

Poisson's
distribution
assumptions

- ① $n \rightarrow \infty$
② events are independent
③ p can be very, very less
sometimes $p = 0$ (p does not really exist)

classmate

Date

Page

→ POISSON'S DISTRIBUTION :

IF $n \rightarrow \infty$ (very large)

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

$e = 2.72$ (constant)

λ = no. of objects per {
time
length
area
volume

→ Scales of measurement

- ① nominal } categorical → shown in bar/pre
② ordinal } data chart. ↳ gives rel. freq.
③ interval and ratio

gives
freq.

① nominal : cannot apply ^(mathematical - quantity) sort, add, sub, multiply
division operations

ex : texts, phone nos.

can be numeric as well (ex : phone no.) → cannot perform any operations on it.

② ordinal : ordered

- ex : grades (A, B, C), quality (high, low), designation
- only sort applicable
(mathematical ops)
- $+$, $-$, $*$, $/$ → not applicable
- exception : likert scale

In case of reviews, we take average reviews

③ interval and ratio

- can apply all mathematical/arithmetic operations
- it is a scale ← ratio
- all numeric data

numerical \rightarrow discrete (count)
 \rightarrow continuous (measure)

① discrete : count, variable

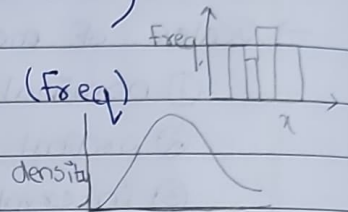
ex: no. of people
 whole no. / integers

② continuous : can be decimal also

if discrete repeats - it can be converted into categorical for better understanding

numerical data \rightarrow ① histogram (freq)

② density plot



③ box plot - to understand

5 point summary

\rightarrow ① univariate \rightarrow bar, histogram, pie

② bivariate

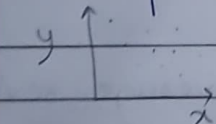
② bi-variate : a) numerical & numerical (vs)

b) num & cat. (vs)

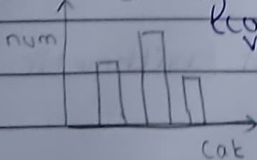
c) C & cat. (vs)

d) Time stamp & num (vs)

• scatterplot \rightarrow covariance or correlation

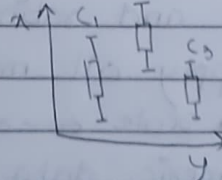


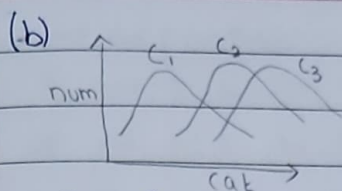
(a) BAR by aggregate



(b)

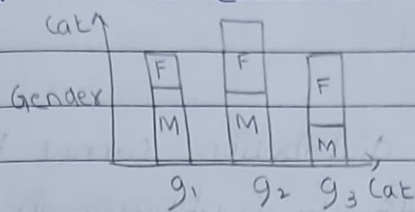
BoxPLOT by group





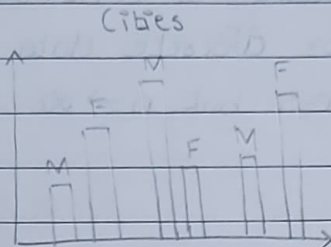
Density plot by group.

(c) cat. vs cat.



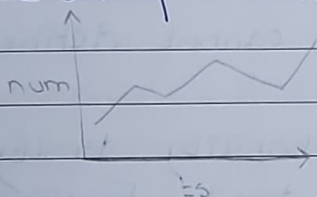
stacked bar chart

F, M → gender



bar chart by group

d) Time stamp vs. numeric.

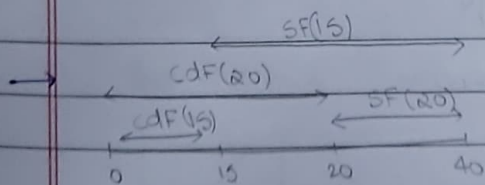


line chart.

metric : noise

seasonality
trends.

- Data analysis → ① descriptive - pop. data available
 → ② inferential data - when population data is not available
 - we consider sample data for analysis

3 ways ① $cdf(20) - cdf(15)$ ② $sf(15) - sf(20)$ ③ $1 - [cdf(15) + sf(20)]$ to find
prob.
b/w 15 & 20

→ Poisson's distribution:

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

$$= F(\lambda)$$

pmf → st. poisson. pmf

cdf → st. poisson. cdf

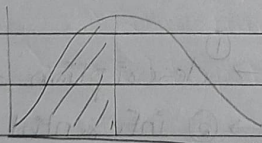
sf → st. poisson. sf

→ Cases when poisson's distribution cannot be applied:

- ① when random variable is continuous in nature
- ② applicable only for discrete data (count)
- ③ when p is not less but $n \rightarrow \infty$

⇒ Continuous prob. dist.

- when random variable is continuous
- since data is continuous, cannot define probability at a given point.
- But you can define probability density using pdf → probability density function ($P(x)/x$)
- Since continuous, no difference between $P(X < 35)$ and $P(X \leq 35) \rightarrow$ ^{so} ~~we~~ we always use $P(X \leq 35)$



$P(X > 35)$ and $P(X \geq 35)$ are same

Formula for bell curve:

$$f(x) = \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{x-\mu}{\sigma} \right)^2}$$

↑
pdf

$e \rightarrow$ exponent $\pi \rightarrow 3.14$ $\sigma \rightarrow$ std. deviation $\mu \rightarrow$ mean \rightarrow normal distribution:

- never touches x -axis.

- $P(x \leq a) = \int_{-\infty}^a f(x) \cdot dx \rightarrow$ st. norm. cdf(a, μ, σ)

- $P(x \geq a) = \int_a^{+\infty} f(x) \cdot dx \rightarrow$ st. norm. sf(a, μ, σ)

- $P(a \leq x \leq b) = \int_a^b f(x) \cdot dx$

3 ways: ① st. norm. cdf(b, μ, σ)- st. norm. cdf(a, μ, σ)

② st. norm. sf(a, μ, σ) - st. norm. sf(b, μ, σ)

③ $1 - [\text{st. norm. cdf}(a, \mu, \sigma) + \text{st. norm. sf}(b, \mu, \sigma)]$

$$Z \rightarrow f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

- $z = \frac{x-\mu}{\sigma}$

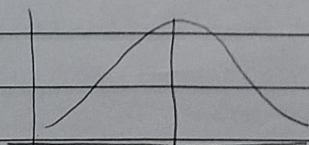
- no units, units cancelled

- if normal distribution followed, $\rightarrow -4 \leq z \leq +4$

- standard normal distribution.

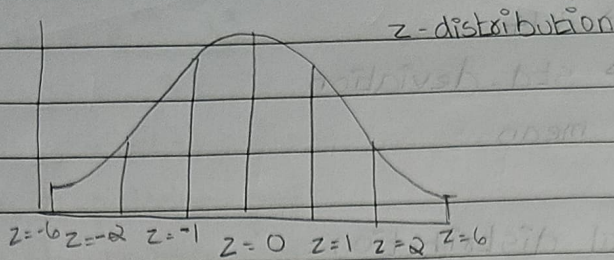
$$z = \frac{x-\mu}{\sigma}$$

~~For~~ $z = x$ when $[\mu = 0, \sigma = 1]$



mean = 0

z/±1



(if variable follow normal dist.)

 $-1 \leq z \leq +1 \Rightarrow 68.2\%$ (area shaded) $-2 \leq z \leq +2 \Rightarrow 95.4\%$ $-3 \leq z \leq +3 \Rightarrow 99.72\%$ $-6 \leq z \leq +6 \Rightarrow 99.999969\%$

area not shaded

 3.14×10^{-6} $= 0.00000314$

if mean = median = 0. } to determine if it
 $k = 0$ } is normal dist.

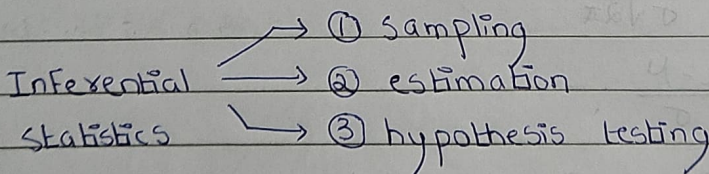
 $1 - 0.00000314$

= area of graph

→ IMP Functions to use:

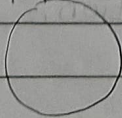
- ① cdf - cumulative probability - ascending (lowest to given pt)
- ② sf - cum. prob. - descending (given pt to highest)
- ③ isf - inverse survival Function (inverse of sf)
- ④ ppf - point percentile Function (inverse of cdf)

⇒ Inferential Statistics:



① Sampling

population



s1 samples

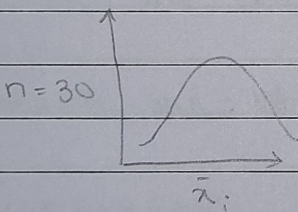
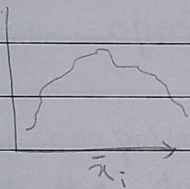
s2

population → parameters
 sample → statistic

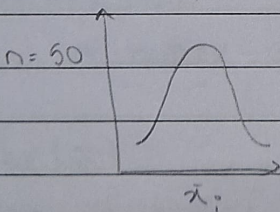
	parameters	Statistic
numerical	mean	μ
	variance	σ^2
	SD	σ
categorical	proportion	p

μ and \bar{x} are not same
for all above parameters \neq statistic

→ Suppose, 100 samples
 $n = \text{sample size}$
 $= 15 \text{ observations}$
 $= 15 \text{ object}$



(after $n=30$
 normal dist. observed)
 $n \geq 30$ approx. normal
 dist.



CLT (Central Limit Theorem)
 aka law of large numbers

CLT :

- i) $n \geq 30$ - approx. normal. dist.
- ii) $\bar{x} \approx \mu$
- iii) S.D. of sample mean $\approx \frac{\sigma}{\sqrt{n}} \rightarrow \text{std. error}$

iii)

Sampling

Probabilistic /
stochastic

- ① Scientific
simple
- a) random sampling
- b) stratified random sampling
- c) cluster sampling
- d) systematic sampling

non-Probabilistic
(not used)

- ① non-scientific
- a) comfortable and
convenient
- b) quota sampling
- c) judgemental sampling
biased.
- d) snowball sampling