## PROJECT REPORT

### 1. Introduction:

## 1.1 Overview

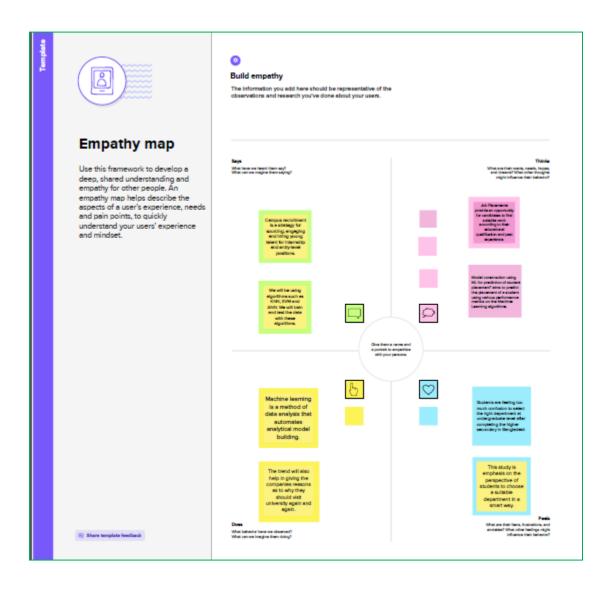
Campus recruitment is a strategy for sourcing, engaging and hiring young talent for internship and entry-level positions. College recruiting is typically a tactic for medium-to large-sized companies with high-volume recruiting needs, but can range from small efforts (like working with university career centers to source potential candidates) to large-scale operations (like visiting a wide array of colleges and attending recruiting events throughout the spring and fall semester). Campus recruitment often involves working with university career services centers and attending career fairs to meet inperson with college students and recent graduates. Our solution revolves around the placement season of a Business School in India. Where it has various factors on candidates getting hired such as work experience, exam percentage etc., Finally it contains the status of recruitment and remuneration details.

#### 1.2 Purpose

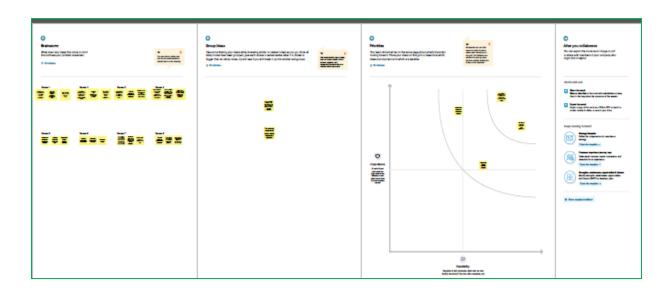
In this study, the target is to analyze student's placement data of last year and use it to determine the probability of campus placement of the present students. For this we have experimented with four different machine learning algorithms i.e. Logistic Regression, Decision Tree, K Nearest Neighbours and Random Forest.

# 2. Problem Definition & Design thinking

# 2.1 Empathy Map



# 2.2 Ideation & Brainstorming Map



## 3. Result

## 3.1 Date Model



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0	
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#### **Activity 1.1: Importing the libraries**

```
In [54]: import numpy as np
         import pandas as pd
         import os
         import seaborn as sns
         import matplotlib.pyplot as plt
         from sklearn import svm
         from sklearn.metrics import accuracy score
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn import metrics
         from sklearn.model selection import cross val score
         from sklearn import preprocessing
         from sklearn.model selection import train test split
         from sklearn.preprocessing import StandardScaler
         import joblib
         from sklearn.metrics import accuracy_score
         import pickle
```

#### **Activity 1.2: Read the Dataset**

Our dataset format might be in . csv, excel files, .txt, .json, etc. We can read the dataset with the help of pandas. In pandas we have a function called read\_csv() to read the dataset. As a parameter we have to give the directory of the csv file.

[55]:	<pre>df = pd.read_csv('collegeplace.csv')</pre>										
[56]:	df.head()										
[56]:		Age	Gender	Stream	Internships	CGPA	Hostel	HistoryOfBacklogs	PlacedOrNot		
	0	22	Male	Electronics And Communication	1	8	1	1	1		
	1	21	Female	Computer Science	0	7	1	1	1		
	2	22	Female	Information Technology	1	6	0	0	1		
	3	21	Male	Information Technology	0	8	0	1	1		
	4	22	Male	Mechanical	0	8	1	0	1		

As we have understood how the data is, let's pre-process the collected data. The download data set is not suitable for training the machine learning model as it might have so much randomness so we need to clean the dataset properly in order to fetch good results. This activity includes the following steps.

- Handling Missing data
- Handling Categorical data
- Handling missing data

#### Activity 2.1: Handling missing values:

Let's find the shape of our dataset first. To find the shape of our data, the df.shape method is used. To find the data type, df.info() function is used.

```
In [58]: df.isnull().sum()
Out[58]: Age
                               0
         Gender
                               0
         Stream
                               0
         Internships
                               0
         CGPA
                               0
                               0
         Hostel
         HistoryOfBacklogs
                               0
          PlacedOrNot
                               0
          dtype: int64
```

## 4. Trailhead Profile:

Team lead: Santha sharmila J

Member 1: Gomathi S

Member 2: Divya Bharathi S

Member 3: Sathya S

## 5. Advantages:

- 1. Students are offered prestigious job roles in a reputed organization before completing the degree.
- 2. Helps the recruiters to find the right fit for the organization without wasting time.
- 3. Formation of a cordial relationship between the company and the college.
- 4. The chances of selection in campus placements are high in comparison to off-campus and pool placements.

#### **Disadvantages:**

- 1. Candidates need to work hard to crack campus placement interviews and as freshers, they require a lot of training for work.
- Often, students at a college don't get their dream companies and they have to settle for the companies that recruit them at the time of campus placement at relatively lesser pay packages.

#### 6. Appendix:

```
from flask import Flask, render_template , request app=Flask(__name__) import pickle import joblib model=pickle.load(open("placement123.pkl",'rb')) ct=joblib.load('placement') @app.route('/') def hello(): return render_template("index.html") @app.route('/guest', methods = ["Post"]) def Guest():
```

```
sen1=request.form["sen1"]
sen2=request.form["sen2"]
sen3=request.form["sen3"]
sen4=request.form["sen4"]
sen5=request.form["sen5"]
sen6=request.form["sen6"]
@app.route('/y_predict', methods = ["POST"])
def y_predict():
    x_test = [[(yo) for yo in request.form.values()]]
    prediction = model.predict(x_test)
    prediction = prediction[0]
    return render_template("secondpage.html",y=prediction)
app.run(debug=True)
```

## 7. Conclusion:

Besides being all goody-goody, campus placements have their disadvantages too.

The remuneration offered to students is meagre, entry-level jobs aren't exactly as promised, joining can be delayed infinitely and a limited number of companies arrive for placement drive, so that restricts opportunities to a bare minimum for only a select few students.