Project

**Banking dataset Analysis and Prediction**

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**Final Report**

**Banking Dataset- Analysis and prediction**

**Abstract:**

In today’s world, where huge amount of data is generated in every field of day-to-day activities, banking sector is one of them. As an outcome of work, various machine learning concept are studied with respect to Bank marketing data classification. Banking is a provision of the services by bank to an individual customer. The dataset is originally collected from UCI Machine learning repository and Kaggle website. The data is related to bank marketing campaigns of banking institution based on phone call. In this work, Python is used as a coding language and Machine learning concept is used as statistical learning for data analysis. The main reason of using machine learning is to build a predictive model to produce the better prediction. The outcome of the result is analysed with supervised Random Forest algorithm for classification purpose. The Customer bank dataset is used for term deposit prediction. This dataset is publicly available at UCI machine learning repository. It contains customer information. This dataset contains 11163 records and 17 attribute. The dataset has two types of prediction either Yes or No. There are 16 input feature and 1 output. The facts are associated with the direct advertising and marketing campaigns of a Portuguese banking institution. The marketing campaigns have been primarily based on telephone calls. Often, a couple of contact to the identical client was required, in order to access if the product would be or now not subscribed by means of the customer or no longer. Term deposits are a major source of income for a bank. A term deposit is a coin’s investment held at a financial organization. Your money is invested for an agreed rate of interest over a hard and fast amount of time, or time period. The financial institution has numerous outreach plans to promote period deposits to their clients together with email marketing, advertisements, telephonic advertising and marketing, and digital advertising and marketing. Telephonic advertising campaigns continue to be one of the simplest ways to reach out to human beings. However, they require big funding as massive call centres are employed to really execute those campaigns. Hence, it's far important to become aware of the customers maximum probable to convert beforehand so they can be mainly target via calls. The data is related to direct marketing campaigns (phone calls) of a Portuguese banking organization. The category purpose is to predict if the patron will join a term deposit.

**Goal:** The classification goal is to predict if the client will subscribe (yes/no) a term deposit (variable y).

**Theme:**

We use theme like classification and regression, data visualization, data manipulation, importing data.

**Keywords:**

Banking, Descriptive analysis, predictive analysis, ggplot, logistic regression, train and test split.

**Research questions:**

Question **1. To cumulate that how many customers taking loan according to their education level?**

#### **Question 2.** Do young customer book more TDs or older generation?

# **Question 3. Use Box Plot to compare the age of customers for the top 5 of the most common employment forms?**

# Question 4. **To calculate which month customer deposit maximum amount of money?**

**Tools: -**

**All formulation and data visualization will be done in Python, R Studio.**

**GitHub account link: -**

[SarabjitKaur774 (github.com)](https://github.com/SarabjitKaur774)

**INTRODUCTION**

Bank is a financial institution, which provide various service to the customer which perform deposit and providing a loan at an interest rate to the various customer. Banks store massive amount of information about their customer to improve the banking strategies and to maintain good relationship between the customers. Customers are the main asset of the bank. Usually, the selected customer is contacted directly through mail, email, personal contact, telephone cellular or any other contact to advertise the new service this kind of marketing called direct marketing. The objective of marketing in banking is to attract the new customers. The collected data from UCI machine learning repository, is related to bank marketing campaign of banking institution the classification goal is to predict if the customer will subscribe the term deposit. The data is related to direct marketing campaigns of a Portuguese banking institution. The marketing campaigns were based on phone calls. Often, more than one contact to the same client was required, in order to access if the product (bank term deposit) would be subscribed ('yes') or not ('no') subscribed. The test.csv which is the test data that consists of 11163 records and 16 features without the target feature. The dataset contains train and test data Features of educate information are Input variables(financial institution client facts): Age , job, marital, schooling, default, balance, housing, mortgage, contact, day, month, length, campaign, pdays, preceding, poutcome. Output variable (preferred target): y-has the purchaser subscribed a term deposit? (Binary: 'sure’, ‘no') and the test information have already been pre-processed. In this paintings Python used as a programming language, high-stage, interpreter, and considerable well-known library are freely to be had sources for all predominant platform from the Python internet site and Machine gaining knowledge of method for records analysis technique and automates analytical constructing model to predict the accuracy of the financial institution patron records. Where every example in a dataset is defined through a fixed of attributes and category algorithm used along with Naive bayes classifier set of rules gave the first-rate performance degree accuracy of the statistics. The financial institution ought to target the ability patron who have spent extensive quantity of time responding the bank calls. The most important item of this work is to find a way to use system gaining knowledge of method, for evaluation and making the prediction using existing dataset in banking marketing for developing effective selection-making understanding and to construct a system learning version the use of class set of rules to expect the accuracy of the facts.

**Objective:**

The main objective of building the model is to describe whether the customer has opted for term deposit. The bank should target the potential customer with considerable amount of time responding to the phone calls. The work implemented resulted in measuring accuracy, precision, recall and F1 score, towards term deposit prediction.

**What is a Term Deposit?**

A Term deposit is a deposit that a bank or a monetary group offers with a fixed price (often higher than just commencing deposit account) in which your cash can be back at a particular adulthood time. For more records on the subject of Term Deposits please click on in this hyperlink from Investopedia. Term deposits are a major source of earnings for a financial institution. A time period deposit is a coin’s investment held at a financial group. Your cash is invested for an agreed price of hobby over a hard and fast amount of time, or term. The financial institution has numerous outreach plans to promote term deposits to their clients which includes email marketing, commercials, telephonic advertising and marketing, and virtual advertising

**LITERATURE REVIEW :**

The Customer financial institution dataset is used for term deposit prediction. This dataset is publicly available at UCI system studying repository. It consists of patron information. The dataset has types of prediction either Yes or No. There are sixteen input function and 1 output. Attribute in the dataset with data type and description is following: **Bank client data: [**1] - age (numeric), [2] - job : type of job (categorical: "admin.", "unknown", "unemployed", "management", "housemaid", "entrepreneur", "student", "blue-collar", "self-employed", "retired", "technician", "services"), [3] - marital : marital status (categorical: "married", "divorced", "single"; note: "divorced" means divorced or widowed), [4] - education (categorical: "unknown", "secondary", "primary", "tertiary"), [5] - default: has credit in default? (Binary: "yes", "no"), [6] - balance: average yearly balance, in euros (numeric), [7] - housing: has housing loan? (Binary: "yes", "no"), [8] - loan: has personal loan? (Binary: "yes", "no"). **Related with the last contact of the current campaign: [**9] - contact: contact communication type (categorical: "unknown", "telephone", "cellular"), [10] - day: last contact day of the month (numeric), [11] - month: last contact month of year (categorical: "Jan", "feb", "mar", …, "nov", "dec"), [12] - duration: last contact duration, in seconds (numeric). **Other attributes: [**13] - campaign: number of contacts performed during this campaign and for this client (numeric, includes last contact), [14] - pdays: number of days that passed by after the client was last contacted from a previous campaign (numeric, -1 means client was not previously contacted), [15] - previous: number of contacts performed before this campaign and for this client (numeric), [16] - poutcome: outcome of the previous marketing campaign (categorical: "unknown", "other", "failure", "success"). **Output variable** (desired target) : [17] - deposit - has the client subscribed a term deposit? (Binary: "yes", "no"). The device gaining knowledge of techniques for evaluation and making prediction the usage of current statistics in banking marketing. The fulfilment charge of banking advertising depends on the result and decision in order to make greater correct prediction statistical device and techniques are used. A specific level for facts analysis and to find, how they can be used together in a manner converting uncooked information to effective choice-making expertise and building the predictive version in this work used selection tree algorithm will help to are expecting the client will subscribe the term deposit. Esslemont et all. Discussed all financial institution advertising marketing campaign are depending on client big statistics, the size of facts source is not possible for human to analyst to give you fulfilling information with the intention to help in choice making method. Data mining model are helping in the performance of the campaigns, in this work used most important data mining technique Naive bayes, logistic regression, and decision tree, the purpose is increasing the campaign effectiveness and identifying the characteristics that effect a success. A data driven approach was suggested to predict the success of bank telemarketing used data mining approach to predict the success telemarketing call for term deposits, data related to Portuguese retail bank it includes the effect of financial crisis, analysed large set of features related to bank client, social and economic characteristics, and product. In the modelling phase a semi-automatic feature had selected, performed with the data prior, and reduce set of the feature. Compare data mining model super vector machine, decision tree, logistic regression, and neural network, using two metrics, the four models were tested, and neural network present the best result, decision tree is a knowledge extraction method were applied to neural network to predict the several key attribute. Finally, the selected model as credible and valuable for telemarketing campaign. Bank direct marketing is an interactive process, for building the good relationship among customers, to study the customer characteristics and behaviour use an effective multi-channel communication. Apart from income growth, which may also increase patron fantastic response, the intention of financial institution advertising and marketing is to growth the client reaction of direct marketing campaign. Customer profiling in, using category approach for financial institution telemarketing, facts mining processes began by way of many agencies to restore the purchaser profiling. Decision tree, random wooded area, and Naive Bayes have been used, for predicting the client profiles and growing the telemarketing sales type is useful for measured accuracy percentage, precision, and consider quotes. Before evaluating the classifiers pre-processing and normalization were performed for conducting the experiments and evaluation system RapidMiner tool was used. Finally, end result show that selection tree is the exceptional classifier for predicting the patron profile and behaviour.

The information that banks receive from their clients, traders, companions, and contractors is dynamic and can be used for distinctive functions, relying on which parameters are used to examine them. Basically, the scope of AI for banking can be grouped into five huge businesses.

It needs to be pressured that due to internal opposition and gift-day financial catastrophe, there are big pressures for European banks to increase an economic asset. To treatment this

problem, one followed technique is provided appealing lengthy-time period deposit programs with suitable interest costs, specifically through the usage of directed advertising and marketing and marketing campaigns. Also, the same drivers are urgent for a reduction in costs and time. Thus, there may be a need for a development in efficiency: lesser contacts must be performed, but an approximately number of successes (customers subscribing the deposit) ought to be saved.

It is applicable to refer in short to the preceding research and research in the related regions of the problem to find out and to top off the research gaps. The following are the same studies performed through the eminent authors and practitioners on the location of service great of banks.

**(Dhandabani, 2010)** Examined the character of linkage among carrier first-rate and clients

loyalty in Indian retail banking. Study used confirmatory component analysis to discover the carrier great dimension. The effects dimensions are reliability, Responsiveness, Knowledge and restoration, and Tangibles. The service fine dimensions result in patron pride and the patron ‘pleasure ends in consumer’s loyalty. The structure equation model reveals that there's no significant direct linkage among provider excellent and customers loyalty. At the equal time, the provider satisfactory has a significant oblique effect on

purchaser’s loyalty particularly through purchaser’s satisfaction.

**(Maya Basant Lohani, 2012)** Examined on provider exceptional in selected banks and measured in 5 dimensions through the use of SERVQUAL scale advanced via Parasuraman et al (1998 and found out that there exists a small perceptual difference regarding standard provider quality with respective banks. The study of located that bank have more awareness at the tangible element like a computerization, physical centres, and many others. To draw the customers.

**(Hertz 113-114)** Knowledge of industries in which the financial institution’s customers operate. Often a financial institution’s mortgage portfolio might be focused on mainly specialised industries along with real assets, delivery, and natural resources. Evaluating the nature of those portfolios can additionally require a know-how of the corporation and reporting practices of these industries. The shape of the banking enterprise and the Bank’s role and reputation in the marketplace. For instance, if the financial group has a terrible rating, it is able to now not have got admission to better extraordinary loans, thereby taking extra credit score rating hazard.

**(Jain, 2012)** in their study “Customers Perception on Service Quality in Banking Sector: with Special references to Indian personal banks in Moradabad place” try to learn and apprehend the customer belief concerning carrier quality and to examine and apprehend the one-of-a-kind dimension of carrier pleasant in banks.

**J. Joshua Selvakumar (2010)** research the effect of carrier Quality on patron satisfaction in public region and personal region banks. The look at examines the impact of provider high-quality determinants at the diploma the perceived and real carrier high-quality, customer delight can be extremely progressed.

**Dr. Manasa Nagabhushanam (2010)** conducted a research study on service quality of banks in India. The study encompasses the service quality of all the banks i.e., public sector, private sector, and foreign banks and measures the attributes on SERVQUAL scale. The study was conducted to analyse the expected and perceived gap among customers and bankers.

**Ms. Nisha Malik and Mr. Chand Prakash Saini (2011)** studied Private sector banks quality and customers satisfactory by conducting an empirical study of two Private sector Banks. The aim of proposed study was to find out perception of HDFC and ICICI bank customers regarding to the service quality parameter and gap analysis of expected and acknowledged quality parameters. Ans also revels the relationship between psychographics factors and satisfaction levels of rural and urban customers.

**DESIGN AND METHODOLOGY**:

Firstly, our dataset is uncleaned like there are some duplicate values. First, we will clean our dataset to perform further operations. In methodology part we will use regression model to find out relationship between dependent and independent variables. We will also use descriptive and predictive analysis to find out how many customers are ready to subscribe the term deposit. On this Banking related dataset, we will also use classification and clustering. By using these different models, we will collect different results for sales data. Below we create diagram of system design and methodology: -

Model Implementation

Data Visualization

Training Data

Predicted result & Analysis

Validate Model

Train Model

Banking Dataset

Data Preprocessing

Testing Data

**System Design and Methodology**

The system design includes various stages like data collection, pre-processing, making training, testing data, implementing algorithm and last stage is predicted result. The accumulated uncooked statistics can be incomplete or noisy. The data should go through pre-processing section to smooth the information before the use of the facts forgetting to know, another step for schooling version is characteristic extraction. In next step we visualize the dataset and then split the dataset into training and testing. When we seeking to expect the output first, we want to teach the model the use of a dataset and model attempt to research the data to make correct prediction. Test information is unbiased of education records, if a model suit to the educate facts, then it also suits to the test records, least overfitting has taken vicinity. Random forest model is most used set of rules in ML maximum of time this algorithm offers the satisfactory accuracy end result. Training the version using ML algorithm with the education records. The technique of trained version is evaluated with the test data is validate records, sooner or later, the model gives the first-rate predicted result.

1. **Python:** Python is a high-level, interpreted, and popular-cause programming language. Python is used by a software developer as a help language, it is easy to examine and its syntax may be very easy code and consists of lot of code library, easy to build fashions for machine gaining knowledge of. This program includes fewer traces of code than the alternative programming language. Various companies used Anaconda, is the most popular Python distribution broadly used for machine studying and records technological know-how.
2. **Machine Learning:** Machine mastering is a software of synthetic intelligence, is a method for facts evaluation and automates constructing model it learns from the previous facts based totally at the ideas it identifies facts sample and take selection on minimal human intervention. Machine mastering particularly concerned with sample and accuracy. Most industries working on gadget gaining knowledge of method to examine big number of facts which include financial service, Government, Healthcare, Retail, and Transportation.
3. **Supervised:** The majority of machine learning uses supervised learning . This work is licensed under a Creative Commons Attribution 4.0 International License collect the data and produce the output data based on previous experience. The task of learning function that maps an input and output variable and use an algorithm to learn the mapping function from the input to the output, the process of an algorithm learning from the training data. Supervised learning classified into two groups, classification, and regression.
4. **Unsupervised:** Unsupervised learning algorithm are used when the information used for training the machine that is neither classified nor labelled and algorithm allowed to act on the data without guidance, this algorithm mainly deals with hidden structure from unlabelled data and this algorithm does not give the right output. Unsupervised learning algorithm are less accurate compared to supervised learning, Unsupervised learning classified into two groups, clustering, and association problems.
5. **Random Forest:** Random Forest is a supervised, flexible, straightforward learning algorithm used for classification and regression. This random forest consists of multitude of decision tree and results are aggregated, random forest collect the classification and select the most voted prediction as the result, this algorithm reduce the risk of overfitting. Random forest algorithm is reduced overfitting, high accuracy and estimates missing data.
6. **Naive Bayes:** It is handiest class technique primarily based on Bayes theorem used for solving class hassle with an assumption of impartial among predictors and calculate the possibility of an event associated with preceding understanding. A Naive Bayes classifier assumes that the presence of a particular features in a category is unrelated to the presence of any other feature.

**DATA DICTIONARY:**

This dataset contains 11163 records and 17 attribute from which we have 10 categorical attributes attribute and 7 numeric variables. First, I took each attribute one by one to check the datatype and assigned the appropriate datatype to the features. After this, I check the number of levels contains in each attribute and count of values corresponding to each level.

**CATEGORICAL ATTRIBUTE :**

**Table 1: Categorical Attribute.**

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| **Attribute** | **Description** | **Type of Attribute** | **Number of levels** | **Count** |
| Job | Type of job | Character | 12 | management 2566 |
|  |  |  |  | blue-collar 1944 |
|  |  |  |  | technician 1823 |
|  |  |  |  | admin. 1334 |
|  |  |  |  | services 923 |
|  |  |  |  | retired 778 |
|  |  |  |  | self-employed 405 |
|  |  |  |  | student 360 |
|  |  |  |  | unemployed 357 |
|  |  |  |  | … |
| Marital | Marital status | Character | 3 | married 6351 |
|  |  |  |  | single 3518 |
|  |  |  |  | divorced 1293 |
|  |  |  |  |  |
| Education | Customer education | Character | 4 | secondary 5476 |
|  |  |  |  | tertiary 3689 |
|  |  |  |  | primary 1500 |
|  |  |  |  | unknown 497 |
|  |  |  |  |  |
| Default | Has credit in default? | Binary (yes, no) | 2 | no 10994 |
|  |  |  |  | yes 168 |
| Housing | Has housing loan? | Binary (yes, no) | 2 | no 5881 |
|  |  |  |  | yes 5281 |
| Loan | Has personal loan? | Binary (yes, no) | 2 | no 9702 |
|  |  |  |  | yes 1460 |
| Contact | Contact communication type | Character | 3 | cellular 8042 |
|  |  |  |  | unknown 2346 |
|  |  |  |  | telephone 774 |
| Month | Last contact month of year | Character | 12 | May-24 |
|  |  |  |  | aug 1519 |
|  |  |  |  | jul 1514 |
|  |  |  |  | jun 1222 |
|  |  |  |  | nov 943 |
|  |  |  |  | apr 923 |
|  |  |  |  | feb 776 |
|  |  |  |  | oct 392 |
|  |  |  |  | jan 344 |
|  |  |  |  | sep 319 |
|  |  |  |  | mar 276 |
|  |  |  |  | dec 110 |
|  |  |  |  |  |
| Poutcome | Outcome of the previous marketing campaign | Character | 4 | unknown 8326 |
|  |  |  |  | failure 1228 |
|  |  |  |  | success 1071 |
|  |  |  |  | other 537 |
|  |  |  |  |  |
| Deposit | Has the client subscribed a term deposit? | Binary (yes, no) | 2 | no 5873 |
|  |  |  |  | yes 5289 |
|  |  |  |  |  |

**NUMERICAL ATTRIBUTE:**

For numerical attributes I have five number summary that is mean, standard deviation minimum, maximum.

**Table 2: Numeric Attribute**

|  |  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- | --- |
| **Column1** | **count** | **mean** | **std** | **min** | **25%** | **50%** | **75%** | **max** |  |
| age | 11162 | 41 | 12 | 18 | 32 | 39 | 49 | 95 |  |
| balance | 11162 | 1529 | 3225 | -6847 | 122 | 550 | 1708 | 81204 |  |
| day | 11162 | 16 | 8 | 1 | 8 | 15 | 22 | 31 |  |
| duration | 11162 | 372 | 347 | 2 | 138 | 255 | 496 | 3881 |  |
| campaign | 11162 | 3 | 3 | 1 | 1 | 2 | 3 | 63 |  |
| pdays | 11162 | 51 | 109 | -1 | -1 | -1 | 21 | 854 |  |
| previous | 11162 | 1 | 2 | 0 | 0 | 0 | 1 | 58 |  |

**DETAILED DATA DICTIONARY :**

In this, I worked on every variable of this dataset. First of all, I convert the datatype of attribute into appropriate datatype. After that, I check five number summery and also checked unique values corresponding to each attribute. Then I preformed the EDA (Exploratory data analysis), level encoder, splitting train and test, sampling, and classification models.

**DATA VISUALIZATION :**

Data visualisation helps to give insight of a dataset to understand the structure of data. It extracts the relevant information of each attribute and also shows the relationship between different features.

1. **Visualization of Different Job Profile**

Chart, bar chart

Description automatically generated

**Figure 1: Graph of different job profile**

**Explanation:** This horizontal bar graph represents the information corresponding to customers with different job profile. It seems like a greater number of people working in management profiles have subscribed to term deposits because the highest count of customer belongs to Management field, that is above 2500. However, the lowest count of customers belongs to job type which is unknown that is nearly 60.

1. **Customers with marital status and housing loan**

Chart, bar chart

Description automatically generated

**Figure 2: Relationship between marital status and housing loan**

**Explanation:** This bar graph shows the information about how many customers take the housing loan from the bank according to their marital status. Here ‘Yes’ represent the clients who take the housing loan from bank and ‘No’ represent the who don’t take any housing loan. The married clients have almost equal number of counts corresponding to ‘Yes’ or ‘No’ that is records as above 3000 and approximately 3800 respectively. The divorced clients of bank take less housing loan. That’s why the bank is interested more on married people and single than divorced.

1. **Graph of Education background**

**Chart, bar chart

Description automatically generated**

**Figure 3: Education Background of Customers**

**Explanation:** This graph represents the information of customers education background. Client whose education background is secondary are in high numbers, that is near to 5500. The second highest counts of education with tertiary education, which is approximately 3800. Education type which is unknown has the lowest counts as compared to other type of education background.

1. **Relationship between education background and month**

Chart, bar chart

Description automatically generated

**Figure 4: Graph of Education and month**

**Explanation :** This graph represents the information about the relationship between education and month of the year. It seems like a greater number of people whose level of education is secondary has subscribed to term deposits in the may month as compared to other months. However, in December, very less people subscribed the term deposit.

# **Plotting pie chart of Housing loan**

Chart, pie chart

Description automatically generated

**Figure 5: Pie Chart of housing loan**

**Explanation:** this pie chart gives the information about the housing loan like how many clients take the house loan from bank and how many not take housing loan. Here ‘Yes’ represent the clients who take the housing loan from bank and ‘No’ represent the who don’t take any housing loan. Only 168 records corresponding to 'Yes' and 10994 to 'No'.

# **To cumulate that how many customers taking loan according to their education level?**

**Chart, box and whisker chart

Description automatically generated**

**Figure 6: Relationship between Laon and education**

**Explanation:** This bar graph shows the highest numbers of loan taken by clients whose education level is Secondary and followed by Primary. On the other hand, customers whose education background is unknown.

# **Use Box Plot to compare the age of customers for the top 5 of the most common employment forms**

**Chart, box and whisker chart

Description automatically generated**

**Figure 7: Top-5 client categories**

**Explanation:** The plot shows that among the top-5 client categories, the most senior customers represent the management, and the largest number of outliers is among the management and technician.

# **To calculate which month customer deposit maximum amount of money?**

Chart, bar chart

Description automatically generated

**Figure 8: Month of Subscription**

# May got a slightly more subscribers than the other months. Regardless how many people is contacted the subscription average is almost the same with the exception of December and January. These months were got the fewest subscriptions.

# **Correlation Matrix**

Chart

Description automatically generated

**Figure 9: Correlation matrix**

**Explanation:** The chart illustrates the relation between different variable using the correlation matrix. This matrix shows that Duration attribute (0.45) is highly and more positively corelated with the target variable n1\_deposit. However, campaign attribute is the most negatively correlated feature to n1\_deposit (target variable).

## **TRAIN TEST SPLIT METHOD :**

In train-test-split method the entire dataset is split into training and testing sets to measure the accuracy of your model. The training set contain 70% data and testing set contain 30% records. Then I apply the over-sampling technique to make my target variable balance.

## **Modelling**

In the modelling part there are four methods Logistic Regression, KNN, Naive Bayes, Random Forest that I have applied on this dataset to find the best results.

**Logistic Regression:**  This model is supervised learning classification algorithm. It is used to predict a binary outcome based on a set of independent variables. A binary outcome is one in which there are most effective feasible scenarios - either the occasion takes place (1) or it does no longer happen (0). Independent variables are those variables or elements which may additionally affect the outcomes (or dependent variable).

**KNN:** The KNN stands for “K-Nearest Neighbour”. It is a supervised machine learning algorithm. This algorithm is basically used to solve both classification as well as regression problem statements. This is quite intuitive and use distance measures to find the ‘K’ nearest neighbours to a new, unknown data point to make a prediction.

**Naive Bayes:**  Naive Bayes is a classification technique based on Bayes' Theorem with an assumption of independence among predictors. Naive Bayes makes use of the Bayes’ Theorem and assumes that all predictors are unbiased. In different words, this classifier assumes that the presence of one unique characteristic in a category doesn’t have an effect on the presence of some other one.

**Random Forest:** It is a Supervised Machine Learning Algorithm that is used widely in Classification and Regression problems. It builds decision trees on different samples and takes their majority vote for classification and average in case of regression.

Random Forest is a way used in modelling predictions and conduct (behaviour) analysis and is constructed on decision trees. It includes many decision trees representing a distinct example of the type of classification into the random forest. The random forest method considers the instances in my view, taking the one with most of the people of votes as the chosen prediction.

**ACCURACIES OF MODELS:**

I check the accuracy of each model to find the best optimum models for this dataset.

**Table 3: Comparing the accuracy of each model**

|  |  |  |
| --- | --- | --- |
| **Column1** | **Model** | **Train\_test\_split** |
| 0 | Logistic Regression | 77.69483428 |
| 1 | KNN | 76.29143028 |
| 2 | Naive Bayes | 75.18662287 |
| 3 | Random Forest | 81.45715139 |

To check the accuracy of my dataset corresponding to each model. To find the best fit model for the dataset. I have found that the random forest is the best fit model for my dataset with highest accuracy 81% as compared to other models.

# **Confusion matrix corresponding to Random Forest Classifier Mode**

Chart, treemap chart

Description automatically generated

**Figure 10: Confusion Matrix**

**Explanation:** Confusion matrix is one of the easiest metrics utilized for observing the accuracy of a classification model. This is the most popular method used to evaluate [logistic regression](https://intellipaat.com/blog/what-is-logistic-regression/).

The actual values correspond to 0 and predicted values 0 is 43.12% is true and the actual value correspond to 0 and predicted correspond to 1 is false (9.85%).

1. **ROC Curve**

Chart, line chart, scatter chart

Description automatically generated

**Figure 11 : ROC Curve**

**Explanation:** ROC (Receiver operating characteristic) curve plots the true positive rate of a test versus its false positive rate for different cut-off points of a parameter. The accuracy of a test is measured by the area under the ROC curve (AUC) that is 0.89.

**CLASSIFICATION REPORT :**

**Table 4: Classification Report**

It is a performance evaluation metric in machine learning. It is used to show the precision, recall, F1 Score, and support of your trained classification [model](https://thecleverprogrammer.com/2020/11/27/machine-learning-algorithms-with-python/).

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| **Column1** | **precision** | **recall** | **f1 score** | **support** |
| 0 | 0.82 | 0.83 | 0.83 | 1774 |
| 1 | 0.81 | 0.81 | 0.81 | 1575 |

**Explanation:**

**Precision :** Correct positive predictions relative to total positive predictions that is 0.82

**Recall :** Correct positive predictions relative to total actual positives that is 0.83

**F1 score :** The F1 is the weighted harmonic mean of precision and recall. The closer the price of the F1 score is to 1.0, the higher the predicted performance of the model is.

**Support :** Support is the quantity of actual occurrences of the magnificence inside the dataset.

**CONCLUSION**

In banking field, the large amount of data is generated continuously, and this data can be utilized to extract relevant information, to predict whether a client will subscribe to a term deposit. The primary purpose of the usage of machine learning is to build a predictive model to supply the better prediction. This dataset is publicly available at UCI machine learning repository or Kaggle website. It contains customer information like their age, job, education, and marital status etc. The dataset has two types of prediction either Yes or No. There are 16 input attribute and 1 output variable. The predominant goal of constructing the version is to describe whether or not the client has opted for time period deposit. After implementing (Supervised algorithm) Random Forest used for classification purpose, the algorithm provides 81% accuracy for the dataset by measuring accuracy, precision, recall, f1-score. So, we can say after the implementation, the result obtained was satisfactory.

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