

Prediction of Bitcoin Price Using Bi-LSTM Network

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Abstract— Machine Learning and Artificial Intelligence based money exchanging have pulled in enthusiasm in the recent years with the introduction of Bitcoins. The cost of Bitcoins has increased in a large scale and it is fairly difficult to predict the future cost per Bitcoin. In this study, we utilize a machine learning and deep learning model to analyze the digital currency market to predict the cost of Bitcoin per day. We dissect everyday information for 1,691 cryptographic forms of money for the period between November 2017 and April 2019. The study shows that straightforward exchanging procedures assisted by best in class AI algorithms have met the standard benchmarks. Our outcomes also show that non-inconsequential, basic algorithmic instruments can help in envision of momentary development of the cryptographic money. The proposed system uses a Bi-directional LSTM for forecasting the bitcoin prices. The proposed model was able to trace the test dataset with Mean Absolute Percentage Error of 13%. The model is helpful for the user to take decision on investing in Bitcoins.

Keywords- Bitcoins, cryptocurrency, LSTM, Deep learning.

I. INTRODUCTION

Bitcoin is online money developed in the year 2009. Bitcoin is operated in the basis of decentralized power to reduce the exchange expenses when compared to the online installments [7]. Bitcoin are virtual and cloud managed open record which are not controlled by any authority or any individual. As the most punctual digital money to meet across the board fame and achievement.

Bitcoin has enlivened a large group of branches and Local People. Cost of the bitcoin is dependent on the size of mining system. When the mining expense rises, bitcoin cost must also increase. Bitcoin prediction monitors the hash rate each second in order to solve the hashing puzzle [8]. On October 23, 2019, hash rate of a second records 114 quintillion. Data analysis and prediction of bitcoin price issues has raised [9]. Researcher have utilized the machine learning algorithms to inverse and analyze the bitcoin prediction.

The prevalence of digital forms of money has soar in 2017 because of a few back to back a very long time of overly exponential development of their market capitalization. Today, there are multiple, 500 effectively exchanged cryptographic forms of money promoting over \ \$300 billion, with a pinnacle of the market capitalization totaling more than \ \$800 billion in Jan. 2018. Major cryptographic forms of money can be purchased utilizing fiat cash in various online trades and afterward be utilized in their chance to purchase less well known digital currencies.

In this paper a protocol or a model to detect the bitcoin activities is proposed. This system is capable of providing most of the essential features required to predict the level of bitcoin in feature. With the rise of machine learning, artificial intelligence and other relevant digital fields of information technology and computers, it becomes feasible to automate this process and to save some of the intensive amount in bitcoin.

The main Objective of this bitcoin price prediction algorithm to help the investors weather to go for this investment or not. We will use different algorithms of machine learning so that we will get accuracy data and we can help well. The data we are using is last 5 years data and main objective is to use previous data and make a better future outcome using that data. We are using BI-LSTM model and different analysis tools to find accuracy. The time series graph is there in the output that will help the investor to know whether to invest or not.

II. BACKGROUND AND RELATED WORK

Kawasmi et al. [1] presents and uncovers to us a course of action of-systems building model and instrument for a D-CETI (Decentralized Carbon Emissions Trading Infrastructure) and with base or pointing on insurance and system security and suspect goals. The lead or yields and the

structure are completed and done and as a response for the issue of trading carbon releases covertly and synchronously out of the trading administrators. Security and Insurance of the trading pro and the carbon credits of them are the principal necessities associated by the structure and part of the D-CETI. The appropriated lead, and the perceived, decentralized and the structure of official of various systems are the two basic characteristics of D-CETI which perceive and make it comparable by the standard carbon trading plans and shows. D-CETI relies on Bitcoin which is a conveyed propelled money and has no central position. It is an Open Transactions that has a system that broadcasts the usage of cryptography in trades associated with budgetary. The frameworks and structure of D-CETI is surveyed, differentiated and the mechanisms, architecture, and procedures for five other carbon releases trading stages.

Casagrande et al. [2] presents a technique based on computing and uses NLP (Natural Language Processing) and substance mining strategies for helping essentials works in isolating and showing goals from documents that are printed. It is based on developing an NLP-based target elicitation approach within the settings of KAOS objective arranged essentials building strategy. The dynamic associations and upwards to downwards working among destinations are determined by means of normally collecting logical orders from isolated goals. It uses wise metering system for a relevant investigation for exploring the approach proposed. The Smart metering structure is among the noteworthy subsystem of the best in class period of power systems (shrewd cross sections). Goals and plans are expelled by logically parsing the sentence structure of the goal related articulations in overviews of research formation. The yield and results of this important investigation shows that the portrayed methodology is an effective technique to show targets for non-simple structures, and explicitly for the investigation raised systems that are non-simple or complex.

Casagrande et al. [3] shows and elaborates us an interesting self-loader framework space information investigation strategy. The strategy is completely founded on the iterative and repetition obtaining investigation and finding of a huge group of bibliometric information, age of area scientific classifications, and making of space models. This technique was applied on a keen network contextual analysis through assortment and examination of in excess of 6000 records and we have yielded. We have found or we understood that our strategy produces area models of similar quality or particular quality to the conventional physically created space models in a more practical thing. This technique was applied on a shrewd framework contextual analysis through assortment and investigation of in excess of 6000 archives and we have yielded. We have found or we understood that our strategy produces space models of practically identical quality or unmistakable quality to the customer physically delivered area models in a more financially savvy way.

Fisher et al. [4] explains that the social employments of the people and a get-together might help in understanding the social impacts of the get-together. It represents a strategy, steps or figuring of applying relational association assessment to help the task of depicting and finding makers in Usenet newsgroups. We register and imagine or disclosures frameworks made by instances of answers and returns for every essayist and researchers in picked newsgroups and media and find that second-degree feeling of self-propelled frameworks part with us from between different sorts of writer, newsgroups and media. The obtained results show that the newsgroups vary similarly among the people or individuals and their occupations or businesses that they have.

Newsgroups may be depicted by masses which consolidate request and conversational newsgroups, answer newsgroups, fire newsgroups and social assistance newsgroups. The system and revelations find uses for the two experts attempting to depict and seeing different sorts of social cyberspaces and insurances similarly as individuals attempting to perceive or finding particular collaboration assistants and substance makers. Identifying or Finding bunches or groups, societies in massive and sizeable authentic diagrams, for instance, huge social or information systems is an issue of good-sized intrigue and interest of a factor[5].

Practically speaking, one often alternatives a goal and acknowledging ability that catches the instinct or thoughtless of a machine group as set of hubs with better internal community and out of doors availability, and in a while one applies heuristics, estimation algorithms, or other techniques to extricate a set of hubs which might be identified with the goal work and that "resemble" or same exquisite networks for the use of intrigue and discoveries.

Right now, look into of discover a scope of machine community popularity techniques for you to look at or discover a difference in them and to comprehend their relative exhibitions or yield and the planned predispositions inside the bunches they distinguish.

We investigate and find some everyday goal works that are applied to formalize the idea of a gadget and associating community, and we analyze or locate some wonderful classes of estimate calculations and strategies that intend to enhance such target capacities. Also, as opposed to simply fixing a goal or locating an aspect and demanding for an approximation to the nice bunch of any variable size, this paper doesn't forget a size-settled edition of the development problem. Considering it as a community first-rate and as a component of its length gives a plenty higher and one of the best focal factors with which to take a look at community discovery calculations and strategies, for the reason that target capacities and cluster and estimation calculations frequently have non-obtrusive length-subordinate conduct.

III. OVERVIEW OF THE PROPOSED METHOD

Machine learning might be utilized to find the top-quality loads of this relapse equation to such an extent that it depicts the connection among abilities and objective notwithstanding possible. The calculation gains from a lot of models through limiting the mix-up of the relapse line with appreciate to the best possible estimations of those models, with the point it'll depict the connection of future or concealed cases as appropriate. The proposed method was depicted in the Figure 1.

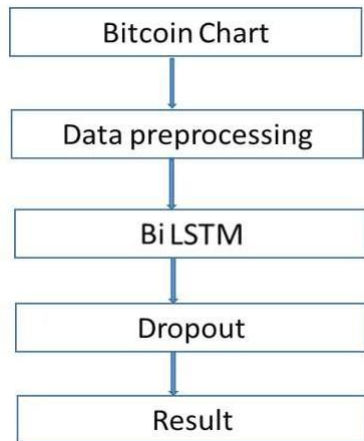


Figure 1. Proposed Method

A. BiLSTM algorithm

Tedious Neural Networks are a gathering of machine learning systems, which needs to wind up a generally utilized strategy for removing styles from fleeting arrangements [7], making it plausible amazing for anticipating time assortment like the Bit coin value design.

Conventional RNN, which makes time collection assessment less powerful, especially for longer enter and additionally yield arrangements. This is known as the evaporating slope inconvenience, which might be illuminated by method for presenting a BiLSTM (Bi-directional Long Short-Term Memory).

A BiLSTM is same a RNN with a versatile memory and it can keep its country during that time with the guide of the usage of non-direct gating contraptions which modify the data own into and out of the telephone [2], as may be recognizable in parent 2. This considers the transient part of the real factors to be preferred thought about rather over the usage of an extraordinary RNN, which can be the goal in the astonishing results of stock business community desire won in before take a gander.

IV. DATA ANALYSIS AND EXPERIMENT RESULTS

To start with, the records are recovered and pre-handled. After this condition of an AI is introduced. At that point, the marvelous component scaling strategy can be analyzed, the astonishing blend of highlights may be chosen and one of a kind combos of assortment lengths and forecast delays are tried.

A. Dataset

We first started with Bit coin market data that was publicly available on Kaggle2. This dataset consists of historical data of Bit coin from 1st January 2012 to 14th September 2020, and is parted into an increment of one-minute. In this way, around 1,574,274 minutes are there in this time frame. This data includes the information on the highest value, the lowest value, the opening value, the closing value, the volume traded, and the weighted price of each timestamp.

In order build an initial model and to iterate quickly, the polarity trends in the market were analysed by us at beginning. If the price decreased or even stayed the same, the dataset was labelled as false and if it went up at the end of minute timestamp, the dataset was labelled as true.

B. Data Processing

After filtering the data, it yields a 2-D tensor of 'x' samples by 'y' characteristics. A time series transform to convert the obtained result into a set of windows data having window size 'u'=50 was used by us. It gave a 3-D tensor of shape (x-u) samples by y characteristics by 'u' day window size. For example, our first point of data 'x'=0 had a 2-D tensor of 'x' characteristics from 0 to 49 days each. Then this data was normalized by us. Ultimately, we segregated the obtained data in the form of input and output data. To make it to the output data, we removed the last day.

The final and end-goal of designing a neural network based on crypto currency is to predict its price variation in real time. Keeping the goal in mind, we wanted to initiate with a dataset that was temporally and highly resolved. If the desired data was available to us on a second by second or even on a minute by minute timescale, our prediction model would have been more accurate and we could be much ahead of the market. This would lead to millions of data points and the neural network used by us excels at such dataset size. But as we hinted to above, it was realized that there's issue with highly resolved data as well.

After having a look at our minute dataset it was understood that there would not be any fluctuation on minute timescale, or even if there would be any, it would be very small and noisy. The graph obtained shown that approximately all the 1.5 million minutes were within the third bin and highlighted price changes under 0.003%. So, our model was unable to learn the

price fluctuation because it was mostly suited to the noisy data and any significant fluctuation was omitted. It should also be noted that at this initial point our 'y-values' were not actually binned out yet, so based on our assumption we finalized to transform the minute wise dataset into a daily wise dataset.

To predict the end or closing price of Bitcoin, a bidirectional RNN having 3 layers was used by providing a set of data from earlier days. The BILSTM model was used and code for filtering crypto currency data was included. After assembly, the use of each of the three methods, the E1 and E2 mistakes of forecasts as for the genuine estimation of the test set are determined, and it can also be seen in the given Table 3. The differentiation in general execution may practically be found in Table 1 as the forecasts of the methodology are towards the genuine qualities than the ones of the elective techniques. Because of this conclusive outcomes the methodology of scaling the abilities and target can be used in the remainder of the examinations.

TABLE I. SCALING TECHNIQUE AND THEIR BLUNDER REGARDING GENUINE ESTIMATIONS OF THE TEST SET

Featurescale	Targetscale	E1error	E2error
True	True	63.89	13002.66
True	False	235.86	202790.29
False	False	372.54	323430.24

The E2 mistake's perception of the expectation of test set of various capacity combos utilizing a forecast put off of zero and an arrangement span of 1. The most reduced E2 mistakes are at cycle 8, power set amount five. The result is obtained in the form of graph. The red line of the curve indicate the real price of bit coin while the blue line represent the predicted price. From graph, it is clear that our model worked almost fine in predicting the price of bit coin.

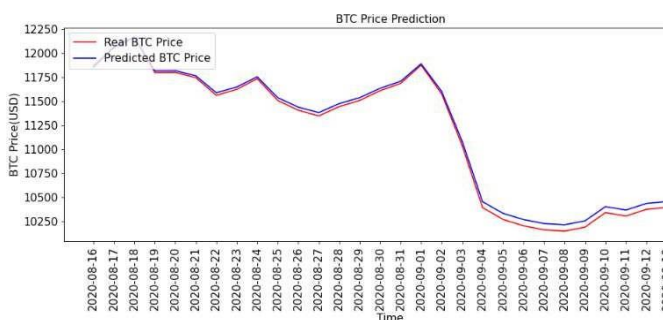


Figure 2. Bitcoin Prediction

The Bitcoin prediction of 16th Aug 2020 to 14th Sep 2020 is shown in Figure 2. To predict whether we should invest in the bit coin market or not, we need to have a look at the slope of the curve. If the slope of the curve is positive, we can go for it while if its zero or negative, it is recommended not to invest.

The proposed model predict the Bitcoin investment with highest accuracy of increment.

V. CONCLUSION

Bit coin is one of the leading cryptocurrency in the world at present. No doubt there are many investors who intend to invest in it. But due to uncertainty in its rise and fall, many suffer great loss at one hand and earn huge profits at the other. With this concern we have come up with idea of predicting the bit coin price up to some extent. Machine Learning Algorithms promise to deal with such predictions. Time series prediction is one of the common way to predict the bit coin prices at length of certain regular intervals. We have analyzed the data of various crypto currencies over the time interval from January 2012 to September 2020. On the basis of the data analyzed we are able to make decision whether to invest in the bitcoin or not based on the nature of the graph. The positive trend of the graph shows high increment in the value of bit coin and negative is just opposite of it. The proposed method of BILSTM with optimized hyperparameter gives better prediction than linear regression, and XG Boost methodology.

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