

random variable maps a sample
space to a real #

↳ random number that is based
on the outcome of
experiment

↳ discrete random variable
can take either a finite
or countable # of values

Countable^{list} \rightarrow 1-1 correspondence
between items in the
list and the positive

integers

$$\underline{\underline{X}} \quad \{ \underline{\underline{x_1, x_2, \dots}} \} : r$$

↳ generate a random number $\underline{\underline{X}}$

$$p(x) = P(\underline{\underline{X}} = x)$$

$$p(3) = P(\underline{\underline{X}} = 3)$$

$$1 \geq p(x) \geq 0$$

$$x \notin r \Rightarrow p(x) = 0$$

$$x \in r \Rightarrow p(x) > 0$$

$$\sum_{x_i \in r} p(x_i) = 1$$

$$\underline{\underline{X}} \rightarrow \{1, 2, 3\}$$

$$p(1) = .5$$

$$p(2) = .4$$

$$F(0) = 0 \Rightarrow \text{there are no values in } r \leq 0 = p(1)$$

$$\underline{\underline{F(1) = .9 = \sum_{x_i \in \{1, 2, 3\}} p(x_i)}}$$

$$p(3) = .1$$

$$F(2) = p(1) + p(2)$$

$$x_i \leq 1$$

$$= .9$$

$$F(3) = 1.0$$

$$F(2.5) = .9$$

cdf cumulative distribution function
 $x \rightarrow r$

$$F(x) = P(\underline{X} \leq x)$$

discrete with $r = \{x_1, \dots\}$

pmf. p

$$F(x) = \sum p(x_i)$$

$$x_i \leq x$$

$$x_i \in r$$

Continuous random variable

can take any value in some range

$$[a, b]$$

or even some
union of these
ranges

pdf is the analog of the pmf

for a continuous random variable

$$P(\underline{X} = x) = 0$$

$$P(\underline{X} \in (c, d)) = \int_c^d \underline{f(x)} dx$$

$$\int_{-\infty}^{\infty} f(x) dx = 1 \quad f(x) \geq 0$$

$$P(\underline{X} \in (x - \Delta x, x + \Delta x)) \approx 2\Delta x f(x)$$

$$\int_{x-\Delta x}^{x+\Delta x} f(x) dx$$

rectangle with
height $f(x)$
and width $2\Delta x$

→ cdf in this case

$$F(x) = P(\underline{X} \leq x) = \int_{-\infty}^x f(y) dy$$

$$\frac{d}{dx} F(x) = f(x)$$

$f(x)$

$$f(x) = \begin{cases} 2x, & 0 < x < 1 \\ 0, & \text{everywhere else} \end{cases}$$

what is the cdf of this

$$F(-1) = \int_{-\infty}^{-1} 0 \cdot dx = 0$$

$$F(3) = 1 = \int_{-\infty}^0 0 \cdot dx + \int_0^1 2x \cdot dx + \int_1^3 0 \cdot dx$$

$$= x^2 \Big|_0^1 = 1$$

$$F(x) = \begin{cases} 0 & , x < 0 \\ x^2 & , x \in [0, 1] \\ 1 & , x > 1 \end{cases}$$

$$P(a < X \leq b) = F(b) - F(a)$$

$$\{X \leq b\} = \{\underline{X} \leq a\} \cup \{a < \underline{X} \leq b\}$$

$$\underline{P(X \leq b)} = \underline{P(\underline{X} \leq a)} + P(a < \underline{X} \leq b)$$

$F(b)$ $F(a)$

Expectation

↳ the expectation of random variable

\underline{X} is the probability weighted
avg value of \underline{X}

$$x_1, \dots, x_n \quad w_1, \dots, w_n \quad \sum w_i = 1$$

$$\sum_{i=1}^n w_i x_i$$

discrete random variable:

$$\underline{X} \rightarrow r = \{x_1, \dots\}$$

pmf p

$$E[\underline{X}] = \sum_{x \in r} p(x_i) \cdot x_i$$

$$\underline{X} \rightarrow \{1, 2, 3\}$$

$$p(1) = .5, \quad p(2) = .4, \quad p(3) = .1$$

$$E[X] = (.5)(1) + (.4)(2) + (.1)(3)$$

$$= 1.6$$

Continuous random \rightarrow density f

$$E[X] = \int_{-\infty}^{\infty} x \cdot f(x) dx$$

$$\underline{Ex} \quad f(x) = \begin{cases} 2x, & x \in (0, 1) \\ 0, & \text{o/w} \end{cases}$$

$$E[X] = \int_0^1 2x \cdot x dx$$

$$+ \int_{-\infty}^0 0 \cdot x dx + \int_1^{\infty} 0 \cdot x dx$$

$$= \int_0^1 2x^2 dx$$

$$= \frac{2}{3} x^3 \Big|_0^1 = \frac{2}{3}$$

property of expectation

↳ random variable

↳ $g(\cdot)$ → want to calculate the expectation of $g(x)$

$$E[g(x)] = \sum_{x_i} g(x_i) \cdot p(x_i) \quad (\text{discrete})$$

$$= \int_{-\infty}^{\infty} \underline{g(x) \cdot f(x)} dx \quad (\text{cont})$$

Variance of a random variable

$$\mu = E[X]$$

$$\text{Var}(X) = E[(X - \mu)^2]$$

= "the average squared distance from the average"

identity $\text{Var}(X) = E[X^2] - E[X]^2$

standard deviation of X is

$$\sqrt{\text{Var}(X)}$$

Let's suppose we have two random variables X and $Y \rightarrow$ discrete and let's they are both defined on the sample space

discrete $p(x, y) = P(X = x \cap Y = y)$

Ex roll two four-sided dice

$X_1 =$ dice 1 outcome

$X_2 =$ dice 2 outcome

$$X = \min(X_1, X_2) \rightarrow \{1, 2, 3, 4\}$$

$$Y = \max(X_1, X_2) \rightarrow \{1, 2, 3, 4\}$$

$$\begin{aligned} P(1,1) &= P(X=1 \text{ and } Y=1) \\ &= P(\text{both } X_1 \text{ and } X_2 \text{ are } 1) \\ &= 1/16 \end{aligned}$$

$$P(2,1) = 0$$

$$P(1,2) = P(\{1,2\} \cup \{2,1\}) = 1/8$$

$$P(1,3) = P(\{1,3\} \cup \{3,1\}) = 1/8$$

$$P(1,4) = 1/8$$

→ joint distribution for continuous random variables

$$[a,b], [c,d] \in \mathbb{R} \quad f(x,y)$$

$$P(X \in [a,b] \cap Y \in [c,d])$$

$$= \int_a^b \int_c^d f(x,y) dy dx$$

Def $g(x, y)$ \underline{X} and \underline{Y} i.i.d. with joint p.d.f. or density f

$$E[g(x, y)]$$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} g(x, y) \cdot f(x, y) dx dy$$

$$= \sum_{y_i} \sum_{x_i} p(x_i, y_i) \cdot g(x_i, y_i)$$

$\Rightarrow \text{Cov}(X, Y)$ $X, Y \sim f$
 $E[X] = \mu_x$ $E[Y] = \mu_y$

$$= E[(\underline{X} - \mu_x)(\underline{Y} - \mu_y)]$$

= "probability weighted avg. of the prod. of signed difference of X and Y from their means"

$$= E[\underline{X}\underline{Y}] - E[\underline{X}]E[\underline{Y}]$$

$$\text{Cov}(X, X) = \text{Var}(X)$$

