

NAALAIYA THIRAN PROJECT - 2022 19ECI01-PROFESSIONAL READINESS FOR INNOVATION, EMPLOYABILITY AND ENTREPRENEURSHIP











Corporate Employee Attrition Analytics

A PROJECT REPORT

Submitted by

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PERI INSTITUTE OF TECHNOLOGY ANNA UNIVERSITY: CHENNAI 600 025 BONAFIDE CERTIFICATE

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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

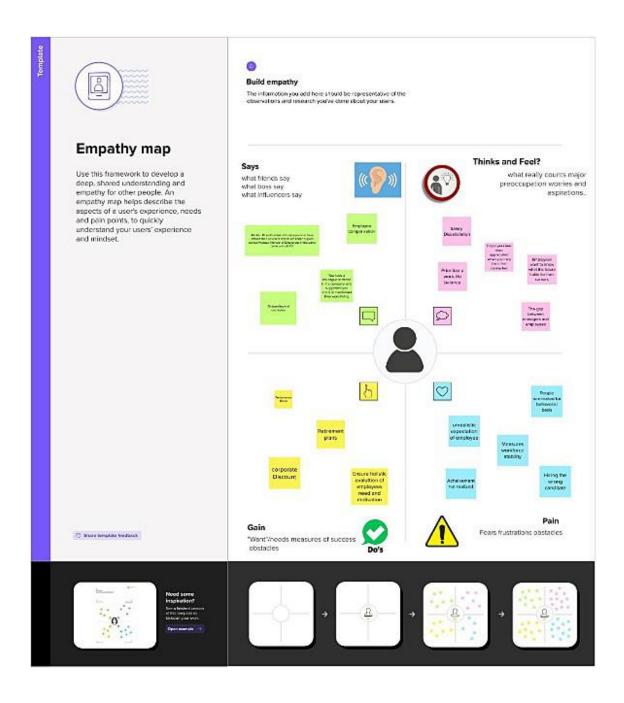
- 1. From Big Data to Deep Data to support people analytics for employee attrition prediction, Nesrine Ben Yahia, Hlel Jihen, Ricardo Colomo-Palacio (2021)
 - 2. Machine Learning Approach for Employee Attrition Analysis.Dr.
- R. S. Kamath | Dr. S. S. Jamsandekar | Dr. P. G. Naik ,Published in International 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.

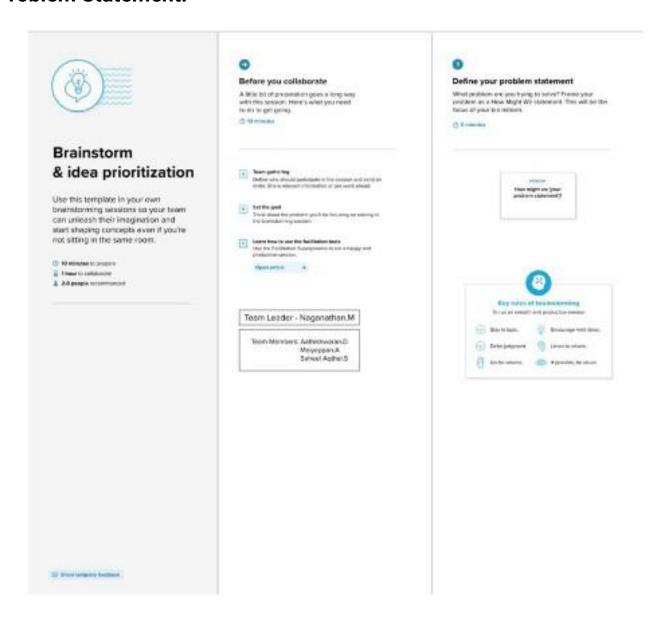
3. IDEATION AND PROPOSED SOLUTION

3.1 Empathy Map Canvas

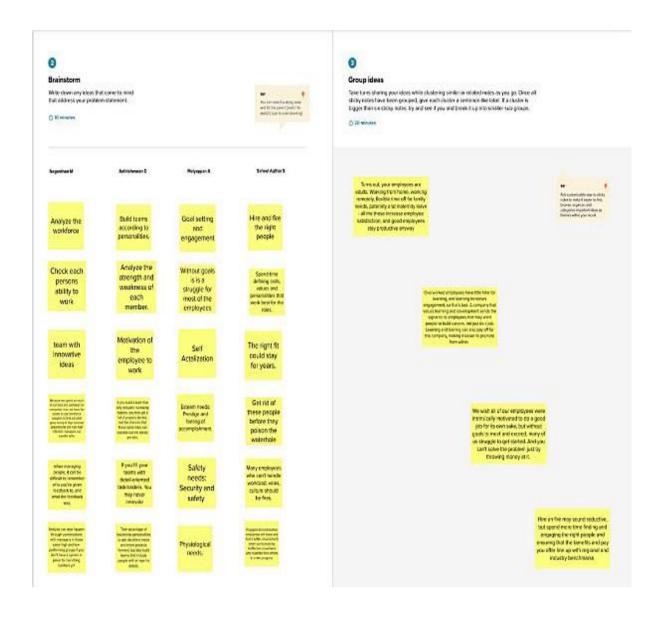


3.2 Ideation & Brainstorming

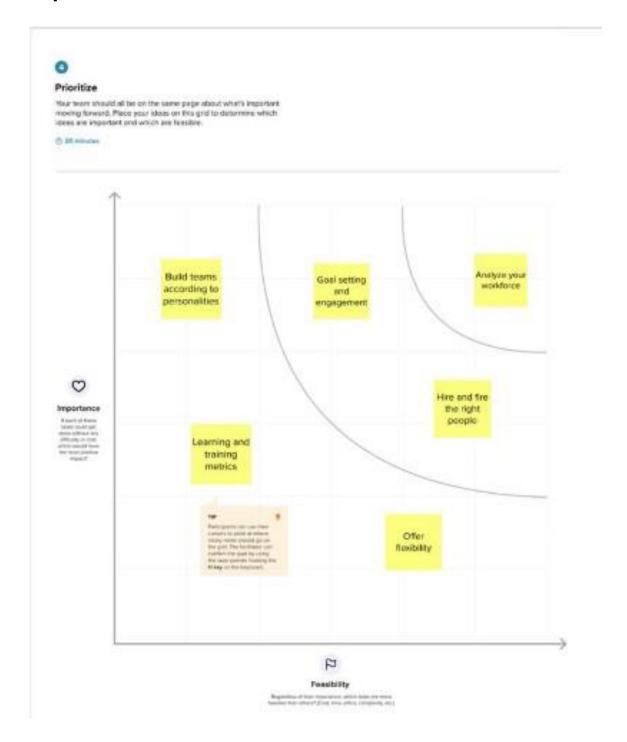
Step-1: Team Gathering, Collaboration and Select the Problem Statement.



Step-2: Brainstorm, Idea Listing and Grouping.



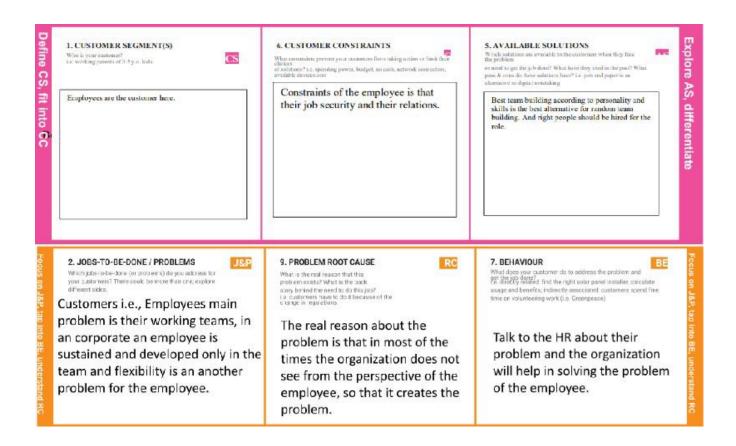
Step-3: Idea Prioritization.

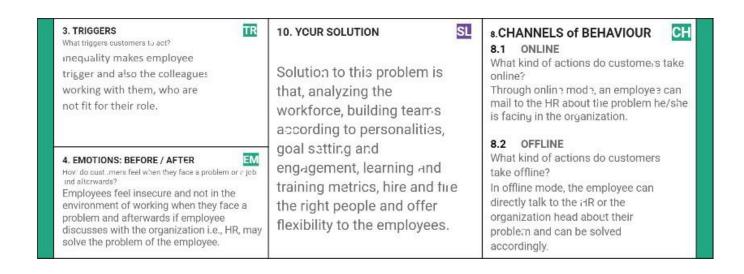


3.3 Proposed Solution

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
	AST 0.0	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
	S.	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
	\$0	about the employee attrition and way to reduce
		the employee attrition
FR-7	Report preparation	Record the outcame of the algorithm in
	540 900	document.

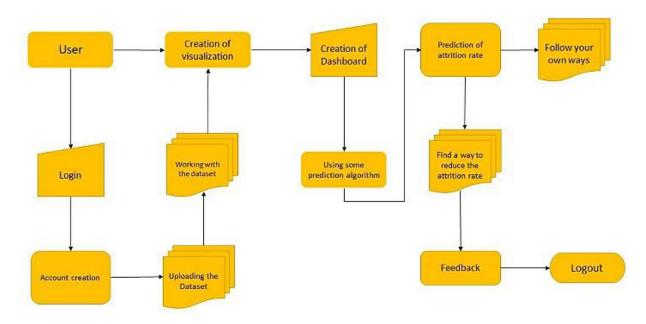
4.2 Non-Functional requirements

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

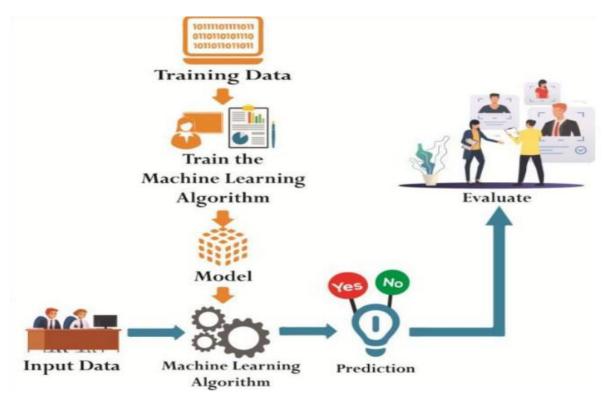
FR No.	Non-Functional Requirement	Description			
NFR-1	Usability	The user can be able to interact with the system user friendly. The system is build with a simple modules and algorithms.			
NFR-2	Security	Access permissions for the particular system information may only be changed by the system's data administrator. The user's data must be having an high security measures.			
NFR-3	Reliability	The database update process must roll back related updates when any update fails. The dataset will not be modified by anyone only user can be able to modify the dataset.			
NFR-4	Performance	The performance of the dashboard is flexible to 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	Functional Requirement (Epic)	User Story Number	User Story / Task	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
	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	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.	ering my email,		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	3	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

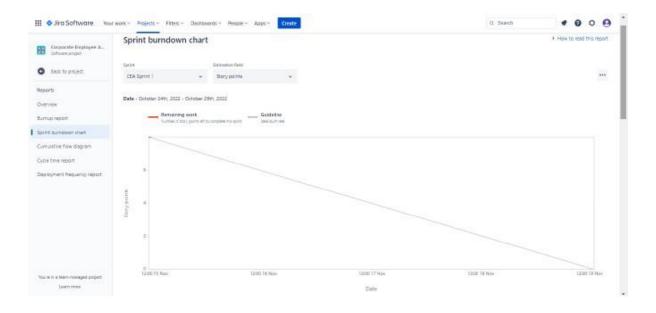
Sprint	Functional Requirement (Epic)	User Story Number	User Story / Task	Story Points	Priority	Team Members
Sprint-3		USN-10	As a user, I can able to test my model	4	Low	Saheel Aqthar S
Sprint-3		USN-11	As a user, I can able to evaluate my performance	3	Medium	Aathishwaran D
Sprint-4		USN-12	As a user, 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

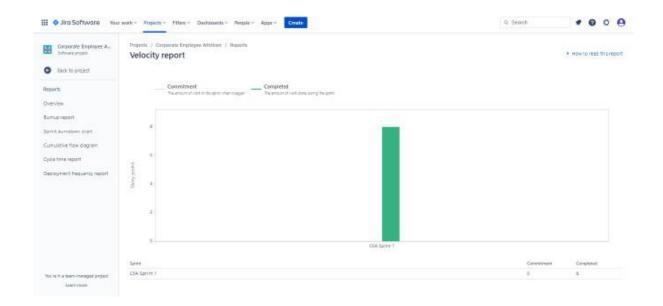
6.2 Sprint Delivery Schedule

Sprint	Totai Story Points	Dutation	Sprint Start Date	Sprint End Date (Planned)	Story Points Completed (as on Planned End Date)	Sprint Release Date (Actual)
Sprint-1	7	ю̀ Days	24 Oc 2022	29 Oc. 2022	7	29 Oci 2022
Sprint-2	13	6 Days	31 Oct 2022	05 i lov 2022	13	05 N₁·√ 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

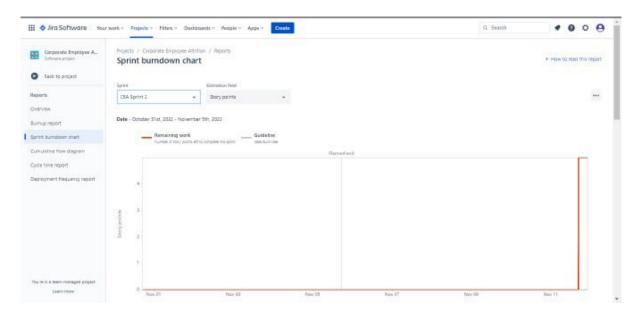
6.3 Reports from JIRA

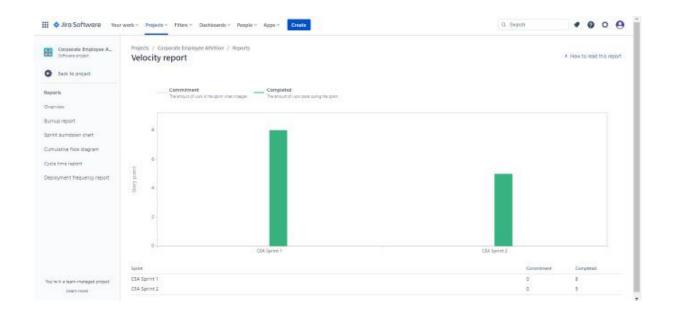
Sprint 1:



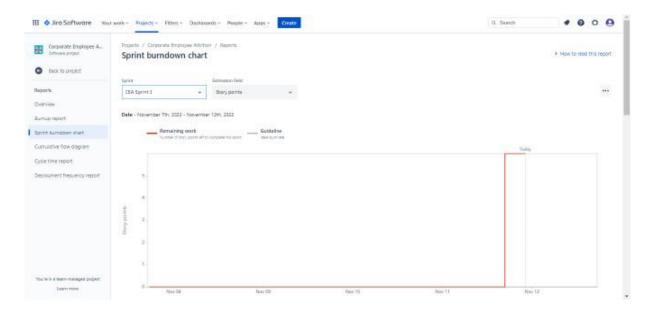


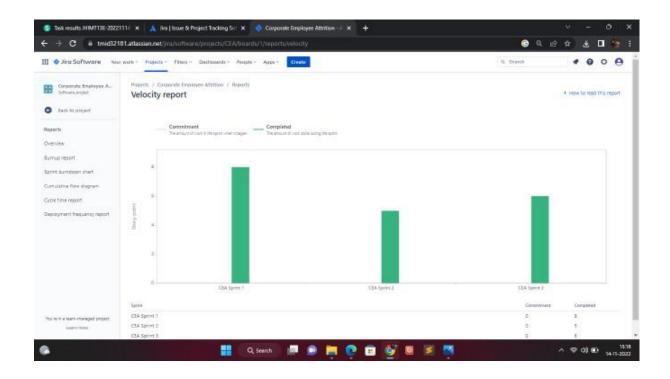
Sprint 2:



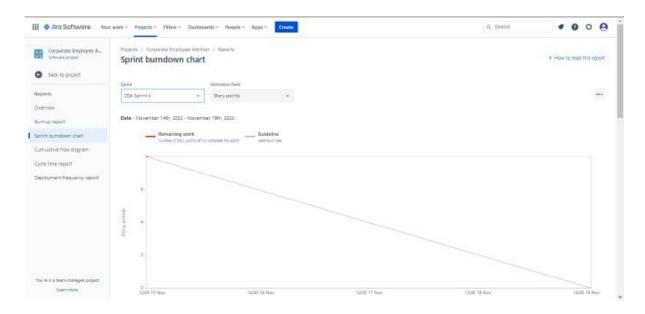


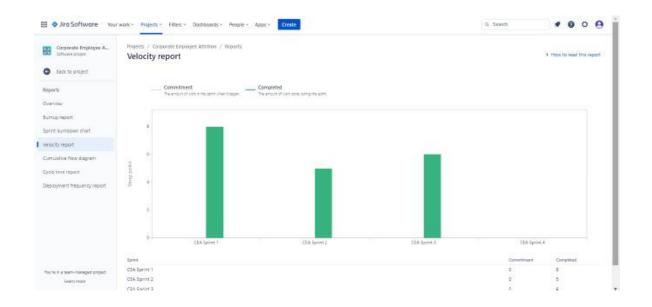
Sprint 3:



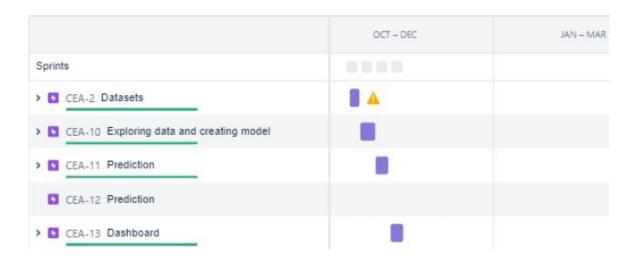


Sprint 4:





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'
,'q'], 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 test = sc.transform(x 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 =
'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

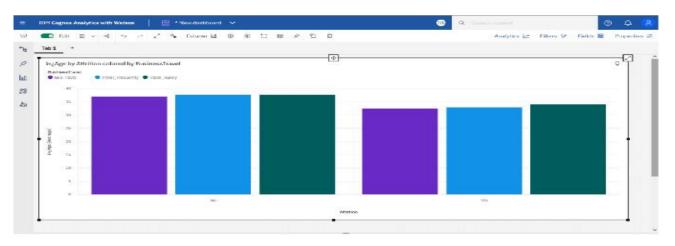
Section	Total Cases	Not Tested	Fail	Pass
Login Page	1	0	0	1
Employee Attrition Details	1	0	0	1

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

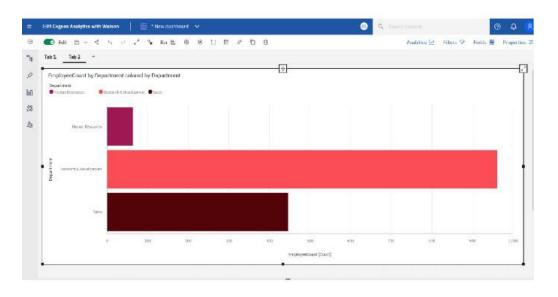
9.1 Performance Metrics

1.ATTRITION STATUS BY AGE:



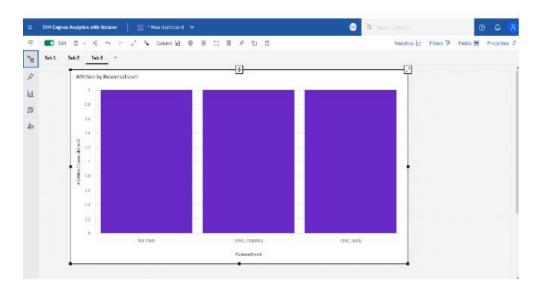
This visualization explains about the attrition status by age prediction through column chart.

2.EMPLOYEE COUNT BY DEPARTMENT:



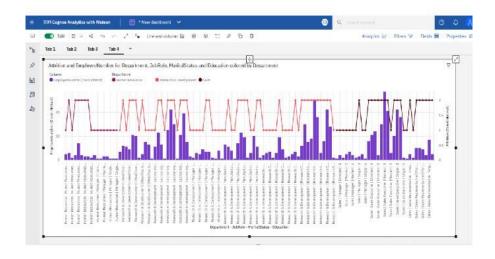
The visualized bar chart will clearly examine the employee count analyised by different departments.

3, ATTRITION BASED ON BUSINESS TRAVEL:



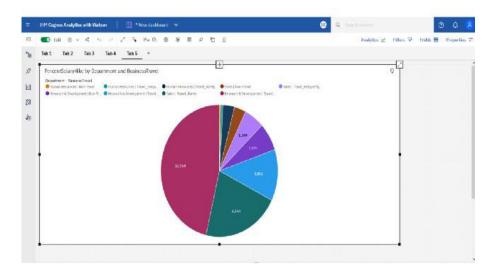
Visualization performed using the waterfall chart to view the attrition based on business travel.

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.

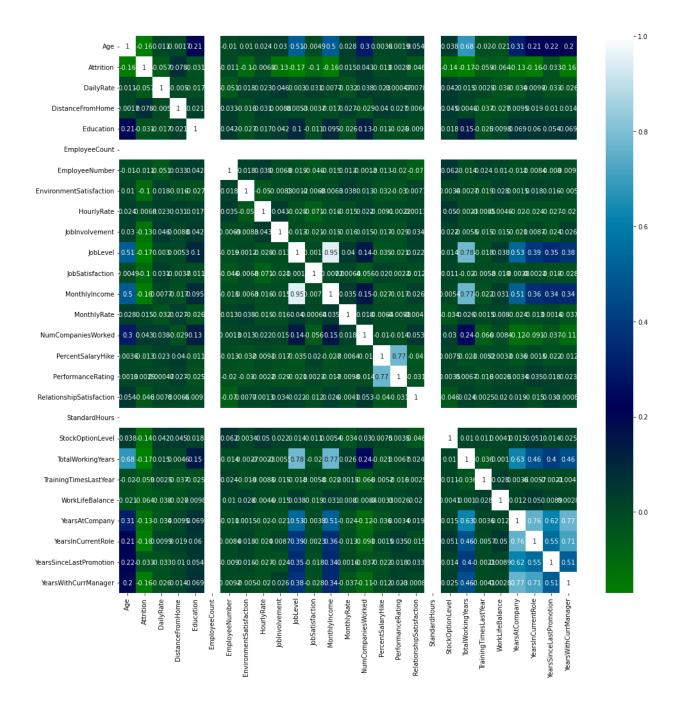
5.ATTRITION BASED ON SALARY HIKE PERCENTAGE:



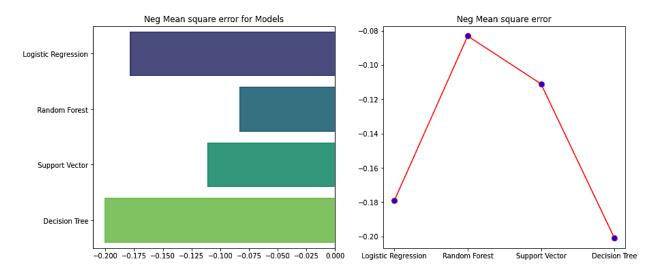
Using the employee data set, I predict the visualization using pie chart which may use to find the attrition between the salary to percentage hike.

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 their job. 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 Project Git Hub

https://github.com/IBM-EPBL/IBM-Project-10300-1659162161