

## LM03 · QUANTITATIVE METHODS · CFA LEVEL I

# Statistical Measures of Asset Returns

Central tendency, dispersion, skewness, kurtosis, correlation

## Why Statistics for Returns?

Returns distributions are rarely normal. We describe them using **central tendency**, **dispersion**, **shape** (skew/kurtosis), and **co-movement** (covariance/correlation). Each LOS corresponds to a different tool used by analysts to quantify risk and reward.

## Learning Outcome Statements

- LOS a** Calculate, interpret, and evaluate measures of central tendency and location to address an investment problem
- LOS b** Calculate, interpret, and evaluate measures of dispersion to address an investment problem
- LOS c** Interpret and evaluate measures of skewness and kurtosis to address an investment problem
- LOS d** Interpret correlation between two variables to address an investment problem

**Exam Tip:** Know the BAI Plus 1-V statistics keystrokes cold. Given a list of returns, you should be able to retrieve  $\bar{X}$ ,  $S_x$  (sample SD),  $\sigma_x$  (pop SD), and  $n$  in seconds.

$$\text{Arithmetic mean } \bar{X} = \frac{\sum X_i}{n} \quad \cdot \quad \text{Population } \mu = \frac{\sum X_i}{N}$$

### WORKED EXAMPLE Mean of 3-Year Returns

Returns over 3 yrs: 10%, -5%, 20%.

Arithmetic mean =  $(10 - 5 + 20)/3 = 8.33\%$

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**GIVEN**

(A) {1,2,3,4,5} (B) {1,2,3,4,5,6}

**SOLUTION**

(A) Median = 3. (B) Median =  $(3+4)/2 = 3.5$ .

## LOS A · MEASURES OF LOCATION

## Quantiles and Box-and-Whisker Plots

### Quantile Family

**Quartiles** 25%, 50%, 75% → split into quarters

**Quintiles** 20%, 40%, 60%, 80% → fifths

**Deciles** 10%, 20%, ..., 90% → tenths

**Percentiles** 1%, 2%, ..., 99% → hundredths

$$Position L_y = (n + 1) \cdot y/100 \text{ — where } y = \text{percentile}, n = \text{number of obs}$$

#### WORKED EXAMPLE

#### First Quartile (25th Percentile)

##### GIVEN

Returns sorted: 3%, 4%, 6%, 9%, 11%, 12%, 14% ( $n = 7$ ).

##### SOLUTION

$L_{25} = (7+1)(25)/100 = 2. \Rightarrow$  2nd item = **4%**. 25% of observations lie at or below this.

### Box-and-Whisker Plot

- Box = interquartile range ( $IQR = Q3 - Q1$ )
- Line inside box = median
- Upper fence =  $Q3 + 1.5 \times IQR$ ; Lower fence =  $Q1 - 1.5 \times IQR$
- Points outside fences = outliers

### Handling Outliers

#### Do Nothing

Values legit → eliminates judgment bias.

#### Delete

**Trimmed mean** — remove extreme x%.

#### Replace

**Winsorized mean** — cap at percentile bounds.

Use cases: rank portfolios by quartile; long-short HF buys top return decile, shorts bottom.

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## LOS B · MEASURES OF DISPERSION

## Range, MAD, Variance, Standard Deviation

$$Range = max - min$$

$$MAD = \frac{\sum |X_{\{i\}} - \bar{X}|}{n}$$

$$\text{Sample variance } s^2 = \frac{\sum (X_{\{i\}} - \bar{X})^2}{(n - 1)} \quad \text{Pop } \sigma^2 = \frac{\sum (X_{\{i\}} - \mu)^2}{N}$$

$$SampleSDs = s \quad PopSD\sigma = \sigma$$

Sample variance uses  $n - 1$  degrees of freedom because one degree is consumed estimating  $\bar{X}$ . This correction produces an *unbiased* estimator.

## WORKED EXAMPLE

## MAD of 4-Year Returns

## GIVEN

4%, 2%, 6%, 8%. Mean = 5%.

## SOLUTION

$$MAD = (|4-5| + |2-5| + |6-5| + |8-5|)/4 = (1 + 3 + 1 + 3)/4 = 2\%$$

## WORKED EXAMPLE

## Sample vs Population SD via BAI Plus

## GIVEN

Same 4%, 2%, 6%, 8%.

## SOLUTION

Keystrokes: [2nd][DATA], [2nd][CLR WRK], 4[ENTER], 2[ENTER], 6[ENTER], 8[ENTER], [2nd][STAT], ensure 1-V mode. Retrieve:  $N=4$ ,  $\bar{X}=5$ ,  $Sx=2.58$  (sample SD),  $\sigma x=2.24$  (pop SD).

**Exam Tip:** Variance is additive; standard deviation is NOT. That's why portfolio risk calculations work in variance space (then take the square root at the end).

## LOS B · DOWNSIDE RISK AND RELATIVE DISPERSION

## Target Semideviation and Coefficient of Variation

## Target Semideviation (Downside Deviation)

$$S_{target} = \sqrt{[\sum_{X_i \leq B} (X_i - B)^2 / (n - 1)]}$$

Counts only observations that fall at or below target B. As B ↑,  $S_{target}$  ↑ because more observations breach the threshold. Denominator uses **full sample n**, not just the downside count.

## Coefficient of Variation (CV)

$$CV = s / \bar{x}$$

Measures *relative* dispersion — risk per unit of mean return. Lower CV = tighter distribution per unit of return (better risk-adjusted).

## WORKED EXAMPLE

## CV Comparison Across Portfolios

## GIVEN

A: mean 10%, SD 9%. B: mean 6%, SD 4%. C: mean 9%, SD 2%.

## SOLUTION

$$CV_A = 9/10 = \mathbf{0.90}$$

$$CV_B = 4/6 = \mathbf{0.66}$$

$$CV_C = 2/9 = \mathbf{0.22}$$

⇒ Portfolio C offers the least risk per unit of return — most attractive.



CHECK YOUR UNDERSTANDING · QUIZ 1

Quiz 1

Select the best answer. Feedback appears after each click.

Q1. Returns are 10%, -5%, 20%. The arithmetic mean return is closest to:

Q2. Quartile 1 of the sorted data {3%, 4%, 6%, 9%, 11%, 12%, 14%} is at position:

Q3. For returns 4%, 2%, 6%, 8%, the mean absolute deviation (MAD) equals:

Q4. Portfolios: A (10%, SD 9%), B (6%, SD 4%), C (9%, SD 2%). Which has the lowest CV?

Q5. A winsorized mean differs from a trimmed mean in that it:

LOS C · SKEWNESS

Shape of the Distribution — Skewness

Normal Distribution

Symmetrical; completely described by 2 parameters:  $\mu$ ,  $\sigma^2$ . Notation:  $X \sim N(\mu, \sigma^2)$ . For symmetry, **mean = median = mode**.

$$\text{Skew} \approx (1/n) \sum (X_i - \bar{X})^3 / s^3 \quad (\text{for } n > 100)$$

| SHAPE                   | ORDERING                | INTERPRETATION    |
|-------------------------|-------------------------|-------------------|
| Skew = 0<br>(symmetric) | mean = median =<br>mode | Normal, symmetric |

|                                 |                      |                                                                                 |
|---------------------------------|----------------------|---------------------------------------------------------------------------------|
| <b>Positive skew</b><br>(right) | mean > median > mode | Long right tail; frequent small losses, few large gains                         |
| <b>Negative skew</b><br>(left)  | mean < median < mode | Long left tail; frequent small gains, few large losses — most financial returns |

**Investor intuition:** Equity returns historically exhibit negative skew — rare but severe drawdowns. Risk management models that assume normality will underestimate tail losses.

## LOS C · KURTOSIS

### Kurtosis — Tail Weight

$$K_{\text{E}} = (1/n) \sum (X_i - \bar{X})^4 / s^4 - 3$$

\quad \quad \quad \text{(excess kurtosis)}

Kurtosis captures combined weight of tails relative to the body.

| TYPE               | KURTOSIS | EXCESS KURTOSIS | DESCRIPTION                                                                 |
|--------------------|----------|-----------------|-----------------------------------------------------------------------------|
| <b>Leptokurtic</b> | $K > 3$  | $K_E > 0$       | Fat tails, peaked center — over-weight head & tails, under-weight shoulders |
| Mesokurtic         | $K = 3$  | $K_E = 0$       | Normal distribution                                                         |
| <b>Platykurtic</b> | $K < 3$  | $K_E < 0$       | Thin tails, flat center — under-weight head & tails, over-weight shoulders  |

**Significance rule:** Excess kurtosis with absolute value > 1 is considered significant. Equity returns typically show  $K_E > 0$  (leptokurtic); a normal-based model will under-state the probability of extreme outcomes.

**Exam Tip:** Securities returns typically exhibit BOTH negative skew AND excess kurtosis. Risk managers focus on the tails, not the mean and SD alone.

## LOS D · COVARIANCE AND CORRELATION

## Co-Movement of Two Variables

## Scatter Plot

Visualizes joint variation in two numerical variables. Identifies outliers, linear / non-linear relationships, or absence of relationship. A *scatter plot matrix* assesses pairwise association among many variables.

$$\text{Cov}(X, Y) = S_{XY} = \frac{\sum (X_i - \bar{X})(Y_i - \bar{Y})}{(n - 1)}$$

$$\text{Correlation } r_{XY} = S_{XY} / (S_X \cdot S_Y)$$

Covariance = joint variability, expressed in *units squared*. Correlation standardizes it to  $[-1, +1]$  and has no units.

## Correlation Properties

1.  $-1 \leq r \leq +1$
2.  $r = 0 \rightarrow$  no linear relationship  $\Rightarrow$  maximum diversification benefit
3.  $r = +1 \rightarrow$  perfect positive correlation  $\Rightarrow$  perfect replication (no diversification)
4.  $r = -1 \rightarrow$  perfect negative correlation  $\Rightarrow$  perfect hedge

## WORKED EXAMPLE

## Scatter Plot Statistics

## GIVEN

$n = 11, \bar{X} = 9, \bar{Y} = 7.5, S_X = 3.32, S_Y = 2.03.$

## SOLUTION

For all four scatter-plot data configurations in the classical *Anscombe quartet*-style example,  $r_{XY} = \mathbf{0.82}$ , yet visual inspection shows different shapes  $\Rightarrow$  always inspect the scatter plot!

## Limitations of Correlation

- Linear association only — misses non-linear structure
- Unreliable when outliers are present

- Correlation does NOT imply causation — may be spurious (chance, or due to a third variable)
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## CHECK YOUR UNDERSTANDING · QUIZ 2

## Quiz 2

Select the best answer. Feedback appears after each click.

**Q1. For a positively skewed distribution, the MOST accurate ordering is:**

**Q2. A distribution with excess kurtosis  $K_E = +2$  is BEST described as:**

**Q3. Correlation of +1.0 between two assets implies:**

**Q4. Covariance between two variables is always expressed in:**

**Q5. Which of the following is a limitation of Pearson correlation?**

## LOS D · CORRELATION IN PORTFOLIO CONTEXT

## How Correlation Drives Portfolio Risk

$$\text{Var}(R_p) = w_A^2 \sigma_A^2 + w_B^2 \sigma_B^2 + 2w_A w_B \sigma_A \sigma_B \rho_{AB}$$

Covariance enters the cross term. Portfolio SD < weighted SD of components *whenever*  $\rho < 1$  — this is the quantitative core of diversification.

## WORKED EXAMPLE

## Two-Asset Portfolio Variance

## GIVEN

$$w_A = w_B = 0.50, \sigma_A = \sigma_B = 20\%, \rho_{AB} = 0.4.$$

**SOLUTION**

$$\text{Var}(R_p) = 0.25(0.04) + 0.25(0.04) + 2(0.5)(0.5)(0.2)(0.2)(0.4)$$

$$= 0.01 + 0.01 + 0.008 = \mathbf{0.028}$$

$$\sigma_p = \sqrt{0.028} = \mathbf{16.73\%}, \text{ versus } 20\% \text{ weighted average} \Rightarrow 327 \text{ bps of diversification.}$$

**Exam Tip:** On the exam, expect a question that switches  $\rho$  from 0.4 to 0.8 (variance rises) or to  $-0.3$  (variance drops). Always plug into the two-asset variance formula and remember the cross term doubles.

## CHECK YOUR UNDERSTANDING · QUIZ 3

## Quiz 3

Select the best answer. Feedback appears after each click.

Q1. For a sample of 10 returns with sum of squared deviations = 0.090, the sample variance is:

Q2. Target semideviation (downside deviation) vs standard deviation will be LARGER when:

Q3. Two assets with  $\rho = 0.40$ , equal weights 50/50, and equal  $\sigma = 20\%$ . Portfolio SD is closest to:

Q4. A data set has mean 8% and standard deviation 12%. The coefficient of variation equals:

Q5. Which statement BEST describes a spurious correlation?

## MODULE COMPLETE

## LM03 · Statistical Measures of Asset Returns Summary

Final Score

0 / 0

## Cheat Sheet

$Mean = \Sigma X_i / n$ ;  $median = middle$ ;  $mode = most\ frequent$

$$L_y = (n + 1) \cdot y/100$$

$$s^2 = \Sigma(X_i - \bar{X})^2 / (n - 1); s = \sqrt{s^2}$$

$$CV = s / \bar{X} \quad \cdot \quad S_{\text{target}} = \sqrt{\frac{\Sigma(X_i - B)^2}{(n-1)}}$$

$$\text{Cov}(X, Y) = \Sigma(X_i - \bar{X})(Y_i - \bar{Y}) / (n - 1)$$

$$r_{XY} = S_{XY} / (S_X S_Y), -1 \leq r \leq +1$$

### Last-Minute Exam Checklist

**Skew:** right-skew mean > median > mode; left-skew mean < median < mode

**Kurtosis:** K=3 normal; K>3 leptokurtic (fat tails); K<3 platykurtic

**Trimmed vs winsorized:** delete vs replace outliers

**CV:** lower = better risk per return (e.g., 2/9 = 0.22)

**Correlation:** +1 replication, -1 hedge, 0 max diversification

**Portfolio var:**  $\rho < 1 \Rightarrow$  diversification benefit