

$$\begin{aligned} \textcircled{1} & , P(X=1) = 0.8 \\ \textcircled{2} & , P(X=2) = 0.2 \end{aligned}$$

$$\sum 1 \cdot P(X=1) = 1 \cdot 0.8 + 2 \cdot 0.2 = 1.2$$

Date _____

~~PRACTICE~~ PRACTICE expectation.

→ mean is also expectation.

The expected value or mean value of a discrete random variable is defined as.

$$E[X] = \sum_{p(u) > 0} u \cdot P(X=u) = \sum_{p(u) > 0} u \cdot p(u)$$

means that if we have random variable 1 and 2 and their respective probab is 0.8 and 0.2 then the mean would be

$$\begin{aligned} \text{mean} &= (1) \cdot P(X=1) + (2) \cdot P(X=2) \\ &= (1) \cdot (0.8) + (2) \cdot (0.2) \\ &= 0.8 + 0.4 \end{aligned}$$

mean. 1.2

•) $E[X]$ is a weighed average of the possible values u that X can take on, each value being weighted by the probability $p(u)$.

Q) X is the outcome when we roll a 4 sided fair die, find $E[X]$ $X =$ the random variable

step 1) X is the outcome since n=4
 X is $[1, 2, 3, 4]$.

step 2)

$$\begin{aligned} \text{now mean} &= P(X=1) \cdot (1) + \\ &P(X=1) \cdot (1) + P(X=2) \cdot 2 + P(X=3) \cdot 3 + P(X=4) \cdot 4. \end{aligned}$$

SOLO

$$X, Y$$

$$(-1)^2 \cdot \frac{1}{4} + 0^2 \cdot \frac{1}{2} + 1^2 \cdot \frac{1}{4} = \frac{1}{4} + 0 + \frac{1}{4} = \frac{2}{4} = \frac{1}{2}$$

$$\sum g(x) \cdot p(x)$$

$$\sum (x^2) \cdot p(x)$$

Date _____

$$1 \cdot \frac{1}{4} + 2 \cdot \frac{1}{4} + 3 \cdot \frac{1}{4} + 4 \cdot \frac{1}{4} \quad (1)^2 \cdot C_p = x=1 \quad 1 \cdot \frac{1}{4}$$

$$\frac{1}{4} + \frac{1}{2} + \frac{3}{4} + 1$$

$$\frac{1+2+3+4}{4}$$

$$\frac{10}{4} = \frac{5}{2}$$

$$\frac{5}{2} = 2.5$$

$$\text{mean} = 2.5$$

If I throw this die for a large number of times then I will get the expected value of a mean value of 2.5

o) Expectation of a function of a random variable

Let X is a discrete random variable and g is a real valued function then the expectation (or expected value) of $Y = g(X)$ is

$Y = g(X)$ \rightarrow new random variable

$$E[g(X)] = \sum_{u: p_X(u) > 0} g(u) \cdot p_X(u)$$

X is a discrete random variable with pmf.

$$u = -1 \quad 0 \quad 1$$

$$p_X(u) = \frac{1}{4} \quad \frac{1}{2} \quad \frac{1}{4}$$

$$Y = X^2 = \text{compute } E(Y).$$

SOLO

$$g(u) = y.$$

$$x^2 = y$$

Date _____

$$E(Y) = 3 \cdot \left(\frac{1}{4}\right) + 1 \cdot \left(\frac{1}{4}\right)$$

$$\frac{1}{4} + \frac{1}{4}$$

$$\frac{2}{4}$$

$$E(Y) = \frac{1}{2}$$

This is useful because by this method we can find out about the expectation of function of X as well.

From previous example we could only find out about the expectation of X .

Also,

$$\sum g(u) \cdot P_X(u)$$

$P_X(u) > 0$.

$$\sum x^2 \cdot P_X(u)$$

why is this?

well we were supposed to

find $E[Y]$ and as we saw in the generic formula

$$E[Y] = \sum g(u) \cdot P_X(u)$$

↳ same function

so in this example ^{of Y} we have

$$E[Y] = x^2.$$

$$\text{so } x^2 \cdot P_X(u)$$

where $g(x)$ is x^2 .

Properties of expected values:

→ If a and b are constants then,

$$E[aX + b] = aE[X] + b$$

holds for all random variables X ,

→ This property is saying that we can take constants outside and then do the calc with them.

→ Proof for discrete random variable: Let $g(X) = aX + b$

$$E[g(X)] = \sum_{u: P(u) > 0} g(u) \cdot P(u)$$

$$E[aX + b] =$$

$$\sum (aX + b) \cdot P(u)$$

SOLO

$$E[X] = \sum u \cdot p_u \quad Z = X + Y \quad E[Z] = \sum p(g(u)) \cdot f(u)$$

Date _____

$$g(u) = ax + b$$

$$\sum (ax + b) \cdot p(x)$$

$$\boxed{\sum p(u) = 1}$$

$$\sum ax \cdot p(x) + \sum b \cdot p(u)$$

$$\underbrace{\sum ax \cdot p(x)}_{E[X]}$$

$$a \sum u \cdot p(u) + b \sum p(u)$$

$$a E[X] + b$$

$$\underbrace{\sum u \cdot p(u)}_{E[X]}$$

$$g(x) = ax + b$$

$$E[ax + b] = \text{proved above.}$$

o) Constant factor.

$$E[aX] = a E[X]$$

o) Constant

→ expectation of a constant is itself.

$$E[b] = b$$

For two random variable's X and Y.

$$E[X + Y] = E[X] + E[Y]$$

$$Y = 3X + 2$$

$$p(k)$$

$$\sum (3k + 2) \cdot p(k)$$

$$\sum (3k) \cdot p(k) + \sum 2 \cdot p(k)$$

$$3 \sum (k) \cdot p(k) + 2$$

$$3 \underbrace{\sum k \cdot p(k)}_{E[X]} + 2$$

SOLO

Date _____

PRACTICE.

Variance and Standard deviation:

1) Variance:

Expected value / mean represented by μ or $E[X]$ yields the weighted average of the possible values of X .

μ is the weighted average of all the possible values of x .

Variance measures the variation (or spread) of these values.

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

This holds for all random variable X .

Simplification:

$$\mu = E[X]$$

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

Standard deviation:

$$\sigma = \text{SD}(X) = \sqrt{\text{Var}(X)}$$

Q1) compute the S.D of fair 4 sided dice.

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

$$\mu = (1) \cdot \frac{1}{4} + 2 \cdot \frac{1}{4} + 3 \cdot \frac{1}{4} + 4 \cdot \frac{1}{4}$$

$$\mu = \frac{1}{4} + \frac{1}{2} + \frac{3}{4} + 1 = 2.5$$

SOLO

Date _____

$$u^2 = (2.5)^2$$
$$u^2 = 6.25$$

$$E[X^2] = \sum_i x_i^2 \cdot p(x_i)$$

$$E[X^2] = (1)^2 \cdot \frac{1}{4} + (2)^2 \cdot \frac{1}{4} + (3)^2 \cdot \frac{1}{4} + (4)^2 \cdot \frac{1}{4}$$

$$\frac{1}{4} + 1 + \frac{9}{4} + 4 = 6.75$$

$$\text{Var}(X) = 6.75 - 6.25$$

$$\text{Var}(X) = 0.5$$

$$\text{S.D.} = \sqrt{\text{Var}(X)}$$

$$\text{S.D.} = \sqrt{0.5}$$

$$\text{S.D.} = 0.707 \text{ or } 1.12 \text{ ans}$$

→ Higher the variance the further the random variable is set apart, lies on a greater region.

→ larger the S.D or variance the random variable is spread over a larger area, the higher the spread.