# UNIT 6 MEASURES OF SKEWNESS AND KURTOSIS

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# 6.0 OBJECTIVES

After going through this Unit, you will be able ta:

distinguish between a symmetrical and a skewed distribution;

compute various coefficients to measure the extent of skewness in a distribution;

distinguish between platykurhc, mesokurtic and leptokurtic distributions; and compute the coefficient of kurtosis.

#### 6.1 INTRODUCTION

In this Unit you will learn various techniques to distingush between various shapes of a frequency distribution. This is the final Unit with regard to the summarisation of univariate data. This Unit will make you familiar with the concept of skewness and kurtosis. The need to study these concepts arises from the fact that the measures of central tendency and dispersion fail to describe a distribution completely. It is possible to have fkquency distributions which differ widely in their nature and composition and yet may have same central tendency and dispersion. Thus, there is need to supplement the measures of central tendency and dispersion. Consequently, in th~s Unit, we shall discuss two such measures, viz, measures of skewness and kurtosis.

#### **6.2 CONCEPT OF SKEWNESS**

The skewness of a distribution is defined as the lack of *symmetry*. In a symmetrical distribution, the Mean, Median and Mode are equal to each other and the ordinate at mean divides the distribution into two eual arts such that one

Cnivariate Data part is mirror image of the other (Fig. 6.1). If some observations, of very

high (low) magnitude, are added to such a distribution, its right (left) tail gets elongated.

, Positively Skewed Distribution | Negatively Skewed Distribution

#### Fig. 6.2

These observations are also known as extreme observations. The presence of extreme observations on the right hand side of a distribution makes it positively skewed and the three averages, viz., mean, median and mode, will no longer be equal. We shall in fact have Mean > Median > Mode when a distribution is positively skewed. On the other hand, the presence of extreme observations to the left hand side of a distribution make it negatively skewed and the relationship between mean, median and mode is: Mean < Median < Mode. In Fig. 6.2 we depict the shapes of positively skewed and negatively skewed distributions.

The direction and extent of skewness can be measured in various ways. We shall discuss four measures.@skewness in this Unit.

#### 6.2.1 Karl ~earhn's Measure of Skewness

In Fig. 6.2 you noticed that the mean, median and mode are not equal in a skewed distribution. The Karl Pearson's measure of skewness is based upon the *divergence of mean from mob* in a skewed distribution.

Since Mean = Mode in a symmetrical distribution, (Mean - Mode) can be taken as an *absolute measure of skewness*. The absolute measure of skewness for a distribution depends upon the unit of measurement. For example, if the mean = 2.45 qetre and mode = 2.14 metre, then absolute measure of skewness will be 2.45 aetre - 2.14 metre = 0.31 metre. For the same distribution, if we change the dt of measurement to centimetres, the absolute measure of skewness is 245

centimetre - 2 14 centimetre = 3 1 centimetre. In order to avoid such a problem Measures Skewness and Kurtosis Karl Pearson takes a relative measure of skewness.

A relative measure, independent of the units of measurement, is defined as the Karl Pearson b Coeficient of Skewness Sk, given by

$$S_{,} =$$
Mean - Mode  $S_{,} d$ .

The sign of Sk gives the direction and its magnitude gives the extent of skewness.

If Sk > 0, the distribution is positively skewed, and if  $S_1 < 0$  it is negatively skewed.

So far we have seen that Sk is strategically dependent upon mode. If mode is not defined for a distribution we cannot find Sk. But empirical relation between mean, median and mode states that, for a moderately symmetrical distribution, we have Mean - Mode = 3 (Mean - Median)

Hence Karl Pearson's coefficient of skewness is defined in tms of median as

**Example 6.1:** Compute the Karl Pearson's coefficient of skewness from the following data:

**Table 6.1** 

#### **Height (in inches)**

#### **Number of Persons**

3 5

Table for the computation of mean and s.d.

## Height (X)

9

```
63
64
65
Total
u = X - 61
- 3
- 2
- 1
0
1
2
3
4
No. of persons V)
10
18
30
42
35
28
16
8
187
fi
- 30 -36
- 30 0
3 5
56
48
32
75
      90
fu2
72
30
0
35
112
1 44
```

Summarisation of

Univariate Data Mean 
$$^{75} = 61 + -= 61.4_{187}$$

To find mode, we note that height is a continuous variable. It is assumed that the height has been measured under the approximation that a measurement on height that is, e.g., greater than 58 but less than 58.5 is taken as 58 inches while a measurement greater than or equal to 58.5 but less than 59 is taken as 59 inches. Thus the given data can be written as

#### Height (in inches) No. of persons

57.5 - 58.5

58.5 - 59.5

59.5 - 60.5

60.5 - 61.5

61.5 - 62.5

62.5 - 63.5

63.5 - 64.5

64.5 - 65.5

By inspection, the modal class is 60.5 - 61.5. Thus, we have

12. Mode = 
$$60.5 + 12 + 7 \times 1 = 61.13$$
  
61.4 -  $61.13 - 0.153$ .  
Hence, the Karl Pearson's coeficient of skewness  $sk = -$ 

Thus the distribution is positively skewed.

# 6.2.2 Bowley's Measure of Skewness

This measure is based on quartiles. For a symmetrical distribution, it is seen that Q, and Q3 are equidistant from median. Thus  $(Q3 - Md) - (M, -Q_3)$  can be taken as an absolute measure of skewness.

 ${f A}$  relative measure of skewness, known as Bowley's coefficient (SQ), is given by

The Bowley's coefficient for the data on heights given in Table 6.1 is computed Measures of Skewness Kurtosis below.

	60.5 - 61.5	
Height (in	61.5 - 62.5	Computation of Q, : No. of persons V)
inches) 57.5 -	62.5 - 63.5	
58.5	63.5 - 64.5	10
58.5 - 59.5	64.5 - 65.5	18
59.5 - 60.5		3 0

42	Cumulative Frequency	135
35	10-	163
28	2 8	179
16	5 8	187
8	100	

<sup>N</sup> Since p = 46.75, the first quartile class is 59.5 - 60.5.

Thus  $_{1a}$ , = 59.5, C = 28, fa, = 30 and h = 1.

#### Computation of M, $(Q_i)$ :

<sup>N</sup> Since  $\mathbf{2}$  = 93.5, the median class is 60.5 - 61.5. Thus

Im = 60.5, C = 58, fm = 42 and h = 1.

# Computation of Q,:

<sup>3N</sup> Since = 140.25, the third quartile class is 62.5 - 63.5. Thus  $_{1a}$ , = 62.5, C = 135, fQ= 28 and h = 1.

Hence, Bowley's

 $62.688 - 2 \times 61.345 + 60.125 = 0.048$ .

coefficient 
$$_{SQ} = \frac{62688 - 60.125 \text{ 6.2.3 Kelly's Measure}}{62688 - 60.125 \text{ 6.2.3 Kelly's Measure}}$$

#### of Skewness

Bowley's measure of skewness is based on the middle 50% of the observations because it leaves 25% of the observaticins on each extreme of the distribution. As an improvement over Bowley's measure, Kelly has suggested a measure based on  $P_{,,}$  and,  $P_{,,}$  so that only 10% of the observations on each extreme are ignored.

Univariate Data Kelly's coefficient of skewness, denoted by S, is given by Summarisation of

Note that  $P_{,,} = M$ , (median).

The value of S<sub>1</sub>, for the data given in Table 6.1, can be computed as given below.

## Computation of P, , :

Since = 10N = 10 1s7 = 18.7. 10th percentile lies in the class 58.5 - 59.5. Thus 100 100

#### Computation of P, :

Since 
$$WN^{-}$$
 = 168.3, 90th percentile lies in the class 63.5 - 64.5. Thus 100 100  $_{1pw}$  = 63.5, C = 163,  $_{fpm}$  = 16 and  $_{h}$  = 1.

Hence, Kelly's coefficient

It may be noted here that although the coefficient S,, So and S, are not comparable, however, in the absence of skewness, each of them will be equal to zero.

#### **Check Your Progress 1**

1) Compute the Karl Pearson's coefficient of skewness from the following data:

Daily Expenditure (Rs.): 0-20 20-40 40-60 60-80 80-100 No. of families: 13 25 27 19 16

2) The following figures relate to the size of Measures of Skewness and Kurtosis capital of 285 companies:

Capital (in Ks. lacs.) No. of companies

1-5 610 11-15 1620 21-25 2630 31-35 ibtal 20 27 29 38 48 53 285

interpret the results.

3) The following measures were computed for a frequency distribution

Karl Pearson's Coefficient of Skewness = - 0.25.

Compute Standard Deviation, Mode and Median of the distribution.

## **6.3 MOMENTS**

The rth moment about mean of a distribution, denoted by **p**,, is given

by **P** he - where 
$$r = 0, 1, 2, 3, 4, \dots N_{i=1}$$

Thus, **rth** moment about mean is the mean of the rth power of deviations of observations from their arithmetic mean. In particular,

1" if 
$$r = 0$$
, we have  $PO = -\sim h(xi - X)O = I$ ,  $N_{i=1}$ 

1" if  $r = 1$ , we have  $PIX i - x = 0$ ,  $N_{i=1}$ 

1" if  $r = 2$ , we have  $N_{i=1} - x = 0$ ,  $N_{i=1}$ 

if 
$$r = 3$$
, we have  $P_3 = L$ ?  $h_{(xi)} - x$ ) and so on.

Univariate Data In addition to the above, we can define *raw moments* as moments about any Summarisation of

arbitrary mean.

Let **A** denote an arbitrary mean, then uth moment about A is defined as

When A = 0, we get various moments about origin.

#### **Moment Measure of Skewness**

The moment measure of skewness is based on the property that, for a symmetrical distribution, all odd ordered central moments are equal to zero.

We note that  $\mathbf{p}_1 = 0$ , for every distribution, therefore, the lowest order moment that can provide an absolute measure of skewness 'is  $\mathbf{p3}$ .

Further, a coefficient of skewness, independent of the units of measurement, is given by

$$_{0}$$
 +& = Y  $_{1}$  , where  $p$ , and  $y$ , are defined as the *first beta* and *first*  $a3=-=$ 

**gamma** coefficients respectively. **P,** is measure of kurtosis as you will come to know in the next Section.

CL; Very often, the skewness is measured in terms of  $P_1 = 3$ , where the sign of skewness is determined by the sign of  $p_1$ .

**Example 6.2:** Compute the Moment coefficient of skewness (P,) from the following data.

Marks Obtained : 0-10 10-20 20-30 30-40 40-50 50-60 60-70 Frequency : 6 12 22 24 16 12 8

Table for the computations of mean, s.d. and **p**,.

				X-3 5		
Class	6	15	<b>-</b> 1			
Intervals	12	25	0		54	- 162 -
0 - 10 10		35	1		48	96
20 20 - 30	024	45	2		22	-22
30 - 40	16	55	3		0	0
40 - 50	12	65	fu		16	16
50 - 60	8	u=- 10		16	48	96
60 - 70	_		- 18 - 24	24	72	216
	100		- 22 0	24		48
Total	Mid		22 0		260	40
Frequenc	values (X)	<b>-</b> 3		0	fu3	
y (f)	5	<b>-</b> 2		fu2		

Since Xfu = 0, the mean of the distribution is 35.

The second moment p, is equal to the variance (oZ) and its positive square root is equal to

$$^{260}$$
 **p2** =-~100=260, and

$$= -Eh[(xi_{-A)-piIr} 1]_{N \text{ (Since } pi}$$

**=GZh(Xi-~)=z-~)** Expanding the term within brackets by binomial theorem, we get

Since the sign of **p3** is positive and **p**, is small, the distribution is slightly positively skewed.

If the mean of a distribution is not a convenient figure like 35, as in the above example, the computation of various central moments may become a cumbersome task. Alternatively, we can From the above, we can write first compute raw moments and then convert them into central moments by using the equations obtained below.

$$pr = -fCIp \sim -lpi + r \sim 2p:-2pi2 - rC3p;-3p1 + """.$$

Conversion of Raw Moments, into Central Moments

In particular, taking r = 2, 3, 4, etc., we get  $p2 = p; -2 \sim lp; 2 + 2 \sim 2p; p \sim 2 = pi - pi2 \text{ (since } p; = 1)$ 3lcasures of Skewness and Kurtosis

We can write

Univariate Data Example 6.3: Compute the first four moments about mean from the following Summarisation of

data.

ClassIntervals: 0-10 10-20 20-30 30-40 Frequency V): 1 **3 4** 2

Table for computations of raw moments (Take A = 25).

Class Intervals f Mid-Value 
$$u = \frac{10}{10}$$
 fu fu2 fu3 fu4

**0 - 10** 1 5 **-** 2 **- 2 4 -8 16 10 - 20 3** 15 **-** 1 **- 3 3 -3 3 20 - 30 4** 25 **0 0 0 0 0 30 - 40 2 3 5 1** 2 2 2 **2** 

Total 10 -3- 9 -9 21

From the above table, we can write

-9x lo3pi= I(, = 900 and

#### **Moments about Mean**

By definition,

# **Check Your Progress 2**

1) Calculate the first four moments about mean for the following distribution. Also calculate **9**, and comment upon the nature of skewness.

Marks: 0 - 20 20 - 40 40 - 60 60 - 80 80 - 100 '

Frequency: 8 28 35 17 12

Measures of Skewness and Kurtosis

2) The first three moment of a distribution about the value 3 of a variable are **2,10** and 30respectively. Obtain **F, p2, p3** and hence **P,.** Comment upon the nature of skewness.

#### 6.4 CONCEPT AND MEASURE OF KURTOSIS

Kurtosis is another measure of the shape of a distribution. Whereas skewness measures the lack of symmetry of the frequency curve of a distribution, kurtosis is a measure of the relative peakedness of its fi-equency curve. Various frequency curves can be divided into three categories depending upon the shape of their peak. The three shapes are termed as Leptokurtic, Mesokurtic and Platykurhc as shown in Fig. 6.3.

Leptokurtic

Mesokurtic

Platykurtic

Fig. 6.3

Summarisation of

Univarinte Data P 4 A measure of kurtosis is given by P 2 = 2, a coefficient given by Karl Pearson. P 2

The value of **p2** = 3 for a mesokurtic curve. When **P2** > 3, the curvt: is more peaked than the mesokurtic curve and is tenned as leptokurtic. Similarly, when **p2** < 3, the curve is less peaked than the mesokurtic curve and is called as platykurtic curve.

**Example 6.4:** The first four central moments of a distribution are 0,2.5,0.7 and 18.75. Examine the skewness and kurtosis of the distribution.

To examine skewness, we compute **p**,.

P4 18.75  $\overline{\phantom{a}}$  3.0, Kurtosis is given by the coefficient **P2** = ---i- =  $\overline{\phantom{a}}$ 

P2 (q2

Hence the curve is mesokurtic.

## **Check Your Progress 3**

1) Compute the first four central moments hm the following data. **Also** find the two beta coefficients.

Value 5 10 15 20 25 30 35

Frequency: 8 15 20 32 23 17 5

2) The first four moments of a distribution are 1,4, 10 and 46 respectively. Compute the moment coefficients of skewness and kurtosis and comment upon the nature of the distribution.

# **6.5 LET US SUM UP** Measures of Skewness and Kurtosis

In this Unit you have learned about the measures of skewness and kurtosis. These two concepts are used to get an idea about the shape of the fiequency curve of a distribution. Skewness is a measure of the lack of symmetry whereas kurtosis is a measure of the relative peakedness of the top of a fiequency curve.

#### 6.6 KEY WORDS

**Skewness:** Departure from symmetry is skewness.

**Moment of Order r:** It is defined as the arithmetic mean of the **rth** power of deviations of observations.

**Coefficient of Kurtosis:** It is a measure of the relative peakedness of the top of a frequency curve.

#### 6.7 SOME USEPUL BOOKS

Elhance, D. N. and V. lhance, 1988, *Fundamentals of Statistics*, Kitab Mahal, Allahabad.

Nagar, A. L. and R. K. Dass, 1983, *Basic Statistics*, Oxford University Press, Delhi.

Mansfield, E., 199 1, *Statistics for Business and Economics: Methods and Applications, W.W.* Norton and Co.

Yule, G U. and M. G Kendall, 1991, *An Introduction to the Theor of Statistics*, Universal Books, Delhi.

# 6.8 ANSWERS OR HINTS TO CHECK YOUR PROGRESS EXERCISES

#### **Check Your Progress 1**

- 1) 0.237
- 2) 0.12, -0.243
- 3) 17.5, 54.38, 51.46

### **Check Your Progress 2**

- 1) 0,499.64, 2579.57, 5891 11.61, 0.053, skewness is positive.
- 2) 5, 6, -14,0.907, since **p3** is negative the distribution is negatively skewed.

# **Check Your Progress 3**

- 1) 0,59.99, 50.18, 8356.64,0.012 (negatively skewed), 2.32 (platykurtic).
- 2) 0,3. Thus the distribution is symmetrical and mesokurtic. Such a distribution is also known as a Normal Distribution.