

Corporate Employee Attrition Analytics

(A PROJECT COMPLETION REPORT)

Team ID:PNT2022TMID54360

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1.INTRODUCTION

1.1 PROJECT OVERVIEW

The overview of this project is to predict the attrition rate for each employee, to find who's more likely to leave the organisation. It will help organization to find ways to prevent attrition or plan in advance the hiring of the new candidate. Attrition proves to be a costly and time-consuming problem for the organization and it also leads to the loss of probability. The scope of the project extends to companies to all industries.

1.2 PROJECT PURPOSE

The key to success in any organization is attracting and retaining top talent. As an HR analyst one of the key task is to determine which factors keep employees at the company and which prompt others to leave. Given in the data is a set of data points on the employees who are either currently working within the company or have resigned. The objective is to identify and improve these factors to prevent loss of good people.

2.LITERATURE SURVEY

2.1 Existing problem

In certain cases employee attrition depends more on employees' satisfaction level as compared to other attributes. In this context, employee attrition presents a critical problem and a big risk for organizations as it affects not only their productivity but also their planning continuity. Employee attrition presents a critical problem and a big risk for organizations as it affects not only their productivity but also their planning continuity. In this context, the salient contributions of this research are as follows.

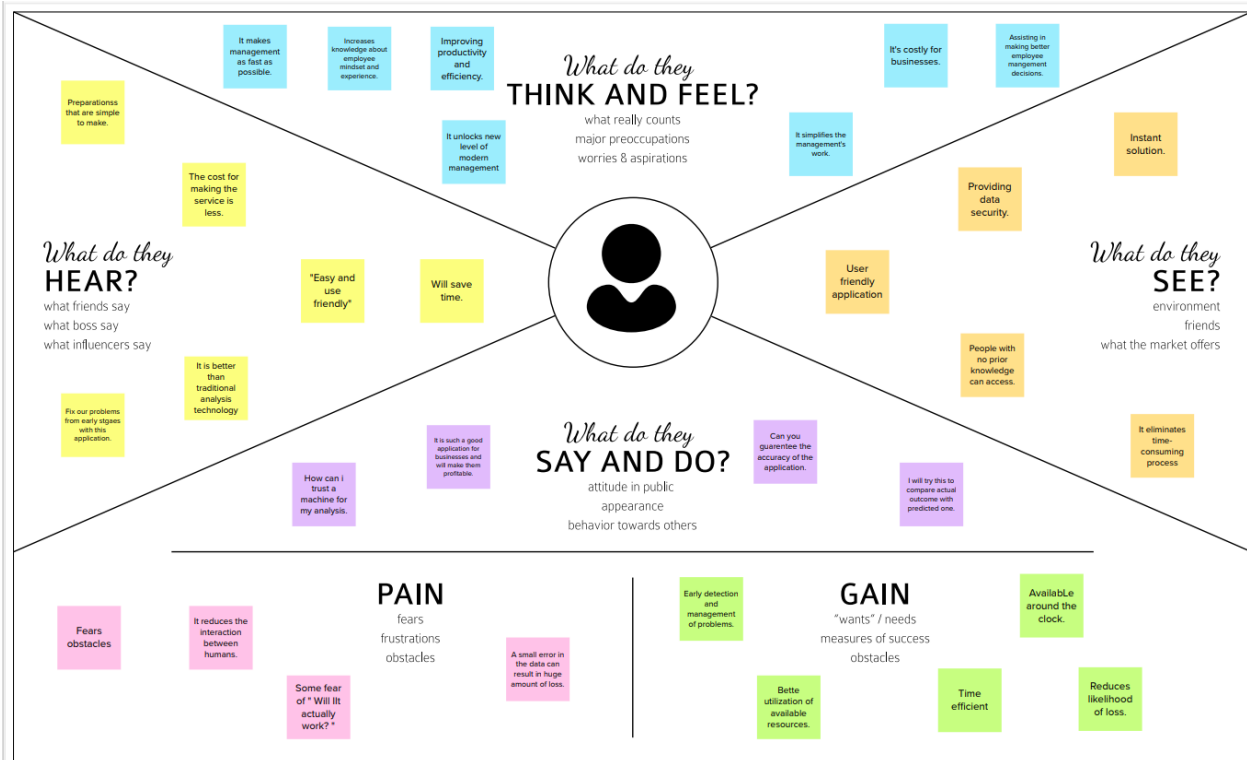
2.2 Problem Statement Definition

Employees are the most important part of an organization. Successful employees meet deadlines, make sales, and build the brand through positive customer interactions. Employee attrition is a major cost to an organization and predicting such attritions is the most important requirement of the Human Resources department in many organizations. In this problem, the task is to predict the attrition rate of employees of an organization.

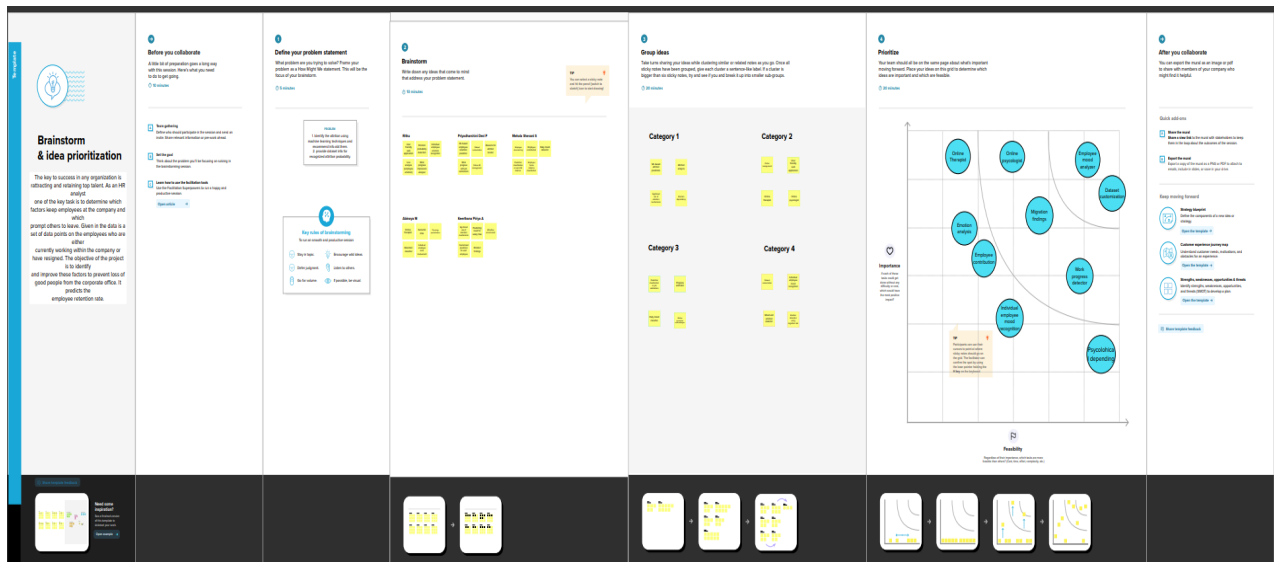
| | |
|---|--|
| What does the platform affect? | Corporate HR managers. |
| What are the boundaries of the problem? | Management that faces the issues of employee attrition. |
| What is the issue? | In organizations, employees are leaving and it might be a concern for the organization. It is more cost-effective to keep the employees a company already has. |
| Why is it important to fix the problem? | It is required for the growth of the organization. It is important to retain the employees. |
| What solution to solve this issue? | An automated system is introduced to identify different reasons for employees leaving the organization by environment, job satisfaction, work-life balance. |
| What methodology used to solve the issue? | Machine learning techniques are used to predict the employee attrition in an organization. |

3.IDEATION & PROPOSED SOLUTION

3.1Empathy Map Canvas



3.2 Ideation & Brainstorming



3.3 Proposed Solution

IBM Corporate Employee Attrition Analysis

★ INTRODUCTION

The Employee Attrition Analysis system is to understand the sentiments of highly disengaged and unhappy employees before it's too late. And thereby, reduce the employee churn by identifying problem areas automatically via AI-powered insights using ML tools.

★ PROBLEM STATEMENT

Nowadays, employee attrition became a serious issue regarding a company's competitive advantage. Employees are the most important part of an organization. Successful employees meet deadlines, make sales, and build the brand through positive customer interactions. It's very expensive to find, hire and train new talents. It's more cost-effective to keep the employees a company already has. A company needs to maintain a pleasant working atmosphere to make their employees stay in that company for a longer period. A few years back it was done manually but it is an era of machine learning and data analytics. Now, a company's HR department uses some data analytics tool to identify which areas to be modified to make most of its employees to stay. Employee attrition is a major cost to an organization and predicting such attritions is the most important requirement of the Human Resources department in many organizations. In this problem, prediction of the attrition rate of employees of an organization is taken place.

★ PROBLEM SOLUTION

The solution to the problem is Machine learning, which is one of the applications of Artificial Intelligence, is being used to implement the proposed system. Employee Attrition Analysis system is going to predict the attrition rate based on employee satisfaction, work environment etc. Finding the sentiment of employee is a challenging task. Other important issues is to identify daily mood recognition, emotion detection, environment, need for hike, employee satisfaction, employee contributor and work progress. To overcome all these issues this predictor has been proposed.

Nowadays a lot of research and work is being implemented in the smart and modern employee sentiment analysis domain. The Gradient Boosting or GBM is an ensemble machine learning algorithm that can be used to build the predictor model. These are the models to predict the employee attrition rate

★ THE BENEFICIAL USERS

- HR
- Management (Company)
- Other job-seekers
- Industrial people
- Team Leads
- Clients

IBM Corporate Employee Attrition Analysis

★ VALUE FOR SOCIETY

Employees are the most important part of an organization that influences a company's economic growth. Many new technologies, such as Machine Learning are being implemented into the management so that it is easier for the HR and the technical team to recognize and improve the environment or giving hikes and subsidiaries.

★ VALUE FOR ENVIRONMENT

- In the employee attrition analysis system, the work progress, emotions on frequent intervals will predict the attrition rate.
- For the employee attrition analysis system, the manager can input the environment data, mood detection data and the system will predict the attrition rate and then the improvements can be made.
- For this system, the user can give details about the environment, work progress, emotion that will give the rate of attrition as output. This can help improve and retain the employees.

★ VALUE FOR BUSINESS

Predicting the employee attrition rate, analysing the mood and emotions makes it easier for the HR and the management as it provides result in minimal effort and time. This action adds a lot of value to the company and the business in society.

★ FORM FACTORS

Our employee attrition analysis system for the attrition rate prediction is in the form of web application to provide this valuable service to the environment and society.

★ IT IS AN OPPORTUNITY ? (By public review)



3.3 Problem Solution fit

Project Title: Corporate Employee Attrition Analysis

Project Design Phase-I - Solution Fit Template

Team ID: PNT2022TMIDxxxxxx

| | | | | |
|--|--|---|---|--|
| Define CS, fit into CC | <div>1. CUSTOMER SEGMENT(S)<div>CS</div></div> <div>HRs of corporate companies</div> | <div>6. CUSTOMER CONSTRAINTS<div>CC</div></div> <div><div><div>• Spending power</div><div>• Budget constraints</div><div>• Resource constraints</div></div></div> | <div>5. AVAILABLE SOLUTIONS<div>AS</div></div> <div>A dashboard to maintain the various reasons for employees leave their jobs, and to provide tips to HRs on tackling different issues related to attrition.</div> | Explore AS, differentiate |
| | | | | |
| Focus on J&P, tap into BE, understand RC | <div>2. JOBS-TO-BE-DONE / PROBLEMS<div>J&P</div></div> <div>To build a supervised machine learning model based on regression algorithms for forecasting the reason for employee attrition based on the following parameters.<div><div><div>• Work progress</div><div>• Employee satisfaction</div><div>• Subsidiaries</div><div>• Distance between offices & homes</div></div></div></div> | <div>9. PROBLEM ROOT CAUSE<div>RC</div></div> <div><div><div>• Manually analyzing reasons for employee attrition is a time - consuming process.</div><div>• Due to this, before identifying the cause the employee would have left the job and valuable human resource is lost.</div><div>• Lots of human effort is required.</div></div></div> | <div>7. BEHAVIOUR<div>BE</div></div> <div><div><div>• The model behaves so as to reduce the human efforts required to consolidate and confess the various reasons for employee attrition by manually analyzing them.</div><div>• It also brings the most important causes for employee attrition to spotlight for the main customers the HRs of the corporates.</div></div></div> | Focus on J&P, tap into BE, understand RC |
| | | | | |

Identify strengths & weaknesses

Identify strengths & weaknesses

4.REQUIREMENT ANALYSIS

4.1 Functional requirement

Following are the functional requirements of the proposed solution

| FR No. | Functional Requirement (Epic) Sub | Requirement(Story/Sub-Task) |
|--------|-----------------------------------|---|
| FR-1 | User Registration | Registration through Form |
| FR-2 | User Confirmation | Confirmation via Email |
| FR-3 | User Authentication | Authenticate the user's attempt to login using the database |
| FR-4 | Retention analysis | Employee attrition analysis by sentiment, work environment, daily contribution etc. |
| FR-5 | Employee management | Validating and managing the registered employee details. |
| FR-6 | Progress management | Add the progress of each employee to the company. |
| FR-7 | Predict button | <p>The predict route is used for prediction and it contains all the codes which are used for predicting our results. Firstly, inside launch function we are having the following things:</p> <ul style="list-style-type: none">● Getting our input and storing it.● Select the necessary attributes for the prediction.● Creating model.● Predicting our results .● Showcase the results with the help of dashboard.● Finally run the application. |

4.2 Non-Functional requirements

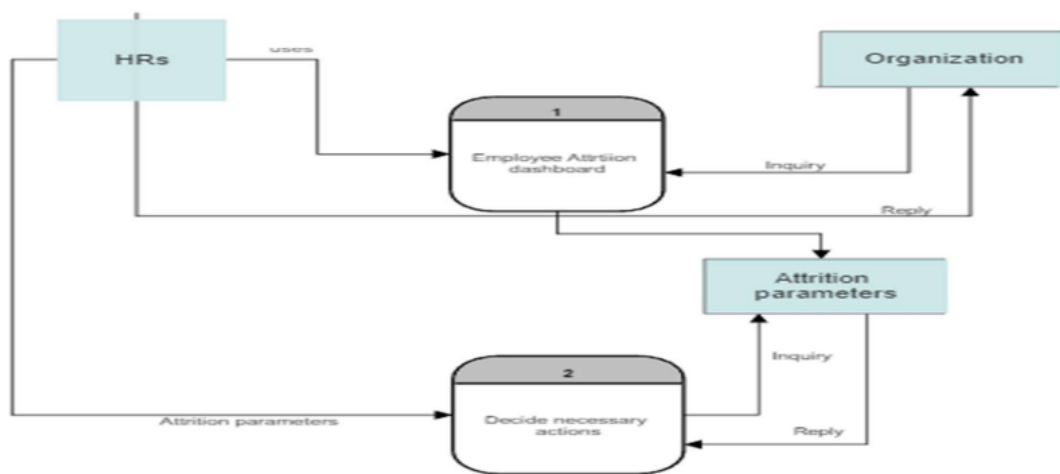
Following are the non-functional requirements of the proposed solution.

| FR No. | Non-Functional | Requirement Description |
|--------|-----------------|---|
| NFR-1 | Usability NFR-2 | This software shall be easy to use for all users with minimal instructions. 100% of the languages on the graphical user interface (GUI) shall be intuitive and understandable by non-technical users. |
| NFR-2 | Security | The user of the system should be provided the surety that their account details are secure. The system will provide security against cross site request forgery. |
| NFR-3 | Reliability | The software shall be operable in all conditions. The system must be less prone to errors. |
| NFR-4 | Performance | The performance of the system must assist the system's quality. |
| NFR-5 | Extensibility | The software shall be extensible to support future developments and add-ons. |
| NFR-6 | Portability | The software shall be 100% portable to all operating platforms. Therefore, this software should not depend on the different operating systems. |
| NFR-7 | Scalability | The system must be able to handle an increase in workload without performance degradation |

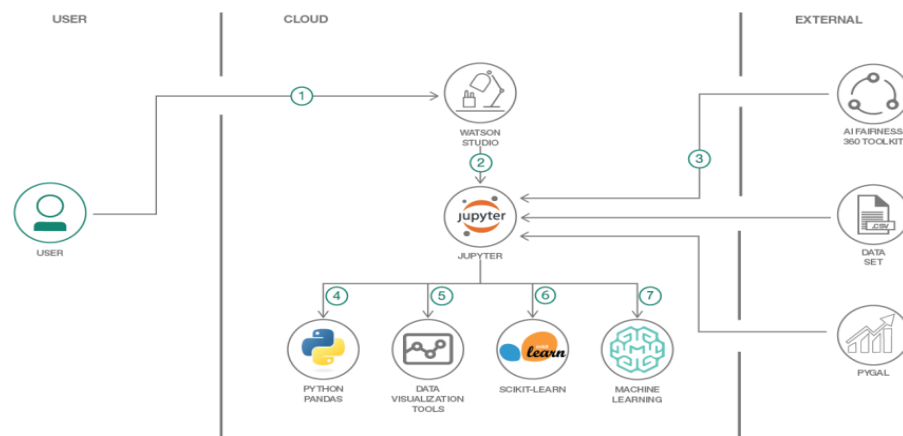
5.PROJECT DESIGN

5.1 Data Flow Diagrams

A Data Flow Diagram (DFD) is a traditional visual representation of the information flows within a system. A neat and clear DFD can depict the right amount of the system requirement graphically. It shows how data enters and leaves the system, what changes the information, and where data is stored.



5.2 Technical Architecture



5.3 User Stories

Use the below template to list all the user stories for the product.

| User Type | Functional Requirement (Epic) | User Story Number | User Story / Task | Acceptance criteria | Priority | Release |
|-----------|-------------------------------|-------------------|---|--|----------|----------|
| Employees | Registration | USN-1 | The employees can register to be a part of the organisation by entering in form and getting authenticated via email. | I can access my account / dashboard | High | Sprint-1 |
| | | USN-2 | As an employee, I will receive confirmation email once I have registered for the application | I can receive confirmation email & click confirm | High | Sprint-1 |
| | | USN-3 | As an employee, I can register for the application through Facebook | I can register & access the dashboard with Facebook Login | Low | Sprint-1 |
| | | USN-4 | As an employee, I can register for the application through Gmail | I can get a verification link through email | Medium | Sprint-1 |
| | Login | USN-5 | As a employee, I can log into the application by entering email & password | I can enter the application | High | Sprint-2 |
| | About | USN-6 | I can click on the "About" to get the idea on employee attrition rate prediction using parameters like work environment, sentiments of employee, daily employee engagement and work progress. | I can get an idea about the project | Low | Sprint-2 |
| | Launch | USN-7 | As a HR, I can upload various analysed parameters from the computer. | I can choose any employee ('s all parameters) from my device | High | Sprint-2 |
| | Predict | USN-8 | As a HR, I can perform prediction using predict button | I can view the employee's parameters on the dashboard along with the attrition rate. | High | Sprint-3 |
| | | USN-9 | I can also upload csv format of employee retention parameters from cloud. | I can view the employee's parameters on the | Medium | Sprint-3 |

6.PROJECT PLANNING & SCHEDULING

6.1 Sprint Planning & Estimation

| TITLE | DESCRIPTION | DATE |
|-----------------------------------|---|-------------------|
| Literature Survey & Information | Gathering Literature survey on the selected project & gathering information by referring technical papers, research publications etc. | 28 SEPTEMBER 2022 |
| Prepare Empathy Map | Prepare empathy map canva to capture the user pains & gains, prepare list of problem statements | 24 SEPTEMBER 2022 |
| Ideation | List the idea by organizing the brainstorming session and prioritize the top 3 ideas based on the feasibility & importance. | 25 SEPTEMBER 2022 |
| Proposed Solution | Prepare the proposed solution document which includes the novelty, feasibility of idea, business model, social impact, scalability of solution etc. | 23 SEPTEMBER 2022 |
| Problem Solution Fit | Prepare problem - solution fit document. | 30 SEPTEMBER 2022 |
| Solution Architecture | Prepare solution - architecture document. | 28 SEPTEMBER 2022 |
| Customer Journey | Prepare the customer journey maps to understand the user interactions & experiences with the application (entry to exit). | 20 OCTOBER 2022 |
| Functional Requirement | Prepare the functional requirement document. | 8 OCTOBER 2022 |
| Data Flow Diagrams | Draw the data flow diagrams and submit for review. | 9 OCTOBER 2022 |
| Technology Architecture | Prepare the technology architecture diagram. | 10 OCTOBER 2022 |
| Prepare Milestone & Activity List | Prepare the milestones & activity list of the project. | 22 OCTOBER 2022 |

6.2 Sprint Delivery Schedule

| Sprint | Duration | Sprint Start Date | Sprint End Date(Planned) | Release Date(Actual) |
|-----------------|-----------------|--------------------------|---------------------------------|-----------------------------|
| Sprint-1 | 6 Days | 24 Oct 2022 | 29 Oct 2022 | 29 Oct 2022 |
| Sprint-2 | 6 Days | 31 Oct 2022 | 05 Nov 2022 | 05 Nov 2022 |
| Sprint-3 | 6 Days | 07 Nov 2022 | 12 Nov 2022 | 12 Nov 2022 |
| Sprint-4 | 6 Days | 14 Nov 2022 | 14 Nov 2022 | 19 Nov 2022 |

7.CODING & SOLUTIONING

To enhance with the coding we added many features and methodologies to bring out an errorless results.

The following features have been added in our project

Business Understanding:

Before solving the problem in the Business domain it needs to be understood properly. Business understanding forms a concrete base, which further leads to easy resolution of queries. We should have the clarity of what is the exact problem we are going to solve.

Analytic Understanding:

Based on the above business understanding one should decide the analytical approach to follow. The approaches can be of 4 types: Descriptive approach (current status and information provided), Diagnostic approach (A.K.A statistical analysis, what is happening and why it is happening), Predictive approach (it forecasts on the trends or future events probability) and Prescriptive approach (how the problem should be solved actually).

Data Requirements:

The above chosen analytical method indicates the necessary data content, formats and sources to be gathered. During the process of data requirements, one should find the answers for questions like 'what', 'where', 'when', 'why', 'how' & 'who'.

Data Collection:

Data collected can be obtained in any random format. So, according to the approach chosen and the output to be obtained, the data collected should be validated. Thus, if required one can gather more data or discard the irrelevant data.

Data Understanding:

Data understanding answers the question "Is the data collected representative of the problem to be solved?". Descriptive statistics calculates the measures applied over data to access the content and quality of matter. This step may lead to

reverting the back to the previous step for correction.

Data Preparation:

Let's understand this by connecting this concept with two analogies. One is to wash freshly picked vegetables and second is only taking the wanted items to eat in the plate during the buffet. Washing of vegetables indicates the removal of dirt i.e. unwanted materials from the data. Here noise removal is done. Taking only eatable items in the plate is, if we don't need specific data then we should not consider it for further process. This whole process includes transformation, normalization etc.

Modelling:

Modelling decides whether the data prepared for processing is appropriate or requires more finishing and seasoning. This phase focuses on the building of predictive/descriptive models.

Evaluation:

Model evaluation is done during model development. It checks for the quality of the model to be assessed and also if it meets the business requirements. It undergoes diagnostic measure phase (the model works as intended and where are modifications required) and statistical significance testing phase (ensures about proper data handling and interpretation).

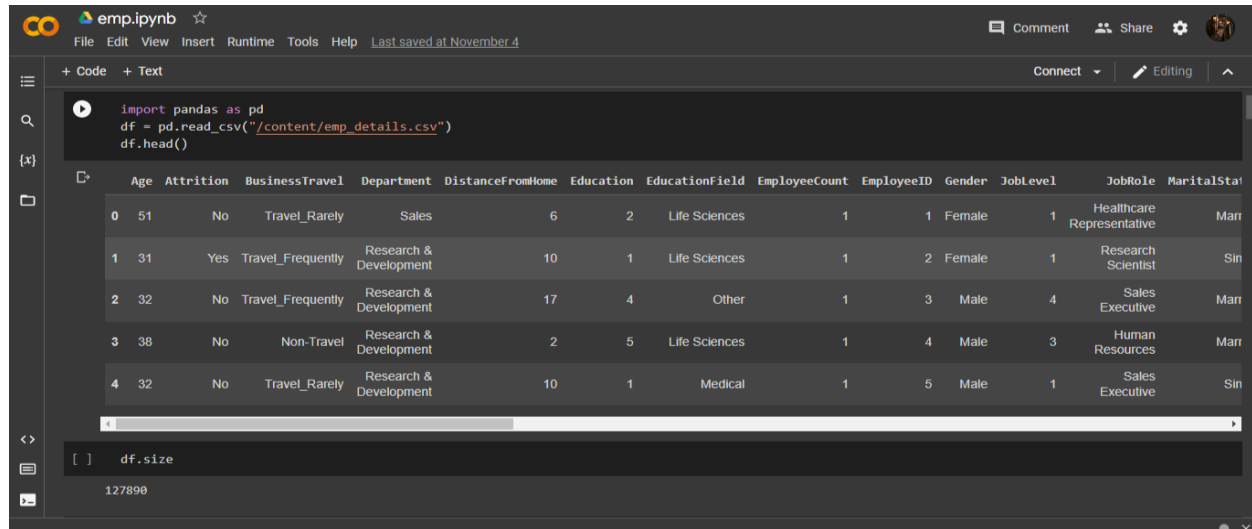
Deployment:

As the model is effectively evaluated it is made ready for deployment in the business market. Deployment phase checks how much the model can withstand in the external environment and perform superiorly as compared to others.

Feedback:

Feedback is the necessary purpose which helps in refining the model and accessing its performance and impact. Steps involved in feedback define the review process, track the record, measure effectiveness and review with refining.

This figure shows the dataframe (which isn't cleaned and sanitized as it has lots of null values)



The screenshot shows a Jupyter Notebook interface with the following code and output:

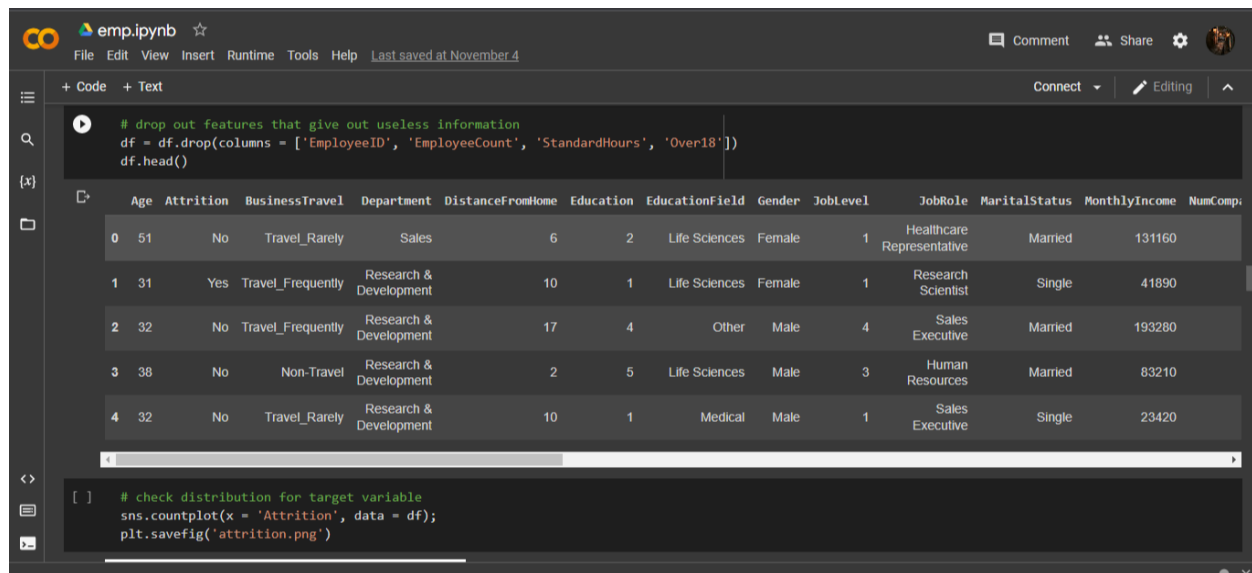
```
import pandas as pd
df = pd.read_csv("/content/emp_details.csv")
df.head()
```

| | Age | Attrition | BusinessTravel | Department | DistanceFromHome | Education | EducationField | EmployeeCount | EmployeeID | Gender | JobLevel | JobRole | MaritalStatus |
|---|-----|-----------|-------------------|------------------------|------------------|-----------|----------------|---------------|------------|--------|----------|---------------------------|---------------|
| 0 | 51 | No | Travel_Rarely | Sales | 6 | 2 | Life Sciences | 1 | 1 | Female | 1 | Healthcare Representative | Married |
| 1 | 31 | Yes | Travel_Frequently | Research & Development | 10 | 1 | Life Sciences | 1 | 2 | Female | 1 | Research Scientist | Single |
| 2 | 32 | No | Travel_Frequently | Research & Development | 17 | 4 | Other | 1 | 3 | Male | 4 | Sales Executive | Married |
| 3 | 38 | No | Non-Travel | Research & Development | 2 | 5 | Life Sciences | 1 | 4 | Male | 3 | Human Resources | Married |
| 4 | 32 | No | Travel_Rarely | Research & Development | 10 | 1 | Medical | 1 | 5 | Male | 1 | Sales Executive | Single |

```
[ ] df.size
```

127890

The below figure shows the dataset after removing unnecessary columns and the rows containing missing values and reordering the same.



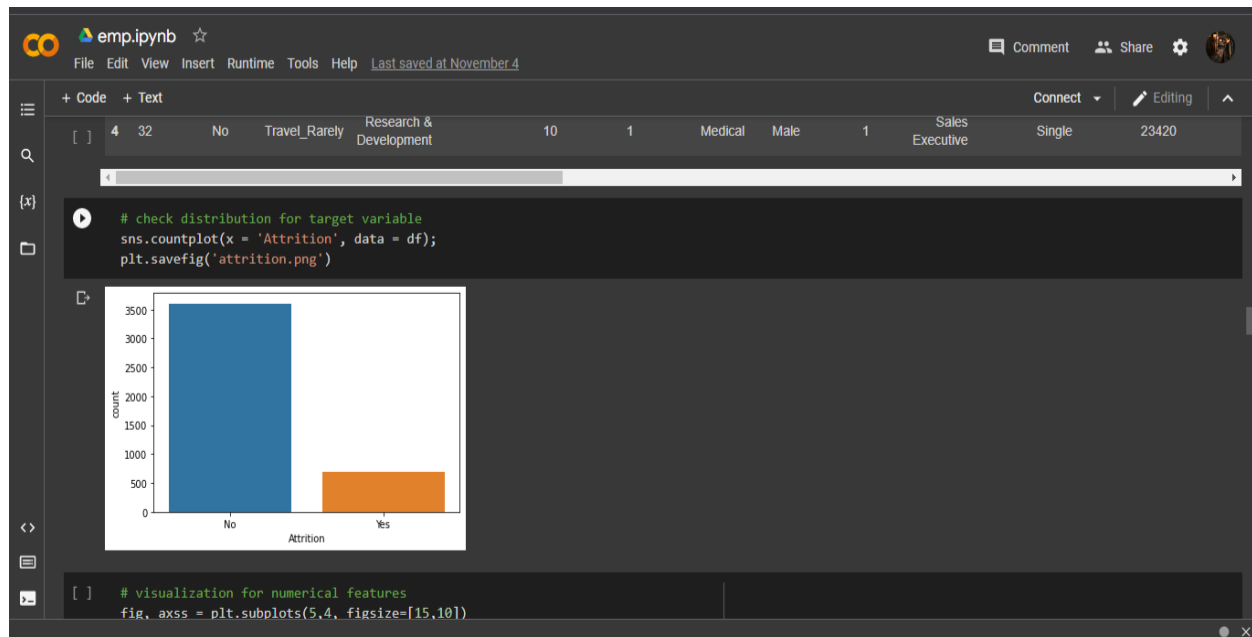
The screenshot shows a Jupyter Notebook interface with the following code and output:

```
# drop out features that give out useless information
df = df.drop(columns = ['EmployeeID', 'EmployeeCount', 'StandardHours', 'Over18'])
df.head()
```

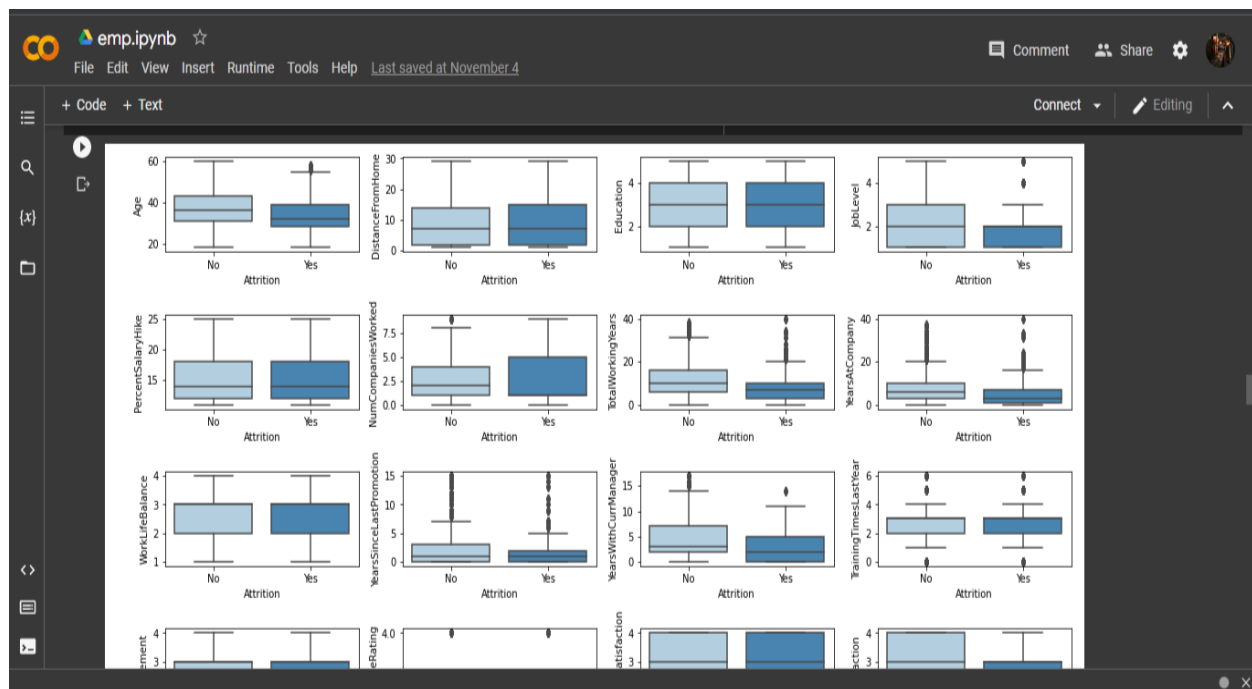
| | Age | Attrition | BusinessTravel | Department | DistanceFromHome | Education | EducationField | Gender | JobLevel | JobRole | MaritalStatus | MonthlyIncome | NumComp... |
|---|-----|-----------|-------------------|------------------------|------------------|-----------|----------------|--------|----------|---------------------------|---------------|---------------|------------|
| 0 | 51 | No | Travel_Rarely | Sales | 6 | 2 | Life Sciences | Female | 1 | Healthcare Representative | Married | 131160 | |
| 1 | 31 | Yes | Travel_Frequently | Research & Development | 10 | 1 | Life Sciences | Female | 1 | Research Scientist | Single | 41890 | |
| 2 | 32 | No | Travel_Frequently | Research & Development | 17 | 4 | Other | Male | 4 | Sales Executive | Married | 193280 | |
| 3 | 38 | No | Non-Travel | Research & Development | 2 | 5 | Life Sciences | Male | 3 | Human Resources | Married | 83210 | |
| 4 | 32 | No | Travel_Rarely | Research & Development | 10 | 1 | Medical | Male | 1 | Sales Executive | Single | 23420 | |

```
[ ] # check distribution for target variable
sns.countplot(x = 'Attrition', data = df);
plt.savefig('attrition.png')
```

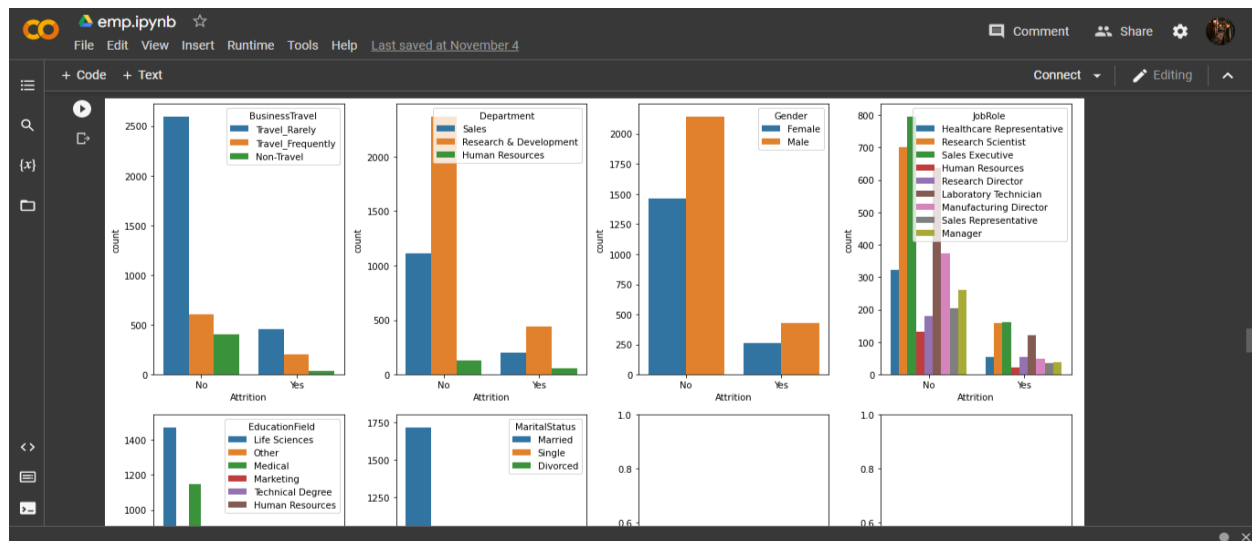
This figure shows the distribution of target variables



The visualisations of the numerical variables is



Similarly of the non numerical variables



And then, we train the model and test for unknown data to predict the attrition rate. And the screenshot is attached herewith.

```
# Logistic Regression
start_time = time.time()
train_pred_log, acc_log, acc_cv_log = fit_ml_algo(LogisticRegression(), X_train,y_train, 10)
log_time = (time.time() - start_time)
print("Accuracy: %s" % acc_log)
print("Accuracy CV 10-Fold: %s" % acc_cv_log)
print("Running Time: %s" % datetime.timedelta(seconds=log_time))
```

Accuracy: 89.8
Accuracy CV 10-Fold: 88.63
Running Time: 0:00:01.753627

```
# SVC
start_time = time.time()
train_pred_svc, acc_svc, acc_cv_svc = fit_ml_algo(SVC(),X_train,y_train,10)
svc_time = (time.time() - start_time)
print("Accuracy: %s" % acc_svc)
print("Accuracy CV 10-Fold: %s" % acc_cv_svc)
print("Running Time: %s" % datetime.timedelta(seconds=svc_time))
```

Accuracy: 88.53
Accuracy CV 10-Fold: 85.91
Running Time: 0:00:00.497278

```
# Linear SVC
start_time = time.time()
train_pred_svc, acc_linear_svc, acc_cv_linear_svc = fit_ml_algo(LinearSVC(),X_train, y_train,10)
linear_svc_time = (time.time() - start_time)
print("Accuracy: %s" % acc_linear_svc)
print("Accuracy CV 10-Fold: %s" % acc_cv_linear_svc)
print("Running Time: %s" % datetime.timedelta(seconds=linear_svc_time))
```

Accuracy: 89.89

Accuracy CV 10-Fold: 88.73

Running Time: 0:00:01.055932

8.RESULTS

The proposed work utilizes the deep learning technique along with some preprocessing steps to improve the prediction of employee attrition. Extensive experiments have been conducted to show the practical value of our work. The prediction accuracy using the original dataset is about 91%, whereas it is about 94% using a synthetic dataset.

9.ADVANTAGES AND DISADVANTAGES

9.1 ADVANTAGES:

- This paper proposed the deep learning technique along with some preprocessing
- steps to improve the prediction of employee attrition.
- To get realistic results, we derived a balanced version from the original.
- Several factors are analyzed to reveal employee attrition intercorrelation and to demonstrate the dominant ones.

9.2 DISADVANTAGES

- It requires very large amount data in order to perform better than other techniques.

10.FUTURE SCOPES

The Attrition Prediction model estimates the attrition risk for your employee populations in real-time, which is recalculated every time an employee submits feedback. The aggregated, segment-level view keeps the accuracy of your predictions high while protecting individual employee identity.

11.CONCLUSION

On the whole, this project was a useful experience. We have gained new knowledge and skills we achieved several of my learning goals. We got insight into professional practice. We learned the different facets of working. We experienced that self exploration, as in many organisations, is an important factor for the progress of projects. Related to our study we learned more about employee attrition rate prediction and the various approaches and algorithms to achieve the same. There is still a lot to discover and to improve. The methods used at the moment are still not standardized and a consistent method is in development. Furthermore we have experienced that it is of importance of each strategy and how other one is better than the current algorithm and in which application. we found that the internship is not one sided, but it is a way of sharing knowledge, ideas and opinions and implementing the same to get results. The internship was also good to find out what our strengths and weaknesses are. This helped me to define what skills and knowledge. We believe that our time spent in learning and surfing regarding various algorithms and the mathematics behind was well worth it and contributed to finding an acceptable solution to build a model and predict the employee's attrition rate. Two main things that we've learned the importance of time-management skills and self-motivation. At last this project has gave us new insights and motivation to pursue a career in machine learning domain.

12.Link to code and executable file

<https://github.com/IBM-EPBL/IBM-Project-42611-1660669649.git>