

# Cointegration Analysis Between BTC Price and Other Cryptocurrencies: A Time Series Approach

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## 1 Abstract

This paper analyzes the cointegration between Bitcoin (BTC) price and three other cryptocurrencies: Solana (SOL), Cardano (ADA), and Binance Coin (BNB). The Engle-Granger method and the Johansen test are used to determine the presence of cointegration. The results show that there is cointegration in the BTC-SOL and BTC-BNB relationships but not in the BTC-ADA relationship. These findings suggest a long-term relationship between BTC-SOL and BTC-BNB prices but not between BTC and ADA. The paper is structured as follows: the State of the Art, covers the research context related to the proposed topic, which is currently trending. The Data and Main Descriptive Statistics section explains the data collection along with a brief analysis. The Methodology section outlines the process used to generate the results analyzed in the following section. Finally, the Conclusions and Annexes sections summarize the key findings along with the code to replicate the exercise.

## 2 State of the Art

To delve into the topic of cointegration, we reference the work of Campbell and Shiller (1987). This study reviews some models and proposed methods for testing cointegration in the context of economic variables. Although it does not focus on cryptocurrencies, it provides an important theoretical framework for understanding cointegration concepts and time series analysis. This article highlights the importance of cointegration in analyzing long-term relationships and is relevant for understanding the theoretical basis of cointegration analysis in general.

To gain a broader understanding of how the stock market functions, the work of Chang et al. (2008) is useful. Despite focusing on short-term adjustments in the U.S. market, it is relevant to consider its discussion on market efficiency, as it warns of possible rational







bubbles in the stock market, along with its analysis of market volatility. It is essential to consider these phenomena in the cryptocurrency market as well. Furthermore, to better understand the dynamics of Bitcoin (BTC), Aleti and Mizrach (2021) provide a solid context by investigating causal relationships, cointegration, and price discovery between Bitcoin spot and futures markets.

Regarding cointegration between Bitcoin and other markets, we can mention the works of Bejaoui et al. (2022) and Lee and Rhee (2022), who examine the relationship between Bitcoin and various macroeconomic variables and assets such as crude oil, the SP500, and natural gas, as well as the price of gold, etc. (The first study analyzes it during the COVID period). Both conclude with evidence of long-term and short-term dependencies between Bitcoin and various assets.

Focusing on our study's approach, there is extensive literature supporting cointegration among different cryptocurrencies. From a theoretical perspective, cointegration is primarily associated with BTC, as it was the first cryptocurrency, and many argue that it significantly influences "newer" coins. In this sense, for example, Göttfert (2019) examines whether there is cointegration between BTC's daily price and five other cryptocurrencies: Ethereum, Ripple, Bitcoin Cash, EOS, and Litecoin, over five different time periods, applying two different cointegration tests: the Engle-Granger approach with the Johansen cointegration test and a Vector Error Correction Model (VECM). The study concludes that Bitcoin's price has a statistically significant long-term impact on all the prices except EOS.

Another example is Gül (2022), who examines cointegration relationships between BTC prices and several major cryptocurrencies: BNB, ADA, DOGE, ETH, DOT, and XRP. In addition to finding cointegration and long-term relationships, the study identifies a bidirectional causal relationship between BNB and ETH. Both studies contribute to the growing body of research exploring relationships among different cryptocurrencies using time series analysis techniques.

## 3 Data and Main Descriptive Statistics

To explore the cointegration between BTC and other cryptocurrencies, we used weekly adjusted price data for the selected currencies: Solana (SOL), Cardano (ADA), and Binance Coin (BNB). This selection was based on comparing two currencies (ADA and BNB) with findings from previous literature and adding an additional one (SOL) that was not covered in the reviewed literature. The data were obtained from Yahoo Finance over a one-year period, starting from 08/05/2023.

BTC price was taken as x1, SOL as y1, ADA as y2, and BNB as y3, obtaining the following







descriptive statistics.

```
Median
      1st Qu.
                           Mean 3rd Qu.
        28009
                 37221
                                   50873
 25832
      1st Qu.
                 Median
                                3rd Qu.
                                          202.87
                 56.90
                                 111.00
         21.78
      1st Qu.
                 Median
                           Mean 3rd Qu.
0.2430 0.2902
                0.3840
                         0.4183
                                 0.5203
                                          0.7282
                 Median
                           Mean 3rd Qu.
        234.5
                 247.1
                          326.9
                                  342.5
                                           606.9
```

Figure 1: Main descriptive statistics by variable

Based on the results in Figure 1, we can generally say that ADA shows a narrower range compared to the other cryptocurrencies. Additionally, most distributions are right-skewed except for y2, where the mean and median are relatively close.

## 4 Methodology

The methodology described below was applied to each cointegration analysis, i.e., SOL-BTC, ADA-BTC, and BNB-BTC. However, we will explain the SOL-BTC case in detail since the methodology remains the same, with only the results varying in the next section. The results are presented in various figures in ??, along with the code used in ??.

First, we transformed the data into a time series, each containing 54 observations (weeks), as shown in Figure 2. Analyzing BTC's order, we performed the Augmented Dickey-Fuller (ADF) unit root test and found that the series is non-stationary, as indicated by the results in Figure 3, where the tau statistic is -1.9435, not statistically significant at even a 10% level.

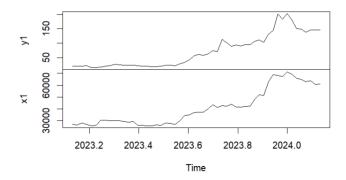


Figure 2: Time series: BTC (x1) and SOL (y1)





```
**********
 Augmented Dickey-Fuller Test Unit Root Test
Test regression trend
Call:
lm(formula = z.diff \sim z.lag.1 + 1 + tt + z.diff.lag)
Residuals:
              1Q Median
-4014.4 -1204.8 -326.8
                            872.4 10653.1
Coefficients:
               Estimate Std. Error t value Pr(>|t|)
                                      1.759
-1.944
(Intercept) 2053.41011 1167.42828
                                                0.0850
                            0.05535
                                                0.0578
               -0.10757
z.lag.1
              104.16058
                           54.33494
z.diff.lag
                0.21129
                            0.13872
                                       1.523
                                                0.1343
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 2549 on 48 degrees of freedom
Multiple R-squared: 0.1037, Adjusted R-squared: 0.
F-statistic: 1.851 on 3 and 48 DF, p-value: 0.1505
Value of test-statistic is: -1.9435 2.0482 1.9624
Critical values for test statistics:
1pct 5pct 10pct
tau3 -4.04 -3.45 -3.15
phi2 6.50 4.88 4.16
phi3 8.73 6.49
```

Figure 3: Augmented Dickey-Fuller Test for the BTC time series

Next, we verify that the series are first-order stationary. To do this, we apply a differentiation and use the unit root test again. In Figure 4, we can visually observe the differentiation for BTC-SOL, and after performing the ADF test on BTC prices, we obtain Figure 5, confirming that the series is first-order stationary at a significance level of 1%, as the new tau statistic is -4.1534 < -2.6.

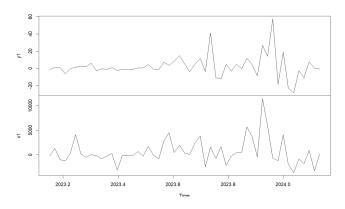


Figure 4: First-order series: BTC (x1) and SOL (y1)

(in **D** d)



```
*************************************
# Augmented Dickey-Fuller Test Unit Root Test #
Test regression none
lm(formula = z.diff \sim z.lag.1 - 1 + z.diff.lag)
Residuals:
Min 1Q Median 3Q Max
-3385.2 -919.6 40.2 1723.2 11345.7
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
                          0.17831 -4.153 0.000131
0.14453 -0.311 0.757343
z.diff.lag -0.04491
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
Residual standard error: 2670 on 49 degrees of freedom
Multiple R-squared: 0.3901, Adjusted R-squared: 0.3652
F-statistic: 15.67 on 2 and 49 DF, p-value: 5.48e-06
Value of test-statistic is: -4.1534
Critical values for test statistics:
1pct 5pct 10pct
tau1 -2.6 -1.95 -1.61
```

Figure 5: ADF Test: First-order BTC series

The rest of the unit root tests for the remaining series can be found in ??. Since both series are of the same order, we can analyze their cointegration. We use the Engle and Granger method, which establishes that: since we are estimating the coefficient, we first use ordinary least squares, then we can perform unit root tests on the residuals, and if they are stationary, we can say that both variables are cointegrated or that a long-term relationship exists between them.

Figure 6 shows the residuals graph of the SOL-BTC model. Visually, we can infer that it meets the condition and that the residuals are stationary. However, to verify this, we perform the ADF test, obtaining in Figure 7 that the residuals are stationary or order 0, since being less than 2, we reject the null hypothesis of non-stationarity.

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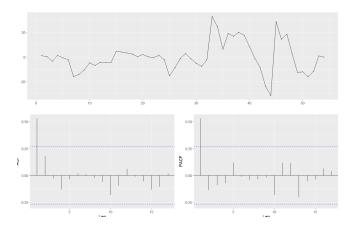


Figure 6: Residuals: model y1~x1

Figure 7: ADF: residuals of the model y1~x1

Since  $\beta$  in the model is an estimate, the ADF test is not the best option to determine whether there is cointegration between the variables, as its critical values would not be properly calculated. Therefore, we complement the analysis using the Johansen test, which provides more reliable results in large samples. The amount of data obtained is considered sufficient to apply this methodology and compare its results.



```
#########################
# Johansen-Procedure #
##############################
Test type: trace statistic , with linear trend
Eigenvalues (lambda):
[1] 0.242101218 0.002971283
Values of teststatistic and critical values of test:
r = 0 | 14.57 | 15.66 | 17.95 | 23.52
Eigenvectors, normalised to first column:
(These are the cointegration relations)
             y1.12
                          x1.12
y1.12 1.000000000 1.000000000
x1.12 -0.003785911 0.004783697
Weights W:
(This is the loading matrix)
          y1.12
                        x1.12
y1.d -0.4506509 -0.003507529
x1.d 26.1484996 -1.066094238
```

Figure 8: Johansen Test: model y1~x1

#### **Results Analysis** 5

The results show that all selected time series are of order 1. Therefore, we can apply the Engle and Granger method. Additionally, for each cointegration analysis, we verify whether a long-term relationship exists between both variables using the Johansen procedure, based on the explanation of the results generated by the command found in QuantStart (n.d.).

For SOL-BTC, as shown in Figure 8 in the section "Values of test statistic and critical values of test", the test statistic value for the null hypothesis  $r \le 1$  (0.15) is lower than the corresponding critical values at significance levels, indicating that the null hypothesis of r <= 1 is not rejected. On the other hand, the test statistic value for the null hypothesis r = 0 (14.57) is also lower than the critical values but is close to the critical value at a 10

The cases of ADA-BTC and BNB-BTC are somewhat different. As seen in Figure 21 and Figure 22, the ADF test indicates that there is no stationarity in the residuals for either model. To enhance the analysis, the Johansen procedure is also applied in both cases.

In the case of ADA-BTC, as shown in Figure 23, the statistics 8.42 and 1.46 are significantly lower than the critical values. Therefore, we do not have statistical evidence to

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reject the null hypothesis of no cointegration between the variables.

Finally, in the case of BNB-BTC, using the results shown in Figure 24, we see that the statistic 16.03 for the test r=0 exceeds the critical value at the 10% significance level, providing statistical evidence of cointegration between these variables or a long-term relationship, unlike the results obtained from the ADF test.

## 6 Conclusions

- 1. Based on the Johansen procedure, there is cointegration in the BTC-BNB price relationship, implying that the prices of these cryptocurrencies tend to move together in the long run. Possible cointegration is also suggested in the BTC-SOL relationship, as its statistic is close to the critical value at a 10
- 2. There is no cointegration between BTC and ADA prices (something to contrast with the literature review findings), implying that the prices of these cryptocurrencies do not tend to move together in the long run.
- 3. The results suggest that BTC has a greater influence on SOL and BNB prices than on ADA prices. These findings could have significant implications for investors seeking a better understanding of the cryptocurrency market.
- 4. As an observation, it is important to mention that the data sample taken may not have been large enough as initially considered, which could affect the obtained results. Additionally, the literature review mentions how high volatility in these markets could change the results over time.
- 5. Since there is cointegration between the two variables, an error correction model may exist. That is, this long-term relationship can be modeled using a Vector Error Correction Model (VECM), which could help predict future movements or understand market dynamics.

### References

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- QuantStart. (n.d.). Johansen test for cointegrating time series analysis in r. https://www.quantstart.com/articles/Johansen-Test-for-Cointegrating-Time-Series-Analysis-in-R/

## 7 Appendix 1

The code used in R software, along with the data utilized for this study, can be found at the following GitHub link:

https://github.com/Sebas20031-DS/Research-and-Writings/tree/main/Time%20Series% 20Cointegration%20Analysis%20BTC%20and%20Altcoins

File name: Cointegration\_BTC\_cryptocurrencies.R

## 8 Appendix 2

**Original Time Series Compared to BTC:** 



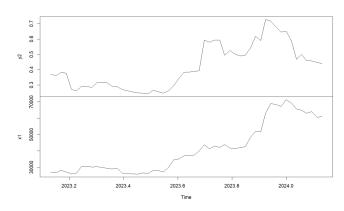


Figure 9: Time series: BTC (x1) and ADA (y2)

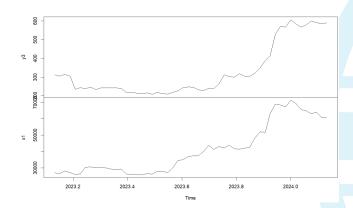


Figure 10: Time series: BTC (x1) and BNB (y3)

**ADF Test for Original Series:** 





Figure 11: Augmented Dickey-Fuller test: SOL time series

Figure 12: Augmented Dickey-Fuller test: ADA time series

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```
************************************
# Augmented Dickey-Fuller Test Unit Root Test #
 Test regression trend
Call:
 lm(formula = z.diff \sim z.lag.1 + 1 + tt + z.diff.lag)
Residuals:
                 1Q Median
 Min 1Q
-52.994 -13.081
                       0.365
                                            98.060
Coefficients:
                Estimate Std. Error t value Pr(>|t|)
                                            0.120
-1.755
2.659
1.327
                                                        0.9049
0.0856
0.0106
 (Intercept)
                  1.00563
                                 8.37110
                                0.03652
0.33013
                -0.06411 \\ 0.87781
 z.lag.1
tt
z.diff.lag
                 0.18390
                                0.13858
                                                        0.1908
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
Residual standard error: 22.55 on 48 degrees of freedom
Multiple R-squared: 0.1942, Adjusted R-squared: 0.1439
F-statistic: 3.857 on 3 and 48 DF, p-value: 0.01496
 Value of test-statistic is: -1.7555 2.9171 3.5811
Critical values for test statistics:
1pct 5pct 10pct
tau3 -4.04 -3.45 -3.15
phi2 6.50 4.88 4.16
phi3 8.73 6.49 5.47
```

Figure 13: Augmented Dickey-Fuller test: BNB time series

## **Differentiated Series:**

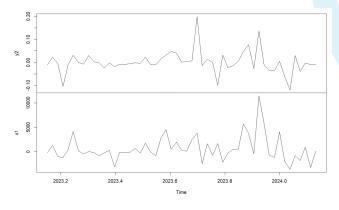


Figure 14: First-order series: BTC (x1) and ADA (y2)

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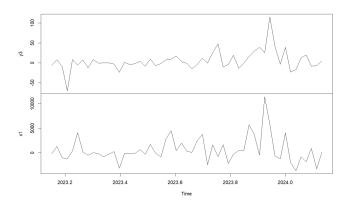


Figure 15: First-order series: BTC (x1) and BNB (y3)

## **ADF Test for Differentiated Series:**

Figure 16: ADF Test: SOL first-order series

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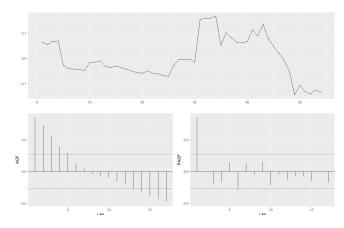


Figure 17: ADF Test: ADA first-order series

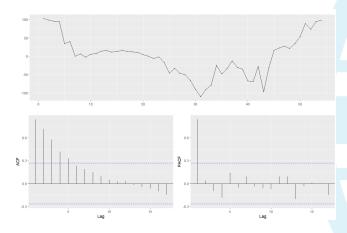
Figure 18: ADF Test: BNB first-order series

#### **Model Residuals:**

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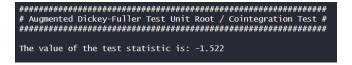


**Figure 19:** Residuals: Model  $y2 \sim x1$ 

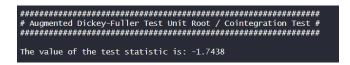


**Figure 20:** Residuals: Model  $y3 \sim x1$ 

#### **ADF Test for Residuals:**



**Figure 21:** ADF Test: Residuals of model  $y2 \sim x1$ 



**Figure 22:** ADF Test: Residuals of model  $y3 \sim x1$ 

### **Johansen Process for Models:**



**Figure 23:** Johansen Test: Model  $y2 \sim x1$ 

```
########################
# Johansen-Procedure #
############################
Test type: trace statistic , with linear trend
Eigenvalues (lambda):
[1] 0.22140267 0.05630886
Values of teststatistic and critical values of test:
test 10pct 5pct 1pct r <= 1 | 3.01 6.50 8.18 11.65
r = 0 | 16.03 | 15.66 | 17.95 | 23.52
Eigenvectors, normalised to first column:
(These are the cointegration relations)
             y3.12
y3.12 1.00000000 1.000000000 x1.12 -0.01269204 -0.005985037
Weights W:
(This is the loading matrix)
y3.12 x1.12
y3.d -0.1303542 -0.02821743
x1.d -1.9103076 -10.69694405
```

Figure 24: Johansen Test: Model  $y3 \sim x1$