =$$

$$\exp\left(-N\left(\frac{\delta^2}{N} + \frac{\delta^4}{N^2} \dots\right)\right) =$$

$$\exp(-\delta^2)$$

All Together Then

$$\frac{1}{\sqrt{2\pi}^{1/2}} e^{-\frac{\delta^2}{2} \cdot 1/2} \left(\frac{1}{\sqrt{N}} \right) d\delta$$

Central Limit Theorem

X_i be r.v. i.i.d. with

$$E(X_i) = 0 \quad \text{var}(X_i) = 1$$

$$Z_N = \frac{1}{\sqrt{N}} \sum_{i=1}^N X_i$$

$$Z_N \rightarrow ??$$

$$U = \begin{cases} 1 & \frac{1}{2} \\ -1 & \frac{1}{2} \end{cases}$$

$$X_i = (-1)^i U$$

Z_N F_N The c.d.f. of Z_N

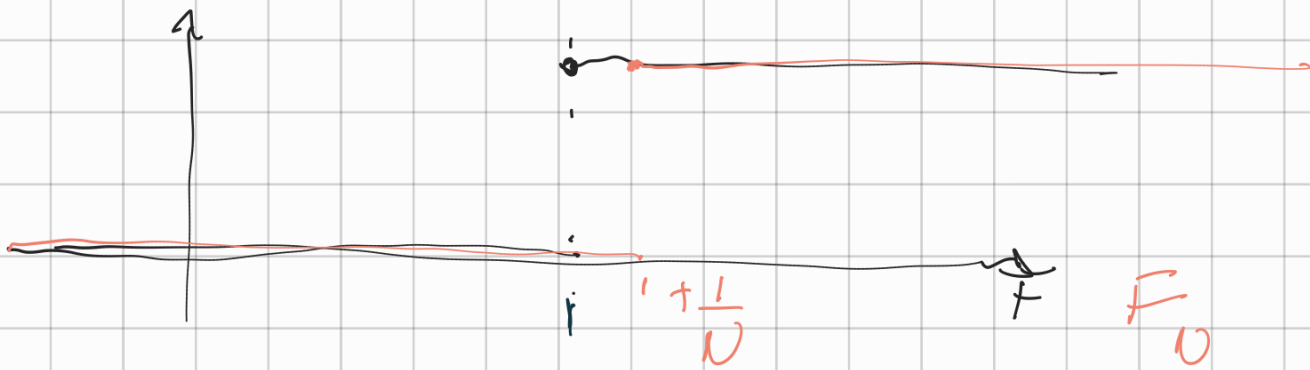
Z F The c.d.f. of Z

$Z_N \Rightarrow Z$, f

$F_N(z) \rightarrow F(z) \quad \forall z \text{ where } F \text{ is cont.}$

$z_0 \quad 1 + \frac{1}{N} \quad \text{prob } 1$

$z \quad 1 \quad \text{prob } 1$



$$F_N(1) = 0$$

$$F(1) = 1$$

$z_N \rightarrow z \text{ in dist.}$

$$P(z_N \leq z) \rightarrow P(z \leq z)$$

for every z such that

$$P(z = z) = 0$$

Geometric r.v. with par $p = \frac{\lambda}{N}$

X_N let X an exp r.v. with par λ

$$\frac{X_N}{N} \Rightarrow X$$

$$P\left(\frac{X_N}{N} > t\right) = \left(1 - \frac{\lambda}{N}\right)^{Nt} \rightarrow e^{-\lambda t} =$$

$$P(X > t).$$

Geometric \rightarrow Exp

Binomial \rightarrow Normal

Continuity Theorem

Let Z_N be r.v. with char. fun. $\phi_N(t)$. Assume

$$\phi_N(t) \rightarrow \phi(t) \quad \forall t \in \mathbb{R}.$$

Then \Downarrow $Z_N \Rightarrow Z$.

$$X_i \quad \phi_X(t) = \left(1 + i\mathbb{E}(X)t - \mathbb{E}(X^2)\frac{t^2}{2} + \dots \right)$$

$$\mathbb{E}(X) = 0 \quad \text{Var}(X) = 1$$

$$\phi_{\frac{X}{\sqrt{N}}} = \left(1 - \frac{t^2}{2N} + O\left(\frac{t^3}{N^{3/2}}\right) \right)$$

$$Z_N = \sum_{i=1}^N \frac{X_i}{\sqrt{N}}$$

$$\phi_{Z_N}(t) = \left(\phi_{\frac{X}{\sqrt{N}}}(t) \right)^N =$$

$$\left(1 - \frac{t^2}{2N} + O\left(\frac{t^3}{N^{3/2}}\right) \right)^N =$$

$$e^{-\frac{t^2}{2}} = \phi_Z(t) \quad \text{with } Z \sim \mathcal{N}(0,1)$$

If X_i are i.i.d. r.v. with

$$E(X_i) = \mu \quad \text{var}(X_i) = \sigma^2$$

Calling

$$Z_N = \frac{1}{\sqrt{N}} \sum_{i=1}^N \frac{X_i - \mu}{\sigma}$$

$Z_N \Rightarrow Z$ where $Z \sim \mathcal{N}(0,1)$.

X_i are i.i.d.

$$\sum_{i=1}^N X_i \sim \mathcal{N}(N\mu, N\sigma^2)$$

$$\frac{1}{N} \sum_{i=1}^N X_i \sim \mathcal{N}\left(\mu, \frac{\sigma^2}{N}\right)$$

$$a_N \rightarrow b_N$$

\Downarrow

$$Na_N \rightarrow Nb_N$$

X_i are The result of measurement with an apparatus. I know

$E(X_i) = \mu$ is The True value

$$\text{Var}(X_i) = \sigma^2$$

I Take N measurement x_i

$$i = 1 \dots N.$$

$$\bar{x} = \frac{1}{N} \sum_i x_i$$

Find δ such that

$$\bar{x} - \delta \leq \mu \leq \bar{x} + \delta$$

with prob 95% .

I need To Take $\delta = 1.96 \frac{\sigma}{\sqrt{N}}$

That with prob. 0.95

$$\bar{x} - 1.96 \frac{\sigma}{\sqrt{N}} \leq \mu \leq \bar{x} + 1.96 \frac{\sigma}{\sqrt{N}}$$

if N is large.

0.5 prob of H.

0.01 ,



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