Bitcoin Analysis

Raw Data

4 CSV files:

- btc_usd: raw data from bitcoin over time by minute from 2012 to 2017
- btc_hacks: data related to reported hacks
- S&P500 and NASDAQ: files with data over time for those indicators

Data importing from csv to PostgreSQL

- Data imported from flat file to a DB.
 - Helps accessing and save data transformation intermedium steps
 - This is done for all 4 CSVs each to a table.



	123 timestamp_value 📆	123 open 7 ‡	123 high 📆	123 low 📆	123 close TI	123 volume_btc 🏋	123 volume_currency \(\forall^2\)	123 weighted_price \\$:
1	1,325,323,200	4.3900	4.3900	4.3900	4.3900	0.45558087	2.000000193	4.3900
2	1,325,323,260	4.3900	4.3900	4.3900	4.3900	0.45558087	2.000000193	4.3900
3	1,325,323,320	4.3900	4.3900	4.3900	4.3900	0.45558087	2.000000193	4.3900
4	1,325,323,380	4.3900	4.3900	4.3900	4.3900	0.45558087	2.000000193	4.3900
5	1,325,323,440	4.3900	4.3900	4.3900	4.3900	0.45558087	2.000000193	4.3900
6	1,325,323,500	4.3900	4.3900	4.3900	4.3900	0.45558087	2.000000193	4.3900
7	1,325,323,560	4.3900	4.3900	4.3900	4.3900	0.45558087	2.000000193	4.3900
8	1,325,323,620	4.3900	4.3900	4.3900	4.3900	0.45558087	2.000000193	4.3900
9	1,325,323,680	4.3900	4.3900	4.3900	4.3900	0.45558087	2.000000193	4.3900
10	1,325,323,740	4.3900	4.3900	4.3900	4.3900	0.45558087	2.000000193	4.3900

btc_usd: Preprocessing

- Convert timestamp to datetime
- Resample data from by min to by day:
 - Low = min(low)
 - High = max(high)
 - Volume = sum(volume)
 - Value = mean(value)
 - Open = first(open)
 - Close = last(close)
- Created some statistics about the day:
 - Median, mode, standard deviation, kurtosis and skewness

	123 timestamp_value 📆	123 open ₹ ‡	123 high 🟋 🕽	123 low T‡	123 close ₹‡	123 volume_btc ₹‡	123 volume_currency 📆	123 weighted_price 📆
1	1,325,323,200	4.3900	4.3900	4.3900	4.3900	0.45558087	2.000000193	4.3900
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9	1,325,323,680	4.3900	4.3900	4.3900	4.3900	0.45558087	2.000000193	4.3900
10	1,325,323,740	4.3900	4.3900	4.3900	4.3900	0.45558087	2.000000193	4.3900

Column Name	#	Data type
to date_	1	<u>timestamp</u>
123 high	2	float8
123 low	3	float8
123 volume_btc	4	float8
123 volume_currency	5	float8
123 weighted_price	6	float8
123 std_weighted_price	7	float8
¹²³ open	8	float8
123 close	9	float8
123 median_weighted_price	10	float8
123 mode_weighted_price	11	float8
123 kurtosis_weighted_price	12	float8
123 skewness_weighted_price	13	float8

btc_usd: Processing

- Create some ratio variables:
 - A ratio between the high and low in the day
 - A ratio between the open and close value
 - A ratio between today and yesterday mean value

Column Name	#	Data type	₩
🕲 date_	12	<u>timestamp</u>	
123 perc_close_open	1	float8	
123 open	2	float8	
123 close	3	float8	
123 perc_high_low	4	float8	
123 high	5	float8	
123 low	6	float8	
123 perc_weighted_price	7	float8	
123 weighted_price	8	float8	
123 std_weighted_price	9	float8	
123 volume_btc	10	float8	
123 volume_currency	11	float8	
123 median_weighted_price	17	float8	
123 mode_weighted_price	18	float8	
123 kurtosis_weighted_price	19	float8	
123 skewness weighted price	20	float8	

NASDAQ and sp500: Preprocessing

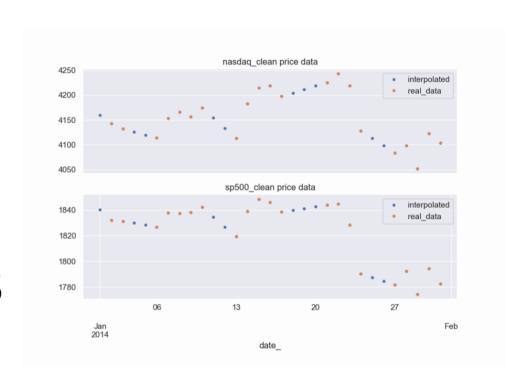
- Clean and convert some variables:
 - Vol
 - Change_percentage
- Resample daily to be able to fill weekends
- Created some ratios:
 - A ratio between the high and low in the day
 - A ratio between today and yesterday mean value



		123 price ₹‡	123 open ₹‡	123 high ₹‡	123 low ₹‡	123 volume_currency 📆	¹²³ perc_price ₹‡
21	2009-07-14 00:00:00	1,799.7300	1,790.7000	1,800.7300	1,782.8500	583,380,000.0000	0.0036
22	2009-07-15 00:00:00	1,862.9000	1,827.6100	1,863.2900	1,824.3100	788,230,000.0000	0.0351
23	2009-07-16 00:00:00	1,885.0300	1,856.0500	1,887.8900	1,854.7000	657,840,000.0000	0.0119
24	2009-07-17 00:00:00	1,886.6100	1,880.3100	1,887.3400	1,873.8100	630,410,000.0000	0.0008
25	2009-07-18 00:00:00	1,929.2600	1,912.9850	1,930.9200	1,909.7350	667,640,000.0000	0.0221069218
26	2009-07-19 00:00:00	1,957.5370	1,942.7120	1,959.3630	1,940.0620	668,710,000.0000	0.0144451931
27	2009-07-20 00:00:00	1,909.2900	1,896.9900	1,909.8900	1,890.0000	618,050,000.0000	0.0120
28	2009-07-21 00:00:00	1,916.2000	1,917.3900	1,917.4600	1,892.1700	650,100,000.0000	0.0036
29	2009-07-22 00:00:00	1,926.3800	1,908.8300	1,934.6100	1,907.2500	676,120,000.0000	0.0053
30	2009-07-23 00:00:00	1,973.6000	1,924.9600	1,979.3400	1,924.7700	869,630,000.0000	0.0245
31	2009-07-24 00:00:00	1,965.9600	1,943.2300	1,966.2600	1,937.6400	644,280,000.0000	-0.0039

NASDAQ and sp500: weekends

- 1st approach: interpolation
 - Suffers from data leakage because we need next Monday
- 2nd approach: linear regression using last 5 days



btc_hacks: Preprocessing

- Cleaning loss_usd
- Resample:
 - Create a variable to count days since the last hack
 - Create a variable which is the value of the last hack divided by the days without hacks



		123 loss_usd 🏋‡	123 lost_by_days_without_hacks \(\forall^2\)	123 days_without_hacks 🏋:
1	2011-06-13 00:00:00	500,000.0000	500,000.0000	0.0000
2	2011-06-14 00:00:00	[NULL]	250,000.0000	1.0000
3	2011-06-15 00:00:00	[NULL]	166,666.666666667	2.0000
4	2011-06-16 00:00:00	[NULL]	125,000.0000	3.0000
5	2011-06-17 00:00:00	[NULL]	100,000.0000	4.0000
6	2011-06-18 00:00:00	[NULL]	83,333.3333333333	5.0000
7	2011-06-19 00:00:00	[NULL]	71,428.5714285714	6.0000
8	2011-06-20 00:00:00	[NULL]	62,500.0000	7.0000
9	2011-06-21 00:00:00	[NULL]	55,555.555555556	8.0000
10	2011-06-22 00:00:00	[NULL]	50,000.0000	9.0000
11	2011-06-23 00:00:00	[NULL]	45,454.5454545455	10.0000
12	2011-06-24 00:00:00	[NULL]	41,666.666666667	11.0000
13	2011-06-25 00:00:00	[NULL]	38,461.5384615385	12.0000

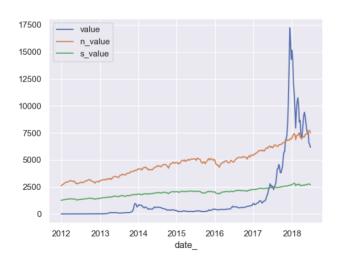
Data fusion in one table

- Data is then joined in one table using the date as a global key.
- Adding prefixes to avoid columns with the same name:
 - b_: btc_usd and btc_hacks
 - s_: sp500
 - n_: nasdaq

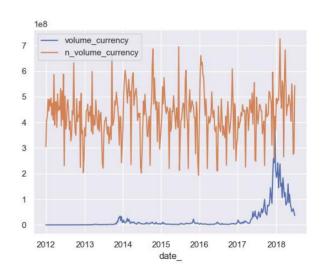
Column Name	#	Data type
¹²³ b_perc_close_open	1	float8
¹²³ b_open	2	float8
123 b_close	3	float8
¹²³ b_perc_high_low	4	float8
123 b_high	5	float8
123 b_low	6	<u>float8</u>
123 b_perc_value	7	float8
123 b_value	8	float8
123 b_std_value	9	float8
¹²³ b_volume_btc	10	float8
123 b_volume_currency	11	float8
🔇 date_	12	<u>timestamp</u>
123 b_median_value	13	float8
123 b_mode_value	14	float8
123 b_kurtosis_value	15	float8
123 b_skewness_value	16	float8
¹²³ n_open	17	float8
¹²³ n_perc_high_low	18	float8
123 n_high	19	float8
123 n_low	20	float8
123 n_perc_value	21	float8
123 n_value	22	float8
123 n_volume_currency	23	float8
123 s_open	24	float8
123 s_perc_high_low	25	float8
123 s_high	26	float8
123 S_low	27	float8
123 s_perc_value	28	float8
123 s_value	29	float8
123 b_loss_usd	30	float8
123 b_lost_by_days_without_hacks	31	float8
123 b_days_without_hacks	32	float8

Dataset analysis

Bitcoin value vs stock value



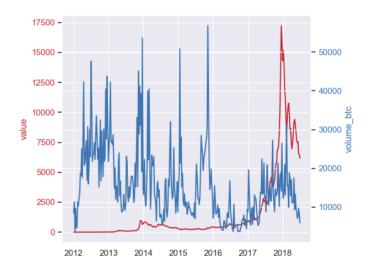
 Bitcoin volume vs Nasdaq Volume



Bitcoin value vs volume

Vs volume in currency
Vs volume in btc units



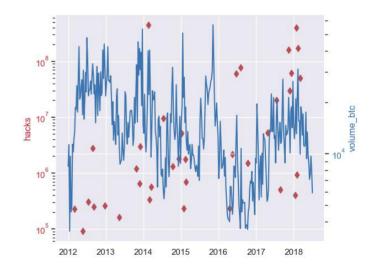


Bitcoin vs Hacks

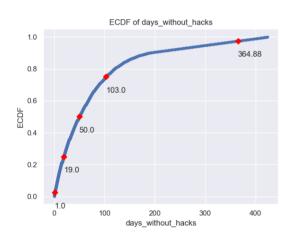
Value vs Hacks

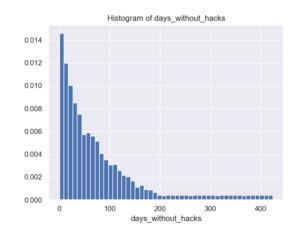


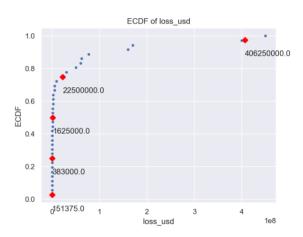
 Volume in btc vs hacks



Bitcoin vs Hacks







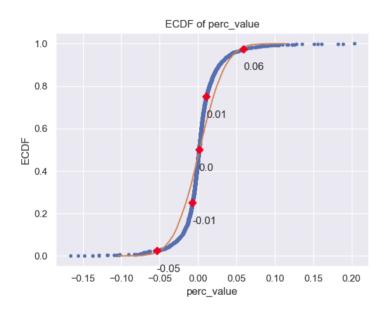
Correlation

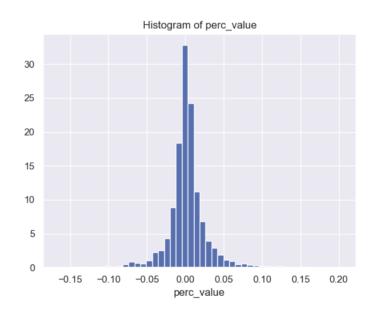
								ation bet										
perc_close_open 1 -0.0	3 0	-0.22	-0.01	-0.01	0.79	-0.01	-0.04	-0.09	-0.05	-0.01	-0.01	-0.04	0.13	-0.04	0.01	0.03	-0.01	-0.03
open -0.03 1	1	0.16	1	1	-0.06	1	0.84	-0.05	0.87	1	1	-0.07	0.05	-0		-0.02		0
dose 0 1	1	0.15	1	1	-0.03	1	0.84	-0.06	0.86	1	1	-0.07	0.05	-0	-0	-0.01	0.63	0
perc_high_low -0.22 0.1	6 0.15	1	0.16	0.14	-0.02	0.15	0.31	0.45	0.29	0.15	0.16	0.02	-0.04	-0.04	-0.03	-0.07	-0.03	-0.08
high -0.01 1	1	0.16	1	1	-0.04	1	0.85	-0.05	0.87	1	1	-0.07	0.05	-0	-0	-0.02		-0
low -0.01 1	1	0.14	1	1	-0.03	1	0.82	-0.07	0.85	1	1	-0.07	0.05		-0	-0.02		0
perc_value 0.79 -0.0	6 -0.03	-0.02	-0.04	-0.03	1	-0.04	-0.07	-0.05	-0.09	-0.04	-0.04	-0.02	-0.03	-0.07		0.01	-0.04	-0.06
value -0.01 1	1	0.15	1	1	-0.04	1	0.84	-0.06	0.86	1	1	-0.07	0.05	-0		-0.02	0.63	0
std_value -0.04 0.8	4 0.84	0.31	0.85	0.82	-0.07	0.84	1	0.09	0.92	0.84	0.84	-0.07	0.03	-0	-0.01	-0	0.45	-0.02
volume_btc -0.09 -0.0	5 -0.06	0.45	-0.05	-0.07	-0.05	-0.06	0.09	1	0.14	-0.06	-0.06	-0.01	-0.02	-0.11	-0.04	0.11	-0.3	-0.16
volume_currency -0.05 0.8	7 0.86	0.29	0.87	0.85	-0.09	0.86	0.92	0.14	1	0.86	0.87	-0.06	0.03	-0.03	-0.01	0.01	0.5	-0.06
median_value -0.01 1	1	0.15	1	1	-0.04	1	0.84	-0.06	0.86	1	1	-0.07	0.05	-0	-0	-0.02		-0
mode_value -0.01 1	1	0.16	1	1	-0.04	1	0.84	-0.06	0.87	1	1	-0.07	0.05	-0	-0	-0.02		-0
kurtosis_value -0.04 -0.0	7 -0.07	0.02	-0.07	-0.07	-0.02	-0.07	-0.07	-0.01	-0.06	-0.07	-0.07	1	-0.39	-0.02	0.01	-0.04	-0.1	-0.03
skewness value 0.13 0.0	5 0.05	-0.04	0.05	0.05	-0.03	0.05	0.03	-0.02	0.03	0.05	0.05	-0.39	1	0.03	-0.01	0.04	0.11	0.04
weekday -0.04 -0		-0.04	-0		-0.07	-0	-0	-0.11	-0.03	-0	-0	-0.02	0.03	1			-0	0.79
day of month 0.01 -0	-0	-0.03	-0	-0	0	-0	-0.01	-0.04	-0.01	-0	-0	0.01	-0.01	-0	1	0.01	-0.01	-0
month 0.03 -0.0	2 -0.01	-0.07	-0.02	-0.02	0.01	-0.02	-0	0.11	0.01	-0.02	-0.02	-0.04	0.04	0	0.01	1	-0.11	0
vear -0.01 0.6		-0.03	0.62	0.63	-0.04	0.63	0.45	-0.3	0.5	0.63	0.63	-0.1	0.11	-0	-0.01	-0.11	1	-0
weekend -0.03 0		-0.08	-0	0	-0.06	0	-0.02	-0.16	-0.06	-0	-0	-0.03	0.04	0.79	-0	0	-0	1
open	dose	oerc_high_low	high	MO	perc_value	value	std_value	/olume_btc	ency	median_value	mode_value	kurtosis_value	value	weekday	day_of_month	month	year	weekend
ose		high_			erc		std	dum	uno	lan	ode	SIS	SSE	wee	o o	=		Wee
oerc_close_open		perc			Ф			>	olume_currency	med	Ĕ	kurtc	skewness_value		day			

Correlation

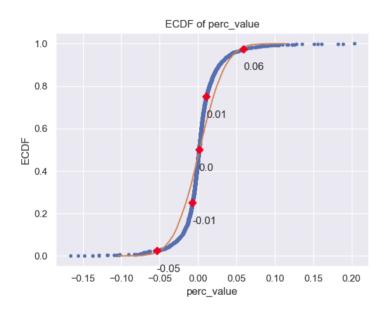


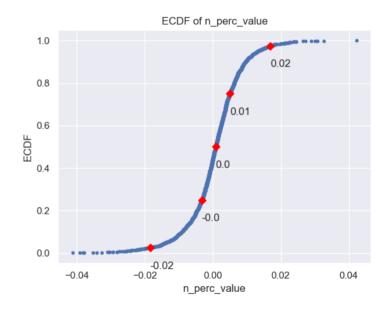
Bitcoin change in value by day





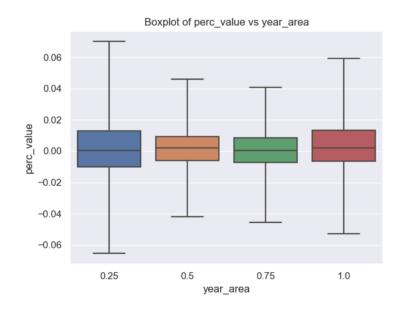
Bitcoin vs Nasdaq change value



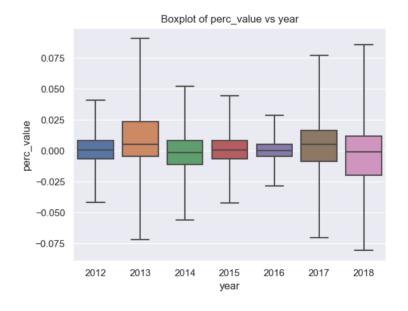


Bitcoin change value by aggregation

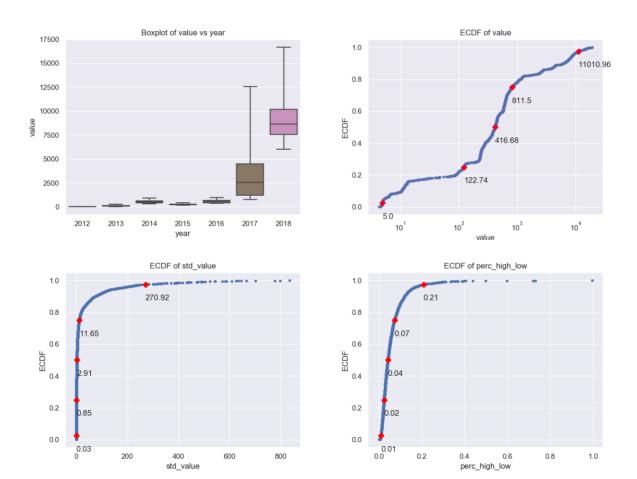
By year area



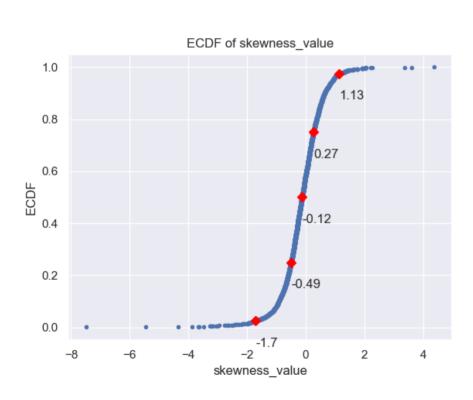
By year value

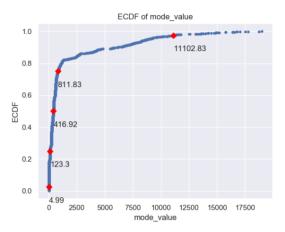


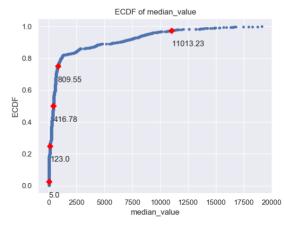
Bitcoin value



Bitcoin: Skewness







Bitcoin: volume

