

# Binomial, normal distribution and sampling distribution

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# Random variable

# PMF and CDF

- Probability Mass Function  $PMF = P(X = x)$
- Cumulative Distribution Function  $CDF = P(X < x)$

# Normal

# Binomial

Assume that you have 5 trials, you need the probability of get exactly three sucess.

## 5 trials and 3 success...

Success and failures	Probability
$E_1 E_2 E_3 F_4 F_5$	$p^3(1-p)^2$
$E_1 E_2 E_4 F_3 F_5$	$p^3(1-p)^2$
$E_1 E_2 E_5 F_3 F_4$	$p^3(1-p)^2$
$E_1 E_3 E_4 F_2 F_5$	$p^3(1-p)^2$
$E_1 E_3 E_5 F_2 F_4$	$p^3(1-p)^2$
$E_1 E_4 E_5 F_2 F_3$	$p^3(1-p)^2$
$E_2 E_3 E_4 F_1 F_5$	$p^3(1-p)^2$
$E_2 E_3 E_5 F_1 F_4$	$p^3(1-p)^2$
$E_2 E_4 E_5 F_1 F_3$	$p^3(1-p)^2$
$E_3 E_4 E_5 F_1 F_2$	$p^3(1-p)^2$

# Binomial distribution

We must said that  $X \sim B(n, p)$

## PMF

$$P(X = x) = \binom{n}{x} P^x (1 - P)^{n-x} \quad (1)$$

## CDF

$$P(X \leq x) = \sum_{i=0}^x \binom{n}{i} P^i (1 - P)^{n-i} \quad (2)$$

# Python



# Normal distribution

Abraham moivre uses the formaul to get binomial probabilities.

# Normal distribution

$$f(X) = \frac{1}{\sqrt{\pi\sigma}} \exp \frac{a}{b} \quad (3)$$



# Poisson Distribution

According to the former binomial distribution  $X \sim b(p, n)$  the two parameter are the shape a form of the distribution. the poisson distribution is the case when the variable follow a binomial distribution with a  $n \rightarrow \infty$

# Frame Title

In the limit case, the occurrence of a only event is only guaranteed in the measure that the space is very small, for instance if the ocurrence of the events is simultaneous, you should not consider a Poisson distribution. the FD we can dervied of a binomial distribution in the following way  $E(x) = np = \lambda$ , thus:

$$\frac{n!}{(n-k)!k!} \left(\frac{\lambda}{n}\right)^k \left(1 - \frac{\lambda}{n}\right)^{n-k}$$

$$\frac{(n-k+1)!}{n^k k!} \left(1 - \frac{\lambda}{n}\right)^n \left(1 - \frac{\lambda}{n}\right)^{-k}$$

$$e = \lim_{n \rightarrow \infty} \left(1 + \frac{1}{n}\right)^n \text{ we must use } t = \frac{n}{k}, \text{ and thus } \frac{n+k}{n} = 1 + \frac{k}{n}$$

$$\lim_{n \rightarrow \infty} = \frac{e^{-k} \lambda^k}{k!}$$

thus a random variable follows a Poisson distribution with a parameter  $\lambda$   
 $X \sim p(\lambda)$  and its FD is rewritten as:

$$p(X = x) = \frac{e^{-\lambda} \lambda^x}{x!}$$



- Population - Parameters
- Sample - Statistics

for instance the mean  $\mu$  and sample mean  $\bar{x}$ . in some books  $\sigma^2$  and  $S^2$  for population and sample variance respectively.



# Distribution of sample statistics

Each sample have different values, then statistics are random variables, but what distribution follow?.

# Distribution mean

## sample

- if  $X \sim N(\mu, \sigma)$  then  $\bar{x} \sim N(\mu, \frac{\sigma}{\sqrt{n}})$
- by CLT if  $n$  is large then  $X$  is approximately normal with  $N(\mu, \frac{\sigma}{\sqrt{n}})$

# How big is?

$n$ ?

30 is a practical value