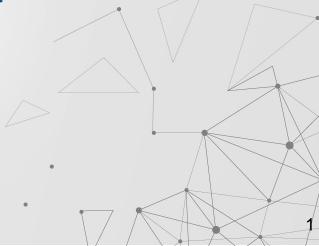
# **Automated Trading Agent** in Cryptocurrency Market

**Group 7 Final Project** 

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# **Agenda**

- 1. Motivation
  Why we applied reinforcement learning into cryptocurrency trading
- 2. Objectives
  The objectives of this study in business and analytics aspect
- Proposed Approach

  How we require data, craft impactful features, and applied applied double actor-critic architecture with recurrent neural network
- **Experiments**Demonstrate our experiment result, and demo how our trading agent works
- 5. Conclusion
  Recap this study, and discuss with future work to improve

#### **Motivation**

## **Pursuing Profits in Cryptocurrency Market**

## **Gigantic Profit Margin in a Short Amount of Time**

Bitcoin worth on \$3451.55 on 2019/2/7, and worth \$12,647,99 on 2019/6/25.It's a **3.6X** growth in about 150 days.

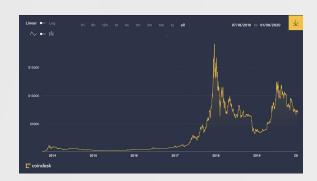
This tremendous growth allures many greedy investors eager to make profit in Cryptomarket.

## **Extremely High Volatility of Crypto as well...**

However, large profit margin comes along with high volatility. The volatility for investor is a kind of double-edge sword. Investors needs to endure high risk of losing a significant amount of investment.

### The Precise Decision Policy is required

In order to hit the jackpot in cryptocurrency market, investors need a precise and reasonable policy to decide when to long or short in the right timing.



### **Objectives**

## **Business & Analytics Goals**

### **Business Goals**

Maximize the value of portfolios (crypto & money) at the end of a given time.

## **Analytics Goals**

 To develop an Automated Trading Agent able to learn the profitable trading policy, without any human interventions.

### **Data Description**

## Bitcoin / Ethereum / Litecoin Daily Price

### BTC / ETH / LTC

We select Bitcoin / Ethereum / Litcoin as our experiment.

These three cryptos are the top three largest

cryptocurrency market capital. The underlying

mechanisms are adapted to other many cryptocurrencies.

Therefore, we expect this study could apply to other scenarios.

Source: coindesk

Column: Open, High, Low, Close, Volume

Training Set: 2016/5/17 - 2018/12/31

Test Set: 2019/1/1 - 2020/1/6



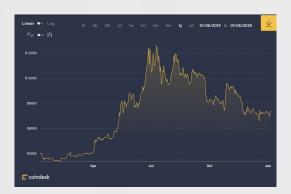




Bitcoin (BTC)

Ethereum (ETH)

Litecoin (LTC)



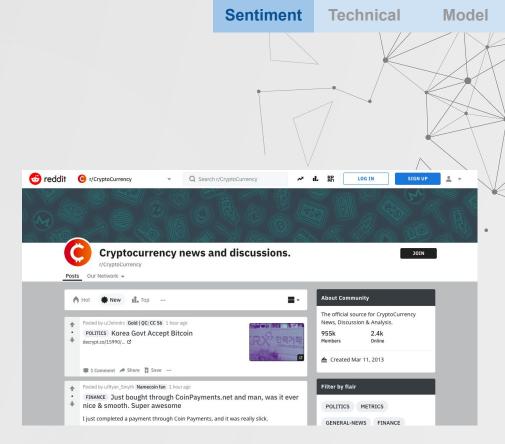
2019 Bitcoin Price Chart

# Proposed Approach Sentiment Analysis

**Source :** Reddit Cryptocurrency posts

Date: 2016/5/17 - 2020/1/6

**Number of Posts**: 1,671,326



**Screenshot of Reddit Cryptocurrency Forum** 

## **Sentiment Analysis**

Introduce a Market Sentiment Index to reflect cryptocurrency market dynamics

- People tend to get greedy when the market is bullish which leads to FOMO phenomenon.
- People sell their currency irrationally in bearish.

#### **Flowchart**



### **Proposed Approach**

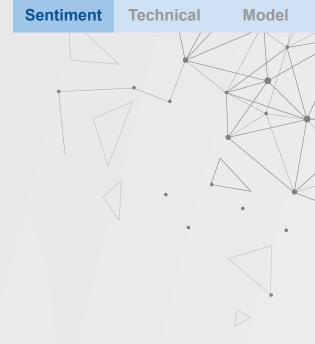
## **Sentiment Analysis**

### **Data preprocessing**

- Remove punctuations, urls ... etc.
- Lemmatizing (playing -> play, looked -> look)
- Lower case

### Sentiment score

- Using pretrained sentiment classifier
- Words Example : Appreciate: Positive, Poor: Negative, Glad: Positive ... etc.
- Sentence Example: "SoftBank, the 62nd largest company in the world and Japan's biggest internet and telecommunications corporation has co-launched a decentralized blockchain development contest for entrepreneurs and startups with Topcoder" -> Positive



### **Proposed Approach**

## **Technical Indicator**

## Up to 7 meaningful indicators

- Moving Average
- Stochastic Oscillator
- MACD
- Relative Strength Index, RSI
- Bollinger Bands
- On Balance Volume
- Aroon



# Moving Average, MA

MA helps smooth out price action by filtering out the "noise" from random short-term price fluctuations.

$$SMA = \frac{A_1 + A_2 + ... + A_{n-1} + A_n}{n}$$

A = average in period nn = number of time periods



## Stochastic Oscillator, KD

A stochastic oscillator is a momentum indicator comparing a particular closing price of a security to a range of its prices over a certain period of time.

$$RSV = \frac{C_n - L_n}{H_n - L_n} \times 100\%$$

$$K_n = \alpha \cdot RSV_n + (1 - \alpha) \cdot K_{n-1}$$

$$D_n = \alpha \cdot K_n + (1 - \alpha) \cdot D_{n-1 \text{ (usually } \alpha = 1/3)}$$

n = trading period(usually n = 9)

C=the last closing price

L = the lowest price over the last n period

 $H = the \ highest \ price \ over \ the \ last \ n \ period$ 



# Moving Average Convergence/ Divergence, MACD

MACD shows the relationship between two moving averages of a security's price.

EMA: Exponential moving average

Difference  $DIF = EMA_{12} - EMA_{26}$ 

Signal  $DEM = MACD = EMA_9$ 

Histogram OSC = DIF - DEM



# Relative Strength Index, RSI

RSI measures the magnitude of recent price changes to evaluate overbought or oversold conditions

$$RS = \frac{EMA_{Up,n}}{EMA_{Down,n}}$$

Relative Strength Index

$$RSI = \frac{EMA_{Up,n}}{EMA_{Up,n} + EMA_{Down,n}} \times 100\%$$



## **Bollinger Bands, B-Bands**

A technical analysis tool defined by a set of lines plotted two standard deviations (positively and negatively) away from a simple moving average (SMA) of the security's price.

$$BOLU = MA(TP, n) + m * \sigma[TP, n]$$
$$BOLD = MA(TP, n) - m * \sigma[TP, n]$$

BOL 
$$U$$
 = Upper Bollinger Band  
BOLD = Lower Bollinger Band  
 $MA$  = Moving average  
 $TP(typical\ price) = (High + Low + Close) \div 3$   
 $n = Number\ of\ days\ in\ smoothing\ period(typically\ 20)$   
 $m = Number\ of\ standard\ deviations(typically\ 2)$   
 $\sigma[TP, n] = Standard\ Deviation\ over last\ n\ periods\ of\ TP$ 



### **Technical Indicator**

## **On-Balance Volume, OBV**

On-Balance Volume uses volume flow to predict changes in stock price.

$$OBV = OBV_{prev} + \begin{cases} volume, & if \ close = close_{prev} \\ 0, & if \ close = close_{prev} \\ -volume, & if \ close = close_{prev} \end{cases}$$

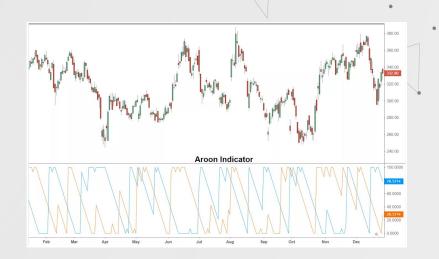
 $OBV = Current \ on - balance \ volume \ level$   $OBV_{prev} = Previous \ on - balance \ volume \ level$   $Volume = latest \ trading \ volume \ amount$ 

## **Aroon Indicator**

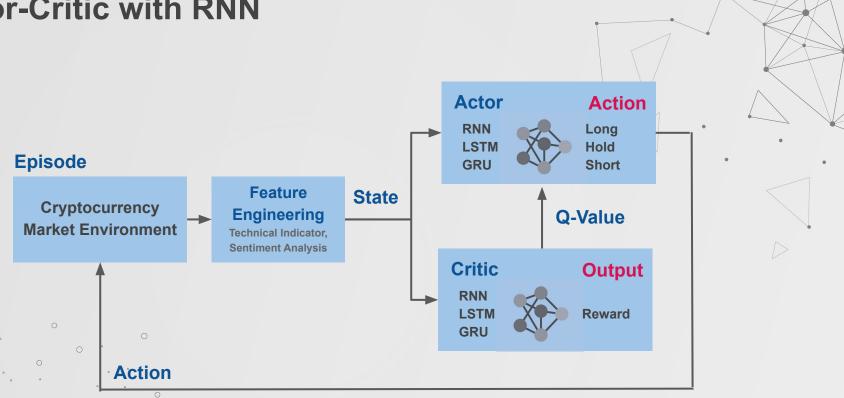
Aroon indicator is used to identify trend changes in the price of an asset, as well as the strength of that trend.

Aroon 
$$Up = \frac{25 - period \ since \ 25 \ period \ high}{25} \times 100$$

$$Aroon \ Down = \frac{25 - period \ since \ 25 \ period \ low}{25} \times 100$$



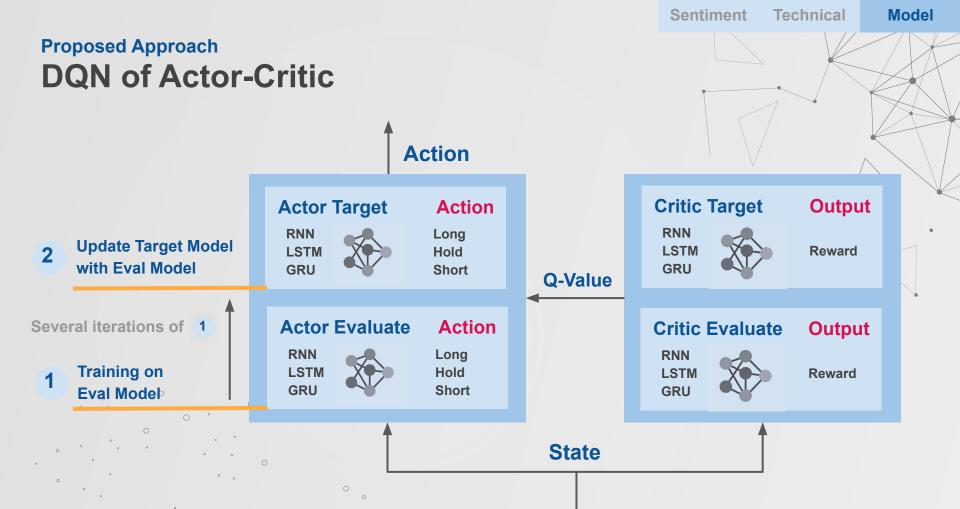
# Proposed Approach Actor-Critic with RNN



Model

**Technical** 

**Sentiment** 



# **Agent Trading Actions**



### Long

Agent expects the price of crypto will rise in the future, therefore he purchase crypto at current price.

Sentiment

**Technical** 

Model



### Hold

Agent considers that there is no profit margin in the near future.



### Short

Agent believes the price of the crypto will decrease in the future. Anticipating to make a profit by buying the crypto at the lower price.

## **Trading Agent Performance**

## **BTC**



**Best Model** 

Model: GRU

**ROI: +3.033%** 



# **Trading Agent Performance**

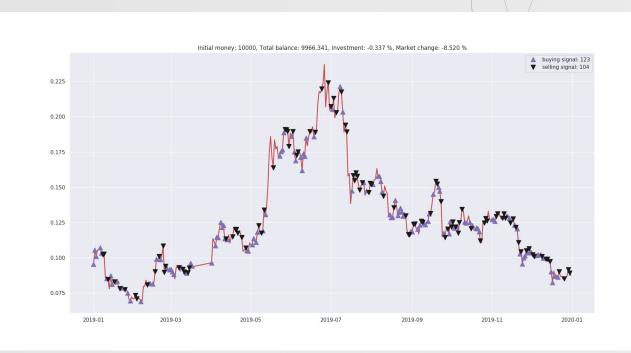
## ETH



**Best Model** 

Model: GRU

**ROI: -0.337%** 



# **Trading Agent Performance**

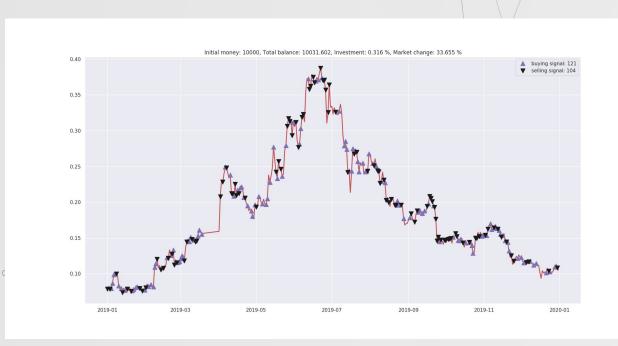
## Litecoin



**Best Model** 

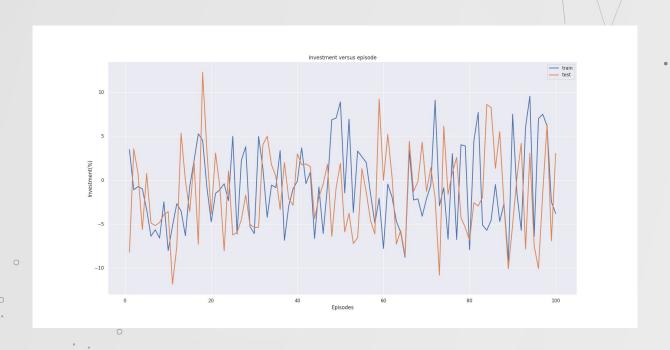
Model: GRU

**ROI: +0.316%** 



# Performance through Episodes: Bitcoin





### **Conclusions**

## **Future Works**

### **Improve Model Architecture**

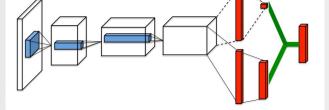
### Attention Mechanism

Enable model to have the ability to assign weights to input feature accordingly.

This mechanism is expected to let the most relevant features to have larger impact to final outcome.

## Dueling Actor-Critic Model

By adjust this type of structure to the network head allows the network to better differentiate actions from one another, and significantly improves the learning.



**Dueling Model** 

### **Conclusions**

## **Future Works**

## **Diversify Cryptocurrencies Investment Portfolios**

Able to trade more or less based on agent's confidence

We simplify cryptocurrencies market actions into only three behavior within one trading unit: Long, Short, Hold. However, the investors in real world have nearly infinite actions that they can trading for more than one units based on their forecasting confidence.

Include other cryptocurrencies into agent investment target

So far, our agent can learn and trade with one crypto at a time. This means our agent can only seek profit opportunity in one specific target. By this constraint, we may miss great amount of opportunities. Therefore, we expected our agent are able to take multiple cryptos into consider at same time, and to maximize profits of investment portfolios.

### **Conclusions**

## Recap

- Applied sentiment analysis and technical indicators to reflect crypto market dynamics.
- Developed an Automated Trading Agent based on actor-critic reinforcement
   learning architecture and various recurrent neural network.
- Our last episode agent are able to achieve up to ROI 3 % on BTC, 0.3% on LTC
   -0.337% on ETH under high volatility crypto market.

# Proposed Approach Sentiment Analysis

				V	
Classifier	Precision	Recall	F1	Accuracy	
Multinomial NB	0.67	0.51	0.56	0.51	
LSTM					