

```
import warnings
warnings.filterwarnings('ignore')
```

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
import plotly.express as px
from google.colab import drive
drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

```
pd.set_option('display.max_columns', None)
burnoutDf=pd.read_excel('/content/drive/MyDrive/dataset/employee_burnout_analysis-AI.xlsx')
burnoutDf
```

Type \	Employee ID	Date of Joining	Gender	Company
0	fffe32003000360033003200	2008-09-30	Female	Service
1	fffe3700360033003500	2008-11-30	Male	Service
2	fffe31003300320037003900	2008-03-10	Female	Product
3	fffe32003400380032003900	2008-11-03	Male	Service
4	fffe31003900340031003600	2008-07-24	Female	Service
...
22745	fffe31003500370039003100	2008-12-30	Female	Service
22746	fffe33003000350031003800	2008-01-19	Female	Product
22747	fffe390032003000	2008-11-05	Male	Service
22748	fffe33003300320036003900	2008-01-10	Female	Service
22749	fffe3400350031003800	2008-01-06	Male	Product

	WFH Setup Available	Designation	Resource Allocation \
0	No	2	3.0
1	Yes	1	2.0
2	Yes	2	NaN
3	Yes	1	1.0
4	No	3	7.0
...

22745	No	1	3.0
22746	Yes	3	6.0
22747	Yes	3	7.0
22748	No	2	5.0
22749	No	3	6.0

	Mental Fatigue Score	Burn Rate
0	3.8	0.16
1	5.0	0.36
2	5.8	0.49
3	2.6	0.20
4	6.9	0.52
...
22745	NaN	0.41
22746	6.7	0.59
22747	NaN	0.72
22748	5.9	0.52
22749	7.8	0.61

[22750 rows x 9 columns]

```
burnoutDf["Date of Joining"] = pd.to_datetime(burnoutDf["Date of Joining"])
```

```
burnoutDf.shape
```

(22750, 9)

```
burnoutDf.info()
```

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

```
RangeIndex: 22750 entries, 0 to 22749
```

```
Data columns (total 9 columns):
```

#	Column	Non-Null Count	Dtype
0	Employee ID	22750 non-null	object
1	Date of Joining	22750 non-null	datetime64[ns]
2	Gender	22750 non-null	object
3	Company Type	22750 non-null	object
4	WFH Setup Available	22750 non-null	object
5	Designation	22750 non-null	int64
6	Resource Allocation	21369 non-null	float64
7	Mental Fatigue Score	20633 non-null	float64
8	Burn Rate	21626 non-null	float64

```
dtypes: datetime64[ns](1), float64(3), int64(1), object(4)
```

```
memory usage: 1.6+ MB
```

```
burnoutDf.head()
```

	Employee ID	Date of Joining	Gender	Company Type	\
0	fffe32003000360033003200	2008-09-30	Female	Service	

1	fffe3700360033003500	2008-11-30	Male	Service
2	fffe31003300320037003900	2008-03-10	Female	Product
3	fffe32003400380032003900	2008-11-03	Male	Service
4	fffe31003900340031003600	2008-07-24	Female	Service

	WFH Setup Available	Designation	Resource Allocation	Mental Fatigue Score \
0	No	2	3.0	3.8
1	Yes	1	2.0	5.0
2	Yes	2	NaN	5.8
3	Yes	1	1.0	2.6
4	No	3	7.0	6.9

	Burn Rate
0	0.16
1	0.36
2	0.49
3	0.20
4	0.52

```
burnoutDf.columns
```

```
Index(['Employee ID', 'Date of Joining', 'Gender', 'Company Type',
      'WFH Setup Available', 'Designation', 'Resource Allocation',
      'Mental Fatigue Score', 'Burn Rate'],
      dtype='object')
```

```
burnoutDf.isna().sum()
```

```
Employee ID      0
Date of Joining  0
Gender           0
Company Type     0
WFH Setup Available  0
Designation      0
Resource Allocation 1381
Mental Fatigue Score 2117
Burn Rate        1124
dtype: int64
```

```
burnoutDf.duplicated().sum()
```

```
0
```

```
burnoutDf.describe()
```

	Designation	Resource Allocation	Mental Fatigue Score
Burn Rate			
count	22750.000000	21369.000000	20633.000000
21626.000000			
mean	2.178725	4.481398	5.728188
0.452005			
std	1.135145	2.047211	1.920839
0.198226			
min	0.000000	1.000000	0.000000
0.000000			
25%	1.000000	3.000000	4.600000
0.310000			
50%	2.000000	4.000000	5.900000
0.450000			
75%	3.000000	6.000000	7.100000
0.590000			
max	5.000000	10.000000	10.000000
1.000000			

```
for i, col in enumerate(burnoutDf.columns):
    print(f"\n\n{burnoutDf[col].unique()}")
    print(f"\n{burnoutDf[col].value_counts()}\n\n")
```

```
['fffe32003000360033003200' 'fffe3700360033003500'
 'fffe31003300320037003900' ... 'fffe390032003000'
 'fffe33003300320036003900' 'fffe3400350031003800']
```

```
fffe32003000360033003200    1
fffe3600360035003500        1
fffe3800360034003400        1
fffe31003000310033003600    1
fffe31003400350031003700    1
```

```
..
fffe33003400340032003400    1
fffe32003100370036003600    1
fffe31003900310035003800    1
fffe32003400320034003200    1
fffe3400350031003800        1
```

Name: Employee ID, Length: 22750, dtype: int64

```
['2008-09-30T00:00:00.000000000' '2008-11-30T00:00:00.000000000'
 '2008-03-10T00:00:00.000000000' '2008-11-03T00:00:00.000000000'
 '2008-07-24T00:00:00.000000000' '2008-11-26T00:00:00.000000000'
 '2008-01-02T00:00:00.000000000' '2008-10-31T00:00:00.000000000'
 '2008-12-27T00:00:00.000000000' '2008-03-09T00:00:00.000000000']
```

'2008-03-16T00:00:00.000000000'	'2008-05-12T00:00:00.000000000'
'2008-01-20T00:00:00.000000000'	'2008-02-23T00:00:00.000000000'
'2008-05-14T00:00:00.000000000'	'2008-02-03T00:00:00.000000000'
'2008-03-17T00:00:00.000000000'	'2008-03-28T00:00:00.000000000'
'2008-05-29T00:00:00.000000000'	'2008-06-27T00:00:00.000000000'
'2008-08-31T00:00:00.000000000'	'2008-01-15T00:00:00.000000000'
'2008-05-04T00:00:00.000000000'	'2008-11-17T00:00:00.000000000'
'2008-09-14T00:00:00.000000000'	'2008-10-09T00:00:00.000000000'
'2008-10-11T00:00:00.000000000'	'2008-09-18T00:00:00.000000000'
'2008-09-16T00:00:00.000000000'	'2008-12-16T00:00:00.000000000'
'2008-05-03T00:00:00.000000000'	'2008-08-04T00:00:00.000000000'
'2008-07-31T00:00:00.000000000'	'2008-06-17T00:00:00.000000000'
'2008-04-28T00:00:00.000000000'	'2008-10-30T00:00:00.000000000'
'2008-02-27T00:00:00.000000000'	'2008-06-22T00:00:00.000000000'
'2008-02-18T00:00:00.000000000'	'2008-06-24T00:00:00.000000000'
'2008-12-08T00:00:00.000000000'	'2008-08-05T00:00:00.000000000'
'2008-04-11T00:00:00.000000000'	'2008-03-26T00:00:00.000000000'
'2008-08-09T00:00:00.000000000'	'2008-08-28T00:00:00.000000000'
'2008-03-21T00:00:00.000000000'	'2008-07-22T00:00:00.000000000'
'2008-05-20T00:00:00.000000000'	'2008-01-23T00:00:00.000000000'
'2008-09-10T00:00:00.000000000'	'2008-05-26T00:00:00.000000000'
'2008-12-22T00:00:00.000000000'	'2008-04-08T00:00:00.000000000'
'2008-02-25T00:00:00.000000000'	'2008-04-24T00:00:00.000000000'
'2008-01-08T00:00:00.000000000'	'2008-11-20T00:00:00.000000000'
'2008-09-11T00:00:00.000000000'	'2008-06-11T00:00:00.000000000'
'2008-02-28T00:00:00.000000000'	'2008-08-20T00:00:00.000000000'
'2008-10-18T00:00:00.000000000'	'2008-08-14T00:00:00.000000000'
'2008-07-17T00:00:00.000000000'	'2008-07-05T00:00:00.000000000'
'2008-02-04T00:00:00.000000000'	'2008-08-01T00:00:00.000000000'
'2008-05-01T00:00:00.000000000'	'2008-05-21T00:00:00.000000000'
'2008-10-21T00:00:00.000000000'	'2008-03-19T00:00:00.000000000'
'2008-09-27T00:00:00.000000000'	'2008-03-12T00:00:00.000000000'
'2008-09-17T00:00:00.000000000'	'2008-02-13T00:00:00.000000000'
'2008-09-19T00:00:00.000000000'	'2008-07-03T00:00:00.000000000'
'2008-10-27T00:00:00.000000000'	'2008-01-22T00:00:00.000000000'
'2008-04-15T00:00:00.000000000'	'2008-10-26T00:00:00.000000000'
'2008-01-31T00:00:00.000000000'	'2008-01-03T00:00:00.000000000'
'2008-03-13T00:00:00.000000000'	'2008-03-27T00:00:00.000000000'
'2008-11-15T00:00:00.000000000'	'2008-08-17T00:00:00.000000000'
'2008-08-08T00:00:00.000000000'	'2008-06-28T00:00:00.000000000'
'2008-05-06T00:00:00.000000000'	'2008-12-17T00:00:00.000000000'
'2008-09-08T00:00:00.000000000'	'2008-07-04T00:00:00.000000000'
'2008-10-28T00:00:00.000000000'	'2008-02-19T00:00:00.000000000'
'2008-02-11T00:00:00.000000000'	'2008-03-02T00:00:00.000000000'
'2008-08-10T00:00:00.000000000'	'2008-01-04T00:00:00.000000000'
'2008-10-12T00:00:00.000000000'	'2008-11-14T00:00:00.000000000'
'2008-09-02T00:00:00.000000000'	'2008-10-04T00:00:00.000000000'
'2008-05-31T00:00:00.000000000'	'2008-03-03T00:00:00.000000000'
'2008-02-21T00:00:00.000000000'	'2008-12-04T00:00:00.000000000'

'2008-09-05T00:00:00.000000000'	'2008-02-24T00:00:00.000000000'
'2008-12-09T00:00:00.000000000'	'2008-01-19T00:00:00.000000000'
'2008-01-26T00:00:00.000000000'	'2008-05-10T00:00:00.000000000'
'2008-05-16T00:00:00.000000000'	'2008-05-07T00:00:00.000000000'
'2008-10-16T00:00:00.000000000'	'2008-07-09T00:00:00.000000000'
'2008-03-11T00:00:00.000000000'	'2008-08-15T00:00:00.000000000'
'2008-08-25T00:00:00.000000000'	'2008-12-14T00:00:00.000000000'
'2008-04-26T00:00:00.000000000'	'2008-04-03T00:00:00.000000000'
'2008-12-19T00:00:00.000000000'	'2008-08-13T00:00:00.000000000'
'2008-03-08T00:00:00.000000000'	'2008-02-05T00:00:00.000000000'
'2008-02-17T00:00:00.000000000'	'2008-04-16T00:00:00.000000000'
'2008-10-24T00:00:00.000000000'	'2008-03-05T00:00:00.000000000'
'2008-09-25T00:00:00.000000000'	'2008-03-01T00:00:00.000000000'
'2008-05-23T00:00:00.000000000'	'2008-09-07T00:00:00.000000000'
'2008-03-23T00:00:00.000000000'	'2008-01-25T00:00:00.000000000'
'2008-12-29T00:00:00.000000000'	'2008-06-15T00:00:00.000000000'
'2008-10-03T00:00:00.000000000'	'2008-01-17T00:00:00.000000000'
'2008-01-30T00:00:00.000000000'	'2008-10-13T00:00:00.000000000'
'2008-02-08T00:00:00.000000000'	'2008-11-25T00:00:00.000000000'
'2008-04-23T00:00:00.000000000'	'2008-11-07T00:00:00.000000000'
'2008-06-20T00:00:00.000000000'	'2008-12-23T00:00:00.000000000'
'2008-11-24T00:00:00.000000000'	'2008-06-21T00:00:00.000000000'
'2008-11-29T00:00:00.000000000'	'2008-08-11T00:00:00.000000000'
'2008-04-29T00:00:00.000000000'	'2008-11-19T00:00:00.000000000'
'2008-12-25T00:00:00.000000000'	'2008-02-14T00:00:00.000000000'
'2008-03-04T00:00:00.000000000'	'2008-10-06T00:00:00.000000000'
'2008-08-16T00:00:00.000000000'	'2008-10-29T00:00:00.000000000'
'2008-07-15T00:00:00.000000000'	'2008-04-21T00:00:00.000000000'
'2008-09-01T00:00:00.000000000'	'2008-01-06T00:00:00.000000000'
'2008-03-20T00:00:00.000000000'	'2008-04-14T00:00:00.000000000'
'2008-02-16T00:00:00.000000000'	'2008-10-10T00:00:00.000000000'
'2008-09-26T00:00:00.000000000'	'2008-06-01T00:00:00.000000000'
'2008-07-11T00:00:00.000000000'	'2008-07-23T00:00:00.000000000'
'2008-07-10T00:00:00.000000000'	'2008-10-05T00:00:00.000000000'
'2008-03-14T00:00:00.000000000'	'2008-06-14T00:00:00.000000000'
'2008-10-23T00:00:00.000000000'	'2008-02-22T00:00:00.000000000'
'2008-05-19T00:00:00.000000000'	'2008-09-20T00:00:00.000000000'
'2008-01-18T00:00:00.000000000'	'2008-07-13T00:00:00.000000000'
'2008-11-04T00:00:00.000000000'	'2008-12-05T00:00:00.000000000'
'2008-07-27T00:00:00.000000000'	'2008-12-07T00:00:00.000000000'
'2008-06-04T00:00:00.000000000'	'2008-09-09T00:00:00.000000000'
'2008-11-01T00:00:00.000000000'	'2008-01-28T00:00:00.000000000'
'2008-04-04T00:00:00.000000000'	'2008-07-06T00:00:00.000000000'
'2008-12-28T00:00:00.000000000'	'2008-07-08T00:00:00.000000000'
'2008-01-21T00:00:00.000000000'	'2008-10-19T00:00:00.000000000'
'2008-01-07T00:00:00.000000000'	'2008-12-24T00:00:00.000000000'
'2008-06-09T00:00:00.000000000'	'2008-09-13T00:00:00.000000000'
'2008-10-14T00:00:00.000000000'	'2008-11-08T00:00:00.000000000'
'2008-12-26T00:00:00.000000000'	'2008-05-08T00:00:00.000000000'

'2008-08-12T00:00:00.000000000'	'2008-08-24T00:00:00.000000000'
'2008-09-21T00:00:00.000000000'	'2008-11-10T00:00:00.000000000'
'2008-01-09T00:00:00.000000000'	'2008-05-18T00:00:00.000000000'
'2008-10-08T00:00:00.000000000'	'2008-09-22T00:00:00.000000000'
'2008-08-06T00:00:00.000000000'	'2008-04-30T00:00:00.000000000'
'2008-12-20T00:00:00.000000000'	'2008-04-13T00:00:00.000000000'
'2008-04-12T00:00:00.000000000'	'2008-11-18T00:00:00.000000000'
'2008-02-15T00:00:00.000000000'	'2008-06-07T00:00:00.000000000'
'2008-11-16T00:00:00.000000000'	'2008-06-26T00:00:00.000000000'
'2008-05-11T00:00:00.000000000'	'2008-09-03T00:00:00.000000000'
'2008-03-06T00:00:00.000000000'	'2008-09-24T00:00:00.000000000'
'2008-04-01T00:00:00.000000000'	'2008-05-25T00:00:00.000000000'
'2008-05-22T00:00:00.000000000'	'2008-01-13T00:00:00.000000000'
'2008-06-06T00:00:00.000000000'	'2008-01-16T00:00:00.000000000'
'2008-03-22T00:00:00.000000000'	'2008-04-20T00:00:00.000000000'
'2008-02-02T00:00:00.000000000'	'2008-10-01T00:00:00.000000000'
'2008-10-07T00:00:00.000000000'	'2008-06-03T00:00:00.000000000'
'2008-11-12T00:00:00.000000000'	'2008-08-26T00:00:00.000000000'
'2008-05-17T00:00:00.000000000'	'2008-12-30T00:00:00.000000000'
'2008-06-19T00:00:00.000000000'	'2008-11-22T00:00:00.000000000'
'2008-05-13T00:00:00.000000000'	'2008-03-30T00:00:00.000000000'
'2008-06-16T00:00:00.000000000'	'2008-04-27T00:00:00.000000000'
'2008-07-01T00:00:00.000000000'	'2008-12-15T00:00:00.000000000'
'2008-09-06T00:00:00.000000000'	'2008-04-19T00:00:00.000000000'
'2008-01-12T00:00:00.000000000'	'2008-12-02T00:00:00.000000000'
'2008-01-24T00:00:00.000000000'	'2008-07-02T00:00:00.000000000'
'2008-08-29T00:00:00.000000000'	'2008-07-29T00:00:00.000000000'
'2008-06-29T00:00:00.000000000'	'2008-01-11T00:00:00.000000000'
'2008-11-09T00:00:00.000000000'	'2008-07-30T00:00:00.000000000'
'2008-08-23T00:00:00.000000000'	'2008-06-05T00:00:00.000000000'
'2008-09-23T00:00:00.000000000'	'2008-06-18T00:00:00.000000000'
'2008-01-14T00:00:00.000000000'	'2008-12-06T00:00:00.000000000'
'2008-01-10T00:00:00.000000000'	'2008-06-13T00:00:00.000000000'
'2008-07-18T00:00:00.000000000'	'2008-07-28T00:00:00.000000000'
'2008-07-26T00:00:00.000000000'	'2008-01-01T00:00:00.000000000'
'2008-08-27T00:00:00.000000000'	'2008-08-30T00:00:00.000000000'
'2008-04-10T00:00:00.000000000'	'2008-07-14T00:00:00.000000000'
'2008-09-28T00:00:00.000000000'	'2008-04-02T00:00:00.000000000'
'2008-10-15T00:00:00.000000000'	'2008-06-30T00:00:00.000000000'
'2008-03-07T00:00:00.000000000'	'2008-10-22T00:00:00.000000000'
'2008-08-02T00:00:00.000000000'	'2008-03-15T00:00:00.000000000'
'2008-03-18T00:00:00.000000000'	'2008-05-28T00:00:00.000000000'
'2008-02-09T00:00:00.000000000'	'2008-08-22T00:00:00.000000000'
'2008-11-02T00:00:00.000000000'	'2008-04-22T00:00:00.000000000'
'2008-11-21T00:00:00.000000000'	'2008-02-12T00:00:00.000000000'
'2008-02-07T00:00:00.000000000'	'2008-07-19T00:00:00.000000000'
'2008-11-23T00:00:00.000000000'	'2008-07-21T00:00:00.000000000'
'2008-08-21T00:00:00.000000000'	'2008-11-11T00:00:00.000000000'
'2008-12-13T00:00:00.000000000'	'2008-04-25T00:00:00.000000000'

```
'2008-11-05T00:00:00.000000000' '2008-08-19T00:00:00.000000000'
'2008-04-17T00:00:00.000000000' '2008-08-07T00:00:00.000000000'
'2008-12-31T00:00:00.000000000' '2008-05-27T00:00:00.000000000'
'2008-09-29T00:00:00.000000000' '2008-05-30T00:00:00.000000000'
'2008-12-18T00:00:00.000000000' '2008-02-20T00:00:00.000000000'
'2008-12-11T00:00:00.000000000' '2008-11-27T00:00:00.000000000'
'2008-07-20T00:00:00.000000000' '2008-11-28T00:00:00.000000000'
'2008-08-03T00:00:00.000000000' '2008-10-20T00:00:00.000000000'
'2008-07-07T00:00:00.000000000' '2008-06-08T00:00:00.000000000'
'2008-03-24T00:00:00.000000000' '2008-12-21T00:00:00.000000000'
'2008-04-09T00:00:00.000000000' '2008-05-05T00:00:00.000000000'
'2008-06-12T00:00:00.000000000' '2008-04-18T00:00:00.000000000'
'2008-01-27T00:00:00.000000000' '2008-10-17T00:00:00.000000000'
'2008-05-09T00:00:00.000000000' '2008-03-29T00:00:00.000000000'
'2008-09-12T00:00:00.000000000' '2008-07-25T00:00:00.000000000'
'2008-04-07T00:00:00.000000000' '2008-05-02T00:00:00.000000000'
'2008-06-02T00:00:00.000000000' '2008-10-02T00:00:00.000000000'
'2008-02-26T00:00:00.000000000' '2008-07-12T00:00:00.000000000'
'2008-02-06T00:00:00.000000000' '2008-06-23T00:00:00.000000000'
'2008-11-06T00:00:00.000000000' '2008-07-16T00:00:00.000000000'
'2008-06-25T00:00:00.000000000' '2008-01-29T00:00:00.000000000'
'2008-02-29T00:00:00.000000000' '2008-03-25T00:00:00.000000000'
'2008-08-18T00:00:00.000000000' '2008-04-05T00:00:00.000000000'
'2008-05-15T00:00:00.000000000' '2008-12-12T00:00:00.000000000'
'2008-10-25T00:00:00.000000000' '2008-04-06T00:00:00.000000000'
'2008-11-13T00:00:00.000000000' '2008-09-04T00:00:00.000000000'
'2008-05-24T00:00:00.000000000' '2008-06-10T00:00:00.000000000'
'2008-03-31T00:00:00.000000000' '2008-12-01T00:00:00.000000000'
'2008-01-05T00:00:00.000000000' '2008-09-15T00:00:00.000000000'
'2008-12-10T00:00:00.000000000' '2008-02-10T00:00:00.000000000'
'2008-12-03T00:00:00.000000000' '2008-02-01T00:00:00.000000000']
```

```
2008-01-06      86
2008-05-21      85
2008-02-04      82
2008-07-16      81
2008-07-13      80
```

```
..
2008-06-27      44
2008-07-06      44
2008-07-04      43
2008-12-24      43
2008-12-07      39
```

Name: Date of Joining, Length: 366, dtype: int64

```
['Female' 'Male']
```


Female 11908
Male 10842
Name: Gender, dtype: int64

['Service' 'Product']

Service 14833
Product 7917
Name: Company Type, dtype: int64

['No' 'Yes']

Yes 12290
No 10460
Name: WFH Setup Available, dtype: int64

[2 1 3 0 4 5]

2 7588
3 5985
1 4881
4 2391
0 1507
5 398
Name: Designation, dtype: int64

[3. 2. nan 1. 7. 4. 6. 5. 8. 10. 9.]

4.0 3893
5.0 3861
3.0 3192
6.0 2943
2.0 2075
7.0 1965
1.0 1791
8.0 1044
9.0 446
10.0 159
Name: Resource Allocation, dtype: int64

```
[ 3.8  5.   5.8  2.6  6.9  3.6  7.9  4.4  nan  5.3  1.8  4.7  5.9  6.7
  4.   7.6  6.3  7.7  6.6  7.4  3.9  3.   8.7  7.3  5.4  6.   7.5 10.
  6.4  5.1  5.6  6.1  3.1  8.   6.8  4.9  9.2  6.5  6.2  8.2  4.1  4.3
  0.8  2.9  2.   9.1  0.   5.7  8.3  5.5  7.   3.3  7.8  7.2  5.2  8.9
  4.5  8.1  8.6  9.5  3.5  4.8  2.4  3.7  1.   8.8  9.3  4.6  9.9  0.5
  2.8  9.   3.4  4.2  1.6  2.7  1.3  3.2  8.4  7.1  9.4  2.1  9.7  2.5
  1.9  1.7  9.6  0.7  0.2  1.2  8.5  9.8  2.2  1.1  0.9  2.3  0.4  1.4
  1.5  0.6  0.3  0.1]
```

```
6.0    470
```

```
5.8    464
```

```
5.9    458
```

```
6.1    457
```

```
6.3    454
```

```
...
```

```
0.5     24
```

```
0.2     23
```

```
0.4     19
```

```
0.1     17
```

```
0.3     13
```

```
Name: Mental Fatigue Score, Length: 101, dtype: int64
```

```
[0.16 0.36 0.49 0.2  0.52 0.29 0.62 0.33 0.56 0.67 0.5  0.12 0.4  0.51
 0.32 0.39 0.59 0.22 0.68 0.57 0.47 0.46 0.61 0.91 0.44 0.6  0.45 0.19
 0.31 0.81 0.42 0.53  nan 0.94 0.37 0.65 0.38 0.15 0.26 0.28 0.71 0.8
 0.63 0.79 0.72 0.34 0.27 0.66 0.04 0.05 0.11 0.41 0.76 0.43 0.85 0.35
 0.   0.55 0.48 0.7  0.18 0.23 0.25 0.75 0.1  0.73 0.58 0.88 0.77 0.3
 0.06 0.03 0.69 0.24 0.74 0.86 0.92 0.78 0.21 0.98 0.02 0.82 0.93 0.83
 0.87 0.64 0.54 0.17 1.   0.08 0.09 0.14 0.13 0.07 0.84 0.99 0.01 0.97
 0.95 0.9  0.96 0.89]
```

```
0.47    475
```

```
0.43    444
```

```
0.41    434
```

```
0.45    431
```

```
0.50    428
```

```
...
```

```
0.98     18
```

```
0.97     17
```

```
0.95     17
```

```
0.96     13
```

```
0.99      8
```

```
Name: Burn Rate, Length: 101, dtype: int64
```

```

burnoutDf= burnoutDf.drop(['Employee ID'],axis=1)

#check skewness
intFloatburnoutDf=burnoutDf.select_dtypes([np.int,np.float])
for i, col in enumerate(intFloatburnoutDf.columns):
    if(intFloatburnoutDf[col].skew() >= 0.1):
        print("\n",col, "feature is positively skewed and value is: ",intFloatburnoutDf[col].skew())
    elif (intFloatburnoutDf[col].skew() <= -0.1):
        print("\n",col, "feature is Negtively skewed and value is:",intFloatburnoutDf[col].skew())
    else:
        print("\n",col, "feature is Normally Distributed and value is:",intFloatburnoutDf[col].skew())

```

Designation feature is Normally Distributed and value is:
0.09242138478903683

Resource Allocation feature is positively skewed and value is:
0.20457273454318103

Mental Fatigue Score feature is Negtively skewed and value is: -
0.4308950578815428

Burn Rate feature is Normally Distributed and value is:
0.045737370909640515

```

burnoutDf['Resource Allocation'].fillna(burnoutDf['Resource Allocation'].mean(),inplace=True)
burnoutDf['Mental Fatigue Score'].fillna(burnoutDf['Mental Fatigue Score'].mean(),inplace=True)
burnoutDf['Burn Rate'].fillna(burnoutDf['Burn Rate'].mean(),inplace=True)

```

```
burnoutDf.isna().sum()
```

Date of Joining	0
Gender	0
Company Type	0
WFH Setup Available	0
Designation	0
Resource Allocation	0
Mental Fatigue Score	0
Burn Rate	0
dtype: int64	

```
burnoutDf.corr()
```

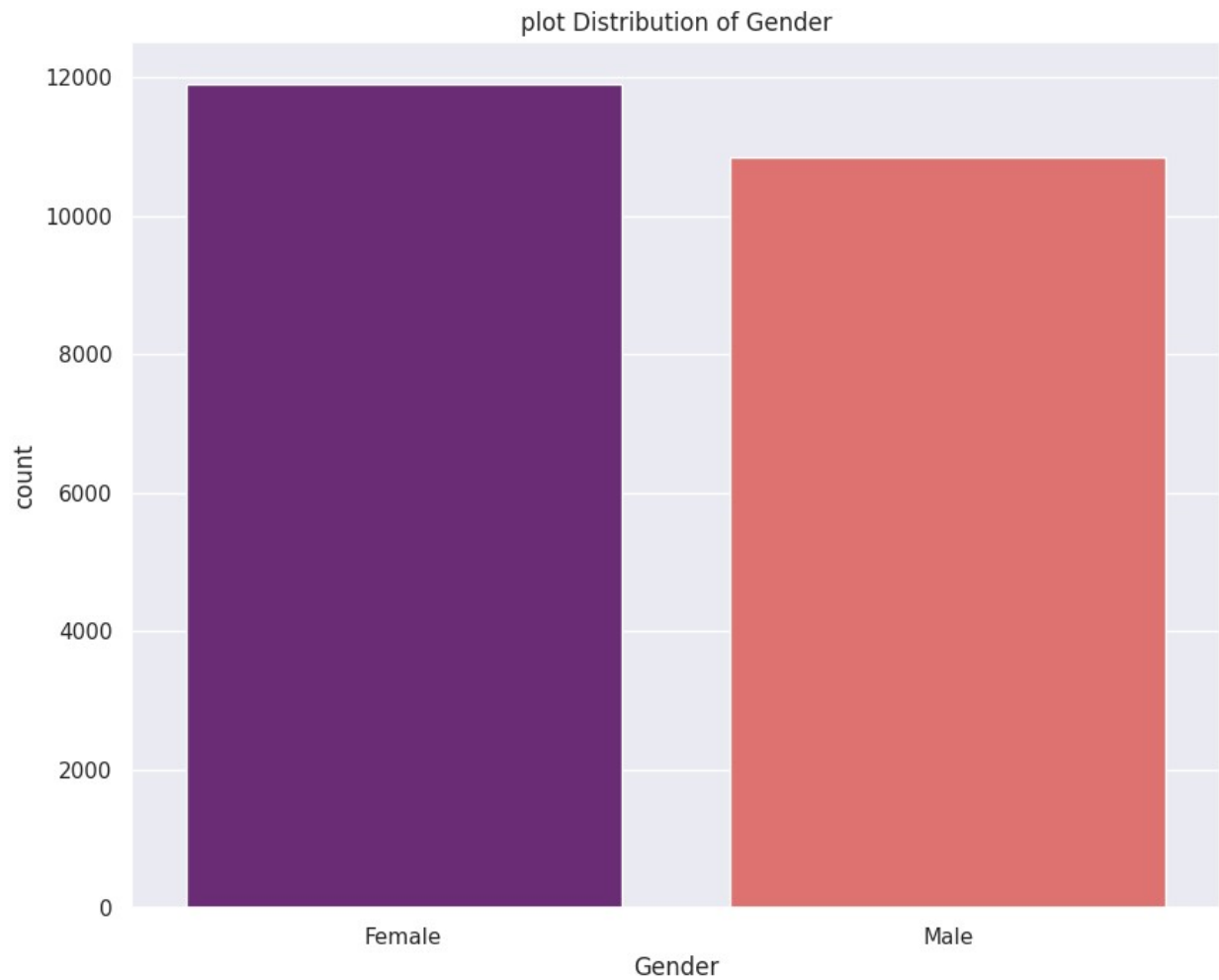
	Designation	Resource Allocation	Mental Fatigue
Score \			
Designation	1.000000	0.852046	0.656445
Resource Allocation	0.852046	1.000000	0.739268
Mental Fatigue Score	0.656445	0.739268	1.000000
Burn Rate	0.719284	0.811062	0.878217

	Burn Rate
Designation	0.719284
Resource Allocation	0.811062
Mental Fatigue Score	0.878217
Burn Rate	1.000000

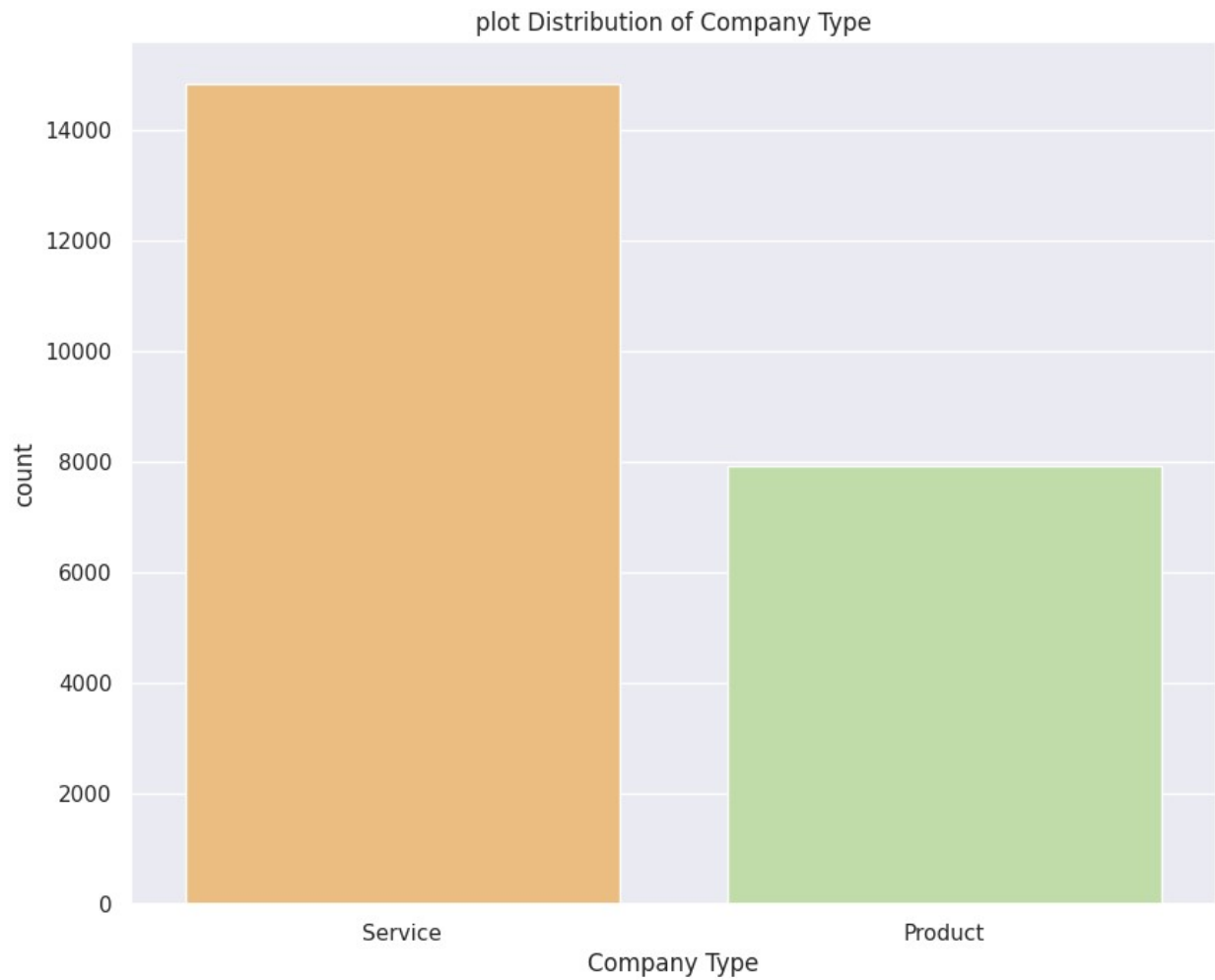
```
corr=burnoutDf.corr()  
sns.set(rc={'figure.figsize':(14,12)})  
fig= px.imshow(corr,text_auto = True,aspect="auto")  
fig.show()
```

```
#distribution of gender
```

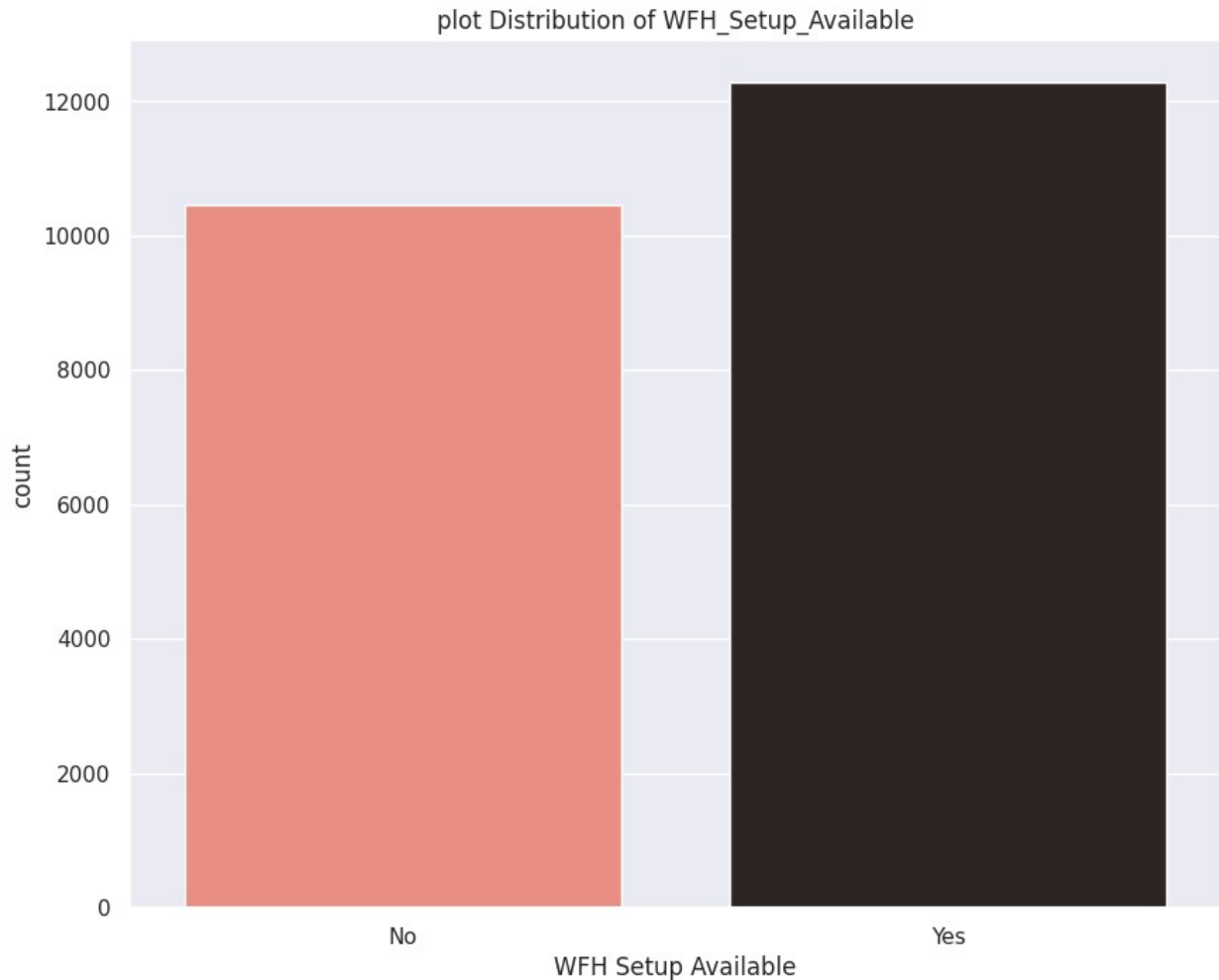
```
plt.figure(figsize=(10,8))  
sns.countplot(x="Gender",data=burnoutDf, palette="magma")  
plt.title("plot Distribution of Gender")  
plt.show()
```



```
#distribution of company types  
plt.figure(figsize=(10,8))  
sns.countplot(x="Company Type",data=burnoutDf, palette="Spectral")  
plt.title("plot Distribution of Company Type")  
plt.show()
```



```
#count plot distribution of WHF set up
plt.figure(figsize=(10,8))
sns.countplot(x="WFH Setup
Available",data=burnoutDf,palette="dark:salmon_r")
plt.title("plot Distribution of WFH_Setup_Available")
plt.show()
```



```
burn_st=burnoutDf.loc[:, 'Date of Joining': 'Burn Rate']
burn_st=burn_st.select_dtypes([int, float])
for i, col in enumerate(burn_st.columns):
    fig = px.histogram(burn_st, x=col, title="plot Distribution of "+col, color_discrete_sequence=['indianred'])
    fig.update_layout(bargap=0.2)
    fig.show()

fig=px.line(burnoutDf, y="Burn Rate", color="Designation", title="Burn rate on the basis of Designation", color_discrete_sequence=px.colors.qualitative.Pastell1)
fig.update_layout(bargap=0.1)
fig.show()

fig=px.line(burnoutDf, y="Burn Rate", color="Gender", title="Burn rate on the basis of Gender", color_discrete_sequence=px.colors.qualitative.Pastell1)
fig.update_layout(bargap=0.2)
fig.show()
```

```
fig=px.line(burnoutDf, y="Mental Fatigue Score",color="Designation",
title="Mental Fatigue vs
Designation",color_discrete_sequence=px.colors.qualitative.Pastell1)
fig.update_layout(bargap=0.2)
fig.show()
```

```
sns.relplot(
    data=burnoutDf, x="Designation", y="Mental Fatigue
Score",col="Company Type",
    hue="Company Type", size="Burn Rate", style="Gender",
    palette=["g", "r"], sizes=(50,200)
)
```

```
<seaborn.axisgrid.FacetGrid at 0x7ba0c96accd0>
```



```
from sklearn import preprocessing
Label_encode = preprocessing.LabelEncoder()

burnoutDf['GenderLabel']=
Label_encode.fit_transform(burnoutDf['Gender'].values)
burnoutDf['Company_TypeLabel']=Label_encode.fit_transform(burnoutDf['Company Type'].values)
burnoutDf['WFH_Setup_AvailableLabel']=Label_encode.fit_transform(burnoutDf['WFH Setup Available'].values)
```

```
gn=burnoutDf.groupby('Gender')
gn= gn['GenderLabel']
gn.first()
```

```
Gender
Female    0
Male      1
Name: GenderLabel, dtype: int64
```



```
ct=burnoutDf.groupby('Company Type')
ct = ct['Company_TypeLabel']
ct.first()
```

```
Company Type
Product      0
Service      1
Name: Company_TypeLabel, dtype: int64
```

```
wsa = burnoutDf.groupby('WFH Setup Available')
wsa = wsa['WFH_Setup_AvailableLabel']
wsa.first()
```

```
WFH Setup Available
No      0
Yes     1
Name: WFH_Setup_AvailableLabel, dtype: int64
```

```
burnoutDf.tail(10)
```

	Date of Joining	Gender	Company Type	WFH Setup Available
Designation \				
22740	2008-09-05	Female	Product	No
3				
22741	2008-01-07	Male	Product	No
2				
22742	2008-07-28	Male	Product	No
3				
22743	2008-12-15	Female	Product	Yes
1				
22744	2008-05-27	Male	Product	No
3				
22745	2008-12-30	Female	Service	No
1				
22746	2008-01-19	Female	Product	Yes
3				
22747	2008-11-05	Male	Service	Yes
3				
22748	2008-01-10	Female	Service	No
2				
22749	2008-01-06	Male	Product	No
3				

	Resource Allocation	Mental Fatigue Score	Burn Rate
GenderLabel \			
22740	6.0	7.300000	0.550000
0			
22741	5.0	6.000000	0.452005

```

1
22742          5.0          8.100000    0.690000
1
22743          3.0          6.000000    0.480000
0
22744          7.0          6.200000    0.540000
1
22745          3.0          5.728188    0.410000
0
22746          6.0          6.700000    0.590000
0
22747          7.0          5.728188    0.720000
1
22748          5.0          5.900000    0.520000
0
22749          6.0          7.800000    0.610000
1

```

```

          Company_TypeLabel  WFH_Setup_AvailableLabel
22740                0                0
22741                0                0
22742                0                0
22743                0                1
22744                0                0
22745                1                0
22746                0                1
22747                1                1
22748                1                0
22749                0                0

```

```

Columns=['Designation', 'Resource Allocation', 'Mental Fatigue Score',
'GenderLabel', 'Company_TypeLabel', 'WFH_Setup_AvailableLabel']
x=burnoutDf[Columns]
y=burnoutDf['Burn Rate']

```

```
print(x)
```

```

          Designation  Resource Allocation  Mental Fatigue Score
GenderLabel \
0                2          3.000000    3.800000
0
1                1          2.000000    5.000000
1
2                2          4.481398    5.800000
0
3                1          1.000000    2.600000
1
4                3          7.000000    6.900000
0

```

```

...
...
22745      1      3.000000      5.728188
0
22746      3      6.000000      6.700000
0
22747      3      7.000000      5.728188
1
22748      2      5.000000      5.900000
0
22749      3      6.000000      7.800000
1

```

```

      Company_TypeLabel  WFH_Setup_AvailableLabel
0                      1                      0
1                      1                      1
2                      0                      1
3                      1                      1
4                      1                      0
...
22745      1      0
22746      0      1
22747      1      1
22748      1      0
22749      0      0

```

```
[22750 rows x 6 columns]
```

```
print(y)
```

```

0      0.16
1      0.36
2      0.49
3      0.20
4      0.52

```

```

...
22745      0.41
22746      0.59
22747      0.72
22748      0.52
22749      0.61

```

```
Name: Burn Rate, Length: 22750, dtype: float64
```

```

from re import X
from sklearn.decomposition import PCA

```

```

pca=PCA(0.95)
x_pca=pca.fit_transform(x)
print("PCA shape of x is: ",x_pca.shape, "and original shape is:
",x.shape)

```

```

print("% of importance of selected features is:",
pca.explained_variance_ratio_)
print("The number of features selected through PCA is:",
pca.n_components_)

PCA shape of x is: (22750, 4) and original shape is: (22750, 6)
% of importance of selected features is: [0.78371089 0.11113597
0.03044541 0.02632422]
The number of features selected through PCA is: 4

#data splitting
from sklearn.model_selection import train_test_split
X_train_pca,X_test,Y_train,Y_test =
train_test_split(x_pca,y,test_size=0.25,random_state=10)

print(X_train_pca.shape,X_test.shape,Y_train.shape,Y_test.shape)

(17062, 4) (5688, 4) (17062,) (5688,)

```

Model Implementation

```

from sklearn.metrics import r2_score

#Random forest regressor
from sklearn.ensemble import RandomForestRegressor

rf_model = RandomForestRegressor()
rf_model.fit(X_train_pca,Y_train)

train_pred_rf = rf_model.predict(X_train_pca)
train_r2 = r2_score(Y_train,train_pred_rf)
test_pred_rf = rf_model.predict(X_test)
test_r2 = r2_score(Y_test,test_pred_rf)

#accuracy score
print("accuracy score of train data: "+str(round(100*train_r2, 4))
+"%")
print("accuracy score of test data: "+str(round(100*test_r2, 4))+"%")

accuracy score of train data: 91.1977%
accuracy score of test data: 83.8698%

#Adaboost regressor
from sklearn.ensemble import AdaBoostRegressor

abr_model = AdaBoostRegressor()
abr_model.fit(X_train_pca,Y_train)

train_pred_adboost = abr_model.predict(X_train_pca)
train_r2 = r2_score(Y_train,train_pred_adboost)
test_pred_adaboost = abr_model.predict(X_test)
test_r2 = r2_score(Y_test,test_pred_adaboost)

```

```
#accuracy score
print("accuracy score of train data: "+str(round(100*train_r2, 4))
+"%")
print("accuracy score of test data: "+str(round(100*test_r2, 4))+"%")

accuracy score of train data: 78.5218%
accuracy score of test data: 78.0277%
```