

```
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_csv('/content/drive/MyDrive/dataset/employee_burnout_analysis.csv')
burnoutDf
```

Type \	Employee ID	Date of Joining	Gender	Company
0	fffe32003000360033003200	9/30/2008	Female	Service
1	fffe3700360033003500	11/30/2008	Male	Service
2	fffe31003300320037003900	3/10/2008	Female	Product
3	fffe32003400380032003900	11/3/2008	Male	Service
4	fffe31003900340031003600	7/24/2008	Female	Service
...
22745	fffe31003500370039003100	12/30/2008	Female	Service
22746	fffe33003000350031003800	1/19/2008	Female	Product
22747	fffe390032003000	11/5/2008	Male	Service
22748	fffe33003300320036003900	1/10/2008	Female	Service
22749	fffe3400350031003800	1/6/2008	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"])
```

#Assigning Shape

```
burnoutDf.shape
```

(22750, 9)

#General Info Of Data

```
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

#Extracting the columns into Data set

```
burnoutDf.columns
```

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

#Checking for Null Values

```
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

#Description Of Parameters

```
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			

#Unique Values

```
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'
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'2008-01-12T00:00:00.000000000'	'2008-12-02T00:00:00.000000000'
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'2008-08-23T00:00:00.000000000'	'2008-06-05T00:00:00.000000000'
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'2008-11-23T00:00:00.000000000'	'2008-07-21T00:00:00.000000000'
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'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'
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'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'
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'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
```

```
#Minimizing irrelevant column
```

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

```
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 Negatively 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 Negatively Skewed and value is: -
0.4308950578815428
```

```
Burn Rate feature is Normally Distributed and value is:
0.045737370909640515
```

```
#Replace the null values with mean
```

```
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

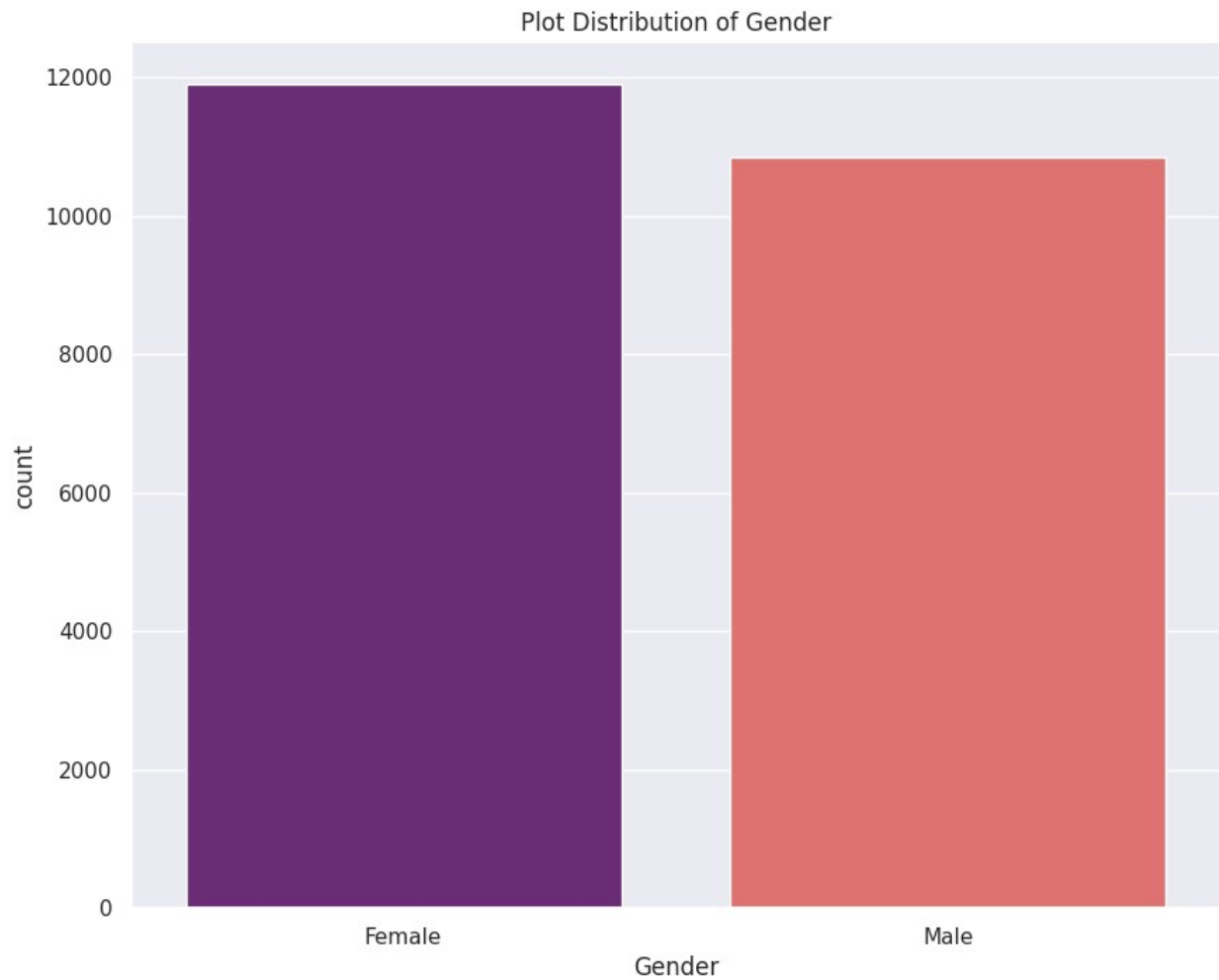
	Burn Rate
Designation	0.719284
Resource Allocation	0.811062
Mental Fatigue Score	0.878217
Burn Rate	1.000000

#Data Visualization

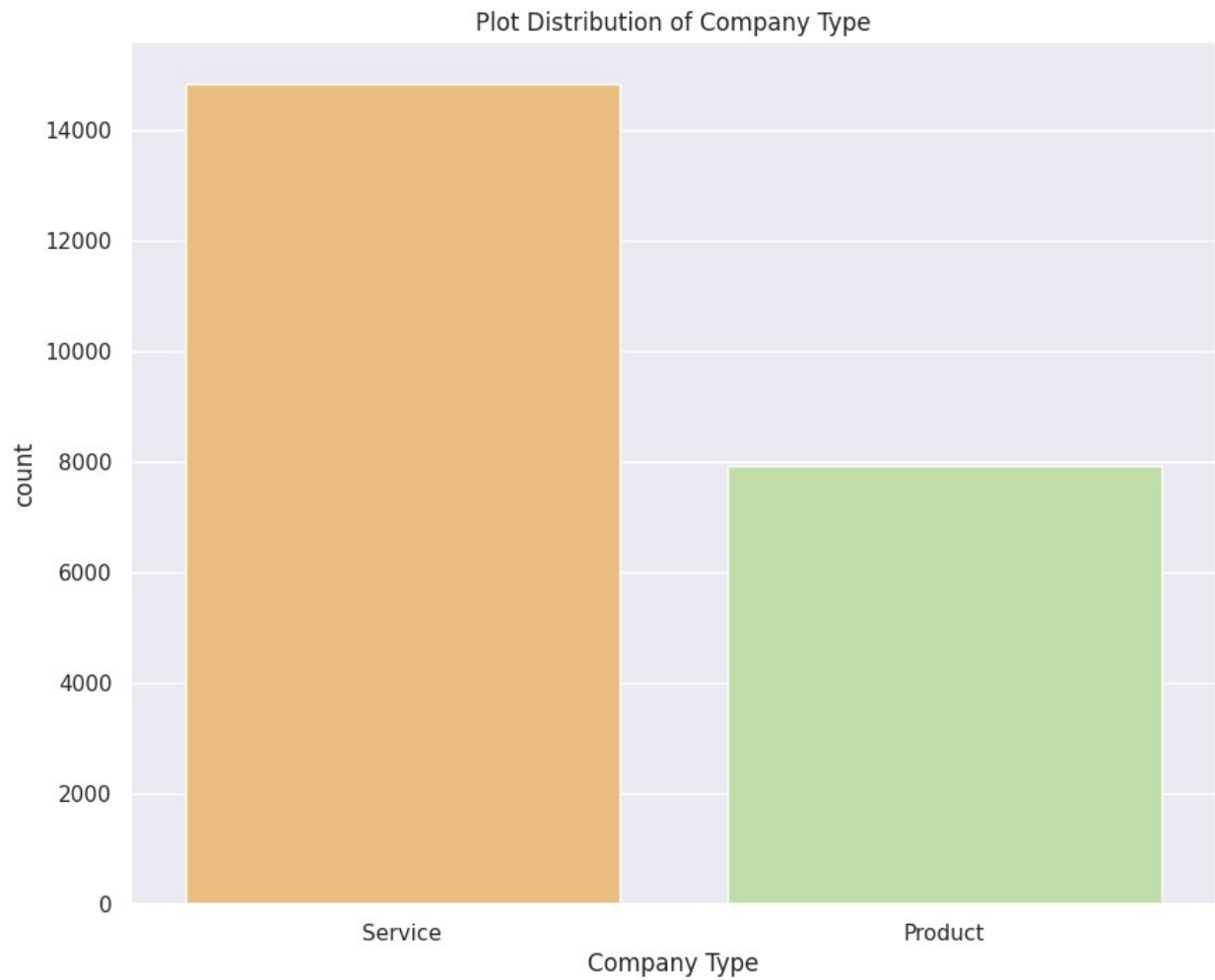
```
corr=burnoutDf.corr()
sns.set(rc={'figure.figsize':(14,12)})
fig = px.imshow(corr,text_auto=True, aspect="auto")
fig.show()
```

#Plot Distribution of Gender

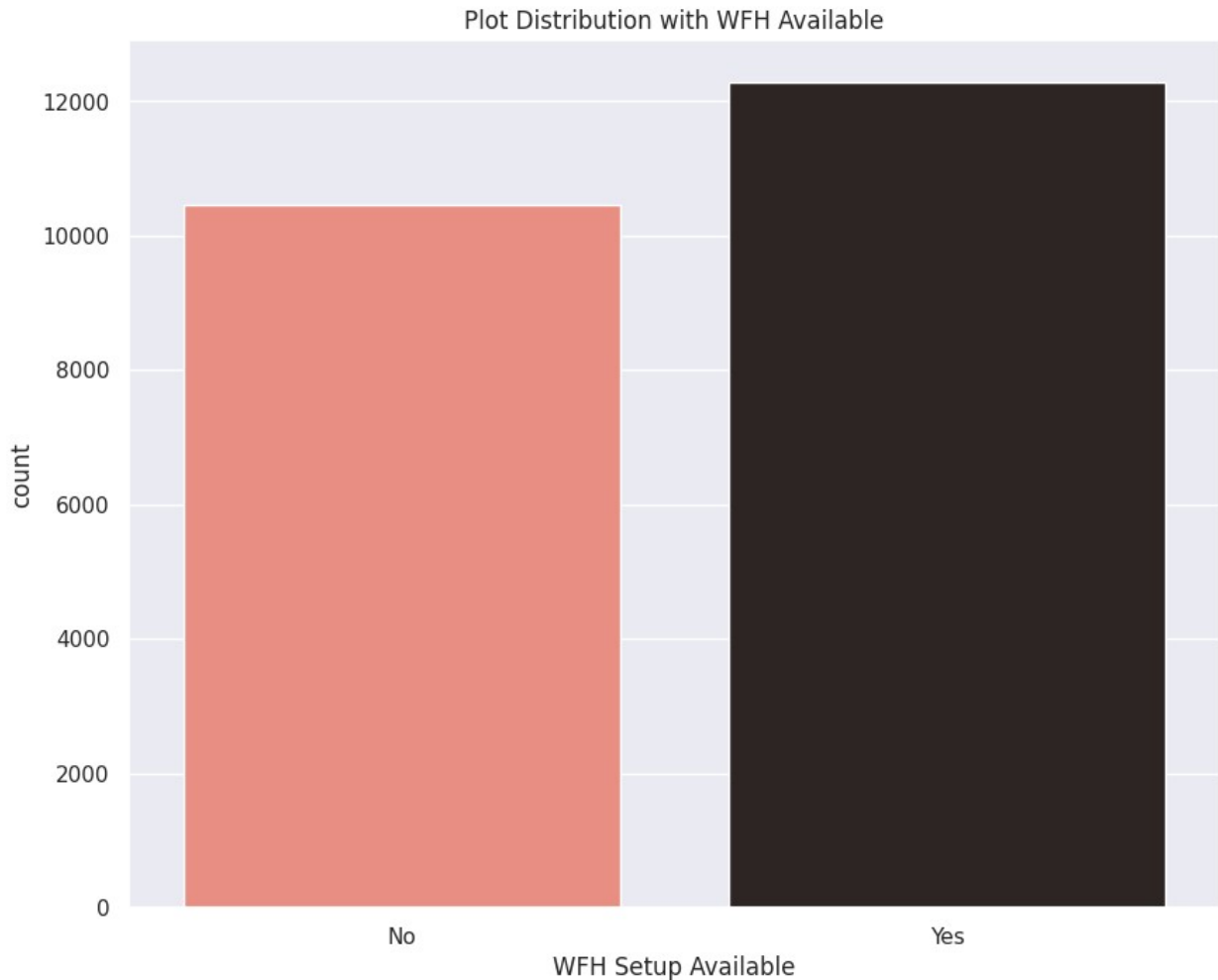
```
plt.figure(figsize=(10,8))
sns.countplot(x="Gender", data=burnoutDf, palette="magma")
plt.title("Plot Distribution of Gender")
plt.show()
```



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



```
plt.figure(figsize=(10,8))
sns.countplot(x= "WFH Setup Available",data=burnoutDf, palette
="dark:salmon_r")
plt.title("Plot Distribution with WFH Available")
plt.show()
```



#Attributes Histogram

```
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()
```

#Burnrate on basis of Designation

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

#Burnrate on basis of Gender

```
fig = px.line(burnoutDf, y="Burn Rate",color="Gender",title="Burn Rate
of Gender",color_discrete_sequence=px.colors.qualitative.Pastell1)
```

```

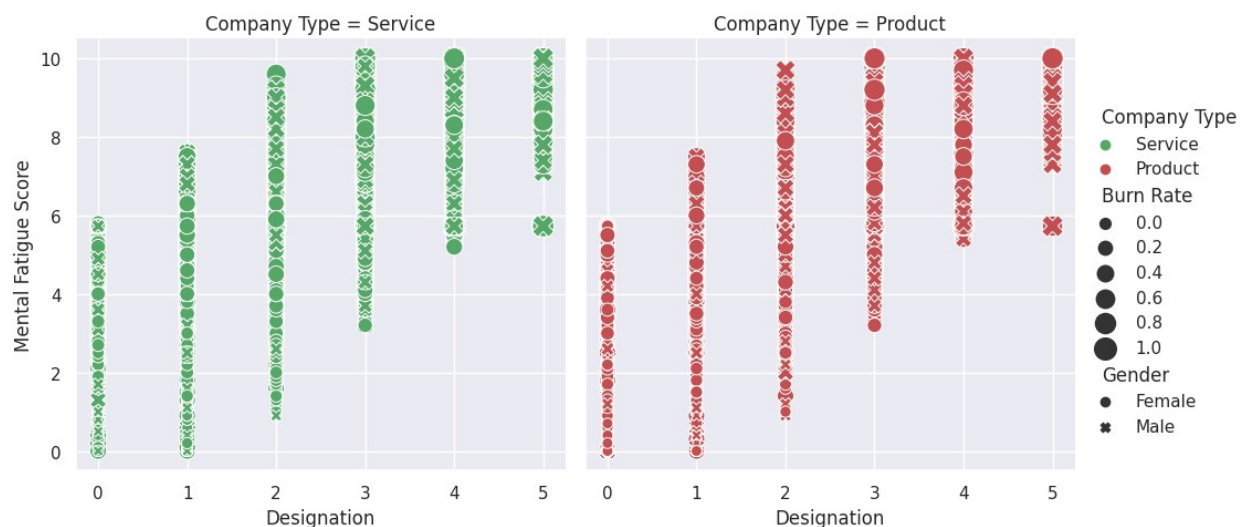
fig.update_layout(bargap=0.2)
fig.show()

#Mental Fatigue vs Designations
fig = px.line(burnoutDf, y="Mental Fatigue Score",color="Designation",title="Mental Fatigue vs Designation",color_discrete_sequence=px.colors.qualitative.Pastel1)
fig.update_layout(bargap=0.2)
fig.show()

#Plot Distribution of "Designation vs mental fatigue"
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 0x7fd8597eb640>

```



```

#Label encoding
from sklearn import preprocessing
Label_encode = preprocessing.LabelEncoder()

#Assigning
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)

```



```
#Checking Assigned Values
```

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

Gender

Female 0

Male 1

Name: GenderLabel, dtype: int64

```
#Checking Assigned Values
```

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

Company Type

Product 0

Service 1

Name: Company_TypeLabel, dtype: int64

```
#Checking Assigned Values
```

```
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

```
# Future selection
```

```
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 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,)

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)

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: 91.1968%
Accuracy score of test data:83.8641%

```