

M362K: April 19th, 2024.

Problem from Sample Exam.

Consider two **independent** r.v.s X and Y w/ cdfs F_X and F_Y , resp.
Define:

$$\begin{aligned} U &:= \max(X, Y) \\ \text{and} \\ V &:= \min(X, Y) \end{aligned}$$

Express F_U and F_V in terms of F_X and F_Y .

→: By def'n: for all $x \in \mathbb{R}$

$$\begin{aligned} F_U(x) &= \mathbb{P}[U \leq x] \\ &= \mathbb{P}[\max(X, Y) \leq x] = \\ &= \mathbb{P}[X \leq x, Y \leq x] = \\ &= \underbrace{\mathbb{P}[X \leq x]}_{F_X(x)} \cdot \underbrace{\mathbb{P}[Y \leq x]}_{F_Y(x)} \quad \text{independence} \\ &= F_X(x) \cdot F_Y(x) \end{aligned}$$

By def'n: for all $x \in \mathbb{R}$

$$\begin{aligned} F_V(x) &= \mathbb{P}[V \leq x] \\ &= \mathbb{P}[\min(X, Y) \leq x] \quad \mathbb{P}[E] = 1 - \mathbb{P}[E^c] \\ &= 1 - \mathbb{P}[\min(X, Y) > x] = \\ &= 1 - \mathbb{P}[X > x, Y > x] = \\ &= 1 - \underbrace{\mathbb{P}[X > x]}_{1 - F_X(x)} \cdot \underbrace{\mathbb{P}[Y > x]}_{1 - F_Y(x)} \quad \text{independence} \\ &= 1 - (1 - F_X(x))(1 - F_Y(x)) \\ &= \cancel{1} - \cancel{1} + F_X(x) + F_Y(x) - F_X(x)F_Y(x) \\ &= F_X(x) + F_Y(x) - F_X(x) \cdot F_Y(x) \end{aligned}$$



More about the Normal Dist'n.

Recall: $Z \sim N(0,1)$

Any normal $X \sim N(\mu, \sigma^2)$ can be expressed as

$$X = \mu + \sigma \cdot Z \quad \text{for some } Z \sim N(0,1)$$

- $E[X] = E[\mu + \sigma \cdot Z] = \mu + \sigma \cdot \underbrace{E[Z]}_{=0} = \mu$
↑
linearity

- $\text{Var}[X] = \text{Var}[\underbrace{\mu}_{\text{shift}} + \sigma \cdot Z] = \sigma^2 \cdot \underbrace{\text{Var}[Z]}_{=1} = \sigma^2$

Problem. Let X be normally distributed w/ mean μ and std dev σ . What is the probability that X falls within 1% of the std dev from its mean?

→: $X \sim \text{Normal}(\text{mean} = \mu, \text{sd} = \sigma)$

$$\mathbb{P}\left[|X - \mu| < \frac{1}{10} \sigma\right] = ?$$

$$\mathbb{P}\left[-\frac{1}{10}\sigma < X - \mu < \frac{1}{10}\sigma\right] =$$

$$= \mathbb{P}\left[-0.1 < \frac{X - \mu}{\sigma} < 0.1\right] =$$

$$= \mathbb{P}[-0.1 < Z < 0.1] =$$

$$= \Phi(0.1) - \underbrace{\Phi(-0.1)}_{1 - \Phi(0.1)} = 2\Phi(0.1) - 1 = 2 \cdot 0.5398 - 1 = 1.0796 - 1$$

$$= 0.0796$$



The Central Limit Theorem.

Let $\{X_1, X_2, \dots\}$ be a sequence of i.i.d. r.v.s w/ $\mu_X = \mathbb{E}[X_1] < \infty$
and $\sigma_X^2 = \text{Var}[X_1] < \infty$.

Define for all $n \in \mathbb{N}$:

$$S_n = X_1 + X_2 + \dots + X_n$$

$$\frac{S_n - n \cdot \mu_X}{\sigma_X \sqrt{n}} \xrightarrow[n \rightarrow \infty]{\mathcal{D}} N(0,1)$$

Problem. Automobile losses are independent and uniformly dist'd between 0 and 15,000.
Calculate the (approximate) probability that the total amount for 200 losses lies between 1,450,000 and 1,700,000.

$$\rightarrow X_i \sim U(0, 15000) \quad i = 1, \dots, 200$$

$$S_{200} = X_1 + X_2 + \dots + X_{200}$$

We need $\mathbb{P}[1.45 \cdot 10^6 < S_{200} < 1.7 \cdot 10^6] = ?$

$$\mu_X = \mathbb{E}[X_1] = 7500$$

In general: $X \sim U(a, b)$

$$\text{Var}[X_1] = \frac{(15000)^2}{12} = 18\,750\,000$$

$$\mathbb{E}[X] = \frac{a+b}{2}$$

$$\sigma_X = 4330.127$$

$$\text{Var}[X] = \frac{(b-a)^2}{12}$$

$$\mathbb{P}\left[\frac{1.45 \cdot 10^6 - 7500 \cdot 200}{4330.127 \cdot \sqrt{200}} < \overset{\sim N(0,1)}{Z} < \frac{1.7 \cdot 10^6 - 7500 \cdot 200}{4330.127 \sqrt{200}} \right] =$$

$$\approx \mathbb{P}[-0.82 < Z < 3.27] = \Phi(3.27) - \Phi(-0.82)$$

$$= 0.9995 - 0.2061 = 0.7934$$

□

Section 4.2.

Exponential Distribution.

A random variable T has the exponential distribution with rate λ , written as

$$T \sim \text{Exp}(\lambda)$$

If T has the pdf given by $f_T(t) = \lambda e^{-\lambda t}$ for $t \geq 0$

Q: What is the cdf of T ?

$$\rightarrow: F_T(t) = \int_0^t f_T(u) du = \int_0^t \lambda e^{-\lambda u} du = \lambda \cdot \left(-\frac{1}{\lambda}\right) e^{-\lambda u} \Big|_{u=0}^t$$

$$F_T(t) = 1 - e^{-\lambda t}$$

Def'n. The survival function of any random variable X is given by

$$S_X: \mathbb{R} \rightarrow [0, 1]$$

$$S_X(x) = 1 - F_X(x) \quad \text{for all } x \in \mathbb{R}$$

Note: For $T \sim \text{Exp}(\lambda)$:

$$S_T(t) = \mathbb{P}[T > t] = e^{-\lambda t}$$

Without proof:

$$\mathbb{E}[T] = \text{SD}[T] = \frac{1}{\lambda}$$