

$$\Rightarrow a) P(|Y_1 - \mu| \leq 1/2)$$

independent random sample  $Y_1, \dots, Y_n$  with unknown mean  $\mu$ ,  $\sigma^2 = 1$

$$Y_1 \sim N(\mu, 1)$$

$$Y_1 - \mu \Rightarrow \frac{Y_1 - (1)\mu}{(1)(1)} \Rightarrow \frac{Y_1 - (1)\mu}{\sigma(1)} \sim N(0, 1)$$

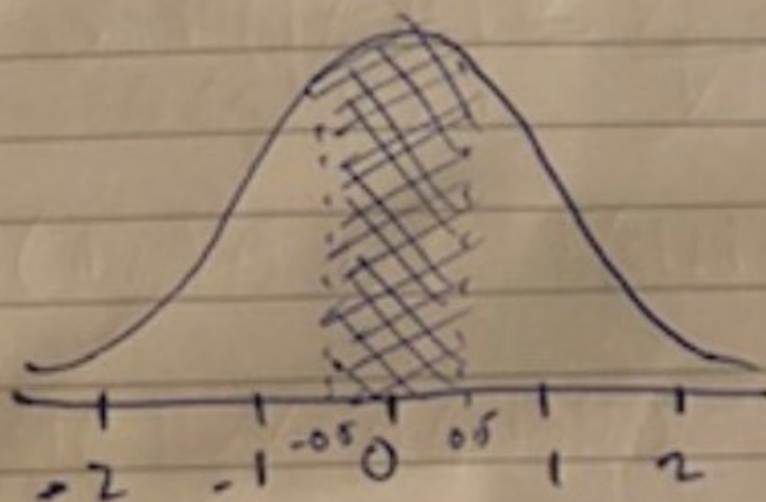
As  $Y_1 - \mu$  is already standardised.

$$P(|Y_1 - \mu| \leq 1/2) \quad \text{XXXXXXXXXX}$$

$$= P(|Z| \leq 0.5) = P(-0.5 \leq Z \leq 0.5)$$

where,  $Z \sim N(0, 1)$  is the standard normal dist<sup>n</sup>.

$$= P(Z \leq 0.5)$$



$$1) b) P(|\bar{Y} - \mu| \leq 1/2)$$

$$\bar{Y} \sim N(\mu, \frac{\sigma^2}{n}) \Rightarrow \bar{Y} \sim N(\mu, \frac{1}{9})$$

Standardising  $\bar{Y}$ , we get

$$\frac{\bar{Y} - \mu}{1/3} \sim N(0, 1) \quad 1/3 \text{ (as) } \sigma/\sqrt{n}$$

$$P(|\bar{Y} - \mu| \leq 1/2) = P\left(\left|\frac{\bar{Y} - \mu}{1/3}\right| \leq \frac{1/2}{1/3}\right)$$

$$P(|Z| \leq 3/2) = P(|Z| \leq 1.5)$$

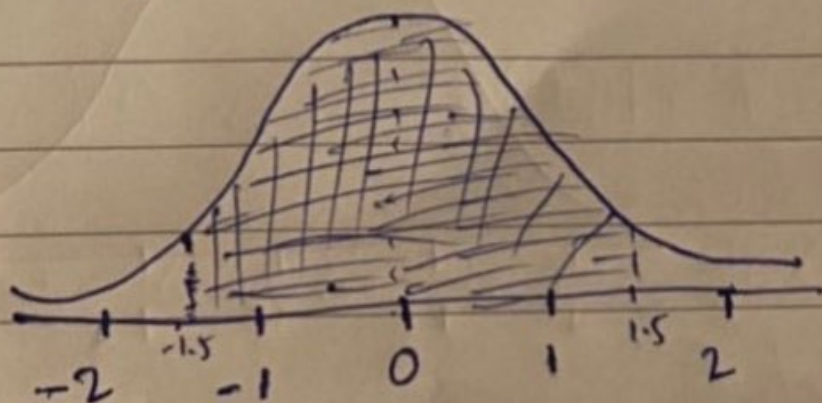
$$P(-1.5 \leq Z \leq 1.5) = P(Z \leq 1.5) - P(Z \leq -1.5)$$

$$= P(Z \leq 1.5) - (1 - P(Z \leq 1.5))$$

$$= 2P(Z \leq 1.5) - 1 = 2(0.9332) - 1$$

$$= 1.8664 - 1$$

$$= 0.8664$$





Q2)

Poisson Random Variable

mean — 2 accidents per week.

$$\lambda = 2 \times 52 = 104$$

$$X \sim \text{Poi}(\lambda = 104)$$

$$P(Y < 100) = \sum_{i=0}^{99} \frac{e^{-\lambda} \lambda^i}{i!} = \sum_{i=0}^{99} \frac{e^{-104} 104^i}{i!}$$

With Central Limit Theorem, we can say  $Y$  is sum of 52 IID Random Variables, with mean & variance  $\mu = \sigma^2 = 2$ , we can approximate  $X \approx Y \sim N(52 \times 2, 52 \times 2)$

$$\begin{aligned} P(X < 100) &\approx P(Y < 100) \\ &= P\left(\frac{Y - 104}{\sqrt{104}} < \frac{100 - 104}{\sqrt{104}}\right) \left(\because z = \frac{Y - \mu}{\sigma}\right) \\ &= P\left(z < \frac{-4}{\sqrt{104}}\right) = P(z < -0.3922) \\ &= 0.34827 \end{aligned}$$