

# 1. Introduction

Welcome to the series of E-learning modules on moments and quantiles. Here we discuss about raw and central moments, their relationships, uses etc. There are many quantiles in general but here in particular we discuss about median, quartiles, deciles and percentiles.

By the end of this session, you will be able to :

- Know about raw and central moments
- Understand the conditions for existence of moments
- Understand the relationship between raw and central moments
- Explain the uses of the moments
- Describe Quantiles, that is, median, quartiles, deciles and percentiles.

## Introduction

- The concept of moments was borrowed from physics
- The characteristics of a frequency distribution are described by its moments
- The  $r^{\text{th}}$  moment of a set of values about any constant is the mean of the  $r^{\text{th}}$  powers of the deviations of the values from the constant
- Moments about any constant can be found
- The moments about the arithmetic mean are called central moments
- The moments about any other constant are called raw moments

If  $X$  is a continuous random variable, then  $r^{\text{th}}$  raw moment is given by,

$\mu_r'$  is equal to Expectation of  $(X^r)$

If the continuous random variable  $X$  has probability density function  $f(x)$ , then  $\mu_r'$  is equal to integral over  $x$ ,  $x^r$  into  $f(x) dx$

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$r^{\text{th}}$  moment about any point  $A$  is given by,

$\mu_r(A)$  is equal to expectation of  $(X - A)^r$

Is equal to integral over  $x$ ,  $(x - A)^r$  into  $f(x) dx$

Suppose in above expression if we take  $A$  is equal to Expectation of  $(X)$  then we get moment about mean or central moment which is denoted by  $\mu_r$  and is given by,

$\mu_r$  is equal to expectation of  $(X - E(X))^r$

This is equal to integral over  $x$ ,  $(x - E(X))^r$  into  $f(x) dx$ .

## 2. Existence of Moments

Expectation of modulus of  $X$  always exists in the extended real numbers.  $E|X|$  is equal to  $R$  union infinity union minus infinity and Expectation of modulus of  $X$  belongs to the closed interval  $[0, \infty)$ ; that is, either Expectation of modulus of  $X$  is a non-negative real number or Expectation of modulus of  $X$  is equal to infinity.

- Expectation of  $X$  exists and is finite that is,  $E|X| < \infty$ .
- Expectation of modulus of  $X$  is less than infinity, implies expectation of  $X$  is less than or equal to expectation of modulus of  $X$  which is finite.
- If  $1 \leq r \leq s$ , then  
Expectation of modulus of  $X$  power  $s$  is less than infinity implies expectation of modulus of  $X$  power  $r$  is less than infinity.

Note that the first two conditions have already been proved in the previous module. Now let us prove the third one.

Given expectation of modulus of  $X$  power ' $s$ ' is less than infinity then we need to prove that expectation of  $X$  power  $r$  also exists, that is it is less than infinity.

Consider integral from minus infinity to infinity modulus of  $x$  power  $r$  into  $f(x) dx$ .

This integral is divided as integral from minus one to one and minus infinity to minus one and one to infinity, which are combined as modulus of  $x$  greater than one.

That is, integral from minus one to one modulus of  $x$  power  $r$  into  $f(x) dx$  plus integral modulus of  $x$  greater than one modulus of  $x$  power  $r$  into  $f(x) dx$

If  $r$  is less than  $s$  then modulus of  $x$  power  $r$  is less than modulus of  $x$  power  $s$  for modulus of  $x$  greater than one

Therefore

integral from minus infinity to infinity modulus of  $x$  power  $r$  into  $f(x) dx$  is less than or equal to integral over minus one to one modulus of  $x$  power  $r$  into  $f(x) dx$  plus integral modulus of  $x$  greater than one, modulus of  $x$  to the power  $s$  into  $f(x) dx$

is less than or equal to integral from minus one to one  $f(x) dx$  plus integral modulus of  $x$  greater than one, modulus of  $x$  power to the power  $s$  into  $f(x) dx$

Since for minus one less than one, modulus of  $x$  power  $r$  is less than one

Integral over minus infinity to infinity modulus of  $x$  power  $r$  into  $f(x) dx$  is less than or equal to one plus integral over modulus of  $x$  greater than one, modulus of  $x$  power  $s$  into  $f(x) dx$  is less than infinity.

Implies, expectation of modulus of  $X$  power  $r$  exists for all  $1 \leq r \leq s$ .

Note that This result states that if the moments of a specified order exist, then all the lower order moments automatically exist. However, the converse is not true. That is we may have a distributions for which all moments of a specified order exist but no higher order moment exist. We prove this with an example.

Let  $X$  be a random variable with probability density function

$f(x)$  is equal to  $2/x^3$ , for  $x$  greater than or equal to one and equal to zero for  $x$  less than one

We have Expectation of  $X$  is equal to integral over one to infinity  $x$  into  $f(x) dx$ .

By substituting  $f(x)$  and simplifying we get,

$2 \int_1^{\infty} x^{-2} dx$

Is equal to minus two divided by  $x$  ranges from one to infinity is equal to two.

Now consider expectation of  $x^2$  is equal to integral over one to infinity,  $x^2 f(x) dx$

This is equal to two into integral from one to infinity  $x^{r-1} dx$ .

Is equal to log  $x$  ranges from one to infinity which is equal to infinity.

Thus for the above distribution, 1<sup>st</sup> order moment exists but 2<sup>nd</sup> order moment does not exist.

Consider another example.

A random variable  $X$  has probability density function

$f(x)$  is equal to  $r+1$  into  $a^r$  divided by  $x+a^r+2$  where  $x$  is greater than or equal to zero and  $a$  is positive.

Now consider  $r^{\text{th}}$  raw moment,

$\mu_r$  is equal to expectation of  $X^r$  is equal to  $r+1$  into  $a^r$  integral over zero to infinity  $x^r$  divided by  $x+a^r+2$   $dx$

substituting  $x$  is equal to  $a$  into  $y$  in the above integral and using beta integral function

Integral over zero to infinity,  $x^{m+1}$  divided by  $1+x^{m+n}$   $dx$  is equal to beta of  $m, n$ .

On simplification,

$\mu_r$  is equal to  $r+1$  into  $a^r$  into beta of  $r+1$  and one is equal to  $a^r$ , where beta  $m, n$  is equal to  $\Gamma(m)\Gamma(n)$  divided by  $\Gamma(m+n)$ .

However

$\mu_{r+1}$  is equal to expectation of  $X^{r+1}$  is equal to  $r+1$  into  $a^r$  integral over zero to infinity  $x^{r+1}$  divided by  $x+a^r+2$   $dx$  tends to infinity.

As the integral is not convergent. Hence in this case only moments up to  $r^{\text{th}}$  order exist and higher order moments do not exist.

# 3. Relationship between Raw and Central Moments

Now let us discuss the relation between raw moments and central moments.

We have already discussed the topic moments in the first semester. Let us revise the same.

The relation between raw and central moments is given by the expression,

$\mu_r$  is equal to  $\mu_r - r C_1 \mu_r - \frac{r C_2}{2} \mu_1^2 + \frac{r C_3}{6} \mu_1^3 - \frac{r C_4}{24} \mu_1^4 + \dots$  plus minus one power four into  $\mu_1$  dash to the power  $r$ .

In particular on substituting  $r$  is equal to two, three and four in above and simplifying we get

$\mu_2$  is equal to  $\mu_2 - \mu_1^2$

$\mu_3$  is equal to  $\mu_3 - 3 \mu_2 \mu_1 + 2 \mu_1^3$  and

$\mu_4$  is equal to  $\mu_4 - 4 \mu_3 \mu_1 + 6 \mu_2 \mu_1^2 - 3 \mu_1^4$ .

Uses of first four moments

- The first moment about zero is the arithmetic mean
- The second central moment is the variance of the distribution
- The third central moment is a measure of skewness
- The fourth central moment is a measure of kurtosis

The measure of skewness is given by,  $\beta_1$  is equal to  $\mu_3^2$  divided by  $\mu_2^3$  and the measure of kurtosis is given by  $\beta_2$  is equal to  $\mu_4$  divided by  $\mu_2^2$ .

Based on the measures of skewness and kurtosis we can write the nature of the distribution.

As  $\beta_1$  is always positive we use the sign of  $\mu_3$  to decide whether the distribution is positively skewed or negatively skewed.

- If  $\mu_3$  is positive, the distribution is positively skewed
- If  $\mu_3$  is negative, the distribution is negatively skewed and
- If  $\mu_3$  is equal to zero the distribution is symmetric

Similarly  $\beta_2$  gives the “peakedness” or flatness of the distribution. That is

- If  $\beta_2$  is less than three, the distribution is platykurtic,
- If  $\beta_2$  is greater than three the distribution is leptokurtic and
- If  $\beta_2$  is equal to three then the distribution has normal curve

# 4. Quantiles and Median

Now let us discuss about the quantiles.

Quantiles are points taken at regular intervals from the cumulative distribution function of a random variable.

Dividing ordered data into 'q' essentially equal-sized data subsets is the motivation for q-quantiles.

Quantiles are the data values marking the boundaries between consecutive subsets.

Put another way, the  $k^{\text{th}}$  q-quantile for a random variable is the value  $x$  such that the probability that the random variable will be less than  $x$  is at most  $k$  by  $q$  and the probability that the random variable will be more than  $x$  is at most  $(q \text{ minus } k)$  divided by  $q$  is equal to one minus  $(k \text{ by } q)$ . There are  $q \text{ minus one}$ , of the  $q$  -quantiles, one for each integer  $k$  satisfying zero less than  $k$  less than  $q$ .

Some q-quantiles have special names

- The second quantile is called the median
- The three-quantiles are called tertiles or terciles represented by  $T$
- The four-quantiles are called quartiles denoted by  $Q$
- The five-quantiles are called quintiles denoted by  $QU$
- The six-quantiles are called sextiles represented by  $S$
- The 10-quantiles are called deciles denoted by  $D$
- The 12-quantiles are called duo-deciles denoted by  $Dd$
- The 20-quantiles are called vigintiles denoted by  $V$
- The 100-quantiles are called percentiles denoted by  $P$
- The 1000-quantiles are called permiles denoted by  $Pr$

The median, quartiles deciles and percentiles most commonly used quantiles.

### Median

Now let us discuss about the median.

We know that median divides the whole distribution into two equal parts.

Hence if a continuous random variable  $X$  has probability mass function  $f$  of  $x$  and if median of the distribution is taken as  $M$ , then

Integral from minus infinity to  $M$   $f$  of  $x$   $d x$  is equal to integral from  $M$  to infinity  $f$  of  $x$   $d x$  is equal to half.

Thus solving

Integral from minus infinity to  $M$   $f$  of  $x$   $d x$  is equal to half or integral from  $M$  to infinity  $f$  of  $x$   $d x$  is equal to half for  $M$ , we get the value of the median.

### Now let us consider Quartiles

Since quartiles divides the distribution into four equal parts, we have three quartiles,  $q$  one,  $q$  two and  $q$  three and are given by the equations,

Integral from minus infinity to  $Q$  one  $f$  of  $x$   $d x$  is equal to one divided by four,

Integral from minus infinity to  $Q$  two  $f$  of  $x$   $d x$  is equal to two into one divided by four is equal

to half and

Integral from minus infinity to  $Q_3$   $f(x) dx$  is equal to three into one divided by four is equal to three by four.

Observe that  $Q_2$  and Median are same.

# 5. Deciles and Percentiles

Now let us consider Deciles and Percentiles.

As deciles divide the distribution into 10 equal parts, we have 9 deciles and  $i^{\text{th}}$  decile is given by the equation,

$\int_{-\infty}^D I f(x) dx$  is equal to  $I$  divided by 10, where  $I$  is equal to one, two etc till nine.

Similarly percentiles divide the distribution into 100 equal parts, we have 99 percentiles and  $I^{\text{th}}$  percentile is given by the equation,

$\int_{-\infty}^P I f(x) dx$  is equal to  $I$  divided by 100, where  $I$  is equal to one, two etc till 99.

The probability distribution of a random variable  $X$  is  $f(x)$  is equal to  $k \sin\left(\frac{1}{5}\pi x\right)$ , where  $0 \leq x \leq 5$ . Determine the constant  $k$  and obtain the median and quartiles of the distribution.

Solution :

As  $f(x)$  is a probability density function,

$\int_0^5 f(x) dx = 1$  implies  $k \int_0^5 \sin\left(\frac{1}{5}\pi x\right) dx = 1$

Let us substitute;  $\frac{1}{5}\pi x = y$  is equal to  $y$ .

As  $x$  is equal to zero  $y$  is equal to zero and as  $x$  is equal to five,  $y$  is equal to  $\pi$  and  $dx = \frac{5}{\pi} dy$ .

Hence we get,

$k \int_0^{\pi} \sin y \cdot \frac{5}{\pi} dy = 1$

By integrating and simplifying, we get

$k$  is equal to  $\frac{\pi}{10}$

Now let us find the median  $M$  as follows.

We know that  $\int_0^M f(x) dx = \frac{1}{2}$  implies  $k \int_0^M \sin\left(\frac{1}{5}\pi x\right) dx = \frac{1}{2}$

Let us substitute  $\left(\frac{1}{5}\pi x\right) = y$  is equal to  $y$

As  $x$  is equal to zero,  $y$  is also equal to zero and as  $x$  is equal to  $M$ ,  $y$  is equal to  $\left(\frac{1}{5}\pi M\right)$  and  $dx = \left(\frac{5}{\pi}\right) dy$

Implies  $\frac{\pi}{10} \int_0^{\frac{\pi M}{5}} \sin y \cdot \frac{5}{\pi} dy = \frac{1}{2}$

On integration we get,

$\frac{\pi}{10} \left[ -\cos y \right]_0^{\frac{\pi M}{5}} = \frac{1}{2}$

Implies  $\frac{1}{10} (1 - \cos \frac{\pi M}{5}) = \frac{1}{2}$

Implies  $M$  is equal to five divided by two

Hence median of the distribution is five divided by two.

We know that the second quartile is same as median. Hence we find first and third quartiles only.

The first quartile  $Q_1$  is given by

$\int_0^{Q_1} f(x) dx = \frac{1}{4}$

Implies  $k \int_0^{Q_1} \sin\left(\frac{\pi x}{5}\right) dx = \frac{1}{4}$

Considering the same transformation as we have done in the case of median we get,

$\frac{5}{\pi} k \int_0^{Q_1} \sin y dy = \frac{1}{4}$

Implies  $\frac{1}{2} [1 - \cos\left(\frac{\pi Q_1}{5}\right)] = \frac{1}{4}$

Therefore  $\cos\left(\frac{\pi Q_1}{5}\right) = \frac{1}{2}$

Is equal to  $\cos\left(\frac{\pi}{3}\right)$

Implies  $Q_1 = \frac{5}{3}$

Similarly the third quartile  $Q_3$  is given by

$\int_0^{Q_3} f(x) dx = \frac{3}{4}$

Implies  $k \int_0^{Q_3} \sin\left(\frac{\pi x}{5}\right) dx = \frac{3}{4}$

Considering the same transformation as we have done in the case of median we get,

$\frac{5}{\pi} k \int_0^{Q_3} \sin y dy = \frac{3}{4}$

Implies  $\frac{1}{2} [1 - \cos\left(\frac{\pi Q_3}{5}\right)] = \frac{3}{4}$

Therefore  $\cos\left(\frac{\pi Q_3}{5}\right) = -\frac{1}{2}$

Implies  $Q_3 = \frac{10}{3}$

Here's a summary of our learning in this session where we have :

- Understood raw and central moments
- Explained the conditions for existence of moments
- Understood the relationship between raw and central moments
- Described the uses of the moments
- Understood Quantiles especially the median, quartiles, deciles and percentiles