

Quartiles

Quartiles are merely particular percentiles that divide the data into quarters, namely:

Q_1 = 1st quartile = 25th percentile (P_{25})

Q_2 = 2nd quartile = 50th percentile
= median (P_{50})

Q_3 = 3rd quartile = 75th percentile (P_{75})

Quartile Example

Using the applicant (aptitude) data,
the first quartile is:

$$n \cdot \frac{P}{100} = (50)(.25) = 12.5$$

Rounded up Q_1 = 13th ordered value = 46

Similarly the third quartile is:

$$n \cdot \frac{P}{100} = (50)(.75) = 37.5 \quad 38 \text{ and } Q_3 = 75$$

Interquartile Range

The **interquartile range** (IQR) is essentially the middle 50% of the data set

$$IQR = Q_3 - Q_1$$

Using the applicant data, the IQR is:

$$IQR = 75 - 46 = 29$$

Z-Scores

- **Z-score** determines the *relative position* of any particular data value x and is **based** on the *mean* and *standard deviation* of the data set
- The Z-score is expressed as the number of standard deviations the value x is from the mean
- A negative Z-score implies that x is to the left of the mean and a positive Z-score implies that x is to the right of the mean

Z Score Equation

$$z = \frac{x - \bar{x}}{s}$$

For a score of 83 from the aptitude data set,

$$z = \frac{83 - 60.66}{18.61} = 1.22$$

For a score of 35 from the aptitude data set,

$$z = \frac{35 - 60.66}{18.61} = -1.36$$

Standardizing Sample Data

The process of subtracting the mean and dividing by the standard deviation is referred to as **standardizing the sample data**.

The corresponding z-score is the standardized score.

Measures of Shape

□ **Skewness**

- Skewness measures the tendency of a distribution to stretch out in a particular direction

□ **Kurtosis**

- Kurtosis measures the peakedness of the distribution

Skewness

- In a symmetrical distribution the mean, median, and mode would all be the same value and $Sk = 0$
- A positive Sk number implies a shape which is skewed right and the
mode < median < mean
- In a data set with a negative Sk value the
mean < median < mode

Skewness Calculation

Pearsonian coefficient of skewness

$$Sk = \frac{3(\bar{x} - Md)}{s}$$

Values of Sk will always fall between -3 and 3

Histogram of Symmetric Data

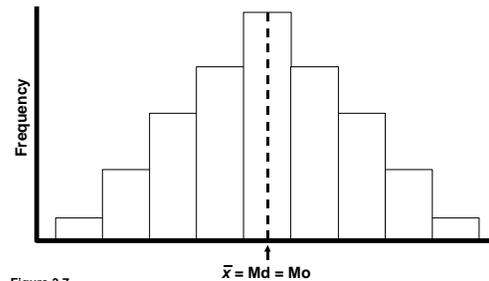


Figure 3.7

Histogram with Right (Positive) Skew

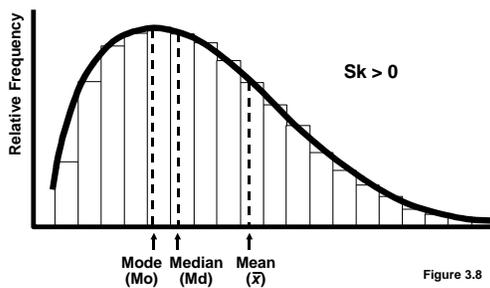


Figure 3.8

Histogram with Left (Negative) Skew

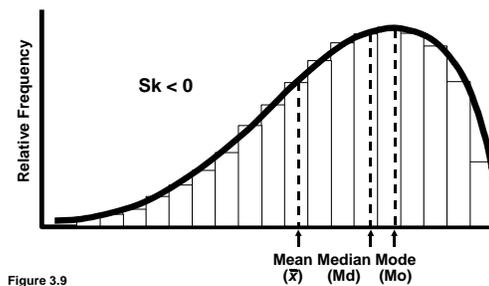


Figure 3.9

Kurtosis

- Kurtosis is a measure of the peakedness of a distribution
- Large values occur when there is a high frequency of data near the mean and in the tails
- The calculation is cumbersome and the measure is used infrequently

Chebyshev's Inequality

1. At least 75% of the data values are between $\bar{x} - 2s$ and $\bar{x} + 2s$, or
At least 75% of the data values have a z-score value between -2 and 2
2. At least 89% of the data values are between $\bar{x} - 3s$ and $\bar{x} + 3s$, or
At least 75% of the data values have a z-score value between -3 and 3
3. In general, at least $(1-1/k^2) \times 100\%$ of the data values lie between $\bar{x} - ks$ and $\bar{x} + ks$ for any $k > 1$

Empirical Rule

Under the assumption of a bell shaped population:

1. Approximately 68% of the data values lie between $\bar{x} - s$ and $\bar{x} + s$ (have z-scores between -1 and 1)
2. Approximately 95% of the data values lie between $\bar{x} - 2s$ and $\bar{x} + 2s$ (have z-scores between -2 and 2)
3. Approximately 99.7% of the data values lie between $\bar{x} - 3s$ and $\bar{x} + 3s$ (have z-scores between -3 and 3)

A Bell-Shaped (Normal) Population

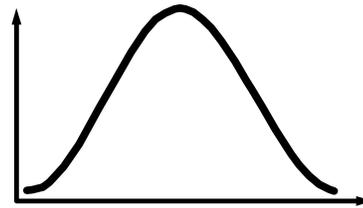


Figure 3.10

Chebyshev's Versus Empirical

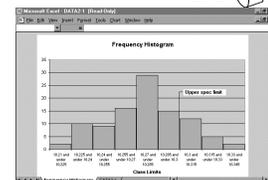
Between	Actual Percentage	Chebyshev's Inequality Percentage	Empirical Rule Percentage
$\bar{x} - s$ and $\bar{x} + s$	66% (33 out of 50)	—	68%
$\bar{x} - 2s$ and $\bar{x} + 2s$	98% (49 out of 50)	75%	95%
$\bar{x} - 3s$ and $\bar{x} + 3s$	100% (50 out of 50)	89%	100%

**$Md = 62$
 $Sk = -.26$**

Table 3.3

Allied Manufacturing Example

Is the Empirical Rule applicable to this data?
Probably yes.
Histogram is approximately bell shaped.



$\bar{x} - 2s = 10.275$ and $\bar{x} + 2s = 10.3284$

96 of the 100 data values fall between these limits closely approximating the 95% called for by the Empirical Rule