

# Measuring Returns, Volatility, and Correlation

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## Volatility, Variance, and Implied Volatility

- The **volatility** of a variable is the standard deviation of that variable's continuously compounded return
- The **variance rate** of a variable is the square of its standard deviation (variance and standard deviation are computed using historical data).
- Risk managers may also compute **implied volatility**, which is the volatility that forces a model price (i.e., option pricing model) to equal the market price.

## Nonnormal Distributions

- The third moment is the skewness and the fourth moment is the kurtosis.
- For a normal distribution, which has thin tails and is symmetric, there is no skewness or excess kurtosis.
- Financial returns often follow a nonnormal distribution, and as such, there is skewness and excess kurtosis.

## Jarque-Bera Test

- The Jarque-Bera (JB) test statistic can be used to test whether a distribution is normal, meaning that there is zero skewness and no excess kurtosis ( $K - 3 = 0$ ).
- If the result falls below the critical value, the null will not be rejected and the distribution will be deemed normal.
- If the result is above the critical value, the null will be rejected. Financial returns are likely to follow a more normal distribution over longer time periods.

## Jarque-Bera Test (cont.)

- The test statistic, where  $T$  is the sample size, is:

$$JB = (T - 1) \left( \frac{\hat{S}^2}{6} + \frac{(\hat{K} - 3)^2}{24} \right)$$

## The Power Law

- The power law is an alternative approach to using probabilities from a normal distribution.
- When  $X$  is **large**, probability of variable  $V$  exceeding  $X$  is:

$$P(V > X) = K \times X^{-\alpha}$$

- $K$  and  $\alpha$  are constants obtained from regression.
- Applies to variables such as incomes, visits to a website, etc.

## The Power Law: Example

- Suppose that  $\alpha = 4$  and  $K = 40$  for a particular financial variable,  $V$ . What is the probability of the variable exceeding 10?

$$P(V > 10) = 40 \times 10^{-4} = 0.4\%$$

## Spearman's Rank Correlation

- Spearman's
  - Nonparametric approach
  - Based on rankings of values
  - Four-step process
    1. Order the set pairs of variables  $X$  and  $Y$  with respect to the set  $X$
    2. Determine the ranks of  $X_i$  and  $Y_i$  for each time period  $i$
    3. Calculate the difference of the variable rankings and square the difference
    4. Sum squares and plug in formula

## Spearman's Rank Correlation (cont.)

Year	X	Y	X Rank	Y Rank	$d_i$	$d_i^2$
2012	-20.0%	10.0%	1	2	-1	1
2014	-10.0%	30.0%	2	4	-2	4
2010	25.0%	-20.0%	3	1	2	4
2013	40.0%	20.0%	4	3	1	1
2011	60.0%	40.0%	5	5	0	0
					Sum =	10

## Kendall's Tau

- Kendall's Tau
  - Ordinal like Spearman's Rank
  - Requires classification of concordant and discordant pairs
    - A concordant pair of observations is when the rankings of two pairs are in agreement:  $X_t < X_{t^*}$  and  $Y_t < Y_{t^*}$  or  $X_t > X_{t^*}$  and  $Y_t > Y_{t^*}$  when  $t \neq t^*$
    - A discordant pair of observations is when the rankings of two pairs are not in agreement:  $X_t < X_{t^*}$  and  $Y_t > Y_{t^*}$  or  $X_t > X_{t^*}$  and  $Y_t < Y_{t^*}$  when  $t \neq t^*$
    - Plug into formula, but a little "brain intensive" for exam