

Data Collection and Preprocessing Phase

Date	15 March 2024
Team ID	740071
Project Title	Work Force Retention System
Maximum Marks	6 Marks

Data Exploration and Preprocessing Report

Dataset variables will be statistically analyzed to identify patterns and outliers, with Python employed for preprocessing tasks like normalization and feature engineering. Data cleaning will address missing values and outliers, ensuring quality for subsequent analysis and modeling, and forming a strong foundation for insights and predictions.

Section	Description
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Data Overview

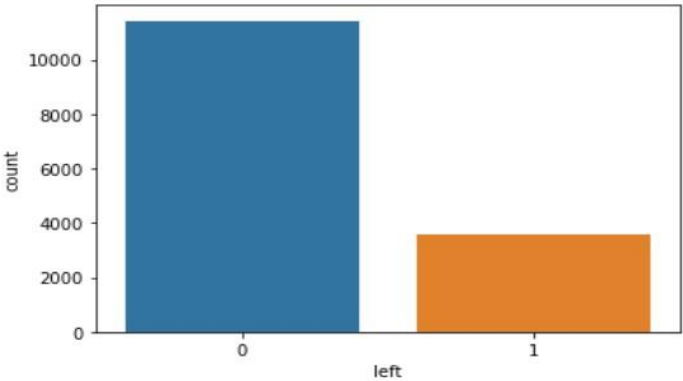
Dimension:
14999 rows × 10 columns
Descriptive statistics:

```
df.describe()
```

	satisfaction_level	last_evaluation	number_project	average_monthly_hours	time_spend_company	Work_accident	left	promotion_last_5years
count	14999.000000	14999.000000	14999.000000	14999.000000	14999.000000	14999.000000	14999.000000	14999.000000
mean	0.612834	0.716102	3.803054	201.050337	3.498233	0.144610	0.238083	0.021268
std	0.248631	0.171169	1.232592	49.943099	1.460136	0.351719	0.425924	0.144281
min	0.090000	0.360000	2.000000	96.000000	2.000000	0.000000	0.000000	0.000000
25%	0.440000	0.560000	3.000000	156.000000	3.000000	0.000000	0.000000	0.000000
50%	0.640000	0.720000	4.000000	200.000000	3.000000	0.000000	0.000000	0.000000
75%	0.820000	0.870000	5.000000	245.000000	4.000000	0.000000	0.000000	0.000000
max	1.000000	1.000000	7.000000	310.000000	10.000000	1.000000	1.000000	1.000000

Univariate Analysis

```
sns.countplot(df['left'])  
  
<AxesSubplot:xlabel='left', ylabel='count'>
```



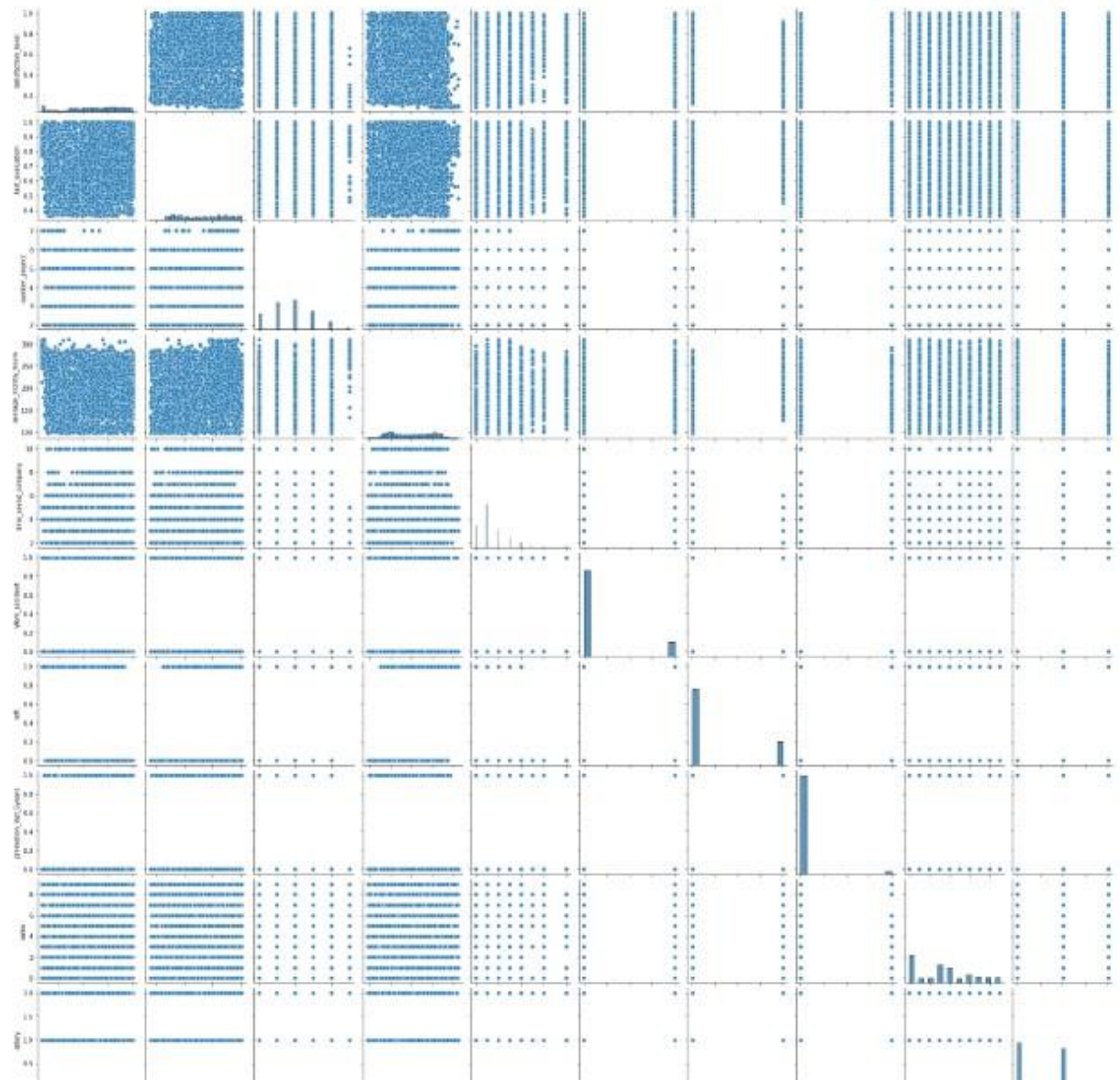
Bivariate Analysis

To find the relation between two features we use bivariate analysis. You can use seaborn package to plot visualisation using two variables of the dataset

Multivariate Analysis

```
sns.pairplot(df)
```

```
<seaborn.axisgrid.PairGrid at 0x2ab22b4e5b0>
```



Loading Data

Loading Data

```
: df = pd.read_csv('HR_comma_sep.csv')
```

```
: df
```

```
:
      satisfaction_level  last_evaluation  number_project  average_montly_hours  time_spend_company  Work_accident  left  promotion_last_year
0          0.38          0.53           2           157              3              0      1
1          0.80          0.86           5           262              6              0      1
2          0.11          0.88           7           272              4              0      1
3          0.72          0.87           5           223              5              0      1
4          0.37          0.52           2           159              3              0      1
...          ...          ...           ...           ...              ...              ...    ...
14994         0.40         0.57           2           151              3              0      1
14995         0.37         0.48           2           160              3              0      1
14996         0.37         0.53           2           143              3              0      1
14997         0.11         0.96           6           280              4              0      1
14998         0.37         0.52           2           158              3              0      1
```

14999 rows × 10 columns

Handling missing Values

```
df.shape
```

```
(14999, 10)
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 14999 entries, 0 to 14998
```

```
Data columns (total 10 columns):
```

#	Column	Non-Null Count	Dtype
0	satisfaction_level	14999 non-null	float64
1	last_evaluation	14999 non-null	float64
2	number_project	14999 non-null	int64
3	average_monthly_hours	14999 non-null	int64
4	time_spend_company	14999 non-null	int64
5	Work_accident	14999 non-null	int64
6	left	14999 non-null	int64
7	promotion_last_5years	14999 non-null	int64
8	sales	14999 non-null	object
9	salary	14999 non-null	object

```
dtypes: float64(2), int64(6), object(2)
```

```
memory usage: 1.1+ MB
```

Feature Engineering Save processed data	Attached the codes in final submission -