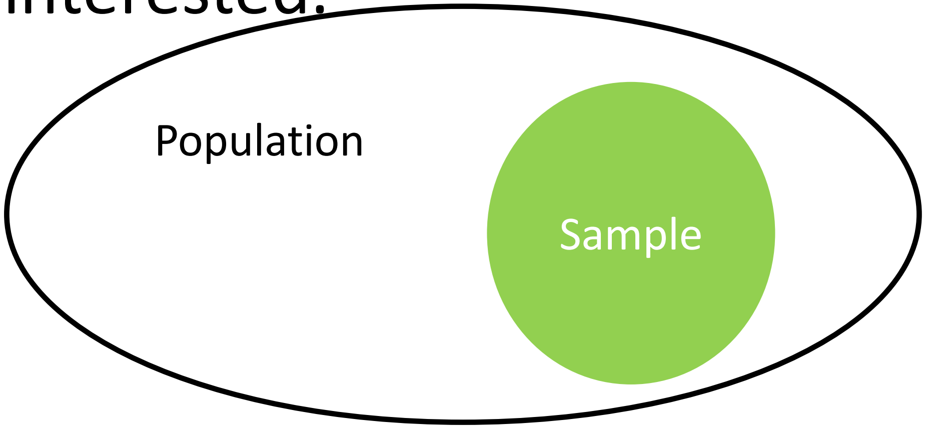
Statistical concepts and market returns

Populations and samples

- The subset of data used in statistical inference is known as a **sample** and the larger body of data is known as the **population.**
  - The **population** is defined as all members of the group in which we are interested.



Parameters and Sample Statistics

A population has parameters, and a sample has statistics.

- Descriptive statistics that characterize population values are called **parameters**.
  - Examples: mean, median, mode, variance, skewness, kurtosis
- Descriptive statistics that characterize samples are known as **sample statistics**.
  - Examples: sample mean, sample median, sample variance
- By convention, we often omit the term "sample" in front of sample statistics, a practice that can lead to confusion when discussing both the sample and the population.

Measurement Scales

Statistical inference is affected by the type of data we are trying to analyze.

- **Nominal scales** categorize data but do not rank them.
  - Examples: fund style, country of origin, manager gender
- **Ordinal scales** sort data into categories that are ordered with respect to the characteristic along which the scale is measured.
  - Examples: “star” rankings, class rank, credit rating
- **Interval scales** provide both the relative position (rank) and assurance that the differences between scale values are equal.
  - Example: temperature
- **Ratio scales** have all the characteristics of interval scales and a zero point at the origin.
  - Examples: rates of return, corporate profits, bond maturity

Weak Scales

Strong Scales

Identifying Scales of Measurement

- **State the scale of measurement for each of the following:**
- Credit ratings for bond issues
- Cash dividends per share
- Hedge fund classification types
- Bond maturity in years

Holding period returns

Holding period returns are a fundamental building block of the statistical analysis of investments.

- Holding period returns (HPR) are calculated as the price at the end of the period plus any cash distribution during the period minus the beginning of period price, all divided by the beginning period price.

$$R_t = \frac{P_t - P_{t-1} + D_t}{P_{t-1}}$$

- For this stock, which is nondividend paying, the HPRs are:

Time	Price	HPR	Time	Price	HPR
0	27.00	—	7	25.90	2.38%
1	25.77	–4.57%	8	27.01	4.28%
2	24.73	–4.04%	9	28.20	4.42%
3	24.32	–1.64%	10	29.52	4.68%
4	24.39	0.28%	11	31.63	7.16%
5	24.71	1.34%	12	35.25	11.43%
6	25.30	2.35%			

Frequency distributions

A tabular display of data summarized into intervals is known as a frequency distribution.

Constructing a frequency distribution:

1. Sort the data in ascending order.
2. Calculate the range of the data, defined as

   Range = Maximum value – Minimum value.
3. Decide on the number of intervals in the frequency distribution, *k*.
4. Determine interval width as Range/*k*.
5. Determine the intervals by successively adding the interval width to the minimum value to determine the ending points of intervals, stopping after reaching an interval that includes the maximum value.
6. Count the number of observations falling in each interval.
7. Construct a table of the intervals listed from smallest to largest that shows the number of observations falling in each interval.

Frequency Distributions

Focus on: Holding Period Returns

- Suppose we have 12 holding period return observations from a non-dividend-paying stock, sorted in ascending order:

  −4.57, −4.04, −1.64, 0.28, 1.34, 2.35, 2.38, 4.28, 4.42, 4.68, 7.16, and 11.43.
- Using k = 4, we have intervals with width of 4.
- The resulting frequency distribution is

Interval	Absolute Frequency
$-4.57 \le$ observation < -0.57	3
$-0.57 \le$ observation < 3.43	4
$3.43 \le$ observation < 7.43	4
$7.43 \le$ observation ≤ 11.43	1

Relative and cumulative frequency

Focus on: Holding Period Returns

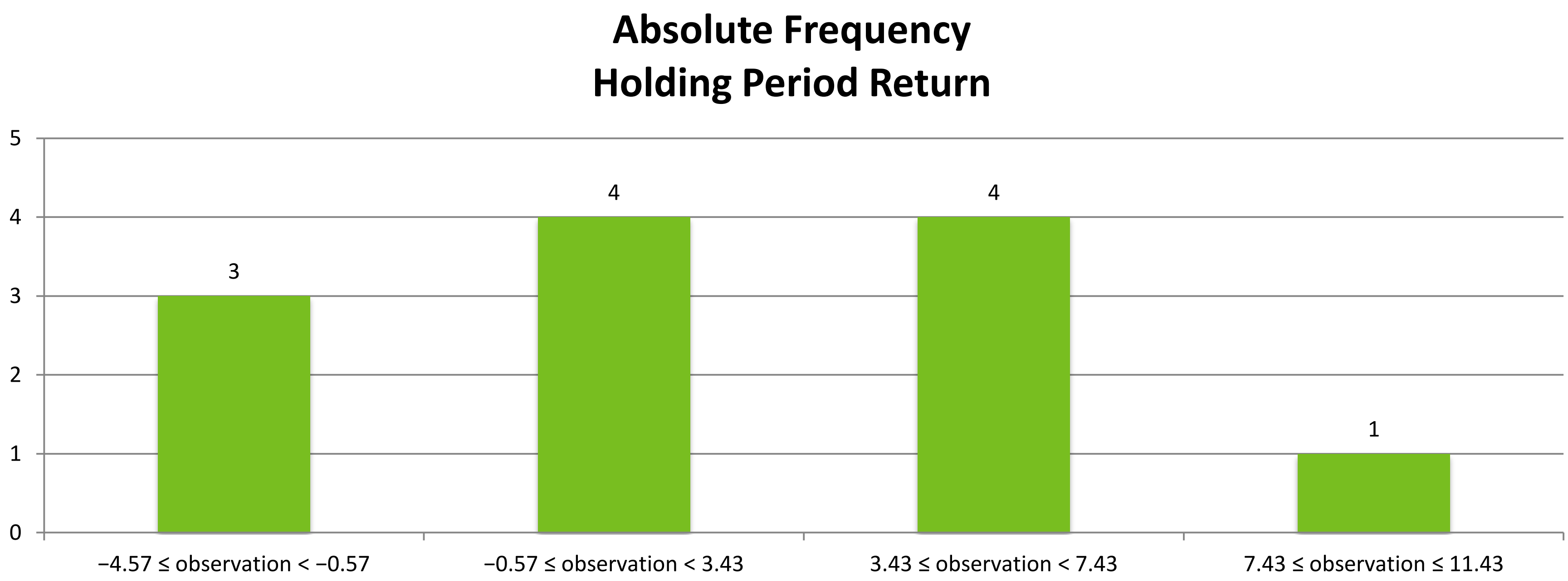
- **Relative frequency** is the absolute frequency divided by the total number of observations.
- **Cumulative (relative) frequency** is the relative frequency of all observations occurring before a given interval.

Interval	Absolute Frequency	Relative Frequency	Cumulative Frequency
$-4.57 \leq$ observation < -0.57	3 $\xrightarrow{\div 12}$	0.250	0.250
$-0.57 \leq$ observation < 3.43	4	0.333	0.583
$3.43 \leq$ observation < 7.43	4	0.333	0.917
$7.43 \leq$ observation ≤ 11.43	1	0.083	1.000

Histograms

Focus on: Holding Period Returns

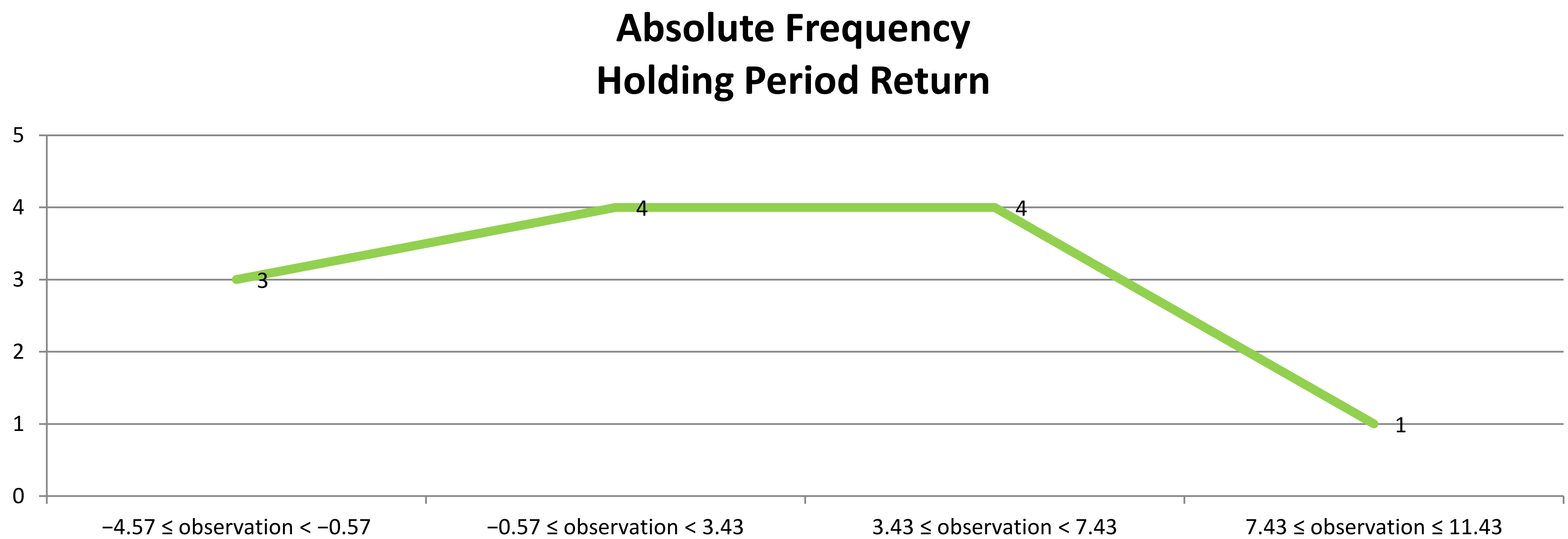
- Histograms are the graphical representation of a frequency distribution.



Frequency Polygon

Focus on: Holding Period Returns

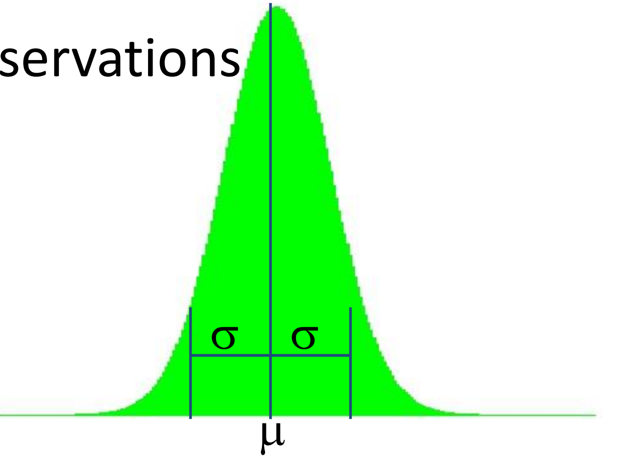
- Frequency polygons are often used to provide higher visual continuity than histograms.



Measures of central tendency

These measures describe where the data are centered.

- Arithmetic Mean
  - The **arithmetic mean** is the sum of the observations divided by the number of observations.
    - **Population mean →** $\mu = \frac{\sum_{i=1}^{N} X_i}{N}$
    - **Sample mean →** $\bar{X} = \frac{\sum_{i=1}^{N} X_i}{N}$
    - The sample mean is often interpreted as center of gravity, for a given set of data.
    - **Cross-sectional data** occur across different observation types at one point in time, and **time-series data** occur for the same unit of observation across time.



Measures of central tendency

Focus on: Cross-Sectional Sample Mean Return

Country	Return	Country	Return
Austria	–2.97%	Italy	–23.64%
Belgium	–29.71%	Netherlands	–34.27%
Denmark	–29.67%	Norway	–29.73%
Finland	–41.65%	Portugal	–28.29%
France	–33.99%	Spain	–29.47%
Germany	–44.05%	Sweden	–43.07%
Greece	–39.06%	Switzerland	–25.84%
Ireland	–38.97%	United Kingdom	–25.66%

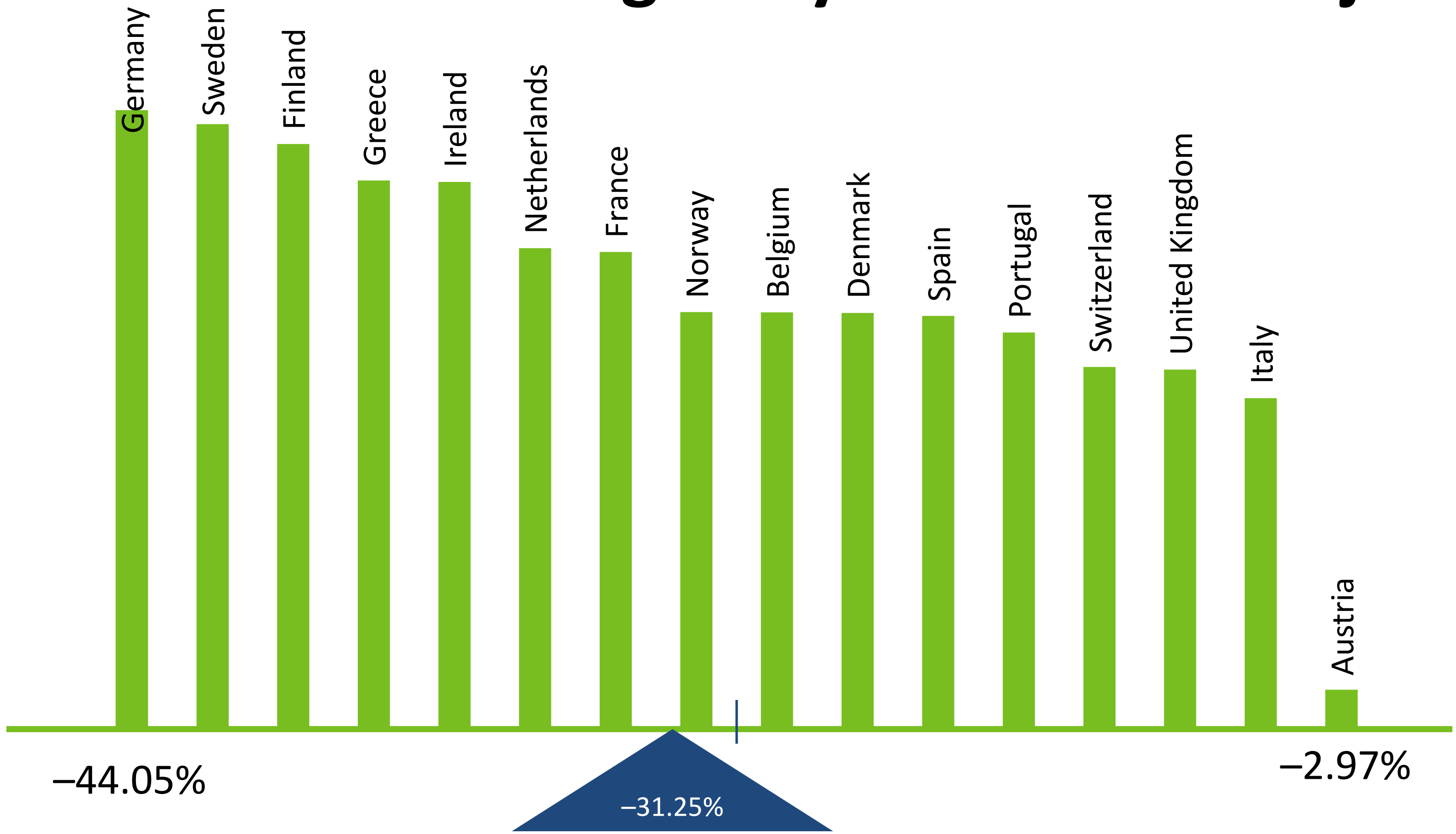
Source: www.msci.com.

$$\bar{X} = \frac{\sum_{i=1}^{N} X_i}{N}$$

$$\bar{X} = \frac{-500.04}{16} = -31.25\%$$

Measures of central tendency

Mean as a center of gravity for the data object



Measures of central tendency

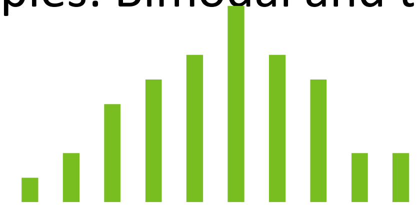
These measures also describe where the data are centered.

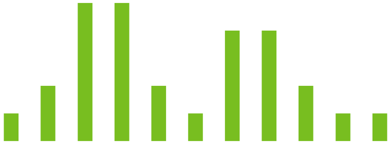
- Weighted Mean → $\bar{X}_W = \sum_{i=1}^{n} w_i X_i$
  - The sum of the observations times each observation's weight (proportional representation in the sample), where the weight is chosen to meet a statistical or financial goal. Example: Portfolio return
- Geometric Mean → $G = \sqrt[n]{\prod_{i=1}^{n} X_i}$
  - Represents the growth rate or compounded return on an investment when *X* is 1 + *R*
- Harmonic Mean → $\bar{X}_H = n \Big/ \sum_{i=1}^{n} 1/X_i$
  - A weighted mean in which each observation's weight is inversely proportional to its magnitude. Example: Cost averaging

Measures of central tendency

These measures also describe where the data are centered.

- The **median** is the middle observation by rank.
  - When we have an odd number of observations, the median will be the closest to the middle. When we have an even number, the median will be the average of the two middle values.
- The **mode** is the most frequently occurring value in a distribution.
  - Distributions are unimodal when there is a single most frequently occurring value and multimodal if there is more than one frequently occurring value.
  - Examples: Bimodal and trimodal


Unimodal


Bimodal

Measures of central tendency

Focus on: Calculating a Median or Mode

$$\text{Median} = \frac{-29.73\% + (-29.71\%)}{2} = -29.72\%$$

Rank	Country	Return	Rank	Country	Return
1	Germany	–44.05%	**9**	**Belgium**	**–29.71%**
2	Sweden	–43.07%	10	Denmark	–29.67%
3	Finland	–41.65%	11	Spain	–29.47%
4	Greece	–39.06%	12	Portugal	–28.29%
5	Ireland	–38.97%	13	Switzerland	–25.84%
6	Netherlands	–34.27%	14	United Kingdom	–25.66%
7	France	–33.99%	15	Italy	–23.64%
8	**Norway**	**–29.73%**	16	Austria	–2.97%

Median: The Case of the Price–Earnings Ratio

ble 10. P/Es for a Client Portfolio

Stock	Consensus Current EPS	Consensus Current P/E
Caterpillar, Inc.	6.34	13.15
Ford Motor Company	1.55	10.97
General Dynamics	6.96	12.15
Green Mountain Coffee Roasters	3.25	25.27
McDonald's Corporation	5.61	17.16
Qlik Technologies	0.17	204.82
Questcor Pharmaceuticals	4.79	13.94

e: Consensus current P/E was calculated as price as of 9 September 2013 divided by consensus EPS as of the same date.

Weighted average

Also known as a weighted mean, the most common application of this measure in investments is the weighted mean return to a portfolio.

- Consider again the country-level data. You have constructed a portfolio that has 50% of its weight in Portugal, Ireland, Greece, and Spain and 50% of its weight in Germany and the UK. Each of the first four countries is equally weighted within the 50%, as are Germany and the UK within their 50%. What is the weighted average return to the portfolio?

$$\bar{X}_W = \sum_{i=1}^{n} w_i X_i$$

Country	Weight	Return	Component Return
Portugal	12.50%	–28.29%	–3.54%
Ireland	12.50%	–23.64%	–2.96%
Greece	12.50%	–39.06%	–4.88%
Spain	12.50%	–29.47%	–3.68%
Germany	25.00%	–44.05%	–1.01%
UK	25.00%	–25.66%	–6.42%
Sum	100%	Weighted Mean =	–32.49%

Measures of dispersion

Dispersion measures variability around a measure of central tendency. If mean return represents reward, then dispersion represents risk.

- Mean Absolute Deviation (MAD) → $\mathrm{MAD} = \frac{\sum_{i=1}^{n} |X_i - \bar{X}|}{n}$
  – The arithmetic average of the absolute value of deviations from the mean.

Measures of dispersion

Dispersion measures variability around a measure of central tendency. If mean return represents reward, then dispersion represents risk.

- Variance is the average squared deviation from the mean.
  - Population variance → $\sigma^2 = \frac{\sum_{i=1}^{n}(X_i - \mu)^2}{n}$
  - Sample variance → $s^2 = \frac{\sum_{i=1}^{n}(X_i - \bar{X})^2}{n-1}$
- Sample variance is "penalized" by dividing by *n* – 1 instead of *n* to account for the fact that the measure of central tendency used, $\bar{X}$, is an estimate of the true population parameter, μ, and so has some uncertainty associated with it.
- Standard deviation is the square root of variance.

Measures of dispersion

Focus on: Sample Standard Deviation

Country	Return	Squared Deviation from Mean
Germany	–44.05%	0.016384
Sweden	–43.07%	0.013971
Finland	–41.65%	0.010816
Greece	–39.06%	0.00610
…	…	…
Austria	–2.97%	0.0780
	Sum=	0.1486
	s^2=	0.0099
	s =	9.95%

Semivariance

We are often concerned with measures of risk that focus on the "downside" of the possible outcomes—in other words, the losses.

- Semivariance is the average squared deviation below the mean.
  - Semideviation is the square root of semivariance.
  - Both are a measure of dispersion focusing only on those observations below the mean.

$$\sum_{for\ all\ Xi<\bar{X}} \frac{(X_i - \bar{X})^2}{n^* - 1}$$

  - Target semivariance, by analogy, is the average squared deviation below some specified target rate, *B*, and represents the "downside" risk of being below the target, *B.*

$$\sum_{for\ all\ Xi<B} \frac{(X_i - B)^2}{n^* - 1}$$

Normal Distribution Function

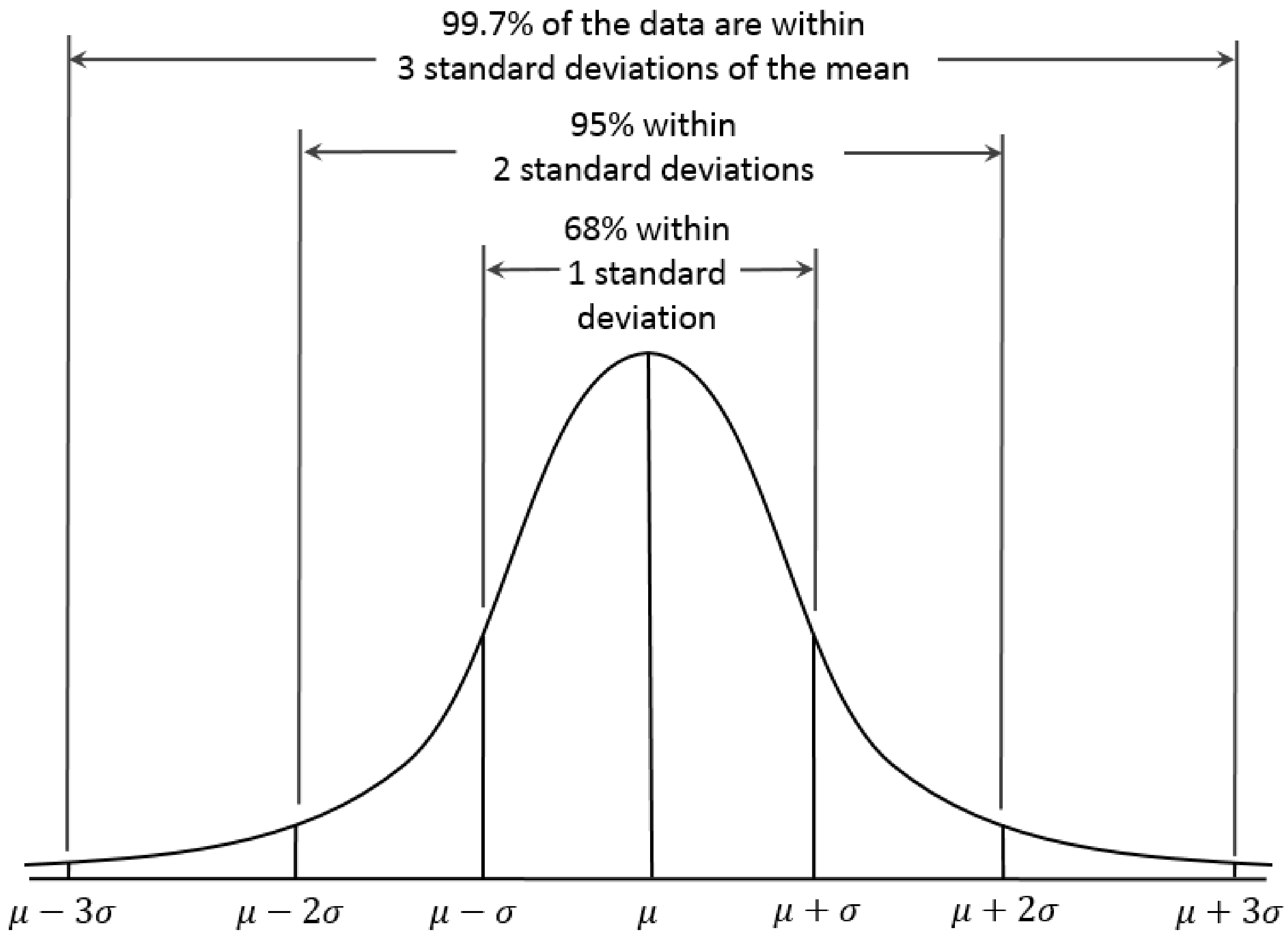
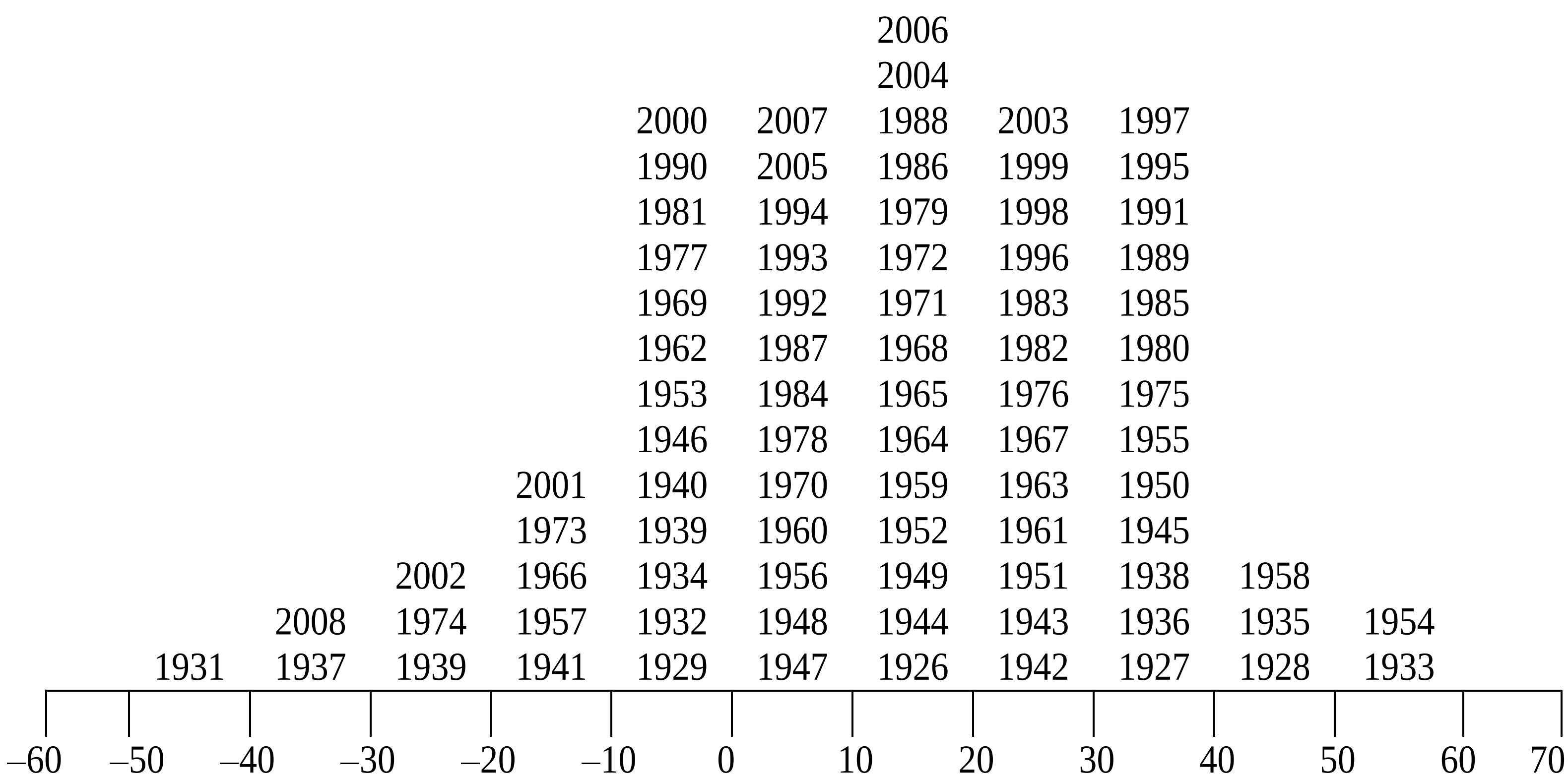


EXHIBIT 5-9 Histogram of U.S. Large Company Stock Returns, 1926-2008



Chebyshev's inequality

This expression gives the minimum proportion of values, *p*, within *k* standard deviations of the mean for any distribution whenever *k* > 1.

$$p \geq 1 - \frac{1}{k^2}$$

k	Interval around the Mean	*p*
1.25	$\bar{X} \pm 1.25s$	0.36
1.50	$\bar{X} \pm 1.50s$	0.56
2.00	$\bar{X} \pm 2.00s$	0.75
2.50	$\bar{X} \pm 2.50s$	0.84
3.00	$\bar{X} \pm 3.00s$	0.89
4.00	$\bar{X} \pm 4.00s$	0.94

Chebyshev's inequality

Focus on: Calculating Proportions Using Chebyshev's Inequality

- For our country data, the mean is –31.25% and the sample standard deviation is 9.95%.
- Lower cutoff at 1.25 standard deviations:

  –31.25% – 1.25 (9.95%) = – 43.6875%
- Upper cutoff at 1.25 standard deviations:

  –31.25% + 1.25 (9.95%) = – 18.8125%

k	Lower Cutoff	Upper Cutoff	Actual *p*	Chebyshev's *p*
1.25	–43.69%	–18.81%	0.875	0.36
1.50	–46.18%	–16.32%	0.938	0.56
2.00	–51.16%	–11.34%	0.95	0.75
2.50	–56.13%	–6.37%	0.97	0.84
3.00	–61.11%	–1.39%	0.997	0.89
4.00	–71.07%	8.57%	1.000	0.94

Chebyshev’s inequality

Applying Chebyshev’s Inequality

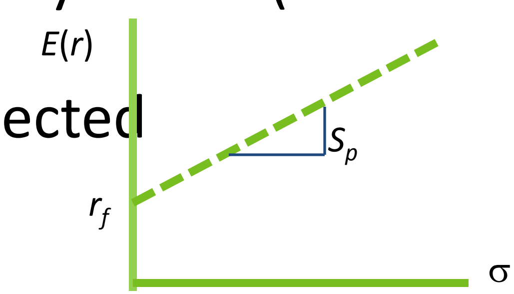
According to Table 22, the arithmetic mean monthly return and standard deviation of monthly returns on the S&P 500 were 0.94 percent and 5.50 percent, respectively, during the 1926–2012 period, totaling 1,044 monthly observations. Using this information, address the following:

1. Calculate the endpoints of the interval that must contain at least 75 percent of monthly returns according to Chebyshev’s inequality.

Combining risk and return

Measures of relative dispersion are used to compare risk and return across differing sets of observations.

- The **coefficient of variation** is the ratio of the standard deviation of a set of observations to their mean value.
  - This ratio can be thought of as the units of risk per unit of mean return.
- The **Sharpe Ratio** is the ratio of the mean excess return (mean return minus the mean risk-free rate) per unit of standard deviation.
  - This ratio can be thought of as units of risky return (excess return) per unit of risk.
  - This will also be the slope of a line in expected return/standard deviation space.



Combining risk and return

Focus on: Coefficient of Variation and the Sharpe Ratio

- Consider a portfolio with a mean return of 25.26% and a standard deviation of returns of 9.95%.

$$CV = \frac{s}{\bar{X}} = \frac{9.95\%}{25.26\%} = 0.3939$$

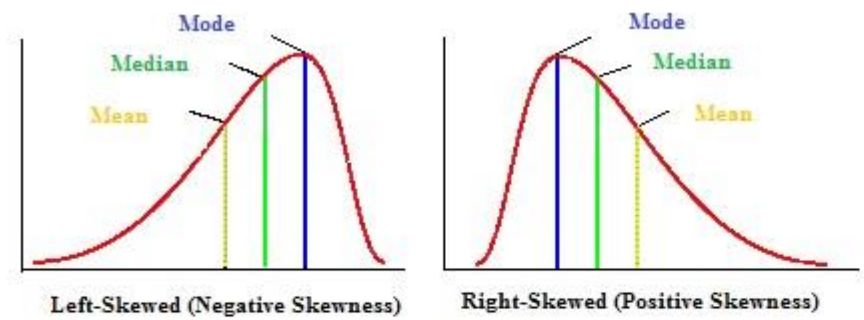
  - The coefficient of variation is

$$S_p = \frac{\bar{R}_p - \bar{R}_f}{s_p} = \frac{25.26\% - 3\%}{9.95\%} = 2.2372$$

  - If the risk-free rate is 3%, then the Sharpe Ratio is

Combining centrality, dispersion, and symmetry

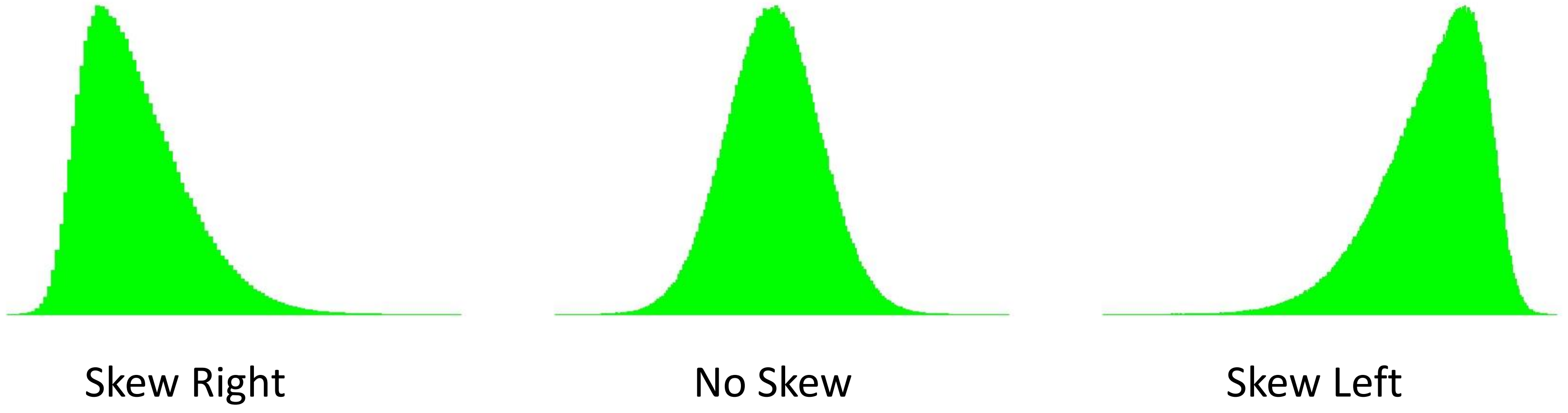
- For a symmetrical distribution, the mean, median, and mode (if it exists) will all be at the same location.
- If the distribution is positively skewed, then the mean will be greater than the median, which will be greater than the mode (if it exists).
- If the distribution is negatively skewed, then the mean will be less than the median, which will be less than the mode (if it exists).



Skewness

The degree of symmetry in the dispersion of values around the mean is known as skewness.

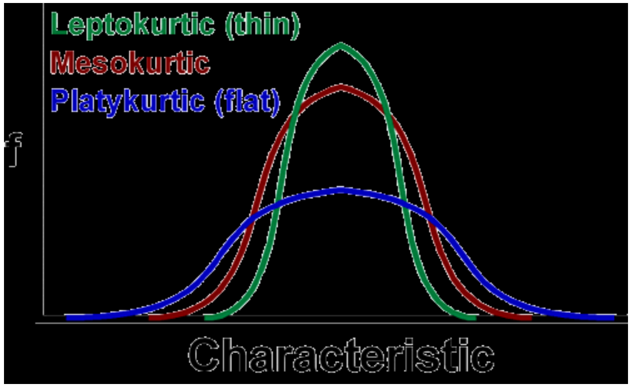
- If observations are equally dispersed around the mean, the distribution is said to be symmetrical.
- If the distribution has a long tail on one side and a “fatter” distribution on the other side, it is said to be skewed in the direction of the long tail.



Kurtosis

- Kurtosis measures the relative amount of “peakedness” as compared with the normal distribution, which has a kurtosis of 3.
  - We typically express this measure in terms of excess kurtosis being the observed kurtosis minus 3.
  - Distributions are referred to as being

1. Leptokurtic (more peaked than the normal; fatter tails)
2. Platykurtic (less peaked than the normal; thinner tails) or
3. Mesokurtic (equivalent to the normal).



Summary

- The underlying foundation of statistically based quantitative analysis lies with the concepts of a sample versus a population.
  - We use sample statistics to describe the sample and to infer information about its associated population.
  - Descriptive statistics for samples and populations include measures of centrality, location, and dispersion, such as mean, range, and variance, respectively.
  - We can combine traditional measures of return (such as mean) and risk (such as standard deviation) to measure the combined effects of risk and return using the coefficient of variation and the Sharpe Ratio.
- The normal distribution is of central importance in investments, and as a result, we often compare statistical properties, such as skewness and kurtosis, with those of the normal distribution.