

PREDICT \$OCEAN PRICE WITH TWEET SENTIMENTS

**OCEAN DATA CHALLENGE :
OCEAN TOKEN SENTIMENT ANALYSIS**

CONTENTS

- ▶ Context
- ▶ The question and its interpretation
- ▶ Data exploration
- ▶ Data reduction with auto encoder
- ▶ LSTM model to predict
- ▶ Conclusions
- ▶ Axes of improvement

CONTEXT

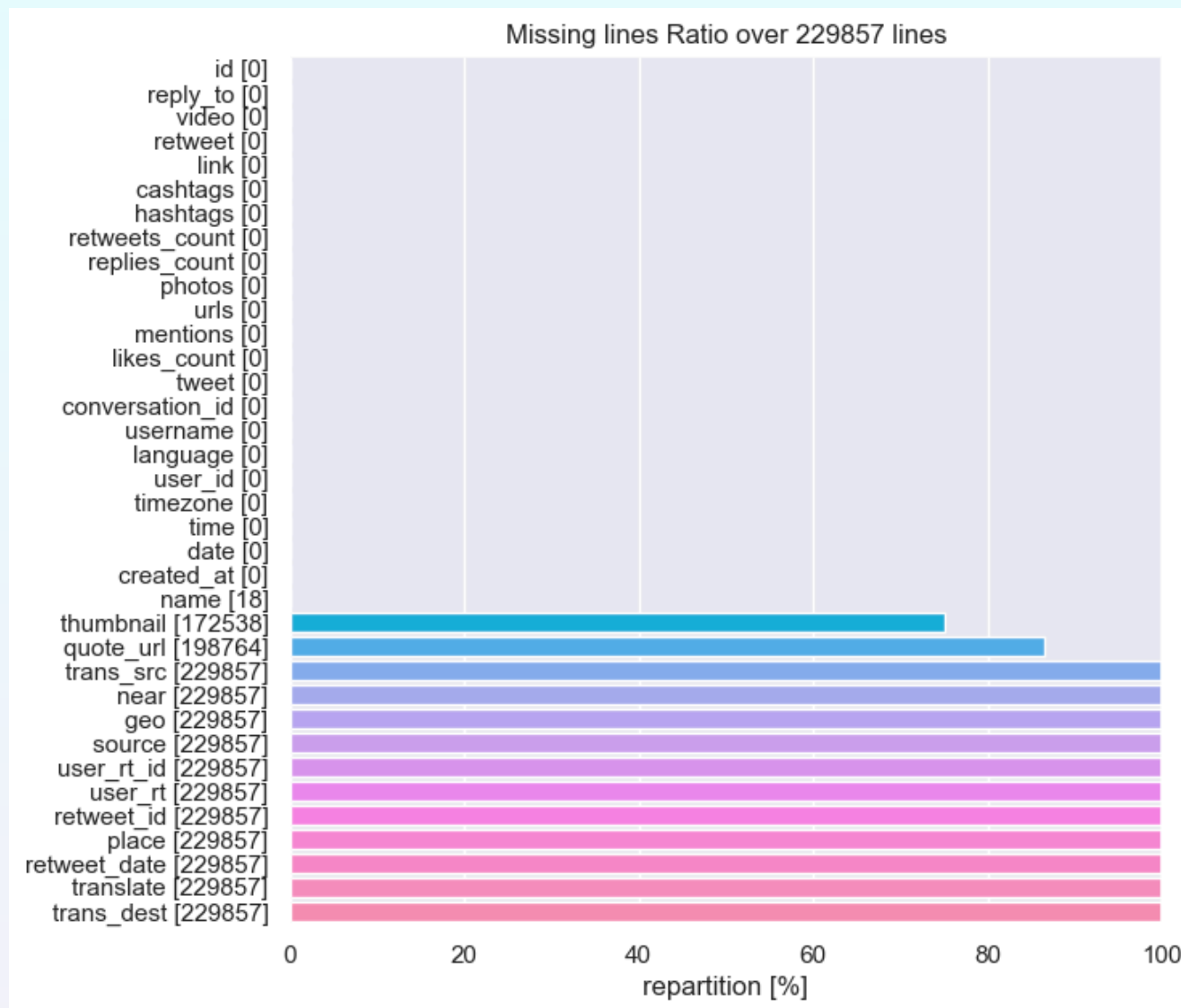
- ▶ Ocean Data Challenge
 - ▶ predict \$OCEAN price with twitter sentiment
- ▶ Data used:
 - ▶ OCEAN token price dataset
 - ▶ OCEAN tweets dataset
 - ▶ not other data!

THE QUESTION AND ITS INTERPRETATION

- ▶ Find correlation with price and:
 - ▶ number of tweets
 - ▶ number of likes
 - ▶ number of retweets
 - ▶ number of individuals
 - ▶ number of influential tweets
- ▶ Model of twitter sentiment
 - ▶ Extract sentiment from tweet with a model to classify tweets
- ▶ Find correlation with price evolution

EXPLORATION : THE DATA : TWEETS

- keep columns with NaN less than 90%



229857 TWEETS

ELIMINATE

DATE UTC+3 != DATE PRICES
=> NEED TO SYNCHRONISE

EXPLORATION : THE DATA : PRICES

NB PRICES = 1024 DAYS

- ▶ Open / High / Low / Close / Volume : classical data
- ▶ Date reference : UTC (checked on Tradingview)
- ▶ Date range synchronise with tweets
- ▶ Add some simple insights
 - ▶ Use price variation : $(\text{Close} - \text{Open}) / \text{Open}$
 - ▶ Classification rules

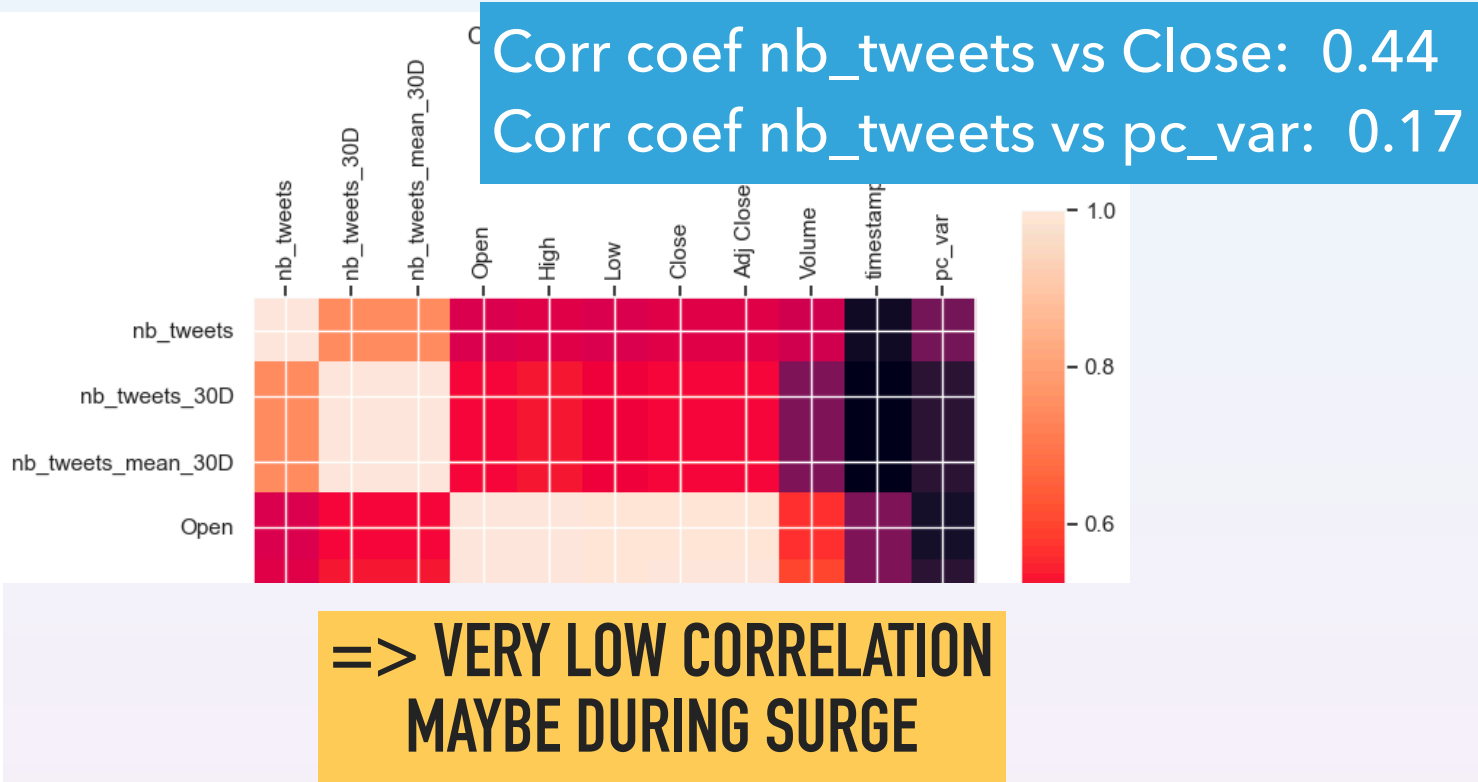
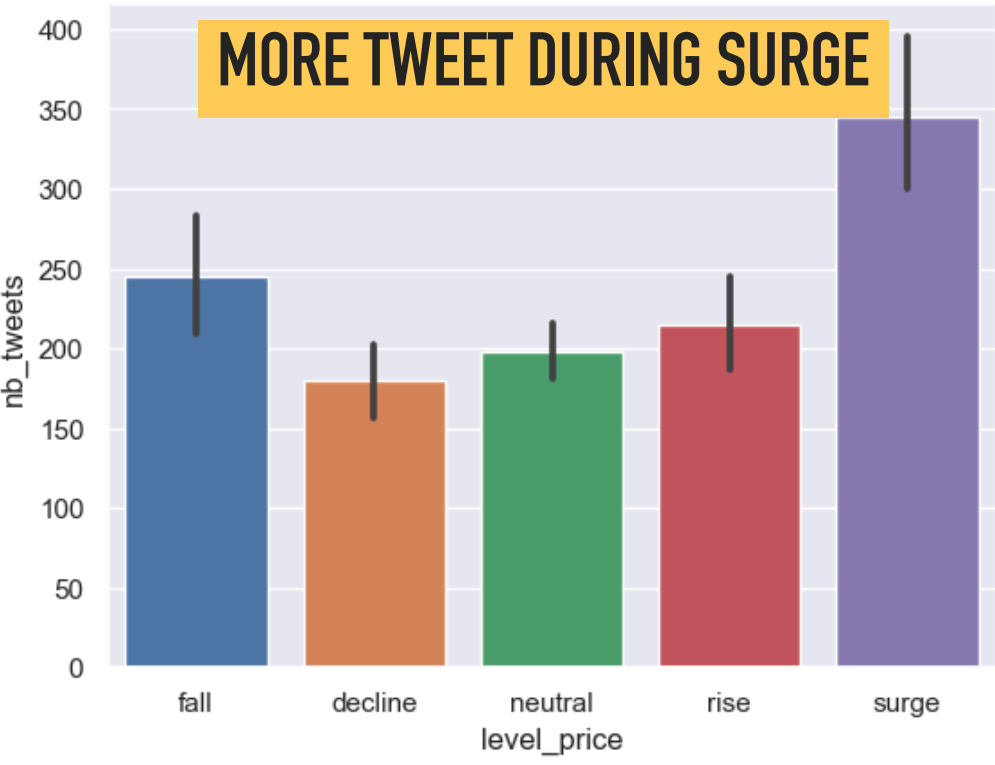
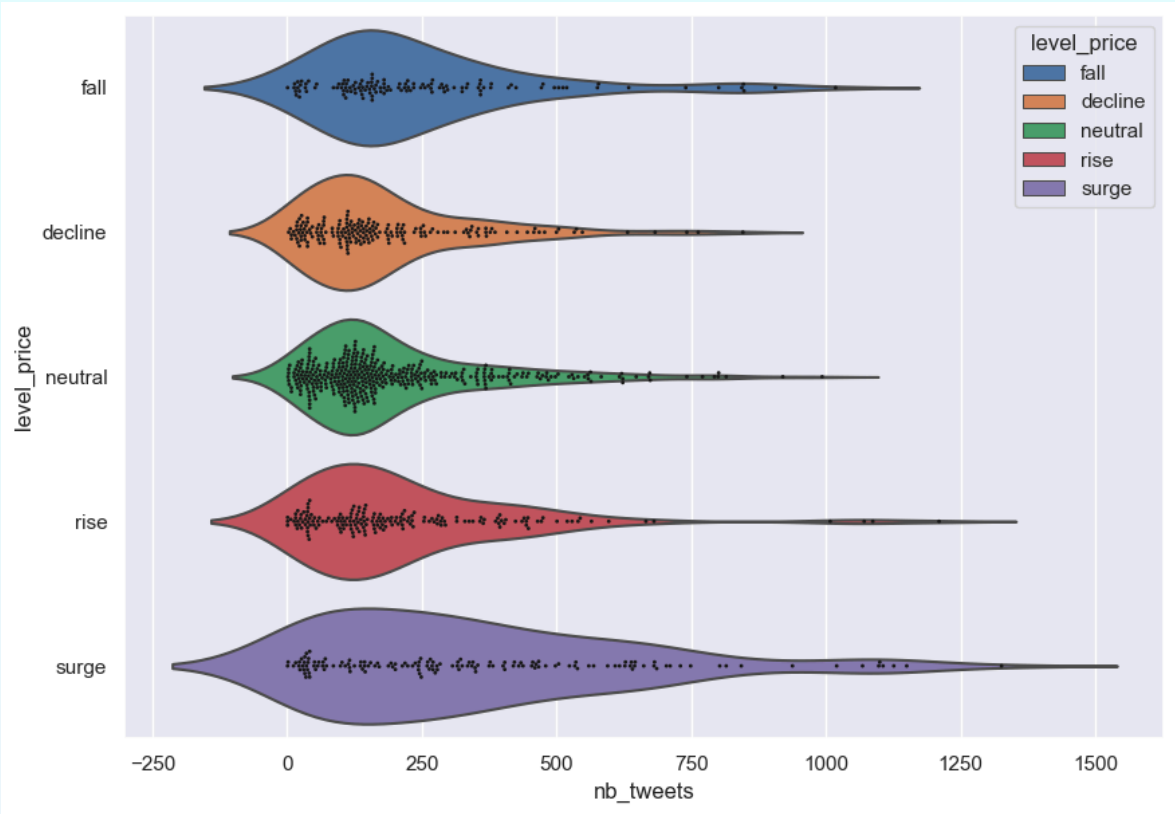
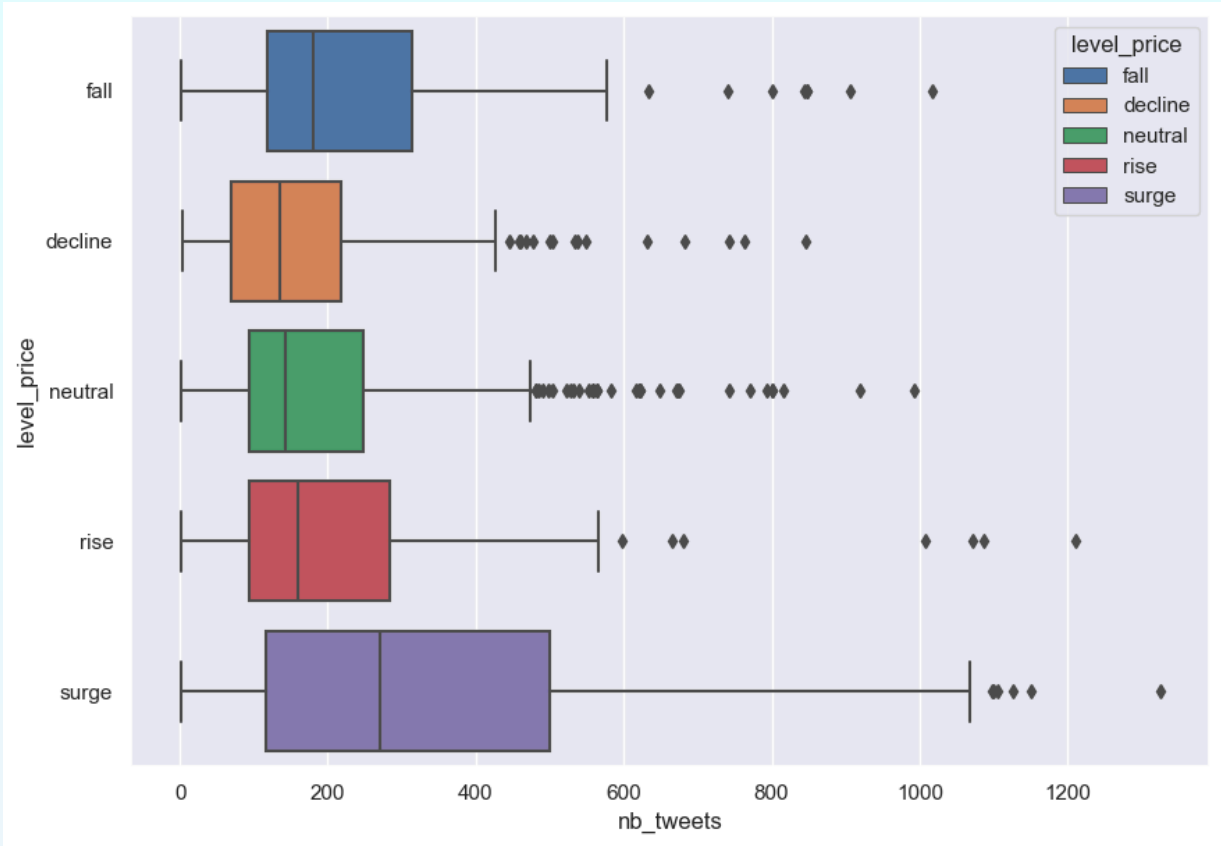
```
Levels of price variation : level_stat_price :  
- neutral : var mean - 1 std < var price < var mean + 1 std  
- bearish : var price < var mean - std  
- bullish : var price > var mean + std
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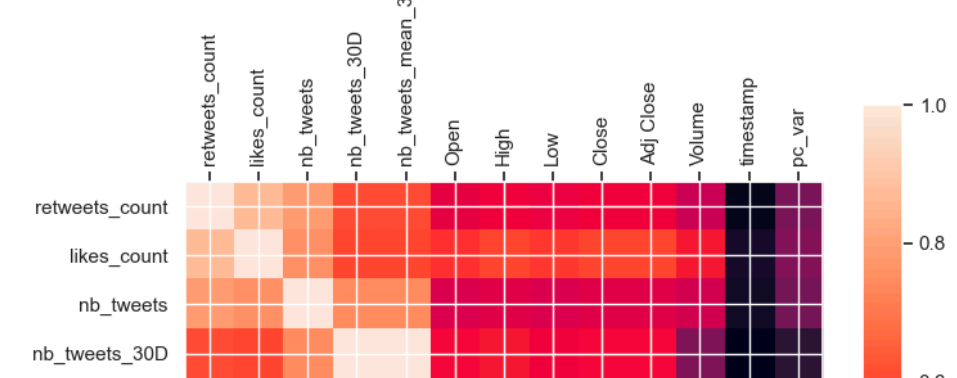
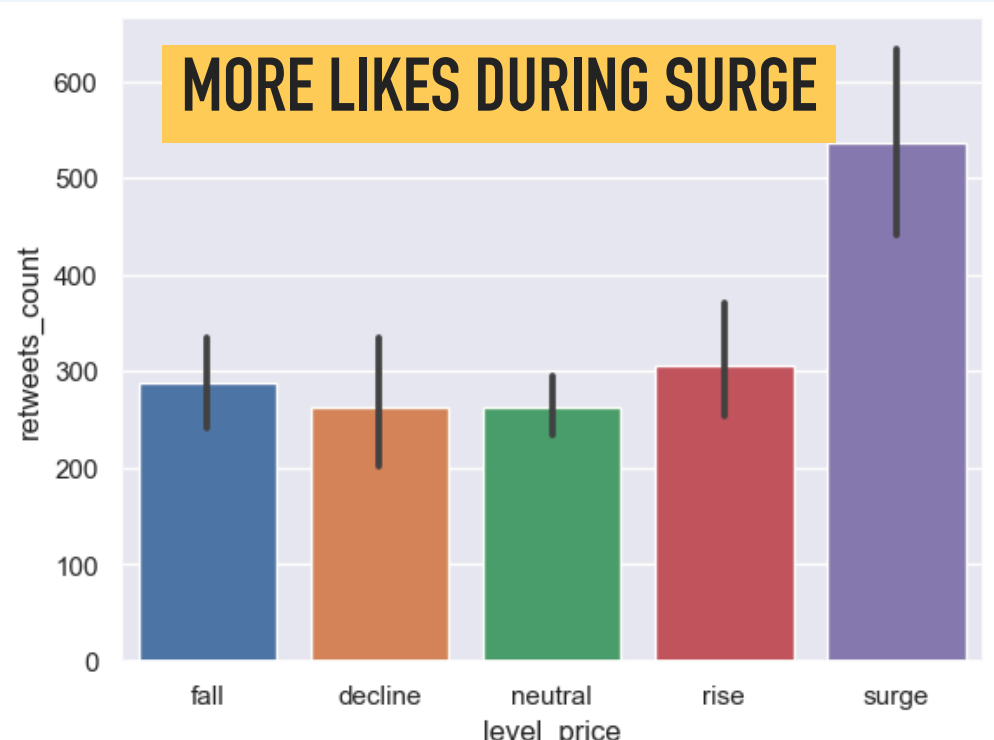
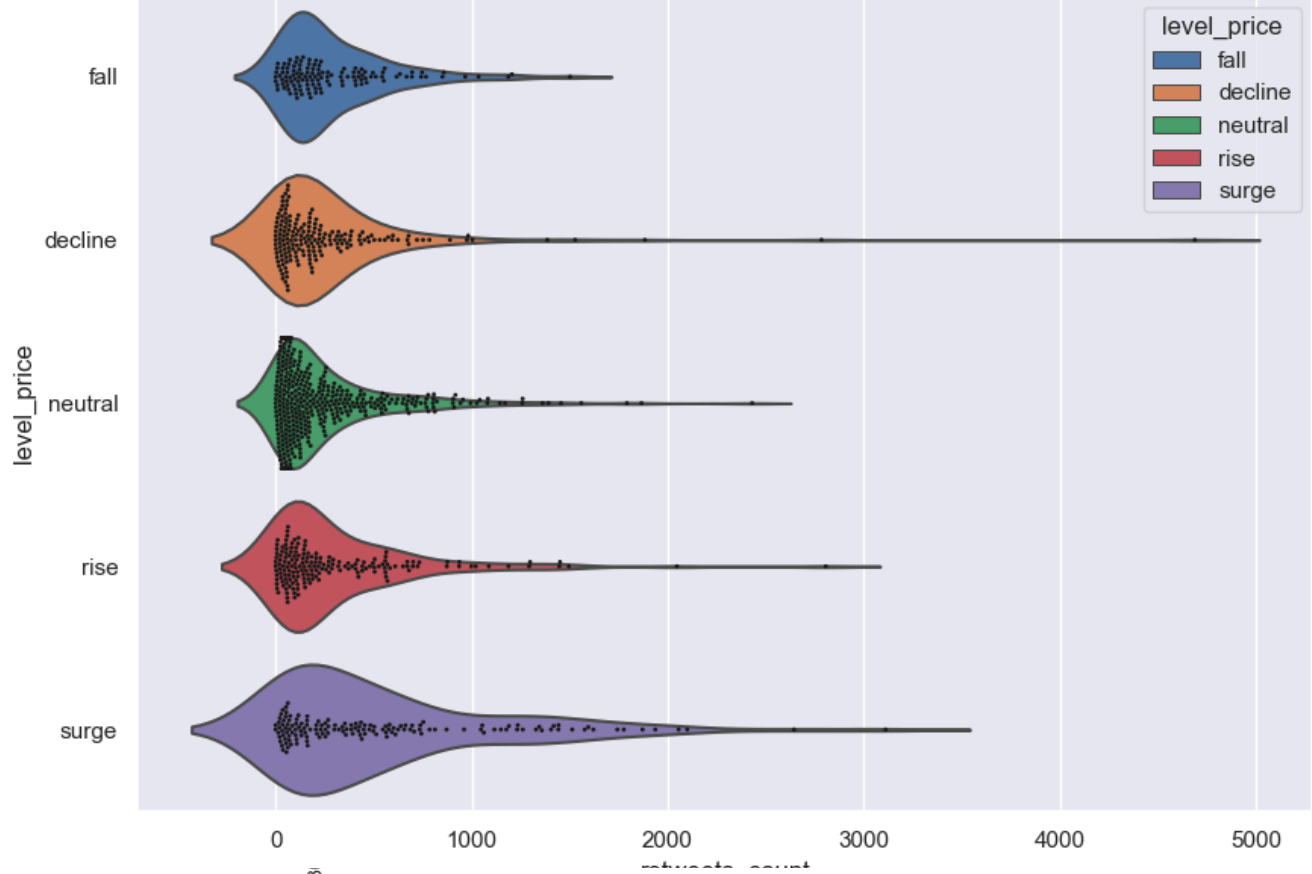
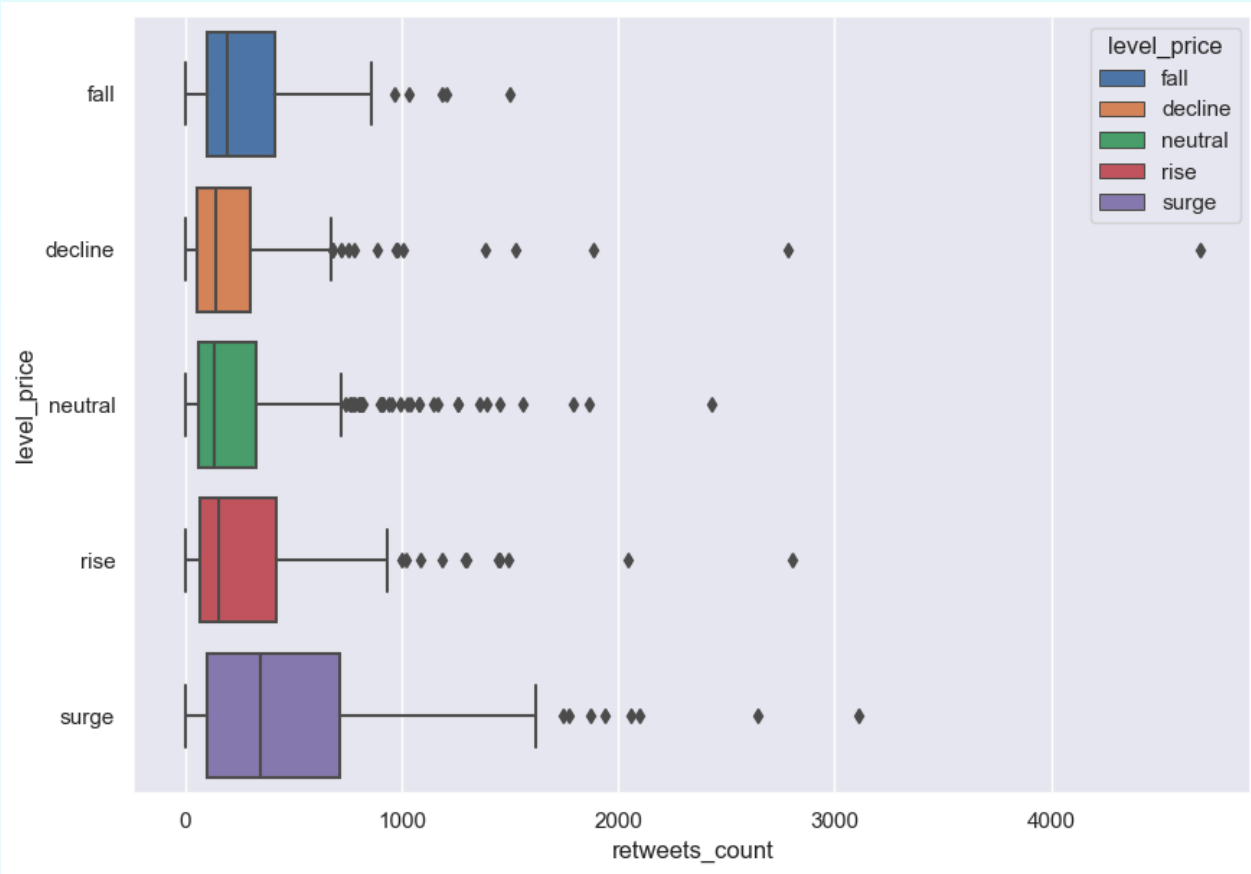
EXPLORATION : MERGE METHOD PRICES & TWEETS

- ▶ Choose to synchronize date for exploration
 - ▶ (no for last part : model prediction)
- ▶ Merge with daily prices :
 - ▶ **Aggregate each day**
 - ▶ with sum
 - ▶ number of tweets / likes / retweets / influential
 - ▶ with nunique
 - ▶ number of individuals
 - ▶ For **influential**, select tweets with **more the 100 likes**

CORRELATION BETWEEN THE PRICE OF \$OCEAN AND THE NUMBER OF TWEETS



CORRELATION BETWEEN THE PRICE OF \$OCEAN AND THE NUMBER OF LIKES



Corr coef nb likes vs Close: 0.59
Corr coef nb likes vs pc_var: 0.21

=> LOW CORRELATION
BUT MORE THAN SIMPLE NB TWWETSMAYBE DURING SURGE

CORRELATION BETWEEN THE PRICE OF \$OCEAN

- ▶ Lower correlation with Nb influential tweets & nb individual

	Close	pc_var	Low	High
nb_influ	0.560370	0.201397	0.537332	0.559453
nb_indiv	0.499816	0.145142	0.476117	0.507835
retweets_count	0.491611	0.180578	0.473187	0.494742
likes_count	0.598143	0.211415	0.573586	0.599180
nb_tweets	0.446507	0.170824	0.421602	0.454216

=> NB LIKES = THE BEST INSIGHT
BUT STILL LOW

MODEL SENTIMENT: SIMPLE TEXTBLOB

- ▶ Use Textblob to extract sentiment in tweets
 - ▶ Desynchronise dates between Tweets and Prices
 - ▶ Tweets = 3 hours before

▶

MODEL SENTIMENT: RESULTS

- ▶ Good to detect neutral
- ▶ Still bad for perish or bullish
 - ▶ predict neutral instead

CONFUSION MATRIX

True label	bearish	neutral	bullish	
	23	71	12	
	114	567	132	
	10	84	8	
		Predicted label		
		bearish	neutral	bullish

MODEL SENTIMENT: RESULTS

- ▶ Sorry no time to finish !!!

