ADA AND BTC ANALYSIS AND TIMESERIES **PREDICTION** Introduction Cardano is a cryptocurrency network and open source project that aims to run a public blockchain platform for smart contracts. Cardano's internal cryptocurrency is called Ada. The development of the project is overseen and supervised by the Cardano Foundation based in Zug, Switzerland. The platform began development in 2015 and was launched in 2017 by Charles Hoskinson, a co-founder of Ethereum and BitShares. According to Hoskinson, he had left Ethereum after a dispute over keeping Ethereum nonprofit. After leaving he co-founded IOHK, a blockchain engineering company, whose primary business is the development of Cardano, alongside the Cardano Foundation and Emurgo. The platform is named after Gerolamo Cardano and the cryptocurrency after Ada Lovelace. The currency debuted with a market cap of 600 million dollars. By the end of 2017, it had a market cap of 10 billion dollars, and reached a value of 33 billion dollars briefly in 2018 before a general tightening of the crypto market dropped its value back to 10 billion dollars. According to Mashable, Cardano claims that it overcomes existing problems in the crypto market: mainly that Bitcoin is too slow and inflexible, and that Ethereum is not safe or scalable. Cardano is considered a third-generation cryptocurrency by its creators. 1 Understand the problem and import the more important libraries I want to compare ADA with BTC (Bitcoin) which all of you should know. If not: https://en.wikipedia.org/wiki/Bitcoin import pandas as pd import numpy as np import matplotlib.pyplot as plt First of all I downloaded both datasets and import them to the project. I took data from Jan 2018 ultil Feb 2021. #Load de data cardano hst = pd.read csv('C:/Users/torre/OneDrive/Escritorio/PERSONAL/DATA SCIENCE/Ca cardano hst.head() **Date** Price Open High Low Vol. Change % **0** Feb 20, 2021 1.149122 0.925955 1.176955 0.914190 1.80B 24.11% **1** Feb 19, 2021 0.925894 0.913695 0.945353 0.880769 839.10M 1.33% **2** Feb 18, 2021 0.913752 0.891719 0.956599 0.891719 904.68M 2.47% **3** Feb 17, 2021 0.891702 0.870390 0.897079 0.823855 750.02M 2.45% Feb 16, 2021 0.870393 0.859853 0.905065 0.834791 1.19% We got the Closed attribute ase Price bitcoin hst = pd.read csv('C:/Users/torre/OneDrive/Escritorio/PERSONAL/DATA SCIENCE/Ca bitcoin hst.head() **Price** Vol. Change % **Date** High Open Low **0** Feb 20, 2021 55923.7 55922.0 57523.8 54124.1 127.85K 0.03% Feb 19, 2021 55906.6 51590.1 56238.5 50816.8 139.43K 8.38% **2** Feb 18, 2021 51582.2 52094.5 52524.0 50941.6 -0.95% 94.35K Feb 17, 2021 52079.2 49161.3 52577.7 49018.1 140.03K 5.92% Feb 16, 2021 49169.7 47934.2 50515.8 47044.4 2.57% cardano hst.shape (1148, 7)In [214... bitcoin hst.shape (1148, 7)Out[214... cardano hst.describe() **Price** Open High Low 1148.000000 1148.000000 1148.000000 1148.000000 count 0.133454 0.133274 0.140585 0.124751 mean 0.160964 0.160209 0.173440 0.144590 std 0.023222 0.023225 0.017774 0.026454 min 25% 0.045895 0.045884 0.047424 0.044232 **50%** 0.081616 0.081616 0.084058 0.078212 **75**% 0.143326 0.143324 0.148790 0.137339 1.149122 1.180000 1.350000 1.050000 max We can get from this data indicators that the historical maximum is 1.35 but never closed like that or even higher that 1.149122. Also we can see that the std (standard deviation) is higher on 'High' than in the other ones so that mean the cryptocurrency market fluctuate more on higher values. bitcoin hst.describe() Out[216... **Price** Open High Low count 1148.000000 1148.000000 1148.000000 1148.000000 10023.177003 9985.070035 10278.126394 9684.443118 7313.543011 6895.392082 std 7186.251726 7540.971388 min 3228.700000 3228.600000 3282.300000 3177.000000 25% 6597.875000 6597.575000 6708.500000 6472.175000 **50**% 8544.700000 8544.800000 8743.650000 8243.850000 10405.225000 10736.575000 75% 10404.775000 10130.675000 55923.700000 55922.000000 57523.800000 54124.100000 2 Data Preparation cardano hst = cardano hst.rename(columns={"Vol.": "Vol","Change %": "Change"}) bitcoin_hst = bitcoin_hst.rename(columns={"Vol.": "Vol","Change %": "Change"}) cardano hst.head() **0** Feb 20, 2021 1.149122 0.925955 1.176955 0.914190 1.80B 24.11% **1** Feb 19, 2021 0.925894 0.913695 0.945353 0.880769 839.10M 1.33% **2** Feb 18, 2021 0.913752 0.891719 0.956599 0.891719 904.68M 2.47% **3** Feb 17, 2021 0.891702 0.870390 0.897079 0.823855 2.45% **4** Feb 16, 2021 0.870393 0.859853 0.905065 0.834791 929.87M 1.19% In [219... import datetime # Convert Date to datetime cardano_hst['Date'] = pd.to_datetime(cardano_hst['Date']) cardano hst['Crypto'] = 'ADA' bitcoin_hst['Date'] = pd.to_datetime(bitcoin_hst['Date']) bitcoin hst['Crypto'] = 'BTC' I'll probably want to concatenate both datasets so thats why I created the Crypto attribute to differentiate rows. cardano hst.head() **Date** Price Open High Low Vol Change Crypto **0** 2021-02-20 1.149122 0.925955 1.176955 0.914190 1.80B 24.11% ADA **1** 2021-02-19 0.925894 0.913695 0.945353 0.880769 839.10M 1.33% ADA **2** 2021-02-18 0.913752 0.891719 0.956599 0.891719 904.68M 2.47% ADA **3** 2021-02-17 0.891702 0.870390 0.897079 0.823855 750.02M 2.45% ADA **4** 2021-02-16 0.870393 0.859853 0.905065 0.834791 929.87M 1.19% ADA In [224... import seaborn as sns sns.lineplot(x='Date', y='Price', data=bitcoin hst) plt.xticks(rotation=25) plt.title('BITCOIN Price TimeSeries') plt.show() BITCOIN Price TimeSeries 50000 40000 30000 20000 10000 2019-09 2020-01 2020-05 5 2018-09 2019-01 Probably we should plot some information grouped by the date sub-attributes so that's why the next code. cardano hst = cardano hst.set index('Date') cardano hst['Year'] = cardano hst.index.year cardano hst['Month'] = cardano hst.index.month cardano hst['Day'] = cardano hst.index.day sns.lineplot(x="Date", y="Price", data=cardano_hst, color='green') plt.xticks(rotation=25) plt.title('CARDANO Price TimeSeries') plt.show() CARDANO Price TimeSeries 1.2 1.0 0.8 0.6 0.4 0.2 0.0 2019-09 2020-01 2019-01 2019-05 2020-05 2020-09 bitcoin hst['Change'] = bitcoin hst['Change'].str.replace('%', bitcoin_hst['Vol'] = bitcoin_hst['Vol'].str.replace('B', '000000000') bitcoin_hst['Vol'] = bitcoin_hst['Vol'].str.replace('M', '000000') bitcoin_hst['Vol'] = bitcoin_hst['Vol'].str.replace('K', '000') bitcoin_hst['Vol'] = bitcoin_hst['Vol'].str.replace('.', '') #bitcoin hst = bitcoin hst.drop(columns=['Date']) <ipython-input-227-afd584413ff3>:5: FutureWarning: The default value of regex will cha nge from True to False in a future version. In addition, single character regular expr essions will*not* be treated as literal strings when regex=True. bitcoin hst['Vol'] = bitcoin hst['Vol'].str.replace('.', '') bitcoin hst.head() Out[228... **Date** Vol Change Crypto Price Open High Low **0** 2021-02-20 55923.7 55922.0 57523.8 54124.1 12785000 0.03 BTC 51590.1 50816.8 2021-02-19 55906.6 56238.5 13943000 8.38 BTC **2** 2021-02-18 51582.2 52094.5 52524.0 50941.6 9435000 -0.95 BTC **3** 2021-02-17 52079.2 49161.3 52577.7 49018.1 14003000 5.92 BTC **4** 2021-02-16 49169.7 47934.2 50515.8 47044.4 14137000 2.57 **BTC** In [229... cardano hst['Change'] = cardano hst['Change'].str.replace('%', '') cardano_hst['Vol'] = cardano_hst['Vol'].str.replace('B', '000000000') cardano hst['Vol'] = cardano hst['Vol'].str.replace('M', '000000') cardano hst['Vol'] = cardano hst['Vol'].str.replace('.', '') #cardano hst = cardano hst.drop(columns=['Date']) <ipython-input-229-9f8d90b4cd53>:4: FutureWarning: The default value of regex will cha nge from True to False in a future version. In addition, single character regular expr essions will*not* be treated as literal strings when regex=True. cardano hst['Vol'] = cardano hst['Vol'].str.replace('.', '') cardano hst.head() High **Price** Vol Change Crypto Year Month Day Open Low Date ADA 2021 **2021-02-20** 1.149122 0.925955 1.176955 0.914190 180000000000 24.11 20 **2021-02-19** 0.925894 0.913695 0.945353 0.880769 83910000000 1.33 ADA 2021 19 **2021-02-18** 0.913752 0.891719 0.956599 0.891719 90468000000 2.47 ADA 2021 18 **2021-02-17** 0.891702 0.870390 2021 0.897079 0.823855 75002000000 2.45 ADA 17 **2021-02-16** 0.870393 0.859853 0.905065 0.834791 92987000000 ADA 2021 2 1.19 16 cardano hst.isnull().any() False Price Open False High False Low False Vol False Change False Crypto False False Year Month False False Day dtype: bool 4 Data Analysis for ADA cardano hst['Month'] = cardano hst.index.month ig, axes = pt.subplots(3, 1, figsize=(11, 10), sharex=True) for name, ax in zip(['Price', 'High', 'Low'], axes): sns.boxplot(data=cardano_hst, x='Month', y=name, ax=ax) ax.set_ylabel('Dollars') ax.set title(name) # Remove the automatic x-axis label from all but the bottom subplot **if** ax != axes[-1]: ax.set xlabel('') Price 1.2 1.0 0.8 0.6 0.4 0.2 0.0 High 1.25 1.00 0.75 0.50 0.25 0.00 Low 1.0 0.8 0.6 0.4 0.2 0.0 10 Month In [238... bitcoin_hst = bitcoin_hst.set_index('Date') bitcoin hst['Month'] = bitcoin hst.index.month ig, axes = pt.subplots(3, 1, figsize=(11, 10), sharex=True) for name, ax in zip(['Price', 'High', 'Low'], axes): sns.boxplot(data=bitcoin hst, x='Month', y=name, ax=ax) ax.set_ylabel('Dollars') ax.set title(name) # Remove the automatic x-axis label from all but the bottom subplot **if** ax != axes[-1]: ax.set xlabel('') Price 50000 40000 30000 20000 10000

High

Low

Month

I am definitely not a trader or a broker but if I wanted to play some money on this I would probably earn more money on January, February, May and December for ADA and the same for BTC but November

4 Time Series Forecasting using

from sklearn.ensemble import RandomForestRegressor

pull data into target (y) and predictors (X)

predictor_cols = ['Open', 'High', 'Low', 'Vol']

RandomForestRegressor

50000

40000

50000

40000

30000

20000

10000

instead of May.

Read the data
train = cardano hst

train_y = train.Price

Create training predictors data
train_X = train[predictor_cols]

my_model = RandomForestRegressor()
my_model.fit(train_X, train_y)

In [154...

30000