

Predict the Shipment Forecast using Time-Series Data in Machine Learning

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Abstract: In the point of large big data and massive increase the rate of time series data flow in the upcoming market program business, mining of related data and real time data[1] are been briefly explained. This paper proposes predicted the value of the trader marketing when we reach the expected the value and increase the rate of accuracy. For this purpose we can use the time series algorithm in machine learning and gets regular item sets by using the corresponding of Map reduce [2], which consumes less space and will not increase the time overhead. The usage of CPU is improved by using the thread calling algorithm and batch algorithm, it meets deep business opportunities and requirement processing or feature model based on the requirements of traders. Thus our results indicates that the model not only explained the time series data stream[4],it also helps traders to get to a confirmation that they can achieve data quickly and achieve accuracy trade off's. This paper proposes a new demand forecasting model which is an extension of the traditional exponential diffusion models [5]. We examined the forecasting performance of the models just after the release of the item when the small number of model calibration data is available. This paper shows that the model which we proposed has the thing of enabling early decision making and best performance.

Keywords: Time Series Data, Machine Learning, Map Reduce, Diffusion Models.

I. INTRODUCTION

In recent years, with the non-discrete expansion of the next generation market programs and the gradual increase of the investment type of concept in some countries, a massive [6] type of time series data that to be processed has been created in the transaction, that are entirely different from compared to the traditional data and the present techniques of very large amount of data, the execution speed and the generation rate increased time. Such that with the less processor and memory resources, ensure of the time series data can be executed/processed effectively and is becoming more and more important and effective. Most of the scholars focus on the real-time and effective data mining and processing association rules, then will be having much academic based research. The processes of literature of big data through the distributed parallel[7] type of calculation, and then it proves that decreasing the late process of the data

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transmission can effectively reduce the impact of late process of the processing the results. The proposed system of literature says that the permanent memory of the data can be use the disk Input Output overhead and increases the rate of data that is for accessing. On the basis of the new data structure, it increases the effectiveness of finding the most frequent item sets using Hah table storage type of technology and the usage or the optimization of the Apriori[8] algorithm. The next literature improves/increases the traditional usage of Apriori algorithm using the MapReduce in parallel, but these increments are only consider to the number of steps, creating a massive number of candidate sets or seeking to scan the database n number times. When the data is very massive, it will generate a very huge candidate set greatly decreasing the efficiency of the algorithm. On the another hand, the distributed program model based on the RTMR[9] method provides us very massive data storage and parallel computing real-timely[10], it makes very full use of the resource and optimizes the CPU utilization, thus provides the better process and data processing model.

II. DEMAND FORECASTING OF NEW ITEMS

Forecasting the new items in shopping media a number of researchers under science area marketing can develop several methods. One of the major and main forecast new items researcher is the development of predicting the forecast method value. Another side of the development is increasing the accuracy of the predicted value when compare with the excepted value.

Prediction for forecasting methods

For the purpose of prediction of forecasting methods utilize the available test data in market. In that market test data can having the sub headers has region, post-Shipment Invoice Date, Customer Code, Technology, SAP Item Code, Qty(Net), SO Date, End Customer Code and plant. Initially the region can access the post shipment data for the shipping date for the user in market for increase their profits. After gave the post shipment date, provider can access the customer information about address for delivery their product in the market. For this purpose, the service provider can discovery the code called Customer code. Customer Code can have all type of data where the customer personal data to customer ordered data. And the third party of this forecast shipment is by where this process can be done by the way of execution. That execution can do by the step by step process.

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By doing this all process called as Technology. SAP Item Code can having the data sku vce number where the product can order and which can product can be placed as well as delivery also. SO Date can having the dates where the customer can took their delivery product.

Region

Region can be allocated by the different places on the map in British rule for trading as shown in the figure.

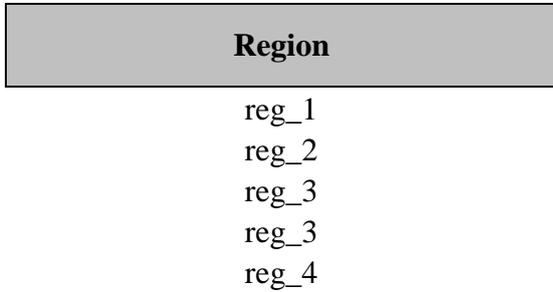


Fig. 2.1.1

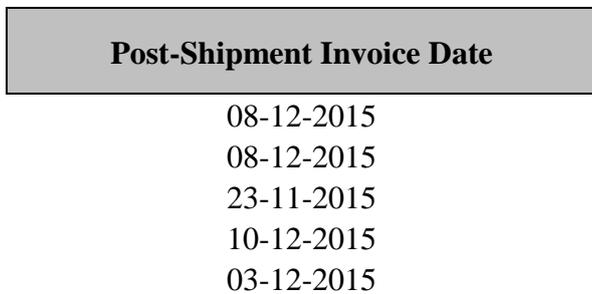


Fig.2.1.2

Post-Shipment Invoice Date

Post-Shipment Invoice Date is the date where the customer can order the data for product as shown in the figure.

Customer Code

Customer Code is the identity for the customer order and product reference as shown in the figure.

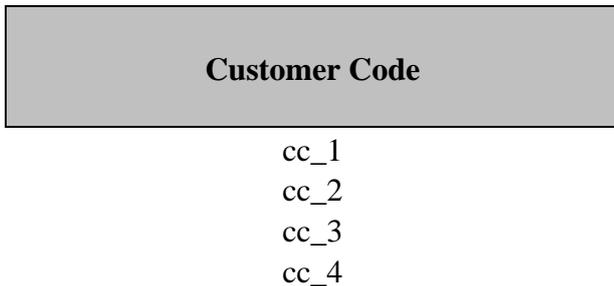


Fig. 2.1.3

III. SYSTEM IMPLEMENTATION

Front-end process

This Front end process can do step by step. Initially, we can read the forecast data for defining the output. After we can take the post-shipment invoice date as initial input values to sales. Then draw the graphs between the region and qty(net) by using the dataframes name as sales. Then after draw the graphs between the plant and qty(net) by

using the dataframes name as sales1. Again draw the graphs between the technology and qty(net) by using the dataframes name as sales2. Apply the Time Series data algorithm to the data. Then draw the times series forecasting graphs. Finally shown the output and the process can do by step by step as shown in the figure.

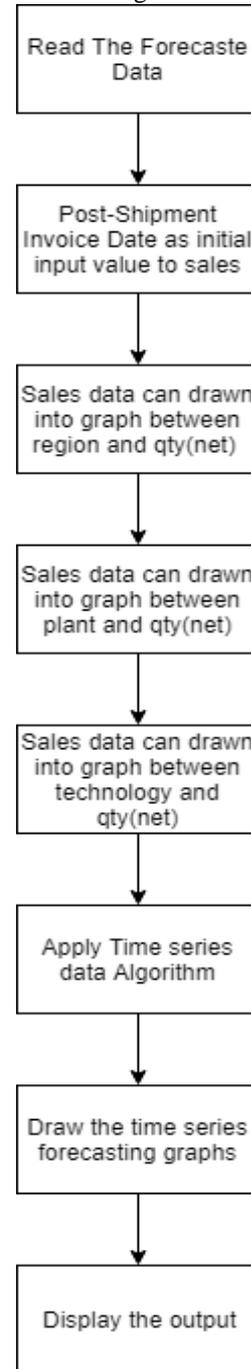


Fig. 3.1.1 Front end system architecture

Back-End Process

This Back-End process can done step by step. Initially, we can import the data from the device where the market time data can be stored. And then we can read the data for defining the output.

After we can creates the dataframes for the coding whose dataframe name called as sales. Then draw the graphs between the region vsqty(net) by using the dataframes name as sales and plant vsqty(net) by using the dataframes name as sales1and technology vsqty(net) by using the dataframes

name as sales2. Apply the Time Series data algorithm to the dataframes (sales). Then predict the value by using the algorithm. Finally compare the value with the expected value. This process can done as shown in the figure.

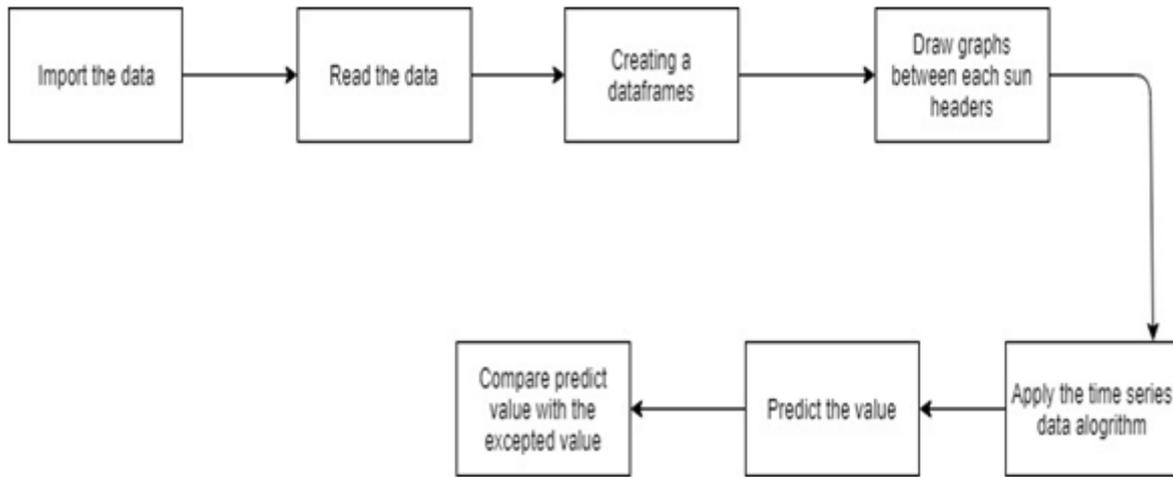


Fig. 3.2.2 Back End process

IV. EXECUTION AND RESULT

Process of the execution

Read the data

For reading the data we can first import the data and then read as shown in the figure.

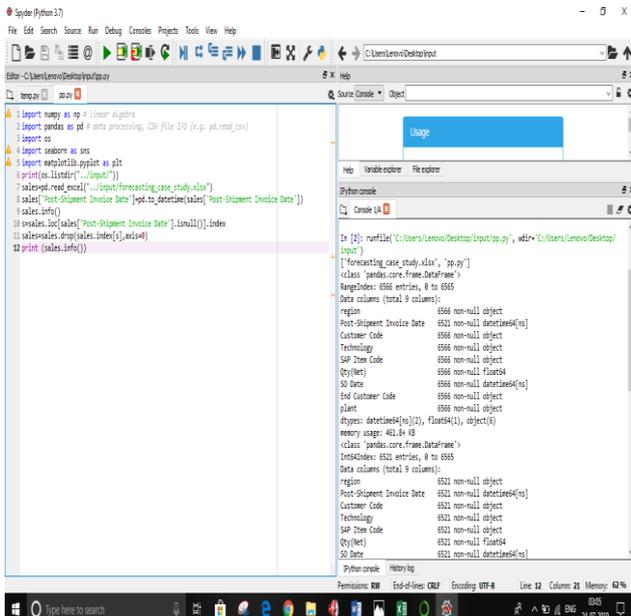


Fig. 4.1.1Importing data and read

Graph between region vsqty(net)

For the drawn the graph we can create the dataframes and run the code as shown in the figure:

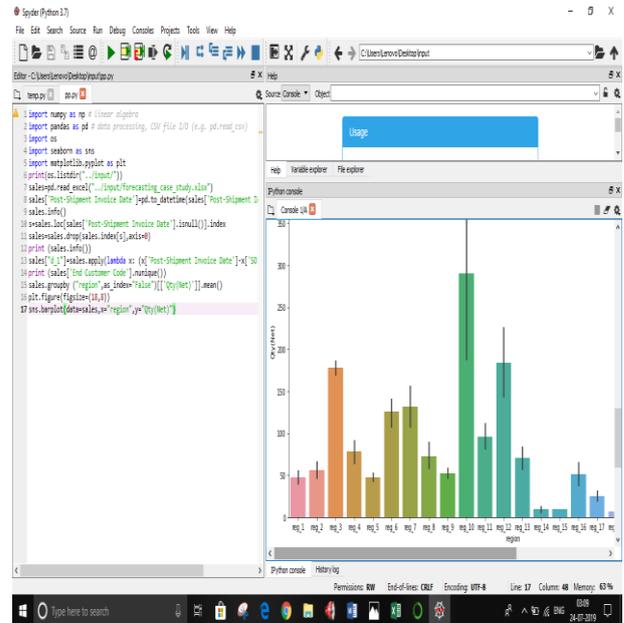


Fig. 4.1.2Creating Data Frames and running the code regionvsqty(net)

Graph between plant vsqty(net)

For the drawn the graph we can create the dataframes and run the code as shown in the figure:

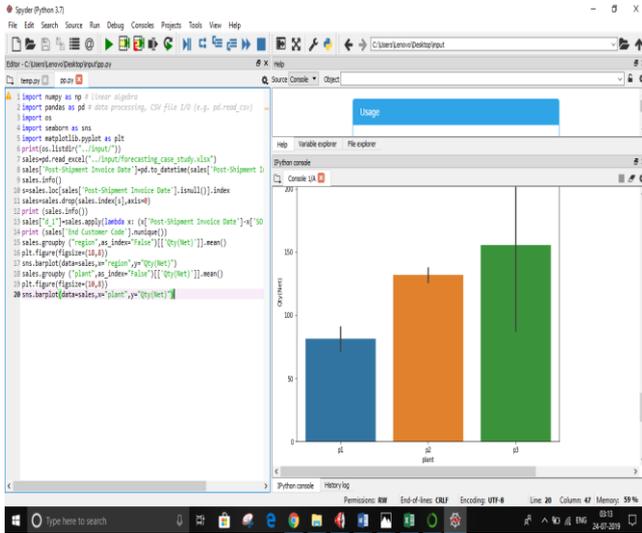


Fig. 4.1.3 plant vsqty(net)

Graph between technology vsqty(net)

For the drawn the graph we can create the dataframes and run the code as shown in the figure:

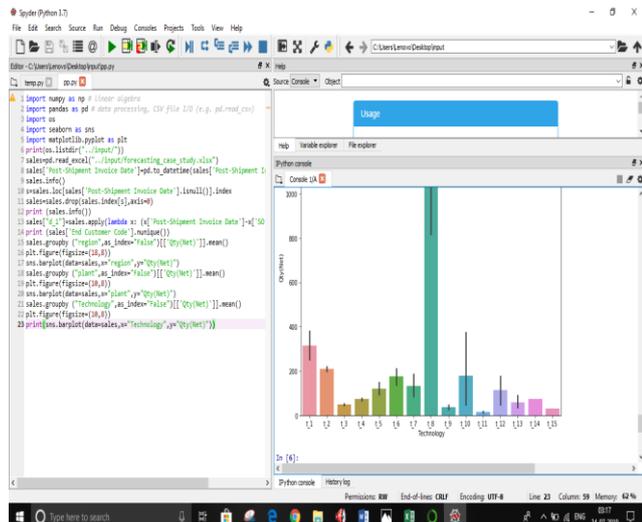


Fig. 4.1.4 Technology vsqty(net)

Plots between date, year, month vsqty(net)

For the drawn the plots between the date, year, month vsqty(net) we can access the dataframes and divided into subplots as shown in the figure.

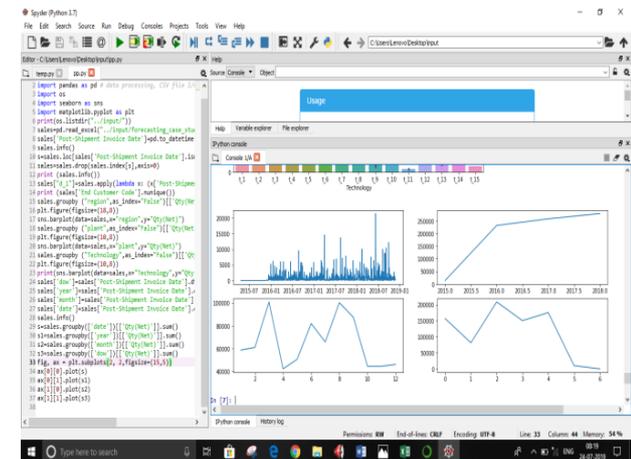


Fig. 4.1.5 Date, year, month vsqty(net)

Plot between Final-date vsqty(net)

When plotting the graph between the final date vsqty(net) we can consider the dataframes where the order can be placed as shown in the figure.

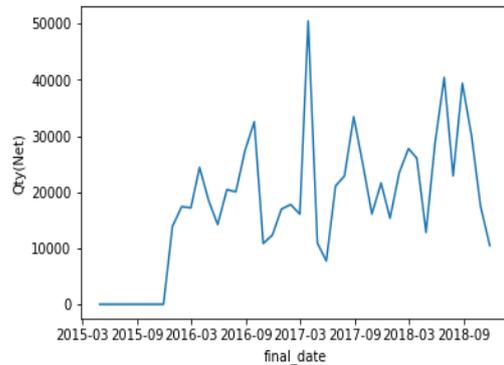


Fig. 4.1.6 Final-date vsqty(net)

Decomposition of Plots

Purpose of decomposition of the plots is analysis where and when net can be raised or dropped.

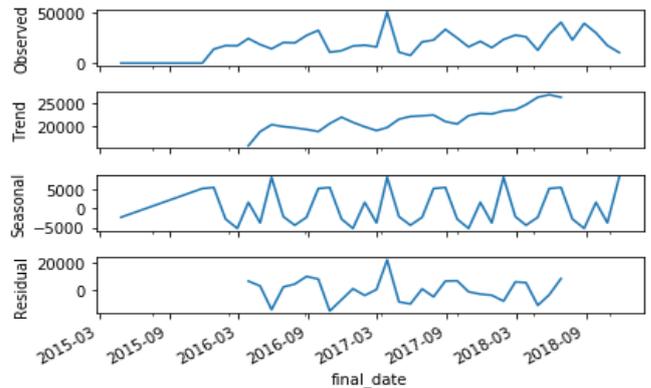


Fig. 4.1.7

Analysing the subplots in each stage

Purpose of this subplots can be expose the each and every sub connecting between the various sub headers as shown in the figure.

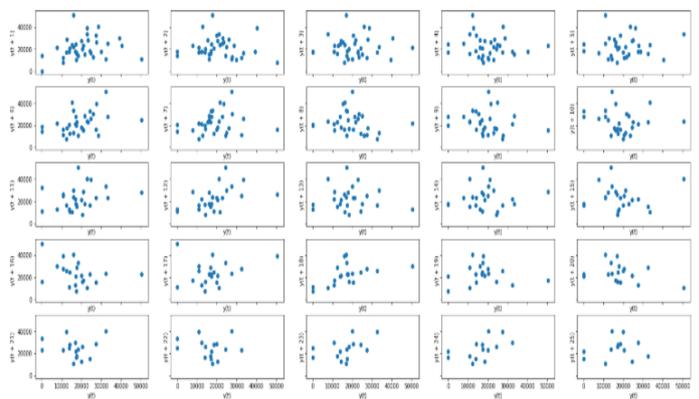


Fig. 4.1.8



Predict and expected values

```

predicted=12910.640685, expected=13873.000000
predicted=12589.460893, expected=17390.000000
predicted=16190.650204, expected=17183.000000
predicted=17346.032379, expected=24394.000000
predicted=22095.342791, expected=18429.000000
predicted=20753.616965, expected=14214.000000
predicted=15716.762068, expected=20430.000000
predicted=18333.905839, expected=20027.000000
predicted=20333.783374, expected=27437.000000
predicted=25127.365073, expected=32511.000000
predicted=31243.362100, expected=10840.000000
predicted=18763.579264, expected=12357.000000
predicted=11746.991819, expected=16948.000000
    
```

Fig. 4.1.9

4.2 Graph between predict and expected value

Accuracy percentage between predict and expected value

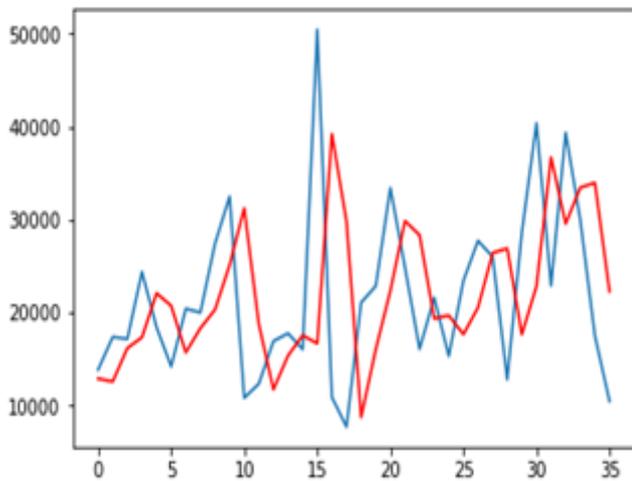


Fig. 4.2.1

Each and every stage accuracy percentage can be raised as shown in the table.

Table . 4.2.2

Si.no	Expected	Predicted	Accuracy
1.	13873.0000	12910.640685	93%
2.	17390.0000	16589.460893	95%
3.	17183.0000	16190.650204	94%
4.	24394.0000	23346.032379	95%
5.	18429.0000	18095.342791	99%
6.	14214.0000	14153.626965	99%

Average Accuracy % = sum of accuracies / no. of accuracies
 = 93+95+94+95+99+99/6
 = 95.833333%

Calculate ARIM values for AIC values

```

ARIMA(0, 0, 0)x(0, 0, 0, 12)12 - AIC:806.8665631096923
ARIMA(0, 0, 0)x(0, 1, 0, 12)12 - AIC:495.41638387315766
ARIMA(0, 0, 0)x(0, 2, 0, 12)12 - AIC:251.75438092929602
ARIMA(0, 0, 0)x(1, 0, 0, 12)12 - AIC:517.61712957488804
ARIMA(0, 0, 0)x(1, 1, 0, 12)12 - AIC:256.700682645606
ARIMA(0, 0, 0)x(2, 0, 0, 12)12 - AIC:254.4969076155974
ARIMA(0, 0, 1)x(0, 0, 0, 12)12 - AIC:768.6612988741803
ARIMA(0, 0, 1)x(0, 1, 0, 12)12 - AIC:477.13233913425313
ARIMA(0, 0, 1)x(0, 2, 0, 12)12 - AIC:233.59764694097188
ARIMA(0, 0, 1)x(1, 0, 0, 12)12 - AIC:538.1541660682681
ARIMA(0, 0, 1)x(1, 1, 0, 12)12 - AIC:258.2919461603527
ARIMA(0, 0, 1)x(2, 0, 0, 12)12 - AIC:272.7841664655487
ARIMA(0, 0, 2)x(0, 0, 0, 12)12 - AIC:746.7996339494779
ARIMA(0, 0, 2)x(0, 1, 0, 12)12 - AIC:457.2864635891052
ARIMA(0, 0, 2)x(0, 2, 0, 12)12 - AIC:210.19086916792423
ARIMA(0, 0, 2)x(1, 0, 0, 12)12 - AIC:540.9050352169527
ARIMA(0, 0, 2)x(1, 1, 0, 12)12 - AIC:260.56372914620295
ARIMA(0, 0, 2)x(2, 0, 0, 12)12 - AIC:275.377656252244
    
```

Final Predicted values

The predicted value can be defined as the region(reg_n) as follows:

For reg_1 sale predicted value is 24971.40000.

For reg_2 sale predicted value is 41391.818587.

For reg_3 sale predicted value is 59569.195500.

For reg_4 sale predicted value is 59542.017689.

```

1 24971.400000
2 41391.818587
3 59569.195500
4 59542.017689
5 43678.215163
6 29062.533240
7 24159.502511
8 21310.387125
9 23753.203919
10 27924.835279
11 18437.264190
12 26088.739908
13 21174.316342
14 12491.070375
15 22959.470665
16 20968.783549
17 3153.953071
    
```

V. CONCLUSION

In order to solve the predicted values of the shopping forecast we can propose the time series data algorithm using machine learning. The purpose of proposed system is when traders can predict the value by using now trader market values and analysis the future marketing values.

By using this times series data algorithm we can increase the accuracy value of the predicted value when compare with the expected value. We can reach the maximum value of the expected value in predicted value.

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