Corporate Employee Attrition Analysis

A PROJECT REPORT

Submitted By

Team ID: PNT2022TMID21456.

Team Leader: NAGANATHAN M (9177119D126),

Team Member: AATHISHWARAN D (9177119D113),

Team Member: MEIYAPPAN A (9177119D125),

Team Member : SAHEEL AQTHAR S (9177119D131).

in partial fulfillment for the award of the degree

of

BACHELOR OF ENGINEERING

in

ELECTRONICS AND COMMUNICATION ENGINEERING.

THIAGARAJAR COLLEGE OF ENGINEERING, MADUARI-625015.

TABLE OF CONTENTS

| CHAPTER NO | TITLE | PAGE NO |
|------------|------------------------------|---------|
| 1 | INTRODUCTION | 4 |
| 1.1 | Project Overview | 4 |
| 1.2 | Purpose | 4 |
| 2 | LITERATURE SURVEY | 5 |
| 2.1 | Existing problem | 5 |
| 2.2 | References | 5 |
| 2.3 | Problem Statement Definition | 6 |
| 3 | IDEATION & PROPOSED SOLUTION | 7 |
| 3.1 | Empathy Map Canvas | 7 |
| 3.2 | Ideation & Brainstorming | 8 |
| 3.3 | Proposed Solution | 11 |
| 3.4 | Problem Solution fit | 11 |
| 4 | REQUIREMENT ANALYSIS | 12 |
| 4.1 | Functional requirement | 12 |
| 4.2 | Non-Functional requirements | 13 |

| CHAPTER NO | TITLE | PAGE NO |
|------------|-----------------------------------|---------|
| 5 | PROJECT DESIGN | 14 |
| 5.1 | Data Flow Diagrams | 14 |
| 5.2 | Solution & Technical Architecture | 14 |
| 5.3 | User Stories | 15 |
| 6 | PROJECT PLANNING & SCHEDULING | 16 |
| 6.1 | Sprint Planning & Estimation | 16 |
| 6.2 | Sprint Delivery Schedule | 17 |
| 6.3 | Reports from JIRA | 17 |
| 7 | CODING & SOLUTIONING | 21 |
| 8 | TESTING | 28 |
| 8.1 | Test Cases | 28 |
| 8.2 | User Acceptance Testing | 28 |
| 9 | RESULTS | 29 |
| 10 | ADVANTAGES & DISADVANTAGES | 32 |
| 11 | CONCLUSION | 33 |
| 12 | FUTURE SCOPE | 33 |
| 13 | APPENDIX | 34 |

1. INTRODUCTION

1.1 Project overview

Employee attrition has become a vital problem across the world. It is one of the crucial issues faced by business leaders within companies where they lose the most talented employees. A good employee is always an asset to the organization and their resignation can lead to various problems like financial losses, overall performance, and loss of acquired knowledge. Furthermore, hiring new employees is far exorbitant, taxing, and time-consuming in comparison to recruiting the existing one. It is very time-consuming to recruit a new employee as it takes him months for training, adjusting to the culture, rules, and environment. Therefore, upcoming trends and technology using Machine Learning Algorithms must be exploited for the benefit of business organizations. Knowing the reason beforehand for the employee attrition, companies can mitigate this loss. This analysis provides a conclusive review of employee attrition from the data set IBM HR Analytics Employee Attrition Performance.

1.2 Purpose

Hardik P. K. (2016), researched on "a study on employee attrition: with special reference to Kerala IT Industry". His research examined the relationship between organizational factors and attrition of IT professional's. The result can conclude that the organizational factors played significant role in predicting the variance in turnover intention (attrition) of Kerala IT professionals. Therefore, the HR managers in IT

organizations may take into consideration the problems with organizational factors of their workers to reduce the turnover intention of the skilled employees.

1. LITERATURE SURVEY

2.1 Existing Problem

The Existing system includes only few attributes for analysis and also deals with qualitative observations and simple statistical analysis. The qualitative observations deal with data and can be observed through human senses. They do not involve measurements or number. Due to the increase in IOT and connected device, we now have access to so much of data and along with it an increase needs to manage and understand data.

2.2 References

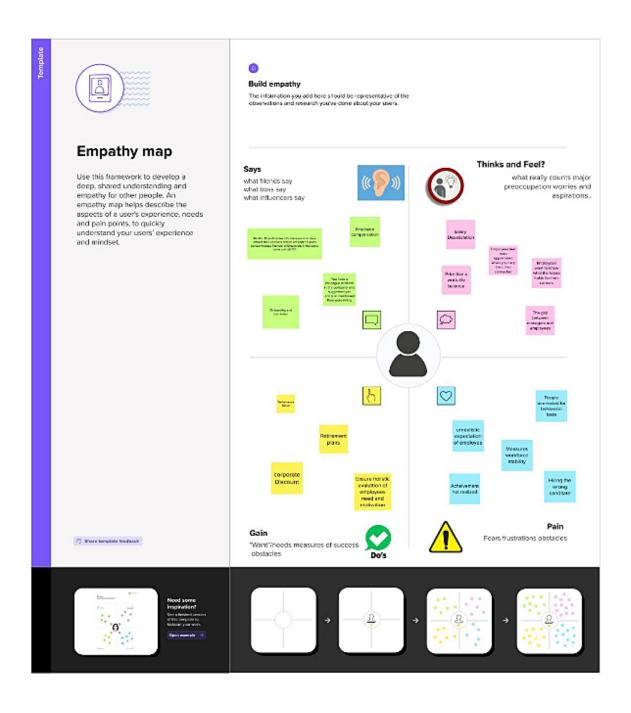
- 1. From Big Data to Deep Data to support people analytics for employee attrition prediction, Nesrine Ben Yahia, Hlel Jihen, Ricardo Colomo-Palacio (2021)
 - 2. Machine Learning Approach for Employee Attrition Analysis.Dr.
- R. S. Kamath | Dr. S. S. Jamsandekar | Dr. P. G. Naik ,Published in International Journal of Trend in Scientific Research and Development (ijtsrd), (March 2019)
- 3. Investigation of early career teacher attrition(ECT) and the impact of induction programs in Western Australia, Janine E.Wyatt, MichaelO'Neill (2021)

2.3 Problem Statement Definition

- To create a dashboard and perform analysis of employee attrition in corporates using IBM Cognos analytics platform.
- To reduce the employee attrition rate through data analytics, data visualization by analysing the major factors that causes attrition.

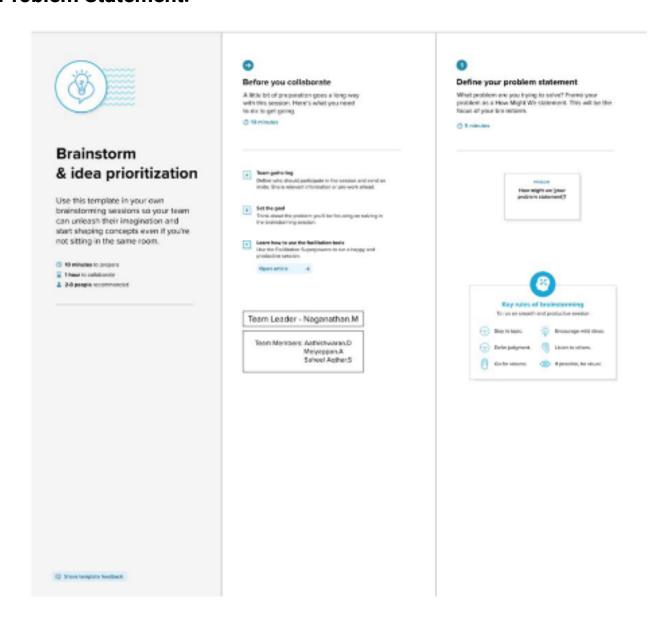
3. IDEATION AND PROPOSED SOLUTION

3.1 Empathy Map Canvas

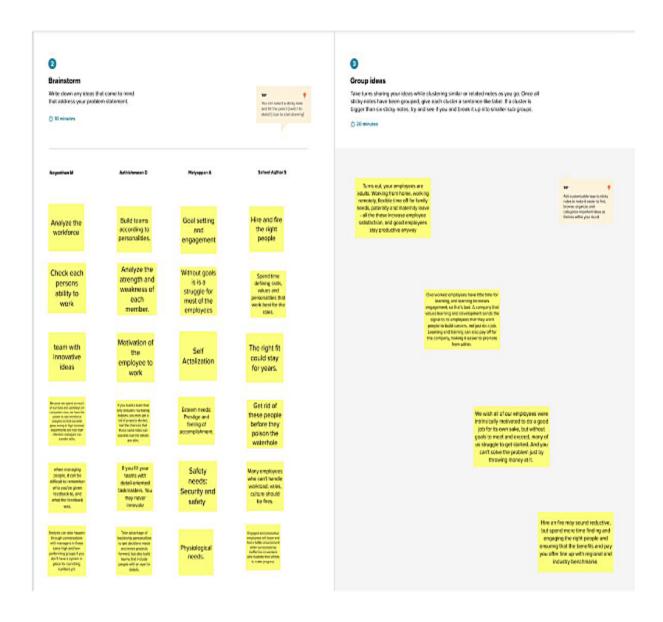


3.2 Ideation & Brainstorming

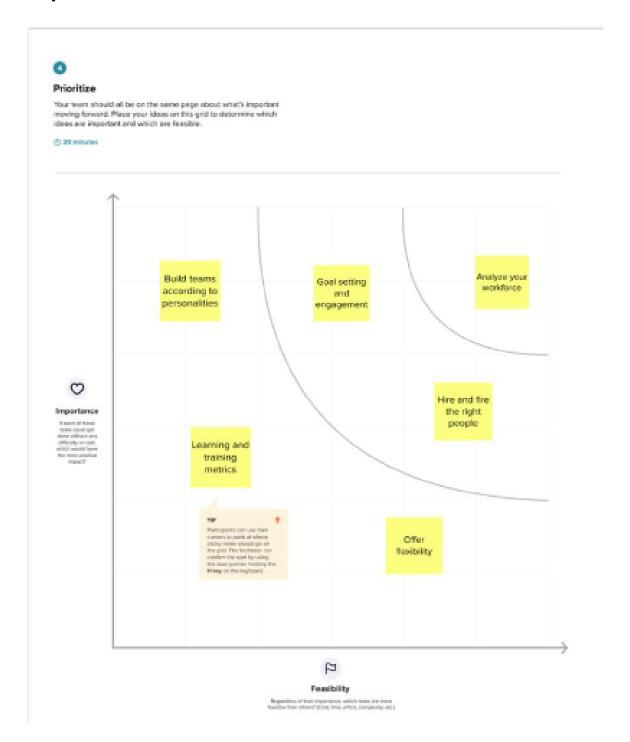
Step-1: Team Gathering, Collaboration and Select the Problem Statement.



Step-2: Brainstorm, Idea Listing and Grouping.



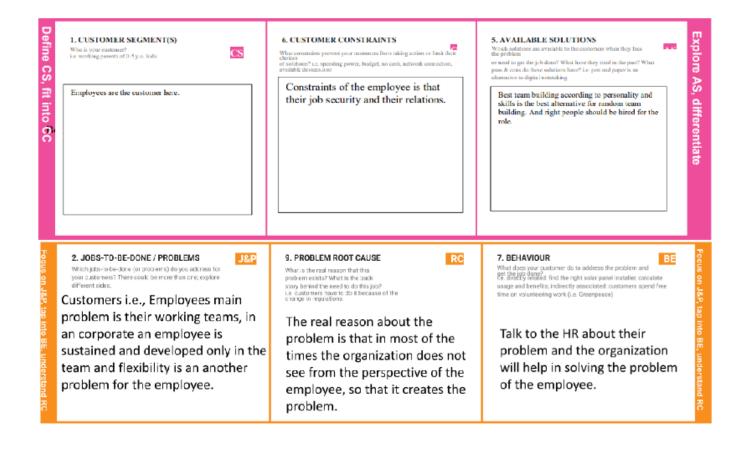
Step-3: Idea Prioritization.

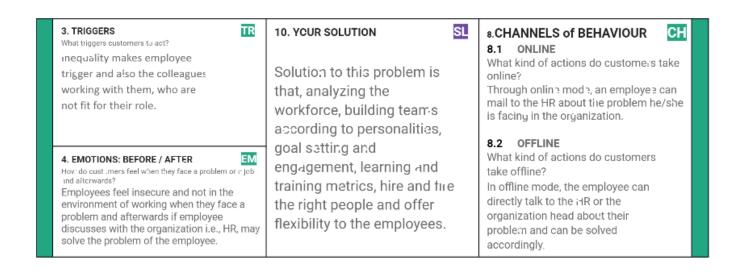


3.3 Proposed Solution

The Existing system includes only few attributes for analysis and also deals with qualitative observations and simple statistical analysis. The qualitative observations deal with data and can be observed through human senses. They do not involve measurements or number. Due to the increase in IOT and connected device, we now have access to so much of data and along with it an increase needs to manage and understand data.

3.4 Problem Solution fit





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 |
| | | Registration through Gmail |
| | | Registration through LinkedIn |
| FR-2 | User Confirmation | Confirmation via Email |
| | | Confirmation via OTP |
| FR-3 | Account Creation | Create an account in the Profile Dashboard |
| FR-4 | Input Credentials | Uploading your dataset |
| | | Analyzing the attrition rate using dashboard |
| FR-5 | Processing Methods | Using IBM Cognos Analytics Dashboard |
| | | Using Prediction algorithm to find attrition rate |
| FR-6 | Output Credentials | Using the Dashboard and Algorithm they know |
| | | about the employee attrition and way to reduce |
| | | the employee attrition |
| FR-7 | Report preparation | Record the outcame of the algorithm in |
| | | document. |
| | | |

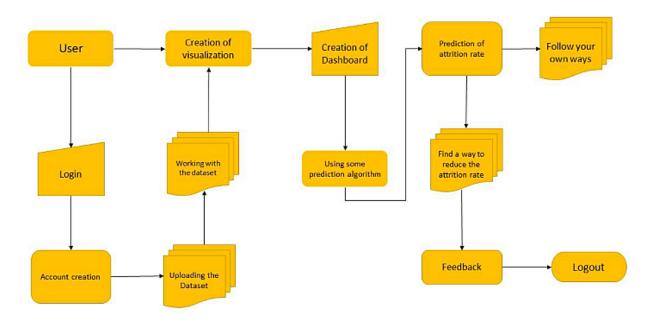
4.2 Non-Functional requirements

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

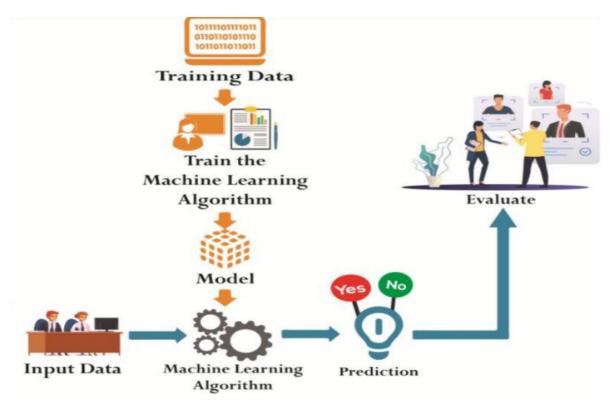
| FR No. | Non-Functional Requirement | Description |
|--------|----------------------------|---|
| NFR-1 | Usability | The user can be able to interact with the system user friendly. The system is build with a simple modules and algorithms. |
| NFR-2 | Security | Access permissions for the particular system information may only be changed by the system's data administrator. The user's data must be having an high security measures. |
| NFR-3 | Reliability | The database update process must roll back all related updates when any update fails. The dataset will not be modified by anyone only the user can be able to modify the dataset. |
| NFR-4 | Performance | The performance of the dashboard is flexibleto every user's. The front-page load time must be no more than 2 seconds for users that access the website using an LTE mobile connection. |
| NFR-5 | Availability | New module deployment mustn't impact front page, dashboard and check out pages availability and mustn't take longer than one hour. The rest of the pages that may experience problems must display a notification with a timer showing when the system is going to be up again. |
| NFR-6 | Scalability | The website attendance limit must be scalable enough to support 200,000 users at atime. The dashboard is scalable for the companies when their employee's dataset is used for analysis. The model can successfully predict the futuristic approach and suggests preventive measures. |

5. PROJECT DESIGN

5.1 Data Flow Diagrams



5.2 Solution & Technical Architecture



5.3 User Stories

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

| User Type | Requirement (Epic) Story Number | | Acceptance criteria | Priority | Release | |
|----------------------------|---------------------------------|--------|---|---|---------|----------|
| Customer (Web user) | Registration | USN-1 | As a user, I can register for the application by entering my email, password, and confirming my password. | I can access my account / dashboard | High | Sprint-1 |
| | | USN-2 | As a user, 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 a user, I can register for the application through Facebook | I can register & access the dashboard with Facebook Login | Low | Sprint-2 |
| | | USN-4 | As a user, I can register for the application through Gmail | I can register & access the dashboard with Gmail Login | Medium | Sprint-1 |
| | Login | USN-5 | As a user, I can log into the application by entering email & password | I can access my account / dashboard | High | Sprint-1 |
| | Dashboard | USN-6 | Uploading the Dataset | I can be able to upload my dataset | High | Sprint 2 |
| | | USN-7 | Working With Dataset | I can be able to access my dashboard | High | Sprint 2 |
| | | USN-8 | Visualization | I can be able to view the visual attrition rate of my dataset | High | Sprint 3 |
| | | USN-9 | Working with Dashboard | I can be able to view the various views of the attrition rate | High | Sprint 3 |
| Customer Care Executive | | USN-10 | Asking Help / Feedback | I can be able to ask help if I can face any issues or problems while using the webpage | Medium | Sprint 4 |
| Administrator | | USN-11 | Managing the Database | I can assure that my data is in secure state | High | Sprint 4 |
| | | USN-12 | Managing the over all process | I can assure that my data and process is going good | High | Sprint 4 |

6. PROJECT PLANNING

6.1 Sprint Planning & Estimation

Product Backlog, Spri. & Schedule, and Estimation (4 Marks)

| Sprint | Functional Requirement (Epic) | Usar Story Number | User Story / Task Story Points Priority | | Friority | Team Members |
|----------|-------------------------------------|----------------------|---|---|----------|------------------|
| Sprint-1 | Registration | USN 1 | As a user, I can register for the application by entering my email, password, and confirming my password. | M | | Naganathan M |
| Sprint-1 | | USN-2 | As a user, I will receive confirmation email once I have registered for the application | 1 | High | Saineel Aqthar S |
| Sprint-2 | | USN-3 | As a user, I can register for the application through Facebook | 2 | Low | Aathishwaran D |
| Sprint-1 | | USN-4 | As a user, I can register for the application through Gmail | 2 | Medium | Meiyappan A |
| Sprint-1 | Login | USN-5 | As a user, I can log into the application by entering email & password | 2 | High | Naganathan M |
| Sprint-2 | Dashboard | USN-6 | As a user, I can able to access the dashboard | 4 | Medium | Saheel Aqthar S |
| Sprint-2 | | USN-7 | As a user, I can able to upload my dataset through dashboard | 2 | High | Aathishwaran D |
| Sprint-3 | | USN-8 | As a user, I can able to done a Data Pre-processing | ż | Medium | Meiyappan A |
| Sprint-3 | | USN-9 | As a user, I can able to build a model for my dataset – Train the model | 4 | Low | Naganathan M |

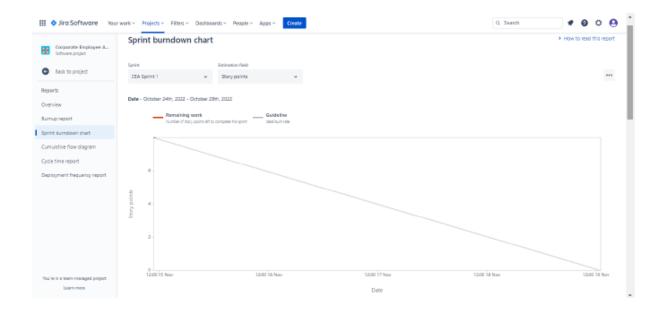
| Sprint | Functional Requirement (Epic) | User Story Number | User Story / Task | Story Points | Priority | Team Members |
|----------|-------------------------------------|----------------------|---|-----------------|----------|-----------------|
| Sprint-3 | | USN-10 | As a user, I can able to test my model | 4 | Low | Saheel Aqthar S |
| Sprint-3 | | USN-11 | As a user, I can able to evaluate my performance | 3 | Medium | Aathishwaran D |
| Sprint-4 | | USN-12 | As a user, I can able find a prediction of my dataset attrition rate using algorithm | 5 | High | Meiyappan A |
| Sprint-4 | | USN-13 | As a user, I can able view the visualization of my dataset in the dashboard | 5 | High | Naganathan M |
| Sprint-2 | | USN-14 | As a user, I can to ask the help to the development team | 3 | Low | Saheel Aqthar S |
| Sprint-4 | Database | USN-15 | As a user, I can assure that my information are in the safe state | 5 | Medium | Aathishwaran D |
| Sprint-2 | Logout | USN-16 | As a user, I can able to logout the page with my presence | 2 | Medium | Saheel Aqthar S |

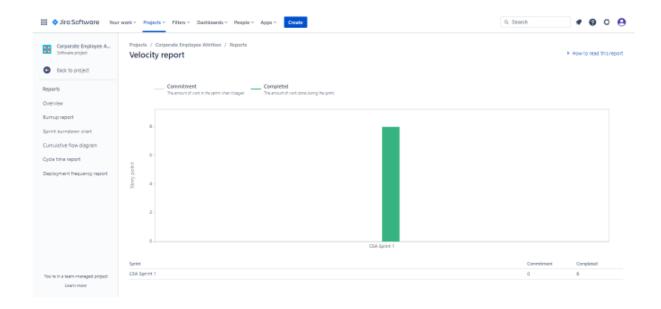
6.2 Sprint Delivery Schedule

| Sprint | Total Story Points | Duration | Sprint Start Date | Sprint End Date (Planned) | Story Points Completed (a.s on Planned End Date) | Sprint Release Date (Actual) |
|----------|-----------------------|----------|-------------------|------------------------------|--|------------------------------|
| Sprint-1 | 7 | 6 Days | 24 Oc 2022 | 29 Oc. 2022 | 7 | 29 Oci 2022 |
| Sprint-2 | 13 | 6 Days | 31 Oct 2022 | 05 Nov 2022 | 13 | 05 Ni·√ 2022 |
| Sprint-3 | 14 | 6 Days | 07 Nov 2022 | 12 Nov 2022 | 14 | 12 Nov 2022 |
| Sprint-4 | 15 | 6 Days | 14 Nov 2022 | 19 Nov 2022 | 15 | 19 Nov 2022 |

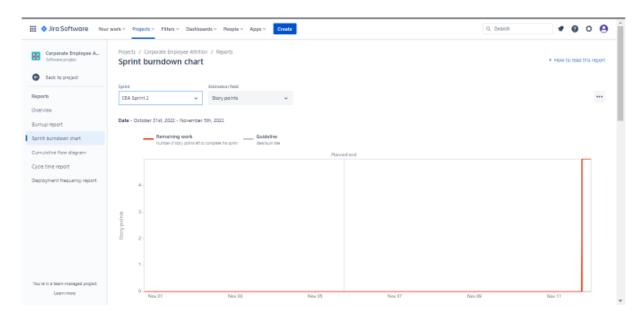
6.3 Reports from JIRA

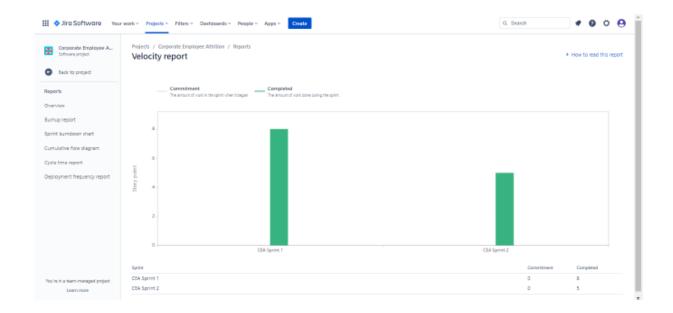
Sprint 1:



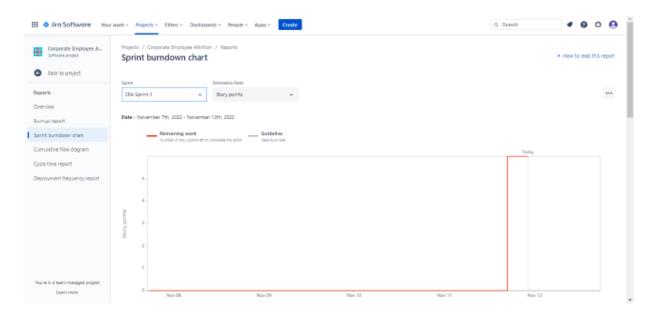


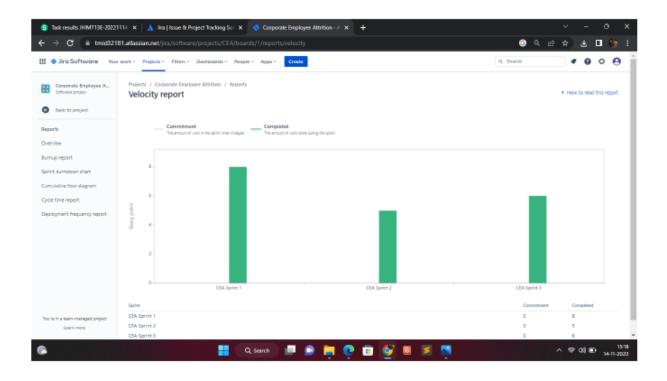
Sprint 2:



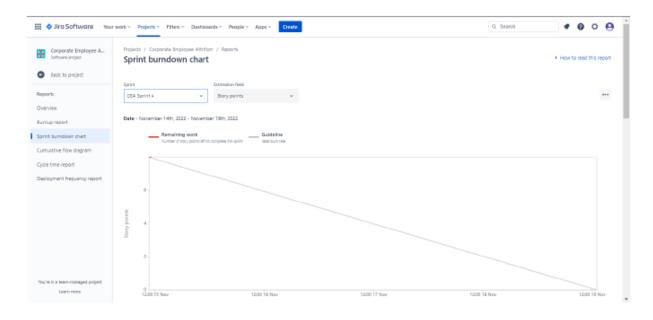


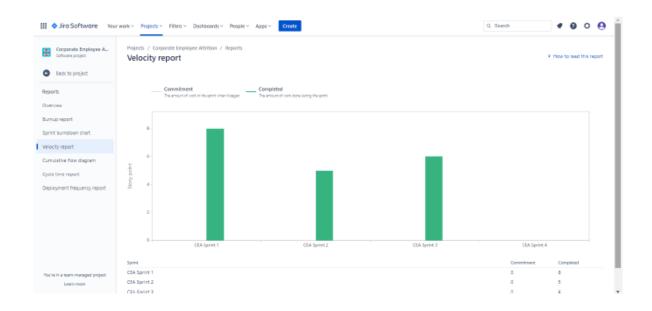
Sprint 3:



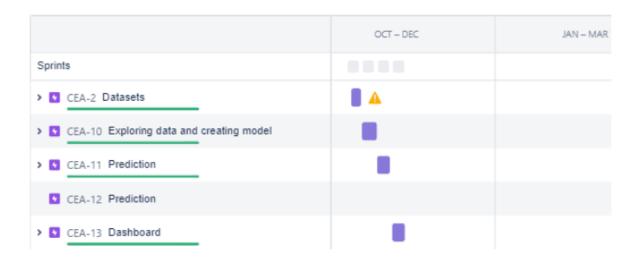


Sprint 4:





Road Map:



7. CODING & SOLUTIONING

```
from google.colab import drive
drive.mount('/content/drive')

#GENERAL
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

```
#FEATURE ENGINEERING
from sklearn.preprocessing import LabelEncoder
from imblearn.over_sampling import SMOTE
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
#MODEL SELECTION
from sklearn.model_selection import KFold
from sklearn.model_selection import cross_val_score
from sklearn.model_selection import GridSearchCV
#MODEL
from sklearn.linear_model import LogisticRegression
from sklearn.ensemble import RandomForestClassifier
from sklearn.svm import SVC
from sklearn.tree import DecisionTreeClassifier
#MODEL SCORES
from sklearn.metrics import confusion_matrix , accuracy_score
,classification_report
#FEATURE IMPORTANCE
from sklearn.inspection import permutation_importance
path = '/content/drive/MyDrive/Colab Notebooks/HR-Employee-Attrition.csv'
df =pd.read_csv(path)
df
df.shape
df.info()
df.select_dtypes('int64', 'float64').columns
cat_cols = df.select_dtypes('object').columns
cat_cols
df.describe().T
df
for cat in cat cols:
    print(cat ,'-> ' , df[cat].unique())
   print()
print("All columns Unique values count")
for col in df:
    print(col, len(df[col].unique()), sep=': ')
plt.figure(figsize =(14,5))
plt.subplot(1, 2, 1)
```

```
sns.countplot(df['Attrition'] , color = 'b' , hue = df['Gender'])
plt.title('Attrition by Gender')
plt.subplot (1, 2, 2)
plt.pie(df['Attrition'].value_counts() ,colors =['r' ,'c'] ,explode =[0,0.1]
, autopct = '%.2f' , labels =['No' , 'Yes'])
plt.title('Attrition')
plt.figure(figsize =(16,4))
plt.subplot(1,3,1)
sns.distplot(df['Age'] , color ='m')
plt.title('Age')
plt.subplot(1, 3, 2)
sns.stripplot(x = 'Gender', y = 'Age', data = df, palette="Set2")
plt.title('Gender vs Age')
plt.subplot(1,3,3)
sns.countplot('Gender', data = df, color ='c')
plt.title('Gender')
plt.tight_layout()
plt.figure(figsize = (14, 13))
plt.subplot(2,1,1)
sns.countplot(y= 'JobRole' , data = df , palette='winter_r')
plt.title('JOB ROLE')
plt.subplot(2,1,2)
sns.countplot(y= 'JobRole' ,data = df ,palette='winter_r' ,hue
=df['Attrition'])
plt.figure(figsize =(14,5))
plt.subplot (1, 2, 1)
sns.countplot('Department', data = df, hue ='Attrition'
, palette='gist_rainbow_r')
plt.subplot(1, 2, 2)
plt.pie(df['Department'].value_counts() ,autopct ='%.2f' ,colors = ['r' ,'c'
,'g'], labels =['Research & Development', 'Sales', 'Human Resources'] , explode
=[0 ,0.1,0])
#HANDLING CATEGORICAL OUTPUT VARIABLE
df['Attrition'].replace({'Yes':1 , 'No':0} , inplace = True)
df['Attrition'].head()
plt.figure(figsize =(14,10))
plt.subplot(2,2,1)
sns.countplot(df['JobSatisfaction'] , hue =df['Attrition'] , palette='Accent_r')
plt.subplot(2, 2, 2)
sns.countplot(df['EnvironmentSatisfaction'] , hue =df['Attrition']
, palette='Accent')
plt.subplot (2, 2, 3)
```

```
sns.countplot(df['JobInvolvement'] , hue =df['Attrition'] , palette='brg_r')
plt.subplot(2, 2, 4)
sns.countplot(df['PerformanceRating'] , hue =df['Attrition']
, palette='twilight_r')
plt.figure(figsize =(20, 8))
sns.boxplot(x = 'JobRole', y = 'MonthlyIncome', data = df, hue = 'Attrition'
, color ='red')
lt.figure(figsize =(12,10))
plt.subplot(2,1,1)
sns.boxplot(x = 'MaritalStatus', y = 'RelationshipSatisfaction', data = df, hue
= 'Attrition', color = 'g')
plt.subplot(2,1,2)
sns.boxplot(df['JobLevel'], df['MonthlyIncome'] , hue = df['Attrition']
, palette='Reds_r')
col = ['YearsInCurrentRole' ,'YearsSinceLastPromotion' ,'YearsWithCurrManager'
, 'YearsAtCompany']
plt.figure(figsize = (10, 10))
for i, c in enumerate(col):
    plt.subplot(2,2,i+1)
    sns.distplot(df[c] ,color ='b')
plt.figure(figsize = (16 , 16))
sns.heatmap(df.corr() , cmap = 'ocean' , cbar = True , annot = True)
no_use = []
for col in df.columns:
    if(len(df[col].unique()) ==1):
        no_use.append(col)
no use
df.drop(columns = no_use , axis = 1 , inplace = True)
df.columns
y_n_type = []
others =[]
for col in df.select_dtypes('object').columns:
    if(len(df[col].unique()) ==2):
        y_n_type.append(col)
y_n_type
df['Gender'].replace({'Male':1 , 'Female':0} ,inplace = True)
df['OverTime'].replace({'Yes':1 ,'No':0} ,inplace = True)
```

```
others = df.select_dtypes('object').columns
others
le = LabelEncoder()
for col in others:
    df[col] = le.fit_transform(df[col])
df.select_dtypes('object').columns
x = df.drop('Attrition', axis = 1)
y = df['Attrition']
print(x.shape ,y.shape)
sns.countplot(df['Attrition'])
(df.Attrition.value_counts()/1470)*100
smote = SMOTE(sampling_strategy='minority')
x , y = smote.fit_resample(x , y)
print(x.shape , y.shape)
y.value_counts()
sns.countplot(y ,palette='viridis')
plt.title('Now Class is Balanced')
x_train , x_test , y_train ,y_test = train_test_split(x , y, test_size=0.2 ,
random_state= 52)
print(x_train.shape)
#scaling the data
sc = StandardScaler()
x_train = sc.fit_transform(x_train)
x \text{ test} = \text{sc.transform}(x \text{ test})
x_train
k = KFold(n_splits = 5)
lr_model = LogisticRegression()
lr_score = cross_val_score(lr_model , x_train , y_train , cv = k , scoring =
'neg_mean_squared_error')
lr_score.mean()
rf_model = RandomForestClassifier()
rf_score = cross_val_score(rf_model , x_train , y_train , cv = k , scoring =
'neg_mean_squared_error')
rf_score.mean()
svm_model = SVC()
```

```
svm_score = cross_val_score(svm_model , x_train , y_train , cv = k , scoring =
'neg_mean_squared_error')
svm_score.mean()
dt model = DecisionTreeClassifier()
dt_score = cross_val_score(dt_model , x_train , y_train , cv = k , scoring = train , v_train ,
'neg_mean_squared_error')
dt_score.mean()
plt.figure(figsize = (14, 6))
plt.subplot(1, 2, 1)
x = ['Logistic Regression', 'Random Forest', 'Support Vector', 'Decision Tree']
y = [lr_score.mean() , rf_score.mean() , svm_score.mean() , dt_score.mean()]
plt.title('Neg Mean square error for Models')
sns.barplot(y, x, palette="viridis")
plt.subplot(1, 2, 2)
plt.plot(x ,y,marker = 'o' ,color = 'r',mfc = 'b' ,ms =8 )
plt.title('Neg Mean square error')
#we obtained less less -ve mena sq error for SVC and random forest
#lets try building model with both of them
model_params ={
           'RandomForestClassifier':
                       'model':RandomForestClassifier(),
                       'param':
                         'n_estimators':[10 ,50 ,100,130],
                          'criterion':['gini' ,'entropy'],
                         'max_depth':range(4,8,1),
                          'max_features':['auto' ,'log2']
                       }
           },
            'SVC':
                       'model':SVC(),
                       'param':
                                   'C':[1,20],
                                   'gamma': [1, 0.1],
                                   'kernel':['rbf']
}
scores =[]
```

```
for model_name , mp in model_params.items():
    model_sel = GridSearchCV(estimator= mp['model'] ,param_grid= mp['param']
, cv = 4 , return train score=False)
    model sel.fit(x train, y train)
    scores.append({
        'model':model_name,
        'best_score':model_sel.best_score_,
        'best_params':model_sel.best_params_
    })
scores
svm_model = SVC(C=20 , gamma=0.1 , kernel='rbf')
svm_model.fit(x_train,y_train)
ytest_pred = svm_model.predict(x_test)
ytrain pred = svm model.predict(x train)
accuracy_score(y_test , ytest_pred)
print(classification_report(y_test , ytest_pred))
print(classification_report(y_train , ytrain_pred))
sns.heatmap(confusion_matrix(y_test , ytest_pred) , annot = True , cmap ='ocean')
sns.heatmap(confusion_matrix(y_train , ytrain_pred) , annot = True , cmap
= 'Spectral_r')
from sklearn.inspection import permutation_importance
perm_importance = permutation_importance(svm_model, x_test, y_test)
perm_importance
perm importance.importances mean
df.columns
cols = ['Age', 'BusinessTravel', 'DailyRate', 'Department',
       'DistanceFromHome', 'Education', 'EducationField', 'EmployeeNumber',
       'EnvironmentSatisfaction', 'Gender', 'HourlyRate', 'JobInvolvement',
       'JobLevel', 'JobRole', 'JobSatisfaction', 'MaritalStatus',
       'MonthlyIncome', 'MonthlyRate', 'NumCompaniesWorked', 'OverTime',
       'PercentSalaryHike', 'PerformanceRating', 'RelationshipSatisfaction',
       'StockOptionLevel', 'TotalWorkingYears', 'TrainingTimesLastYear',
       'WorkLifeBalance', 'YearsAtCompany', 'YearsInCurrentRole',
       'YearsSinceLastPromotion', 'YearsWithCurrManager']
features = np.array(cols)
plt.figure(figsize = (14, 10))
sorted_idx = perm_importance.importances_mean.argsort()
sns.barplot( perm_importance.importances_mean[sorted_idx] , features[sorted_idx]
```

```
)
plt.xlabel("Permutation Importance")
plt.title('FEATURE IMPORTANCE')
```

8. TESTING

8.1 Test Cases

1. Purpose of Document

The purpose of this document is to briefly explain the test coverage and open issue of corporateemployee attrition at the time of the release.

8.2 User Acceptance Testing

2. Defect Analysis

This report shows the number of resolved or closed bugs at each severity level, and how they were resolved.

| Resolution | Severity 1 | Severity 2 | Severity 3 | Severity 4 | Subtotal |
|----------------|------------|------------|------------|------------|----------|
| By Design | 3 | 2 | 0 | 0 | 5 |
| Duplicate | 4 | 0 | 2 | 0 | 6 |
| External | 3 | 2 | 0 | 0 | 5 |
| Fixed | 1 | 0 | 1 | 0 | 2 |
| Not Reproduced | 0 | 3 | 3 | 0 | 6 |
| Skipped | 0 | 0 | 3 | 2 | 5 |
| Won't Fix | 0 | 0 | 1 | 0 | 1 |
| Totals | 11 | 7 | 10 | 2 | 30 |

3. Test Case Analysis

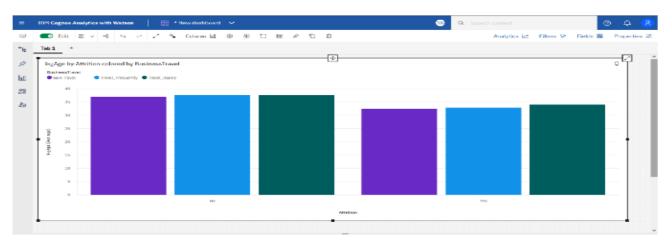
| Section | Total Cases | Not Tested | Fail | Pass |
|----------------------------|-------------|------------|------|------|
| Login Page | 1 | 0 | 0 | 1 |
| Employee Attrition Details | 1 | 0 | 0 | 1 |

| Database | 2 | 0 | 0 | 2 |
|---------------------|---|---|---|---|
| Dashboard | 1 | 0 | 0 | 1 |
| Visualize the data | 8 | 0 | 0 | 8 |
| Logistic Regression | 4 | 0 | 0 | 4 |

9. RESULTS

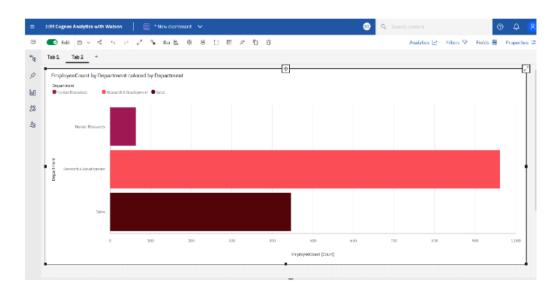
9.1 Performance Metrics

1.ATTRITION STATUS BY AGE:



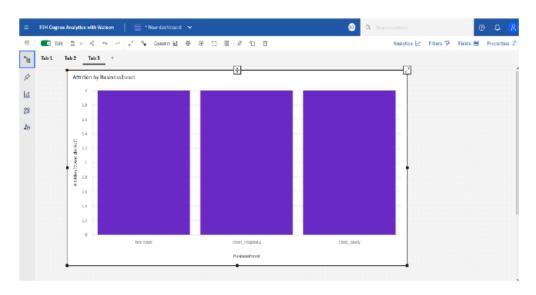
This visualization explains about the attrition status by age prediction through column chart.





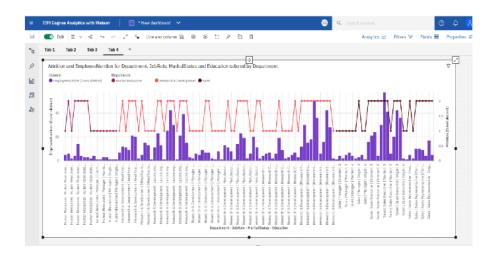
The visualized bar chart will clearly examine the employee count analyised by different departments.

3.ATTRITION BASED ON BUSINESS TRAVEL:



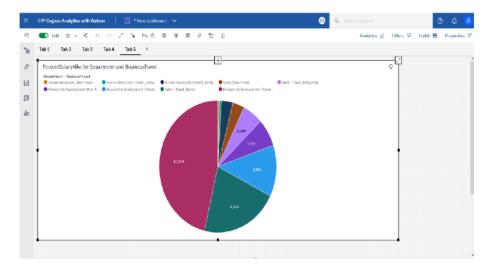
Visualization performed using the waterfall chart to view the attrition based on business travel.

4.ATTRITION BASED ON DEPARTMENT, JOBROLE, EDUCATION & MARITAL STATUS:



With the help of employee data set, the above visualized Line and Column chart explicit the attrition of employees based on department, job role, education and marital status which helps to analyze further implementation.

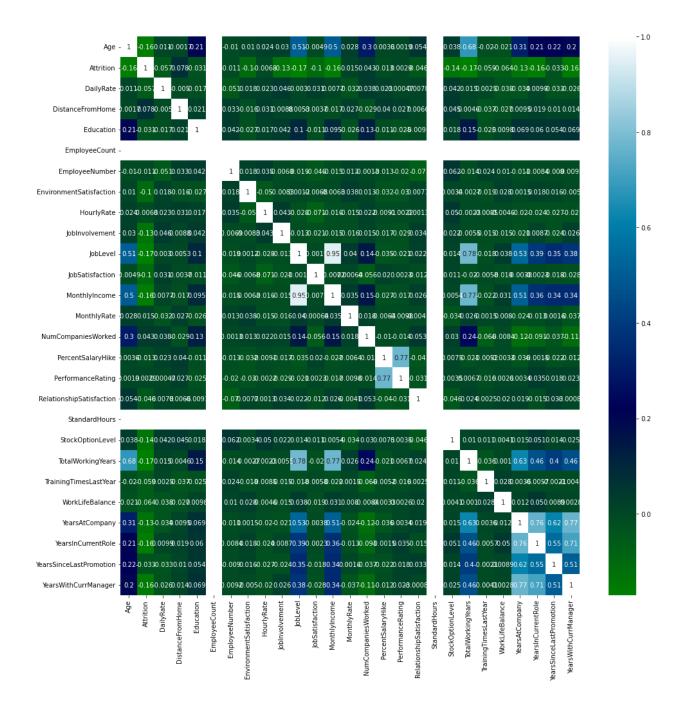
5.ATTRITION BASED ON SALARY HIKE PERCENTAGE:



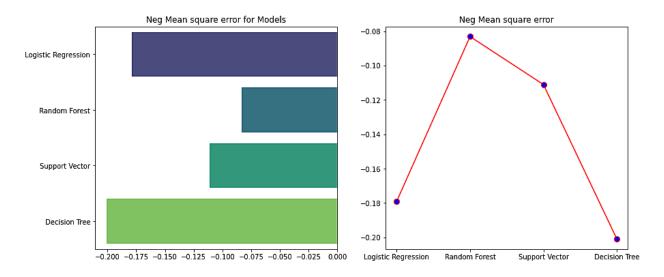
Using the employee data set, I predict the visualization using pie chart which may use to find the attrition between the salary to percentage hike.

9.2 Performance Analysis:

CORRELATION MATRIX



MODEL SELECTION



MODEL OPTIMAL PARAMETER SELECTION USING GRID SEARCH CV

We obtained less less -ve mena sq error for SVC and random forest (NEARLY SAME). Building optimal parameter model with both of them

```
In [48]:
           scores =[]
           for model_name , mp in model_params.items():
               model_sel = GridSearchCV(estimator= mp['model'] ,param_grid= mp['param'] ,cv = 4 ,return_train_score=False)
               model_sel.fit(x_train,y_train)
               scores.append({
                   'model':model_name,
                   'best_score':model_sel.best_score_,
                   'best_params':model_sel.best_params_
               })
           scores
Out[48]: [{'model': 'RandomForestClassifier',
            'best_score': 0.8788032454361056,
            'best_params': {'criterion': 'gini',
             'max_depth': 7,
'max_features': 'log2',
             'n_estimators': 100}},
           {'model': 'SVC',
            'best_score': 0.9284989858012171,
            'best_params': {'C': 20, 'gamma': 0.1, 'kernel': 'rbf'}}]
```

SELECTED MODEL -> SVC MODEL AND LINEAR REGRESSION

10. ADVANTAGES & DISADVANTAGES

10.1 Advantages

Data Collection:

The study is conducted among working IT professionals of two different categories. This categorization mainly was focused on experience level and role in the organization. It was important to know the views of candidates who seek for the job for various reasons as well as the views of interviewers involved in the process of hiring the candidates. The research study involves reference of both primary and secondary data. Primary Data Primary data is collected through a field survey with the help of a structured self-administrated Questionnaire. The survey consisted of close ended questions by the means of convenience sampling. The scaling technique installed in the questionnaire is 5-point rating scale. Total 120 respondent were IT professionals belonging to the organizations from Nagpur, Pune and Mumbai cities in Maharashtra. Secondary Data Secondary data is collected by referring to the Journals, research papers and published data in the form of books and newspapers.

Type of Research:

The research paper adopted the descriptive research design methodology. Sample Design, Sample Size and Sampling Method The sample selected for the study is an Indian Information Technology Industry. The nature of the sample is restricted to working professionals in Information Technology sector and is collected through the convenience sampling technique. The sample size was 120 respondents.

11. CONCLUSION

Employees as well as organizations must be clear with their expectations regarding the job profile. Any sort of mismatch leads to discrepancy and employees may fail to perform at theirjob. This eventually leads to attrition. Organizations should state the requirements and expectations unambiguously. This helps candidates decide upon to accept the job position or not. This eventually avoids further conflicts in the employment terms.

12. FUTURE SCOPE

Research findings suggest that attrition reasons in IT organizations primarily revolve around professional growth and challenges in the organization. Although economic factors happen to the most influential factor, professionals may settle for second best criteria of their preference that is career growth and supportive work policies in the organization. On the other hand, candidates who aspire to have a better job than the one in hand are more interested in securing the next job. Young talent wants to work on latest technology and functional domain. IT professionals who are young career makers are less influenced by Brand name or geographical area. Most of the IT professionals look for challenging role and position in the organization. Candidates as well as senior professionals believe that challenging work motivate them to maintain the interest in the work life. Employees as well as organizations must be clear with their expectations regarding the job profile. Any sort of mismatch leads to discrepancy and employees may fail to perform at their job. This eventually leads to attrition. requirements Organizations should state the and expectations unambiguously. This helps candidates decide upon to accept the job position or not. This eventually avoids further conflicts in the employment terms. Further this research can make more detailed conclusions over "mapping of candidates' expectations with organizations' requirement" by collecting the data focusing on all the steps of recruitment and selection process.

13. APPENDIX

13.1 Source Code

https://github.com/IBM-EPBL/IBM-Project-26015-1659979775/blob/main/Final%20Deliverables/Final_Code.ipynb

13.2 Project Git Hub

https://github.com/IBM-EPBL/IBM-Project-26015-1659979775