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# AD 685 QUANTITATIVE METHODS FOR FINANC TERM PROJECT

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## 1. Data selection

For this project I selected The Coca-Cola Company (KO). It is an American multinational beverage corporation headquartered in Atlanta, Georgia. A quick search on Yahoo Finance shows its Market Cap (intraday) is USD 299.663 billion (retrieved 18 Jun 2025). According to the project instructions, KO falls under the “large-cap” category (market capitalization between \$100 billion and \$200 billion). Therefore, I have chosen to use the SPDR S&P 500 ETF (SPY) as the benchmark. In addition to the benchmark, I selected SPDR Gold Shares (GLD) as my second explanatory variable (denoted as X2). GLD is an exchange-traded fund (ETF) designed to track the price of gold bullion. As the Coca-Cola Company operates on a global scale and is subject to commodity and macroeconomic fluctuations, the inclusion of gold returns provides a relevant economic signal.

## 2. Calculate daily earnings and related statistics

Because most of the estimations will refer to returns rather than prices, I choose to use logarithmic return:

$$r_t = \ln\left(\frac{P_t}{P_{t-1}}\right)$$

where  $P_{t-1}$  and  $P_t$  are the starting and ending prices.

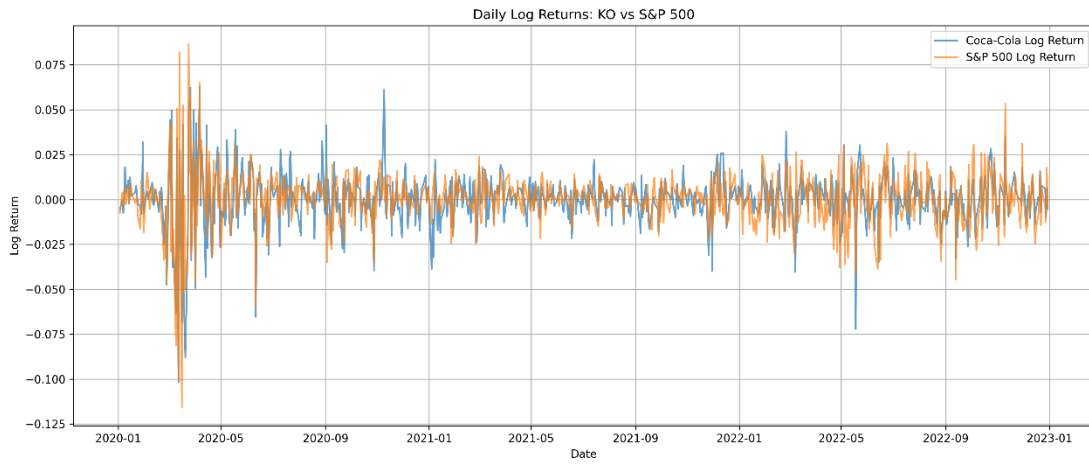
Log-returns are additive and symmetric, closely approximate simple returns for small moves, and exhibit distributions closer to normality, making them ideal for multi-period quantitative asset analysis. The price data for GLD and SPY are also converted to daily logarithmic yields to maintain consistency with the yield calculation method used by KO.

By using Python, I obtained a series of statistics for the stock returns (see Table 1).

<b>Name</b>	The Coca-Cola Company
<b>Ticker</b>	KO

<b>Mean</b>	0.000193
<b>Median</b>	0.000958
<b>Sample standard deviation</b>	0.015624
<b>Minimum</b>	-0.101728
<b>Maximum</b>	0.062783
<b>Sample skewness</b>	-0.841427
<b>Sample kurtosis</b>	6.966093
<b>Starting date</b>	2020/1/3
<b>Final date</b>	2022/12/30

**Table 1** Stock Return Statistics



**Figure 1:** Daily Log Returns KO vs SPY

### 3. OLS regression and Newey–West estimator

#### 3.1 Ordinary Least Squares Regression

Ordinary Least Squares (OLS) regression is a method for estimating the relationship between a dependent variable and one or more independent variables by minimizing the sum of squared residuals (differences between observed and predicted values).

$$y_i = \beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \cdots + \beta_k x_{ik} + \varepsilon_i$$

OLS finds the set of coefficients  $\beta_0, \beta_1, \dots, \beta_k$  that minimize:

$$\sum_{i=1}^n (y_i - \hat{y}_i)^2$$

### 3.2 Newey–West estimator

In time-series regressions, residuals often display heteroskedasticity and autocorrelation, violating classical OLS assumptions. This makes standard errors invalid, compromising inference.

The Newey–West estimator corrects this by adjusting the OLS covariance matrix to provide robust standard errors.

The general formula is:

$$\widehat{Var}_{NW}(\widehat{\beta}) = (X'X)^{-1}\hat{S}(X'X)^{-1}$$

where:

$$\hat{S} = \Gamma_0 + \sum_{h=1}^q w_h (\Gamma_h + \Gamma_h^T)$$

$q$ : maximum lag order (e. g., **Newey–West (1 lag)** uses  $q = 1$ )

$$w_h = 1 - \frac{h}{q+1} : \text{Bartlett weights}$$

$\Gamma_h$ : sample lag – hhh autocovariance of residuals

For the special case of **Newey–West (1 lag)**:

$$\hat{S}_{lag=1} = \Gamma_0 + 0.5(\Gamma_1 + \Gamma_1^T)$$

Therefore, the robust covariance becomes:

$$\widehat{Var}_{NW(1)}(\widehat{\beta}) = (X'X)^{-1}[\Gamma_0 + 0.5(\Gamma_1 + \Gamma_1^T)](X'X)^{-1}$$

## 4. Regression Results

### 4.1 Separate data

First, I separated all the data into two periods:

- ◆ First 2 years. This data will be used to “teach” or establish our model.
- ◆ Last year. This data will be used to test our model and to “forecast”.

The results are shown on the table below:

KO-train (2020-2021)		KO-test (2022)	
Mean	0.000147	Mean	0.000286
Median	0.000751	Median	0.001399
Std Dev	0.016984	Std Dev	0.012482
Min	-0.101728	Min	-0.072169
Max	0.062783	Max	0.037942
Skewness	-0.81508	Skewness	-0.835994
Kurtosis	6.624366	Kurtosis	4.678076

**Table 2** KO’s two periods’ statistics

Compared to 2020–2021, KO's log returns in 2022 showed slightly higher average returns, but lower volatility and fewer extreme movements. The return distribution remained slightly left-skewed, but became more stable and closer to normal, as indicated by lower kurtosis. Overall, KO returns in 2022 were calmer and less risky than in the prior two years.

SPY-train (2020-2021)		SPY-test (2022)	
Mean	0.000754	Mean	-0.000863
Median	0.001648	Median	-0.001815
Std Dev	0.016144	Std Dev	0.015298
Min	-0.115887	Min	-0.044456
Max	0.086731	Max	0.053497
Skewness	-1.061486	Skewness	-0.001245
Kurtosis	13.253851	Kurtosis	0.33286

**Table 3** SPY’s two periods’ statistics

In 2022, SPY's average and median log returns turned negative, contrasting with the positive returns in 2020–2021. Volatility stayed at a similar level, but extreme movements became much less frequent, as seen from the sharp drop in kurtosis. The distribution also became more symmetric, with skewness moving closer to zero. Overall, SPY returns in 2022 were weaker but more stable.

GLD-train (2020-2021)		GLD-test (2022)	
Mean	0.000341	Mean	-0.000031
Median	0.000926	Median	0.000388
Std Dev	0.010643	Std Dev	0.009619
Min	-0.05519	Min	-0.030162
Max	0.04739	Max	0.030235
Skewness	-0.614508	Skewness	0.133872
Kurtosis	3.869966	Kurtosis	0.629863

**Table 4** GLD's two periods' statistics

GLD's log returns in 2022 showed slightly lower volatility compared to 2020–2021.

The average return turned slightly negative, though the median remained positive, suggesting minimal directional change. Skewness shifted from left-skewed to mildly right-skewed, and kurtosis decreased notably, indicating fewer extreme price moves. Overall, GLD returns in 2022 were more balanced and less volatile than in the previous two years.

## 4.2 Stock returns & Intercept

```

=== OLS ===
                                OLS Regression Results
=====
Dep. Variable:                  y      R-squared:                  0.002
Model:                        OLS      Adj. R-squared:             -0.000
Method:                       Least Squares      F-statistic:              0.7580
Date:                         Thu, 19 Jun 2025      Prob (F-statistic):       0.384
Time:                         20:09:43      Log-Likelihood:          1339.8
No. Observations:              504      AIC:                     -2676.
Df Residuals:                  502      BIC:                     -2667.
Df Model:                      1
Covariance Type:               nonrobust
=====
                                coef      std err          t      P>|t|      [0.025      0.975]
-----
const                -2.3085         2.652      -0.871      0.384      -7.518       2.901
x1                   3.129e-06      3.59e-06       0.871      0.384      -3.93e-06      1.02e-05
=====
Omnibus:                112.389      Durbin-Watson:           2.092
Prob(Omnibus):           0.000      Jarque-Bera (JB):        908.295
Skew:                   -0.718      Prob(JB):                5.84e-198
Kurtosis:                9.418      Cond. No.                2.59e+09
=====

```

**Figure 2:** KO vs Intercept (OLS)

```

=== Newey-West (lag = 1) ===
                                OLS Regression Results
=====
Dep. Variable:                  y      R-squared:                  0.002
Model:                        OLS      Adj. R-squared:             -0.000
Method:                       Least Squares      F-statistic:              0.5620
Date:                         Thu, 19 Jun 2025      Prob (F-statistic):        0.454
Time:                         20:09:43      Log-Likelihood:           1339.8
No. Observations:              504      AIC:                      -2676.
Df Residuals:                  502      BIC:                      -2667.
Df Model:                      1
Covariance Type:               HAC
=====
               coef      std err          t      P>|t|      [0.025      0.975]
-----
const          -2.3085         3.080        -0.750      0.454        -8.359         3.743
x1              3.129e-06      4.17e-06         0.750      0.454        -5.07e-06      1.13e-05
=====
Omnibus:                  112.389      Durbin-Watson:              2.092
Prob(Omnibus):             0.000      Jarque-Bera (JB):           908.295
Skew:                     -0.718      Prob(JB):                   5.84e-198
Kurtosis:                  9.418      Cond. No.                   2.59e+09
=====

```

**Figure 3: KO vs Intercept (Newey-West)**

Whether using ordinary OLS or Newey-West, the p-values for Intercept are much greater than 0.05 (0.384 and 0.454) and are not significant.

### 4.3 Stock returns & Benchmark returns & Intercept

```

=== OLS ===
                                OLS Regression Results
=====
Dep. Variable:                KO_logRet      R-squared:                  0.571
Model:                        OLS      Adj. R-squared:             0.571
Method:                       Least Squares      F-statistic:              669.3
Date:                         Thu, 19 Jun 2025      Prob (F-statistic):        2.05e-94
Time:                         19:37:12      Log-Likelihood:           1552.9
No. Observations:              504      AIC:                      -3102.
Df Residuals:                  502      BIC:                      -3093.
Df Model:                      1
Covariance Type:               nonrobust
=====
               coef      std err          t      P>|t|      [0.025      0.975]
-----
Intercept       -0.0005         0.000        -0.912      0.362        -0.001         0.001
SPY_logRet       0.7953         0.031        25.871      0.000         0.735         0.856
=====
Omnibus:                  75.347      Durbin-Watson:              1.873
Prob(Omnibus):             0.000      Jarque-Bera (JB):           551.441
Skew:                     -0.377      Prob(JB):                   1.80e-120
Kurtosis:                  8.069      Cond. No.                   62.0
=====

```

**Figure 4: KO vs SPY & Intercept (OLS)**

=== Newey-West (lag = 1) ===

OLS Regression Results						
=====						
Dep. Variable:	KO_logRet	R-squared:	0.571			
Model:	OLS	Adj. R-squared:	0.571			
Method:	Least Squares	F-statistic:	250.8			
Date:	Thu, 19 Jun 2025	Prob (F-statistic):	4.15e-46			
Time:	19:37:12	Log-Likelihood:	1552.9			
No. Observations:	504	AIC:	-3102.			
Df Residuals:	502	BIC:	-3093.			
Df Model:	1					
Covariance Type:	HAC					
=====						
	coef	std err	t	P> t	[0.025	0.975]
-----						
Intercept	-0.0005	0.001	-0.880	0.379	-0.001	0.001
SPY_logRet	0.7953	0.050	15.837	0.000	0.697	0.894
=====						
Omnibus:	75.347	Durbin-Watson:		1.873		
Prob(Omnibus):	0.000	Jarque-Bera (JB):		551.441		
Skew:	-0.377	Prob(JB):		1.80e-120		
Kurtosis:	8.069	Cond. No.		62.0		
=====						

**Figure 5: KO vs SPY & Intercept (Newey-West)**

In both the OLS and Newey-West regressions, the intercept terms are not statistically significant, with p-values of 0.362 and 0.379, respectively. In contrast, the SPY log return is highly significant in both models, with p-values well below the 0.05 threshold.

#### 4.4 Stock returns & CRSP returns & Intercept

=== OLS ===

OLS Regression Results

Dep. Variable: KO\_logRet

R-squared: 0.518

Model: OLS

Adj. R-squared: 0.517

Method: Least Squares

F-statistic: 538.7

Date: Thu, 19 Jun 2025

Prob (F-statistic): 1.67e-81

Time: 19:37:12

Log-Likelihood: 1523.1

No. Observations: 504

AIC: -3042.

Df Residuals: 502

BIC: -3034.

Df Model: 1

Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
Intercept	-0.0006	0.001	-1.080	0.281	-0.002	0.000
CRSP_logRet	0.0074	0.000	23.210	0.000	0.007	0.008

Omnibus: 93.962

Durbin-Watson: 1.834

Prob(Omnibus): 0.000

Jarque-Bera (JB): 785.599

Skew: -0.528

Prob(JB): 2.57e-171

Kurtosis: 9.024

Cond. No.: 1.67

**Figure 6: KO vs CRSP & Intercept (OLS)**



```

=== Newey-West (lag = 1) ===
                                OLS Regression Results
=====
Dep. Variable:                KO_logRet    R-squared:                0.518
Model:                        OLS          Adj. R-squared:           0.517
Method:                      Least Squares  F-statistic:              227.0
Date:                        Thu, 19 Jun 2025  Prob (F-statistic):       1.35e-42
Time:                        19:37:13       Log-Likelihood:           1523.1
No. Observations:            504           AIC:                     -3042.
Df Residuals:                502           BIC:                     -3034.
Df Model:                    1
Covariance Type:             HAC
=====

```

	coef	std err	t	P> t	[0.025	0.975]
Intercept	-0.0006	0.001	-1.033	0.302	-0.002	0.001
CRSP_logRet	0.0074	0.000	15.067	0.000	0.006	0.008

```

=====
Omnibus:                    93.962    Durbin-Watson:           1.834
Prob(Omnibus):              0.000    Jarque-Bera (JB):         785.599
Skew:                       -0.528    Prob(JB):                 2.57e-171
Kurtosis:                   9.024     Cond. No.                  1.67
=====

```

**Figure 7: KO vs CRSP & Intercept (Newey-West)**

In both the OLS and Newey-West regressions, the intercept terms are not statistically significant, with p-values of 0.281 and 0.302, respectively. In contrast, the CRSP log return is highly significant in both models, with p-values far below the 0.05 threshold.

#### 4.5 Stock returns & Benchmark returns & Gold returns & Intercept

```

=== OLS ===
                                OLS Regression Results
=====
Dep. Variable:                KO_logRet    R-squared:                0.572
Model:                        OLS          Adj. R-squared:           0.570
Method:                      Least Squares  F-statistic:              334.4
Date:                        Thu, 19 Jun 2025  Prob (F-statistic):       5.66e-93
Time:                        19:37:13       Log-Likelihood:           1553.1
No. Observations:            504           AIC:                     -3100.
Df Residuals:                501           BIC:                     -3087.
Df Model:                    2
Covariance Type:             nonrobust
=====

```

	coef	std err	t	P> t	[0.025	0.975]
Intercept	-0.0005	0.000	-0.925	0.355	-0.001	0.001
SPY_logRet	0.7926	0.031	25.457	0.000	0.731	0.854
Gold_logRet	0.0267	0.047	0.566	0.571	-0.066	0.120

```

=====
Omnibus:                    73.909    Durbin-Watson:           1.874
Prob(Omnibus):              0.000    Jarque-Bera (JB):         552.457
Skew:                       -0.349    Prob(JB):                 1.09e-120
Kurtosis:                   8.081     Cond. No.                  96.0
=====

```

**Figure 8: KO vs SPY & Gold & Intercept (OLS)**

```

=== Newey-West (lag = 1) ===
                                OLS Regression Results
=====
Dep. Variable:                KO_logRet    R-squared:                0.572
Model:                        OLS          Adj. R-squared:           0.570
Method:                       Least Squares    F-statistic:             131.4
Date:                         Thu, 19 Jun 2025    Prob (F-statistic):       1.35e-46
Time:                         19:37:14          Log-Likelihood:           1553.1
No. Observations:             504              AIC:                     -3100.
Df Residuals:                 501              BIC:                     -3087.
Df Model:                     2
Covariance Type:              HAC
=====
               coef      std err          t      P>|t|      [0.025      0.975]
-----
Intercept      -0.0005      0.001      -0.885      0.376      -0.001      0.001
SPY_logRet       0.7926      0.052     15.201      0.000       0.690      0.895
Gold_logRet      0.0267      0.080       0.335      0.738      -0.130      0.184
=====
Omnibus:                73.909    Durbin-Watson:           1.874
Prob(Omnibus):           0.000    Jarque-Bera (JB):        552.457
Skew:                   -0.349    Prob(JB):                 1.09e-120
Kurtosis:                8.081    Cond. No.                 96.0
=====

```

**Figure 9:** KO vs SPY & Gold & Intercept (Newey-West)

In both the OLS and Newey-West regressions, the intercept terms are not statistically significant, with p-values of 0.355 and 0.376, respectively. The SPY log return remains highly significant across both models, with p-values well below 0.05, confirming its strong predictive power for KO's returns. However, the Gold log return is not statistically significant in either model (p-values of 0.571 and 0.738), suggesting that gold returns do not provide meaningful additional explanatory power for KO's daily return.

#### 4.6 Stock returns & CRSP returns & Gold returns & Intercept

```

=== OLS ===
                                OLS Regression Results
=====
Dep. Variable:                KO_logRet    R-squared:                0.518
Model:                        OLS          Adj. R-squared:           0.516
Method:                      Least Squares  F-statistic:              269.1
Date:                        Thu, 19 Jun 2025  Prob (F-statistic):       4.29e-80
Time:                        19:37:14       Log-Likelihood:           1523.2
No. Observations:            504           AIC:                     -3040.
Df Residuals:                501           BIC:                     -3028.
Df Model:                    2
Covariance Type:             nonrobust
=====
               coef      std err          t      P>|t|      [0.025      0.975]
-----
Intercept      -0.0006      0.001      -1.091      0.276      -0.002      0.000
CRSP_logRet     0.0073      0.000     22.799      0.000      0.007      0.008
Gold_logRet     0.0253      0.050      0.505      0.614      -0.073      0.124
=====
Omnibus:                92.115    Durbin-Watson:           1.835
Prob(Omnibus):          0.000    Jarque-Bera (JB):        784.589
Skew:                  -0.502    Prob(JB):                4.25e-171
Kurtosis:              9.029    Cond. No.                158.
=====

```

**Figure 10: KO vs CRSP & Gold & Intercept (OLS)**

```

=== Newey-West (lag = 1) ===
                                OLS Regression Results
=====
Dep. Variable:                KO_logRet    R-squared:                0.518
Model:                        OLS          Adj. R-squared:           0.516
Method:                      Least Squares  F-statistic:              117.7
Date:                        Thu, 19 Jun 2025  Prob (F-statistic):       1.28e-42
Time:                        19:37:15       Log-Likelihood:           1523.2
No. Observations:            504           AIC:                     -3040.
Df Residuals:                501           BIC:                     -3028.
Df Model:                    2
Covariance Type:             HAC
=====
               coef      std err          t      P>|t|      [0.025      0.975]
-----
Intercept      -0.0006      0.001     -1.034      0.302      -0.002      0.001
CRSP_logRet     0.0073      0.001     14.432      0.000      0.006      0.008
Gold_logRet     0.0253      0.086      0.296      0.768     -0.143      0.194
=====
Omnibus:                92.115    Durbin-Watson:           1.835
Prob(Omnibus):          0.000    Jarque-Bera (JB):        784.589
Skew:                  -0.502    Prob(JB):                4.25e-171
Kurtosis:              9.029    Cond. No.                158.
=====

```

**Figure 11: KO vs CRSP & Gold & Intercept (Newey-West)**

In both the OLS and Newey-West regressions, the intercept terms are not statistically significant, with p-values of 0.276 and 0.302, respectively. The CRSP log return is highly significant in both models (p-values < 0.001), confirming its strong explanatory power for KO's daily returns. However, the Gold log return is not statistically significant (p = 0.614 in OLS, and 0.768 in Newey-West), indicating that gold returns do not significantly contribute to explaining KO's return variation.

In summary, no variable changes its significance status when comparing OLS and Newey-West t-statistics.

## 5. Results Analysis

Regression of KO vs SPY, GLD and CRSP				
Sample: Jan 01, 2020 - Jan 01, 2025				
	1	2	3	4
Intercept	-0.0005	-0.0006	-0.0005	-0.0006
(OLS s.e.)	0.0000	0.0010	0.0000	0.0010
(NW s.e.)	0.0010	0.0010	0.0010	0.0010
S&P500	0.7953	-	0.7926	-
(OLS s.e.)	0.0310	-	0.0310	-
(NW s.e.)	0.0500	-	0.0520	-
CRSP	-	0.0074	-	0.0073
(OLS s.e.)	-	0.0000	-	0.0000
(NW s.e.)	-	0.0000	-	0.0010
X <sub>2</sub>	-	-	0.0267	0.0253
(OLS s.e.)	-	-	0.0470	0.0500
(NW s.e.)	-	-	0.0800	0.0860
R <sup>2</sup> Adjusted	0.571	0.517	0.57	0.516
s.e. Reg	0.01113	0.01181	0.01114	0.01182
NOBS	504	504	504	504

**Table 5 4** Estimated Regressions

In this project, I compare two alternative specifications for modeling the daily log returns of The Coca-Cola Company (KO): Model (1), which uses the SPDR S&P 500 ETF (SPY) as the market proxy, and Model (2), which uses CRSP. Although both market return variables are highly statistically significant (p-values < 0.001), I find that Model (1) offers stronger explanatory power with an R-squared of 0.571, compared to 0.517 in Model (2). This suggests that SPY better captures market-related movements in KO returns over the sample period. While I do not observe strong evidence of omitted variable bias in either model, I acknowledge that other macroeconomic factors, such as gold prices or interest rates, could influence KO's

returns. To address this possibility, I estimate additional models that include gold returns as an explanatory variable.

In Models (3) and (4), I incorporate gold returns alongside SPY and CRSP, respectively. In both models, the market return variables remain statistically significant, while the gold return is not (p-values well above 0.05). Model (3) yields the highest R-squared (0.572), which confirms that SPY is a more effective market benchmark in this setting. Based on these results, I select **Model (3)** as the preferred specification for explaining KO's return behavior over the sample period, due to its stronger explanatory power and robustness.