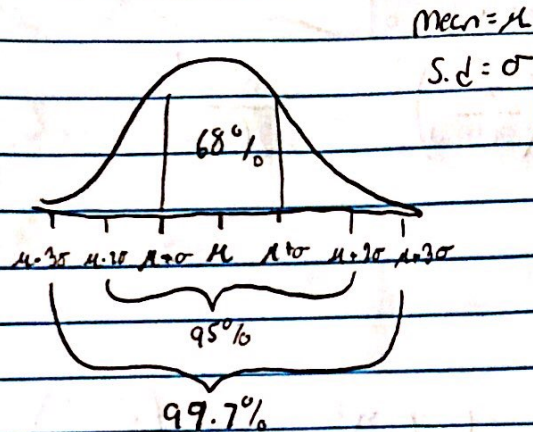


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Empirical Rule



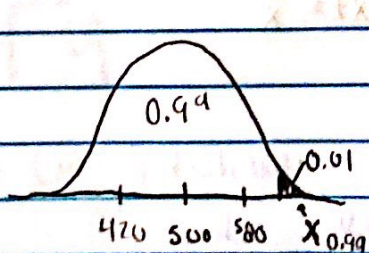
Example: College A and B both use SAT Score for admission criteria. $\mu = 500$ $\sigma = 80$

College A accepts people whose score is above 600.

Q1) What % of high school seniors can get into college A?

$$\begin{aligned} \text{Score is } X \rightarrow P(X > 600) &= P\left(\frac{X - 500}{80} > \frac{600 - 500}{80}\right) \\ &= P\left(Z > \frac{5}{4}\right) = 1 - P(Z \leq 1.25) = 0.1056 = 10.56\% \end{aligned}$$

Q2) College B accepts the top 1%. What is the minimum score required to get accepted by college B?



$$X = 500 + 80Z$$

$$P(X \leq X_{0.99}) = 0.99$$

$$P(500 + 80Z \leq X_{0.99}) = 0.99$$

$$P\left(Z \leq \frac{X_{0.99} - 500}{80}\right) = 0.99$$

$$Z_{0.99} = \frac{X_{0.99} - 500}{80}$$

$$\begin{aligned} X_{0.99} &= 500 + 80Z_{0.99} \\ &= 686 \end{aligned}$$

□ Other common continuous distributions

Uniform

$$a < b \quad f(x) = \frac{1}{b-a} \quad x \in [a, b]$$

$$E(x) = \int_a^b x \frac{1}{b-a} dx = \left(\frac{x^2}{2(b-a)} \right)_a^b = \frac{b^2 - a^2}{2(b-a)}$$

$$= \frac{(b-a)(b+a)}{2(b-a)} = \frac{b+a}{2}$$

$$E[x^2] = \int_a^b x^2 \frac{1}{b-a} dx = \frac{1}{b-a} \left(\frac{x^3}{3} \right)_a^b = \frac{1}{b-a} \left(\frac{b^3 - a^3}{3} \right)$$

$$E(x^2) - E(x)^2 = \frac{b^3 - a^3}{3(b-a)} - \frac{(b+a)^2}{4} = \frac{4b^3 - 4a^3 - 3(b+a)(b-a)}{12(b-a)}$$

$$= \frac{4b^3 - 4a^3 - 3b^2 + 3a^2b - 3ab^2 + 3a^3}{12(b-a)}$$

$$= \frac{b^3 - a^3 + 3a^2b - 3ab^2}{12(b-a)} = \frac{(b-a)^3}{12(b-a)} = \frac{(b-a)^2}{12}$$

Variance

Exponential $\lambda > 0$

$$f(x) = \lambda e^{-\lambda x} \quad x > 0 \quad E(x) = \frac{1}{\lambda}$$

$$\text{Var}(x) = \frac{1}{\lambda^2}$$

Gamma distribution $\alpha > 0, \beta > 0$

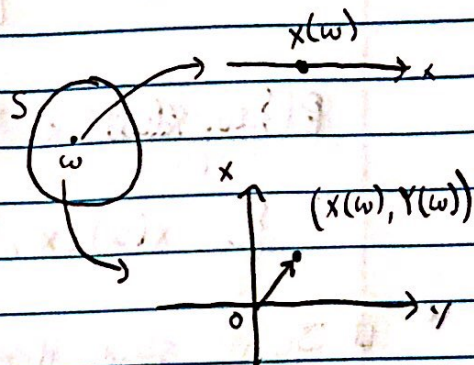
$$f(x) = \frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x} \quad E(x) = \frac{\alpha}{\beta}, \quad V(x) = \frac{\alpha}{\beta^2}$$

gamma

Sum of α independent identical exponential (λ) becomes
gamma(α, λ)

Jointly distributed random variables

Random vector $\begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$



A random vector (X, Y) is a vector
valued function from (S) to \mathbb{R}^2

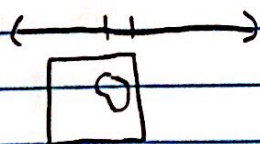
More generally, k -component random vector (X_1, \dots, X_k) is a
function from S to \mathbb{R}^k , $k \geq 2$

Every component of a random vector is a random variable denote the
range of (X, Y) to be $S_{X,Y}$

S_X : range of X

S_Y : range of Y

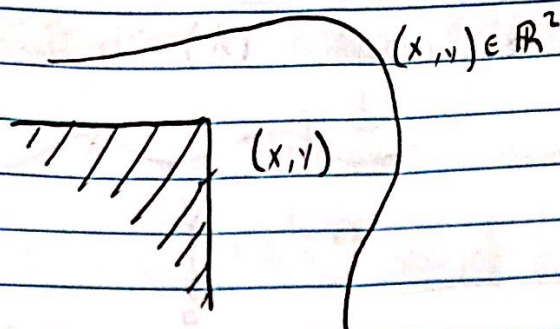
Joint Distribution Function of random variables



The (joint) distribution function of (X, Y) is a function from
 \mathbb{R}^2 to \mathbb{R}^1 is defined to be $F_{X,Y}(x,y) = P(X \leq x, Y \leq y)$



bivariate function



$$P(\{\omega: X(\omega) \leq x, Y(\omega) \leq y\})$$

$$\{\omega: X(\omega) \leq x, Y(\omega) \leq y\}$$

- Joint: study distribution of (X, Y)
- Marginal: we study distributions of components of the random vector
 - ↳ study X
 - ↳ study Y

each component of

- If (X, Y) is discrete we have joint pmf

- If each component of (X, Y) is continuous we have joint pdf

We need to extend domain of P and F accordingly

$$\mathbb{Z}^2 \text{ or } \mathbb{R}^2$$