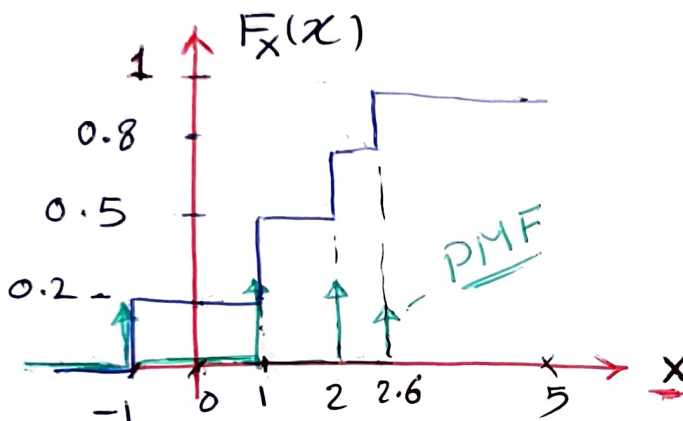


- Continuous R.V.'s

- " Probability Distributions

~~10~~ → 10

Review: Discrete Probability distributions



$$P_X(-1) = \text{non zero}$$

$$P_X(1) = \text{non zero}$$

$$P_X(2) = "$$

$$P_X(2.6) = "$$

$$P_X(0) = F_X(0^+) - F_X(0^-) = 0$$

$$P_X(1) = 0.5 - 0.2 = 0.3$$

$$P_X(5) = 0$$

$$P(2 < x \leq 5) = 1 - 0.8 = 0.2$$

$$P(0 < x) = 0.8$$

$$P(x \leq 0) = F_X(0) = 0.2$$

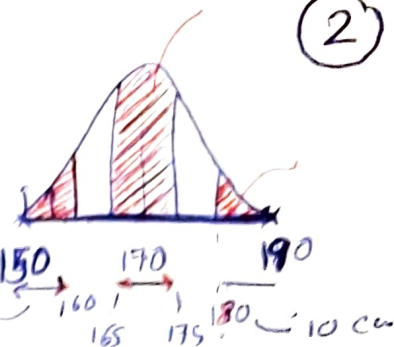
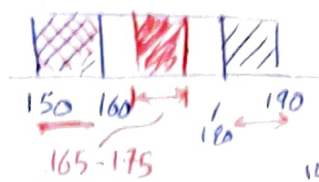
$$P(x < 0) = 0.2$$

$$P(x > 2) \neq P(x \geq 2)$$

$$P(x \geq 2) = P(x > 2) + P_X(2)$$

# Continuous Random Variables

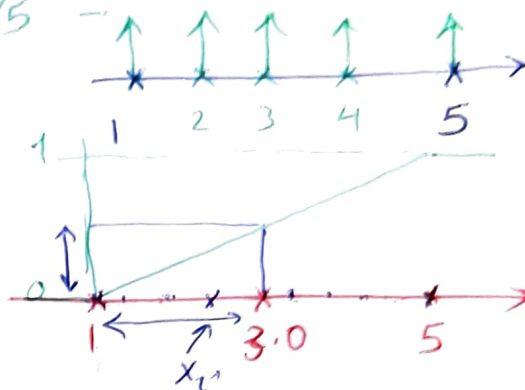
(2)



$$X \in \{1, \dots, 5\} \frac{1}{5}$$

real number

$$Y \in [1, 5]$$

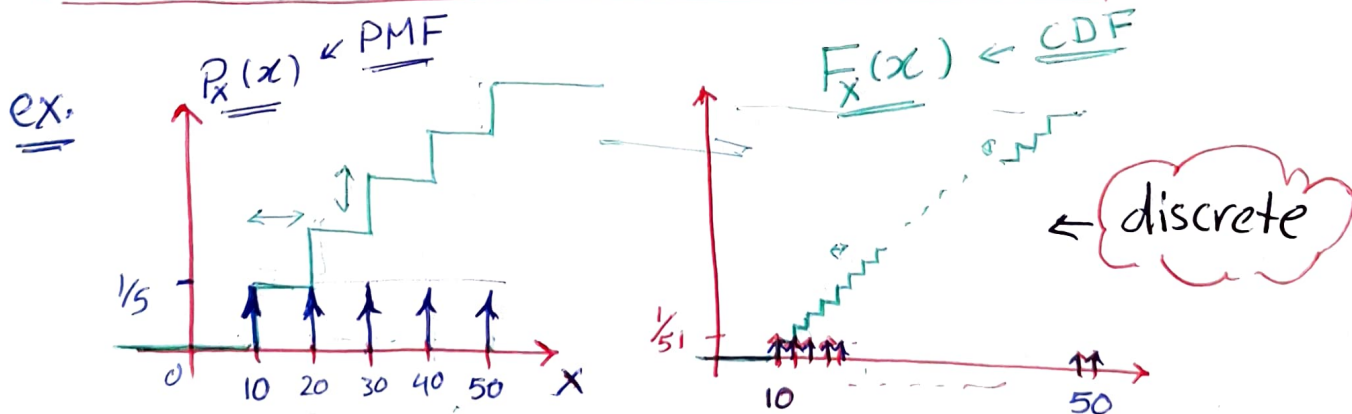


$x_i$  is impossible  $\neq P_X(x_i) = 0$

infinite number of possible outcomes.

$$P_X(x_1 < X \leq x_2) = \checkmark$$

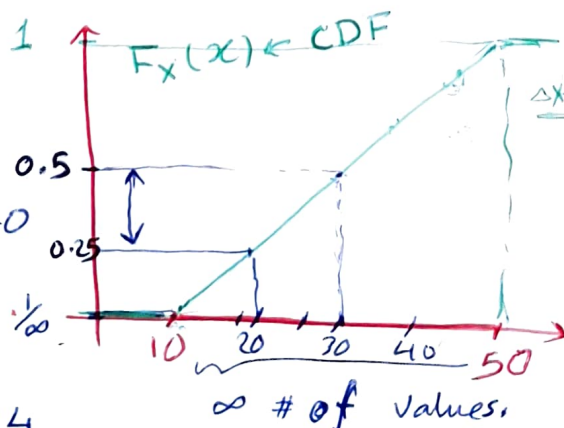
$$P_X(175 \text{ cm}) \stackrel{?}{=} P(174.5 < X \leq 175.5) = \checkmark$$



$$P_X(30) = 0$$

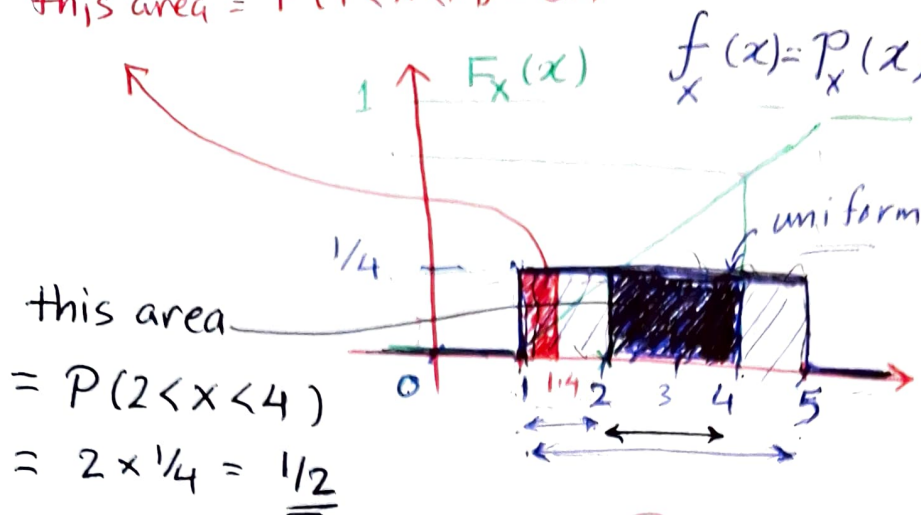
$$P_X(10 < X < 30) = 0.5 - 0 = 0.5$$

$$P_X(20 < X \leq 30) = 0.5 - 0.25 = 0.25$$



continuous

this area  $\equiv P(1 < x < 4) = 0.1$



$$f_x(x) = p_x(x) = \frac{d}{dx} F_x(x)$$

(3)

$$\frac{\Delta y}{\Delta x} = \frac{1}{5-1} = \frac{1}{4}$$

$$f_x(x) = p_x(x);$$

Probability  
Density  
Function

PDF

$$P(2 < x < 4) = \frac{3}{4} - \frac{1}{4} = \frac{1}{2}$$

$$= F_x(4) - F_x(2)$$

$$= \int_2^4 f_x(x) dx$$

$$= F_x(x) \Big|_2^4 = F_x(4) - F_x(2)$$

$$= \int_2^4 f_x(x) dx = \int_2^4 \frac{1}{4} dx = \frac{1}{4} x \Big|_2^4$$

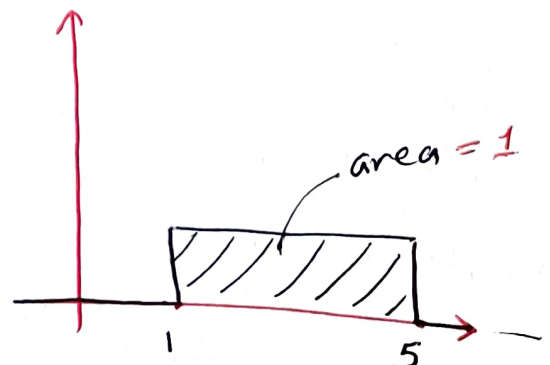
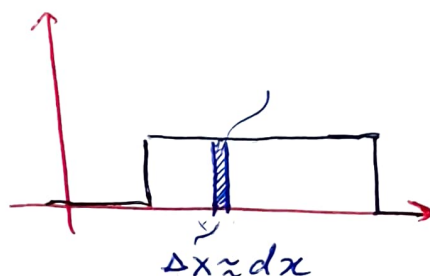
$$= \frac{1}{4} \times 4 - \frac{1}{4} \times 2$$

$$= 1 - \frac{1}{2} = \frac{1}{2}$$

$$f_x(x) = \frac{d}{dx} F_x(x)$$

$$F_x(x) = \int_{-\infty}^x f_x(x) dx$$

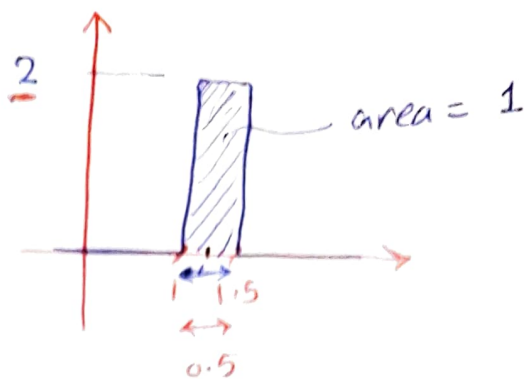
$$\int_{-\infty}^{\infty} f_x(x) dx = 1$$



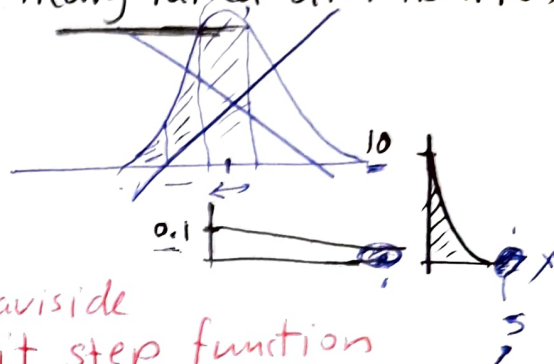
(4)

density  $\times$  range = Prob.

↓  
Uniform  
distribution

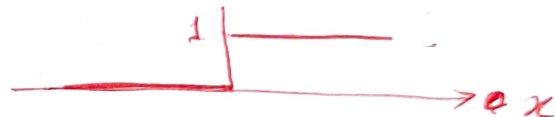
→ exponential distribution

heavy tailed distribution

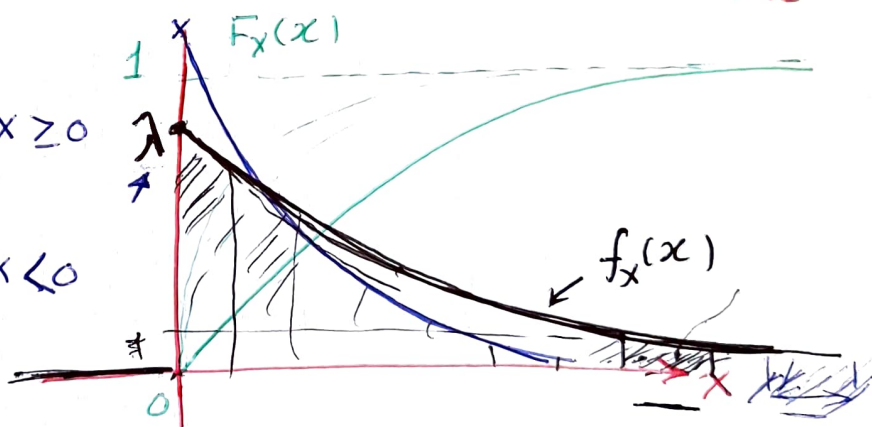


$$F_X(x) = (1 - e^{-\lambda x}) u(x)$$

$$= \begin{cases} 1 - e^{-\lambda x} & ; x \geq 0 \\ 0 & ; x < 0 \end{cases}$$

Heaviside  
unit step function

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x} & ; x \geq 0 \\ 0 & ; x < 0 \end{cases}$$



$$\frac{d}{dx}(1 - e^{-\lambda x}) = 0 - (-\lambda e^{-\lambda x}) = \lambda e^{-\lambda x}$$

$$F_X(x) = \int_{-\infty}^x f_X(\alpha) d\alpha$$

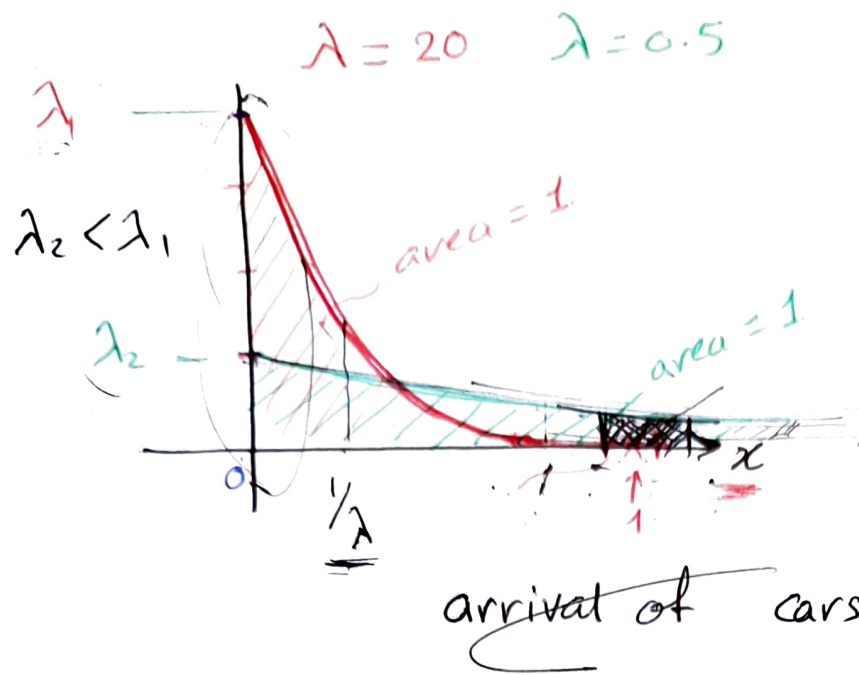
$$\text{for } x < 0 \Rightarrow F_X(x) = 0$$

$$F_X(x) = \int_0^x \lambda e^{-\lambda \alpha} d\alpha$$

$$\text{for } \underline{x \geq 0}$$

$$= \lambda \left( \frac{-e^{-\lambda \alpha}}{\lambda} \right) \bigg|_0^x = -e^{-\lambda x} - (-e^{-\lambda \cdot 0}) = 1 - e^{-\lambda x}$$





~~7 X 40/8~~

(5)

$$f_x(x) = \begin{cases} \lambda e^{-\lambda x} & x \geq 0 \\ 0 & x < 0 \end{cases}$$

25 | 18 | 17 | 23

□ □ □ □

$x \sim 1 \text{ Km}$   
 $P(0.5 < x < 1.5)$  (avg.)

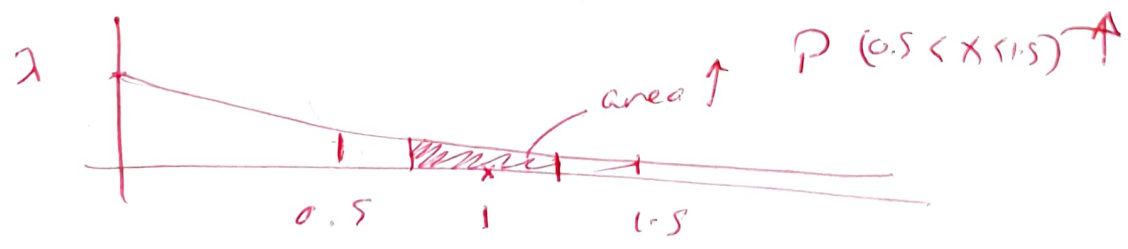
# of cars ~ Poisson dist.

road #1 : 20 cars/Km

$\lambda \equiv$  density cars/Km

road #2 : 0.5 cars/Km  
 ("1 cars per 2 Km")

zero | 1 | 2 | 0 | 0



# Gaussian "Normal" Distribution

6

$$F_X(x) = ?$$

$$P(x_1 < x \leq x_2) = ?$$

$$f_X(x) = \frac{1}{\sqrt{2\pi}\sigma_X} e^{-\frac{(x-\mu_X)^2}{2\sigma_X^2}}$$

$$\frac{0.3}{100}$$

$$\mu_X - 3\sigma_X$$
  

$$170 - 30$$
  

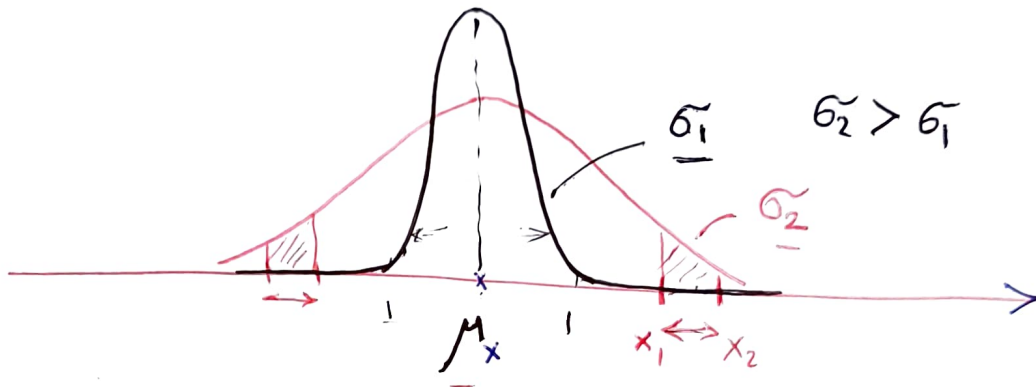
$$(140)$$

$\mu_X$   
↑  
mean value

$$\mu_X + 3\sigma_X$$
  

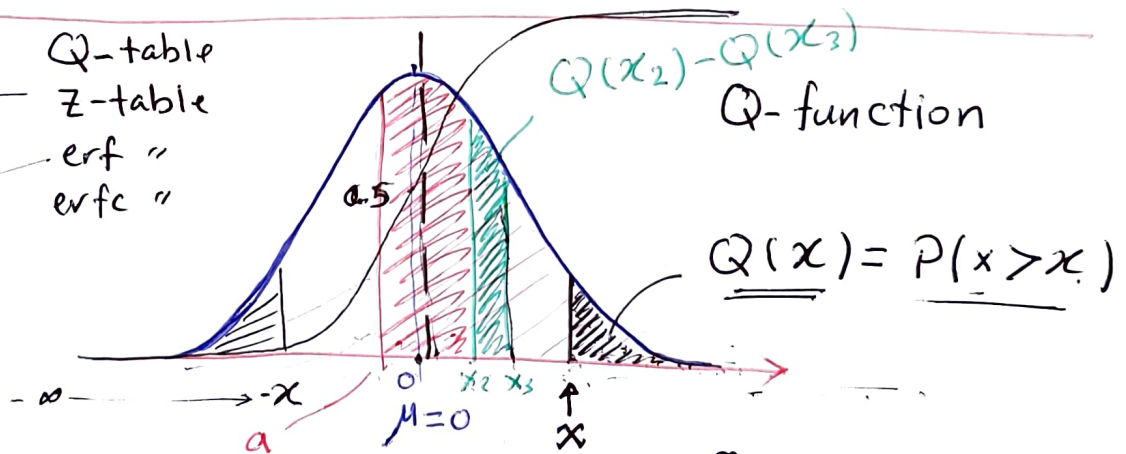
$$\frac{170+90}{100}$$

$\sigma_X^2$  : Variance  
 $\sigma_X$  : standard deviation



| x | 0.0   | 0.1   |
|---|-------|-------|
| 0 | 0.5   | 0.499 |
| 1 | 0.242 | 0.244 |

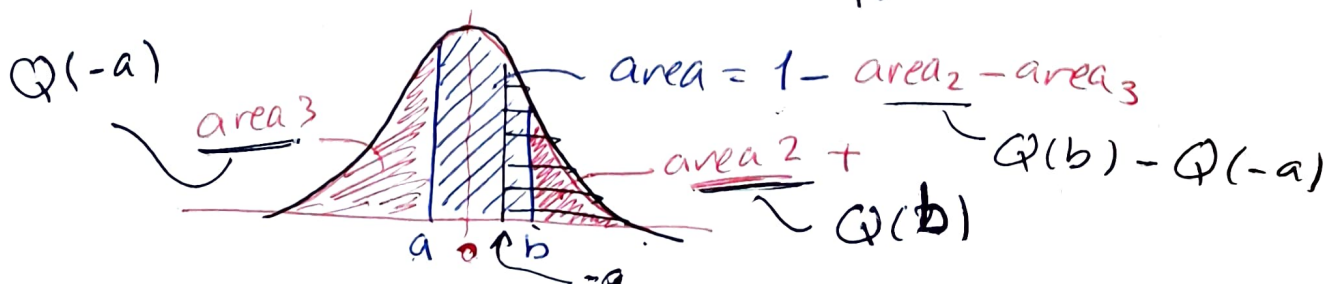
Q-table  
 Z-table  
 erf "  
 erfc "



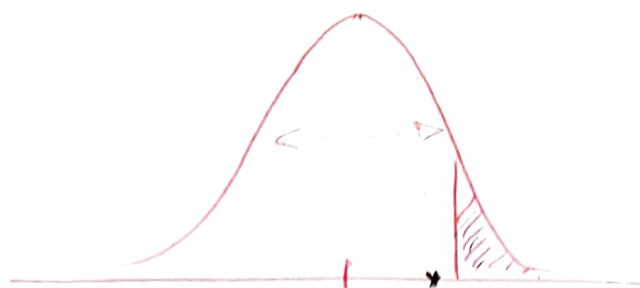
$$Q(x) = P(x > x)$$

$$Q(0) = 0.5$$

$$Q(x) = \int_x^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{a^2}{2}} da$$

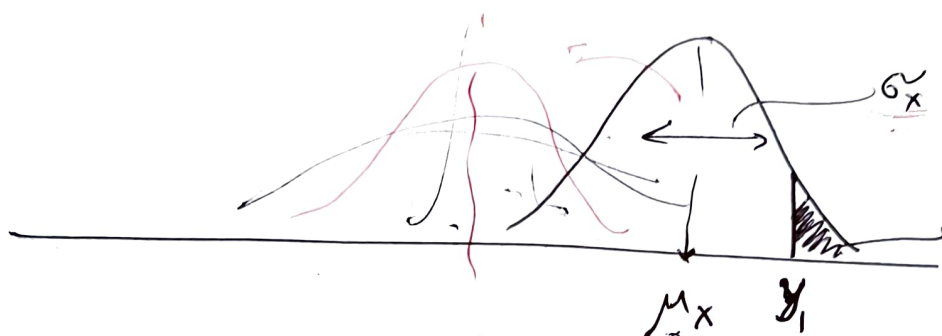


7



$$\mu = 0$$

$$\sigma_x = 1$$

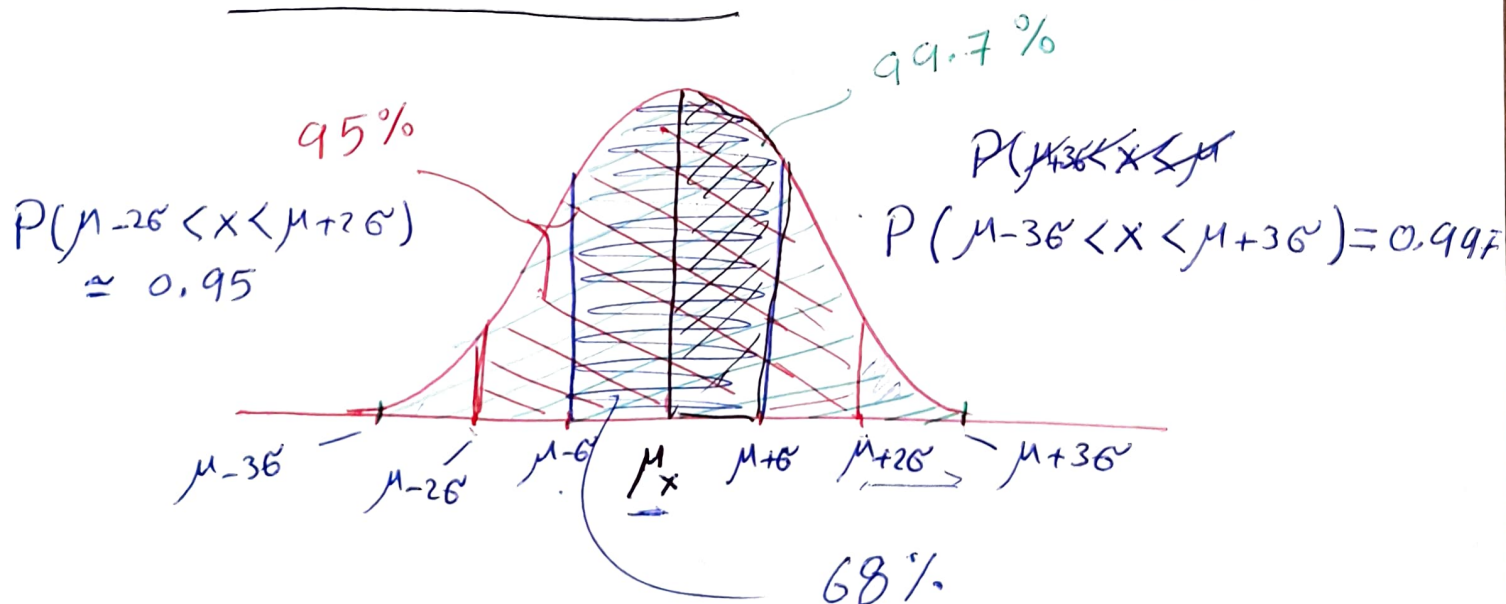


$$Y \sim N(\mu_Y, \sigma_Y^2)$$

$$P(Y > y_1) = Q\left(\frac{y_1 - \mu_Y}{\sigma_Y}\right)$$

68-95-99.7 Rule

six-sigma



$$\mu_x, \sigma_x^2$$

8

Statistical tools / moments

for Continuous Prob. Distributions.

Continuous



$$\mu_x \equiv \bar{x} \equiv \int_{-\infty}^{\infty} x \underbrace{f_x(x)}_{\substack{\uparrow \\ \text{P.D.F.}}} dx$$

P.D.F.

$$\overline{x^2} = \int_{-\infty}^{\infty} x^2 \underbrace{f_x(x)}_{\substack{\downarrow \\ \text{P.D.F.}}} dx$$

$$\sigma_x^2 = \int_{-\infty}^{\infty} (x - \bar{x})^2 f_x(x) dx$$

Discrete distributions

$$\mu_x \equiv \bar{x} = \sum_{i=1}^n x_i \underbrace{P_x(x_i)}$$

$$\sigma_x^2 = \overline{x^2} - \bar{x}^2$$

$$\overline{x^2} = \sum_{i=1}^n x_i^2 P_x(x_i)$$

$$\begin{aligned} \sigma_x^2 &= \overline{(x_i - \bar{x})^2} \\ &= \sum_i \underbrace{(x_i - \bar{x})^2} \underbrace{P_x(x)} \end{aligned}$$



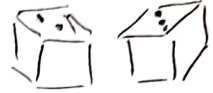
# CLT

## Central Limit theorem

X is an R.V.

Y is an R.V.

indep.

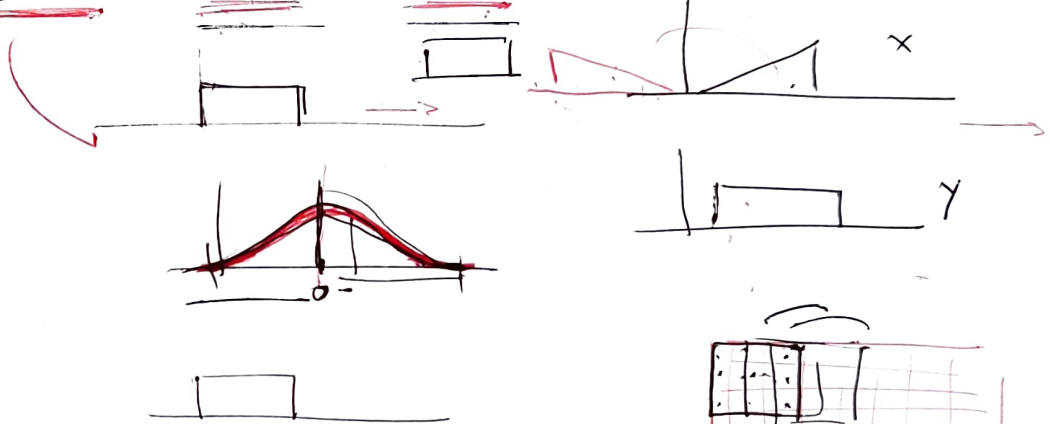


$$Z = X + Y$$

convolution!

CNN  
Convolutional

$$f_z(z) = f_x(x) * f_y(y)$$



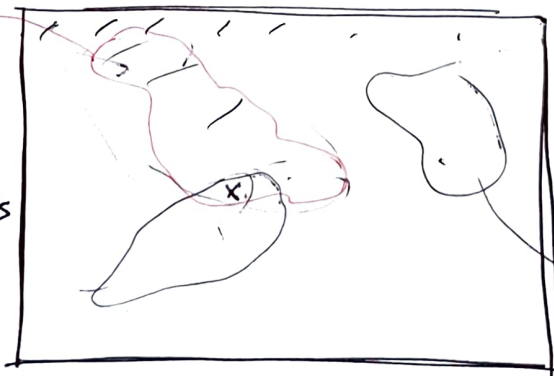
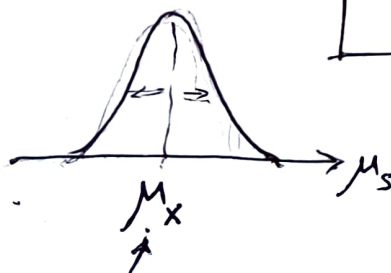
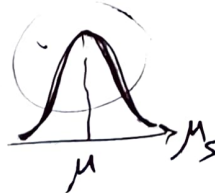
# CLT

sample mean

sample #1

$$\mu_{s1} = \sim 168$$

$$\mu_{s2} = \sim 171$$



Population

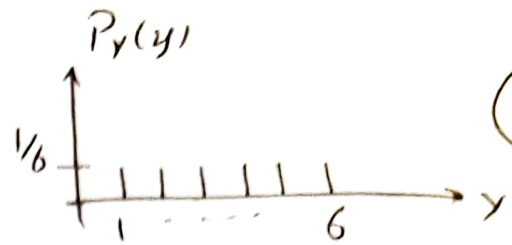
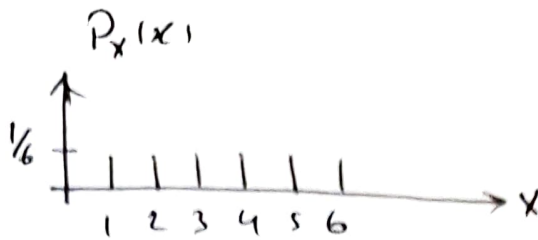
$$(45)$$

$$\mu_{s3} = 173$$

$$170$$

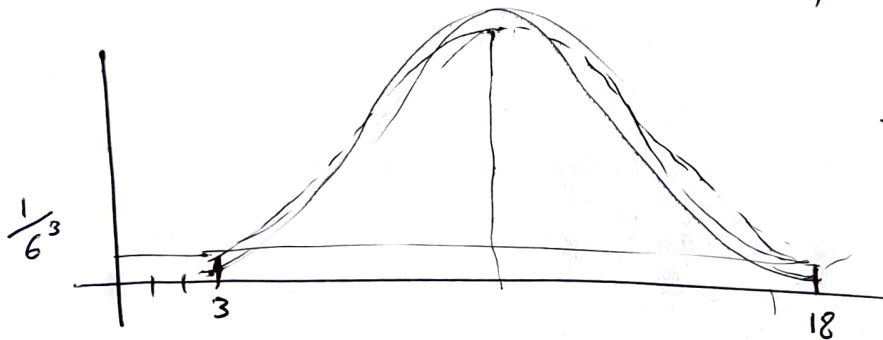
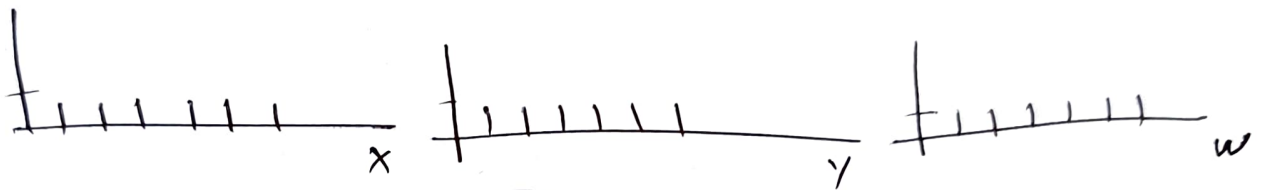
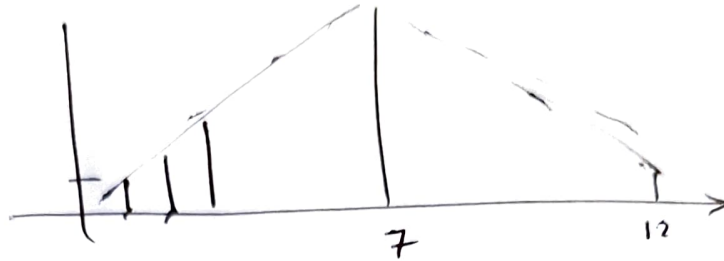
# CLT

ex.



(10)

$$Z = \underline{X} + \underline{Y}$$



$$\underline{Z} = \underline{X} + \underline{Y} + \underline{W} + \dots$$

Poisson

Bernoulli

