

Mean of a Discrete Random Variable

$$\mu = \frac{\sum x_i}{N}$$

The mean of a discrete random variable X , μ_x or μ , is

$$\mu = \sum x P(X=x)$$

Also called expected value or expectation.

Ex Six-sided die $P(X=x) = \frac{1}{6}$ $x=1, 2, 3, 4, 5, 6$

$$\begin{aligned}\mu &= 1\left(\frac{1}{6}\right) + 2\left(\frac{1}{6}\right) + 3\left(\frac{1}{6}\right) + 4\left(\frac{1}{6}\right) + 5\left(\frac{1}{6}\right) + 6\left(\frac{1}{6}\right) \\ &= \frac{1+2+3+4+5+6}{6} \\ &= 3.5\end{aligned}$$

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Note We may also use expected value or expectation in place of "mean".

$$\mu = \frac{\sum x_i}{N} = \frac{19 + 20 + 20 + 19 + 21 + 27 + 20 + 21}{8}$$

$$= \frac{19 + 19 + 20 + 20 + 20 + 21 + 21 + 27}{8}$$

$$= \frac{2 \times 19 + 3 \times 20 + 2 \times 21 + 27}{8}$$

$$= 19\left(\frac{2}{8}\right) + 20\left(\frac{3}{8}\right) + 21\left(\frac{2}{8}\right) + 27\left(\frac{1}{8}\right) \quad \star$$

$$= 19 P(X=19) + 20 P(X=20) + 21 P(X=21) + 27 P(X=27)$$

$$= \sum x P(X=x)$$

Age X	P(X=x)
19	$\frac{2}{8} = 0.250$
20	$\frac{3}{8} = 0.375$
21	$\frac{2}{8} = 0.250$
27	$\frac{1}{8} = 0.125$

Interpretation

$$\mu_x = 2.3$$

In a large number of independent observations of a random variable X , the average value of the observations will approximately equal the mean μ of X .

Further, the larger the number of observations, the closer the average will tend to be to μ .

This principle is referred to as the law of averages or the law of large numbers.

Standard Deviation of a Discrete Random Variable

denoted σ_x or σ

$$\sigma = \sqrt{\sum (x_i - \mu)^2 P(X=x_i)}$$

OR

$$\sigma = \sqrt{\sum x_i^2 P(X=x) - \mu^2}$$

As before

σ^2 is the variance of X .

x	P(X=x)	x P(X=x)	X = # of tellers busy with customers at 1pm
0	0.029	0.000	$\sigma = \sqrt{\sum x^2 P(X=x) - \mu^2}$ $= \sqrt{19.438 - (4.118)^2}$ $= 1.6$
1	0.049	0.049	
2	0.078	0.156	
3	0.155	0.465	
4	0.212	0.848	
5	0.262	1.310	
6	0.215	1.290	

$$\mu = 4.118$$

x^2	$x^2 P(X=x)$
0	0
1	0.049
4	0.312
9	1.395
16	3.392
25	6.550
36	7.740
<hr/>	
19.438	

On average, the number of busy tellers varies around the mean of 4.118 by 1.6.