```
In [1]:
```

import pandas as pd

In [2]:

df = pd.read_csv('survey.csv')

In [3]:

df.head()

Out[3]:

	Timestamp	Age	Gender	Country	state	self_employed	family_history	treatment	work_interf
0	27-08-2014 11:29	37	Female	United States	IL	NaN	No	Yes	0
1	27-08-2014 11:29	44	М	United States	IN	NaN	No	No	Ra
2	27-08-2014 11:29	32	Male	Canada	NaN	NaN	No	No	Ra
3	27-08-2014 11:29	31	Male	United Kingdom	NaN	NaN	Yes	Yes	0
4	27-08-2014 11:30	31	Male	United States	TX	NaN	No	No	Nε
5 rows × 27 columns									

In [4]:

df.tail()

Out[4]:

	Timestamp	Age	Gender	Country	state	self_employed	family_history	treatment	work_in
1254	12-09-2015 11:17	26	male	United Kingdom	NaN	No	No	Yes	
1255	26-09-2015 01:07	32	Male	United States	IL	No	Yes	Yes	
1256	07-11-2015 12:36	34	male	United States	CA	No	Yes	Yes	Son
1257	30-11-2015 21:25	46	f	United States	NC	No	No	No	
1258	01-02-2016 23:04	25	Male	United States	IL	No	Yes	Yes	Son
5 rows × 27 columns									

```
In [5]:
df.shape
Out[5]:
(1259, 27)
In [6]:
df.columns
Out[6]:
'mental_health_consequence', 'phys_health_consequence', 'coworkers',
       'supervisor', 'mental_health_interview', 'phys_health_interview', 'mental_vs_physical', 'obs_consequence', 'comments'],
      dtype='object')
In [7]:
df.duplicated().sum()
Out[7]:
1
In [8]:
df = df.drop duplicates()
```

In [9]:

df.isnull().sum()

Out[9]:

Timestamp	0
Age	0
Gender	0
Country	0
state	514
self_employed	18
family_history	0
treatment	0
work_interfere	264
no_employees	0
remote_work	0
tech_company	0
benefits	0
care_options	0
wellness_program	0
seek_help	0
anonymity	0
leave	0
mental_health_consequence	0
<pre>phys_health_consequence</pre>	0
coworkers	0
supervisor	0
mental_health_interview	0
phys_health_interview	0
mental_vs_physical	0
obs_consequence	0
comments	1094
dtype: int64	

In [10]:

df.info()

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1258 entries, 0 to 1258
Data columns (total 27 columns):
#
    Column
                               Non-Null Count Dtype
                                _____
0
    Timestamp
                               1258 non-null
                                               object
1
    Age
                               1258 non-null
                                               int64
2
    Gender
                               1258 non-null
                                               obiect
3
    Country
                               1258 non-null object
4
    state
                               744 non-null
                                             object
5
    self employed
                               1240 non-null object
6
    family history
                               1258 non-null
                                               object
7
                               1258 non-null
    treatment
                                               object
8
    work interfere
                               994 non-null
                                               object
9
                               1258 non-null
    no employees
                                               object
10 remote_work
                               1258 non-null
                                               object
11 tech company
                              1258 non-null
                                               object
12 benefits
                              1258 non-null
                                               object
13 care_options
                               1258 non-null
                                               object
14 wellness program
                               1258 non-null
                                               object
15
    seek help
                               1258 non-null
                                               object
16 anonymity
                               1258 non-null
                                               object
17
                               1258 non-null
    leave
                                               object
18
    mental health consequence 1258 non-null
                                               object
19
    phys health consequence
                               1258 non-null
                                               object
20 coworkers
                               1258 non-null
                                               object
21
                               1258 non-null
    supervisor
                                               object
22
    mental_health_interview
                               1258 non-null
                                               object
23
    phys_health_interview
                               1258 non-null
                                               object
24 mental_vs_physical
                               1258 non-null
                                               object
25 obs_consequence
                               1258 non-null
                                               object
26 comments
                               164 non-null
                                               object
dtypes: int64(1), object(26)
memory usage: 275.2+ KB
```

In [11]:

```
df.describe()
```

Out[11]:

Age 1.258000e+03 count 7.949129e+07 mean 2.819419e+09 std min -1.726000e+03 25% 2.700000e+01 50% 3.100000e+01 75% 3.600000e+01 1.000000e+11 max

```
In [12]:
df.nunique()
Out[12]:
Timestamp
                              884
Age
                                53
Gender
                                49
                                48
Country
                                45
state
                                 2
self employed
family history
                                 2
                                 2
treatment
                                 4
work interfere
                                 6
no employees
remote work
                                 2
                                 2
tech company
                                 3
benefits
                                 3
care options
wellness program
                                 3
                                 3
seek_help
anonymity
                                 3
                                 5
leave
                                 3
mental health consequence
                                 3
phys_health_consequence
coworkers
                                 3
supervisor
                                 3
                                 3
mental_health_interview
                                 3
phys health interview
                                 3
mental_vs_physical
obs_consequence
                                 2
                              160
comments
dtype: int64
In [13]:
```

```
import matplotlib.pyplot as plt
import seaborn as sns
```

In [14]:

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

In [15]:

In [16]:

```
for i in df cat.columns:
    print(df cat[i].unique())
[nan 'Yes' 'No']
['No' 'Yes']
['Yes' 'No']
'Often' 'Rarely' 'Never' 'Sometimes' nan]
['Jun-25' 'More than 1000' '26-100' '100-500' '01-May' '500-1000']
['No' 'Yes']
['Yes' 'No']
['Yes' "Don't know" 'No']
['Not sure' 'No' 'Yes']
['No' "Don't know" 'Yes']
['Yes' "Don't know" 'No']
['Yes' "Don't know" 'No']
['Somewhat easy' "Don't know" 'Somewhat difficult' 'Very difficult'
 'Very easy']
['No' 'Maybe' 'Yes']
['No' 'Yes' 'Maybe']
['Some of them' 'No' 'Yes']
['Yes' 'No' 'Some of them']
['No' 'Yes' 'Maybe']
['Maybe' 'No' 'Yes']
['Yes' "Don't know" 'No']
['No' 'Yes']
```

In [17]:

```
for i in df_cat.columns:
    print(df_cat[i].value_counts())
```

```
No
       1094
Yes
       146
Name: self employed, dtype: int64
Nο
       767
       491
Yes
Name: family history, dtype: int64
       636
Nο
       622
Name: treatment, dtype: int64
Sometimes
             465
Never
             213
Rarely
             172
Often
             144
Name: work_interfere, dtype: int64
Jun-25
                  289
26-100
                  289
More than 1000
                  282
100-500
                  176
01-May
                  162
500-1000
                   60
Name: no_employees, dtype: int64
No
       882
Yes
       376
Name: remote_work, dtype: int64
Yes
       1031
No
        227
Name: tech_company, dtype: int64
Yes
              477
Don't know
              408
              373
Name: benefits, dtype: int64
            500
No
Yes
            444
Not sure
            314
Name: care_options, dtype: int64
No
              841
Yes
              229
Don't know
              188
Name: wellness_program, dtype: int64
No
              645
Don't know
              363
Yes
              250
Name: seek_help, dtype: int64
Don't know
           819
Yes
              375
No
              64
Name: anonymity, dtype: int64
Don't know
Somewhat easy
                      266
                      206
Very easy
Somewhat difficult
                      126
Very difficult
                       98
Name: leave, dtype: int64
         490
No
Maybe
         477
Yes
         291
Name: mental_health_consequence, dtype: int64
No
         925
         272
Maybe
          61
Name: phys_health_consequence, dtype: int64
Some of them
               773
```

```
260
No
Yes
                 225
Name: coworkers, dtype: int64
Yes
                 516
                 392
No
Some of them
                 350
Name: supervisor, dtype: int64
         1007
No
Maybe
           207
           44
Yes
Name: mental health interview, dtype: int64
Maybe
         557
No
          499
Yes
          202
Name: phys_health_interview, dtype: int64
Don't know
               343
Yes
No
               339
Name: mental_vs_physical, dtype: int64
No
       1074
Yes
        184
Name: obs consequence, dtype: int64
In [18]:
for i in df cat.columns:
    plt.figure(figsize = (14,6))
    sns.countplot(df_cat[i],data=df_cat, palette = 'hls')
    plt.xticks(rotation = 90)
    plt.yticks(rotation = 90)
    plt.show()
  200
  009
  000
 count
400
  300
  200
  100
                       9
                                                              ęs-
                                       family_history
  8
  200
```

In [19]:

print(df['Country'].value_counts())

United States	751
United Kingdom	184
Canada	72
Germany	45
Ireland	27
Netherlands	27
Australia	21
France	13
India	10
New Zealand	8
Poland	7
Switzerland	7
Sweden	7
Italy	7
South Africa	6
Belgium	6
Brazil	6
Israel	5
	4
Singapore	
Bulgaria	4
Austria	3
Finland	3
Mexico	3
Russia	3
Denmark	2
Greece	2
Colombia	2
Croatia	2
Portugal	2
Moldova	1
Georgia	1
Bahamas, The	1
China	1
Thailand	1
Czech Republic	1
Norway	1
Romania	1
Nigeria	1
Japan	1
Hungary	1
Bosnia and Herzegovina	1
Uruguay	1
Spain	1
Zimbabwe	1
Latvia	1
Costa Rica	1
Slovenia	1
Philippines	1
Name: Country, dtype: into	_
wame. country, acype. Into	→

```
In [20]:
```

```
print(df['state'].unique())

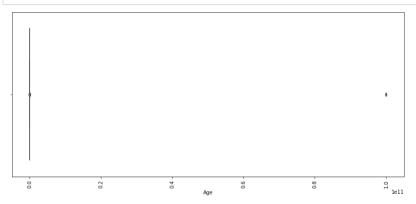
['IL' 'IN' nan 'TX' 'TN' 'MI' 'OH' 'CA' 'CT' 'MD' 'NY' 'NC' 'MA' 'IA' 'PA'
    'WA' 'WI' 'UT' 'NM' 'OR' 'FL' 'MN' 'MO' 'AZ' 'CO' 'GA' 'DC' 'NE' 'WV'
    'OK' 'KS' 'VA' 'NH' 'KY' 'AL' 'NV' 'NJ' 'SC' 'VT' 'SD' 'ID' 'MS' 'RI'
    'WY' 'LA' 'ME']

In [21]:
```

```
df.drop(columns=['Timestamp', 'Country', 'state', 'comments'], inplace = True)
```

In [22]:

```
plt.figure(figsize = (14,6))
sns.boxplot(df['Age'],data=df, palette = 'hls')
plt.xticks(rotation = 90)
plt.yticks(rotation = 90)
plt.show()
```



In [23]:

```
Q1 = df.quantile(0.25)
Q3 = df.quantile(0.75)
IQR = Q3 - Q1
print(IQR)
```

Age 9.0 dtype: float64

```
In [24]:
```

```
df[\sim((df < (Q1 - 1.5 * IQR)) | (df > (Q3 + 1.5 * IQR))).any(axis=1)]
```

Out[24]:

	Age	Gender	self_employed	family_history	treatment	work_interfere	no_employees	remo
0	37	Female	NaN	No	Yes	Often	Jun-25	
1	44	М	NaN	No	No	Rarely	More than 1000	
2	32	Male	NaN	No	No	Rarely	Jun-25	
3	31	Male	NaN	Yes	Yes	Often	26-100	
4	31	Male	NaN	No	No	Never	100-500	
1254	26	male	No	No	Yes	NaN	26-100	
1255	32	Male	No	Yes	Yes	Often	26-100	
1256	34	male	No	Yes	Yes	Sometimes	More than 1000	
1257	46	f	No	No	No	NaN	100-500	
1258	25	Male	No	Yes	Yes	Sometimes	26-100	

1218 rows × 23 columns

←

In [25]:

df.shape

Out[25]:

(1258, 23)

In [26]:

```
df.drop(df[df['Age'] < 0].index, inplace = True)
df.drop(df[df['Age'] > 100].index, inplace = True)
```

In [27]:

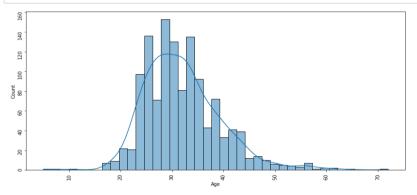
```
df['Age'].unique()
```

Out[27]:

```
array([37, 44, 32, 31, 33, 35, 39, 42, 23, 29, 36, 27, 46, 41, 34, 30, 40, 38, 50, 24, 18, 28, 26, 22, 19, 25, 45, 21, 43, 56, 60, 54, 55, 48, 20, 57, 58, 47, 62, 51, 65, 49, 5, 53, 61, 8, 11, 72], dtype=int64)
```

In [28]:

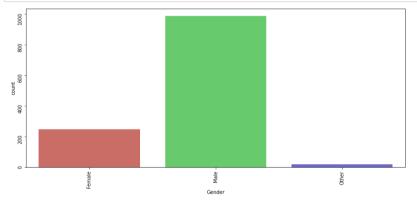
```
plt.figure(figsize = (14,6))
sns.histplot(df['Age'], kde = True, palette = 'hls')
plt.xticks(rotation = 90)
plt.yticks(rotation = 90)
plt.show()
```



In [29]:

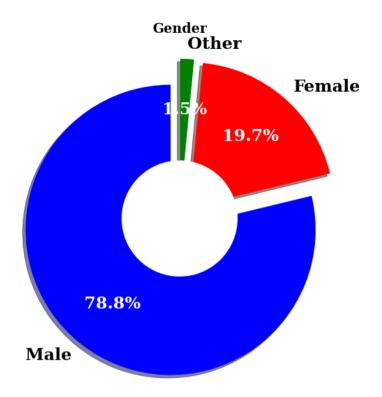
```
In [30]:
```

```
df['Gender'].unique()
Out[30]:
array(['Female', 'Male', 'Other'], dtype=object)
In [31]:
df['Gender'].value_counts()
Out[31]:
Male
          987
Female
          247
Other
           19
Name: Gender, dtype: int64
In [32]:
plt.figure(figsize = (14,6))
sns.countplot(df['Gender'],data=df, palette = 'hls')
plt.xticks(rotation = 90)
plt.yticks(rotation = 90)
plt.show()
```



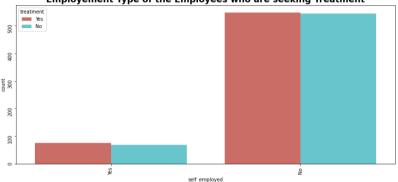
In [33]:

```
gender_data = df['Gender'].value_counts()
explode = (0.1, 0.1, 0.1)
plt.figure(figsize=(14, 10))
patches, texts, pcts = plt.pie(gender_data,
                                 labels = gender data.index,
                                 colors = ['blue', 'red', 'green'],
                                 pctdistance = 0.65,
                                 shadow = True,
                                 startangle = 90,
                                 explode = explode,
                                 autopct = '%1.1f%%',
                                 textprops={ 'fontsize': 25,
                                               'color': 'black',
                                               'weight': 'bold',
                                               'family': 'serif' })
plt.setp(pcts, color='white')
hfont = {'fontname':'serif', 'weight': 'bold'}
plt.title('Gender', size=20, **hfont)
centre_circle = plt.Circle((0,0),0.40,fc='white')
fig = plt.gcf()
fig.gca().add_artist(centre_circle)
plt.show()
```

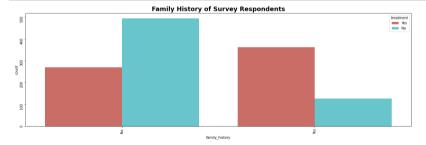


In [34]:

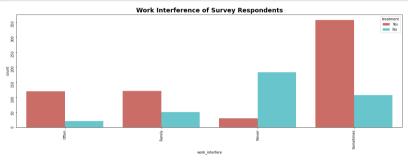




In [35]:

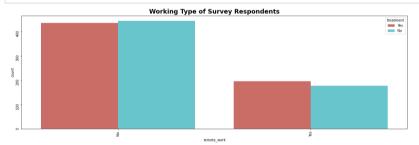


In [36]:

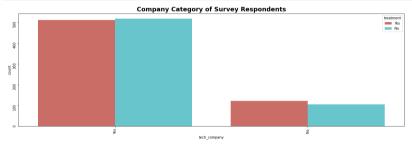


In [37]:

```
plt.figure(figsize = (20,6))
sns.countplot(df['remote_work'], hue = df['treatment'], palette='hls')
plt.title('Working Type of Survey Respondents', fontsize=18, fontweight='bold')
plt.xticks(rotation = 90)
plt.yticks(rotation = 90)
plt.show()
```

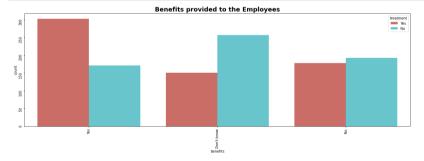


In [38]:

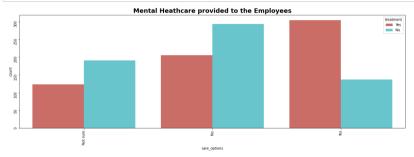


In [39]:

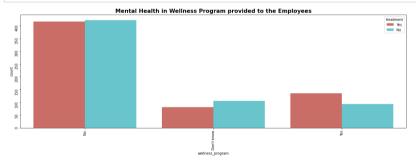
```
plt.figure(figsize = (20,6))
sns.countplot(df['benefits'], hue = df['treatment'], palette='hls')
plt.title('Benefits provided to the Employees', fontsize=18, fontweight='bold')
plt.xticks(rotation = 90)
plt.yticks(rotation = 90)
plt.show()
```



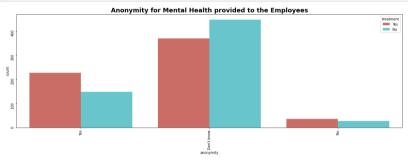
In [40]:



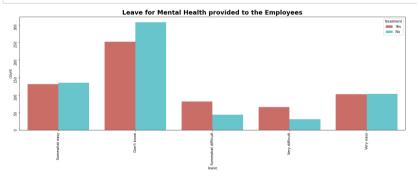
In [41]:



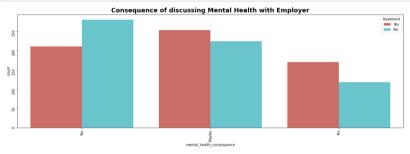
In [42]:



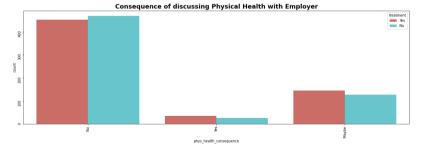
In [43]:



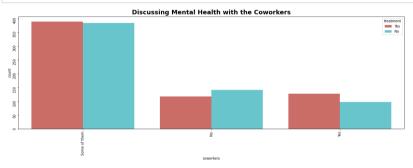
In [44]:



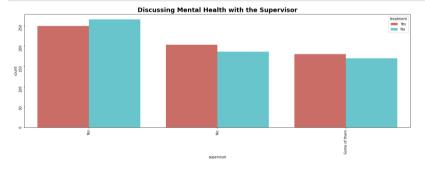
In [45]:



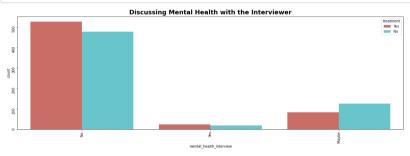
In [46]:



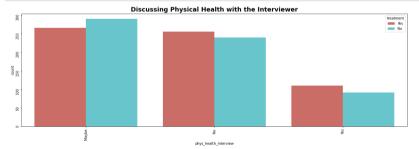
In [47]:



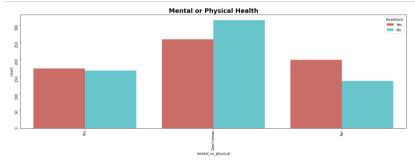
In [48]:



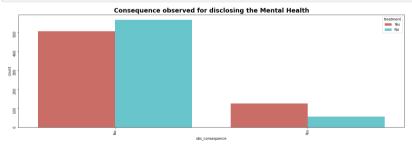
In [49]:



In [50]:



In [51]:



```
In [52]:
```

```
df['work_interfere'] = df['work_interfere'].fillna('Not Avialable')
```

```
In [53]:
```

```
df['self_employed'] = df['self_employed'].fillna('No')
```

```
In [54]:
```

```
df.isnull().sum()
```

Out[54]:

```
Age
                              0
Gender
                              0
self employed
                              0
family history
                              0
                              0
treatment
work interfere
                              0
no employees
                              0
remote_work
                              0
                              0
tech company
benefits
                              0
care_options
                              0
wellness_program
                              0
seek help
                              0
anonymity
                              0
                              a
leave
mental_health_consequence
                              0
phys_health_consequence
                              0
coworkers
                              0
supervisor
                              а
mental_health_interview
                              0
phys health interview
mental_vs_physical
                              0
obs_consequence
                              0
dtype: int64
```

In [55]:

```
\textbf{from} \ \textbf{sklearn.preprocessing} \ \textbf{import} \ \textbf{LabelEncoder}
```

In [56]:

```
object_cols = ['Gender', 'self_employed', 'family_history', 'treatment',
    'work_interfere', 'no_employees', 'remote_work', 'tech_company',
    'benefits', 'care_options', 'wellness_program', 'seek_help',
    'anonymity', 'leave', 'mental_health_consequence',
    'phys_health_consequence', 'coworkers', 'supervisor',
    'mental_health_interview', 'phys_health_interview',
    'mental_vs_physical', 'obs_consequence']
```

In [57]:

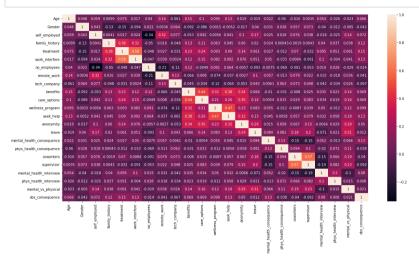
```
label_encoder = LabelEncoder()
for col in object_cols:
   label_encoder.fit(df[col])
   df[col] = label_encoder.transform(df[col])
```

In [58]:

```
corr = df.corr()
```

In [59]:

```
plt.figure(figsize = (20,10))
sns.heatmap(corr, annot = True)
plt.show()
```



In [60]:

```
from sklearn.linear_model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score, recall_score, plot_roc_curve, confusion_matrix,
from sklearn.ensemble import RandomForestClassifier, AdaBoostClassifier, GradientBoostingCla
from xgboost.sklearn import XGBClassifier
import xgboost as xgb
from sklearn import metrics
```

In [61]:

```
In [62]:
```

```
key = ['LogisticRegression','KNeighborsClassifier','DecisionTreeClassifier','RandomForestCla
value = [LogisticRegression(), KNeighborsClassifier(n neighbors = 2, weights = 'uniform'), De
models = dict(zip(key,value))
models
4 I
                                                                                            Þ
Out[62]:
{ 'LogisticRegression': LogisticRegression(),
 'KNeighborsClassifier': KNeighborsClassifier(n neighbors=2),
 'DecisionTreeClassifier': DecisionTreeClassifier(random state=10),
 'RandomForestClassifier': RandomForestClassifier(n estimators=60, random stat
 'GradientBoostingClassifier': GradientBoostingClassifier(random state=20),
 'AdaBoostClassifier': AdaBoostClassifier(),
 'XGBClassifier': XGBClassifier(base score=None, booster='gbtree', callbacks=N
one,
               colsample bylevel=None, colsample bynode=None,
               colsample bytree=None, early stopping rounds=None,
               enable categorical=False, eval metric=None, gamma=None,
               gpu id=None, grow policy=None, importance type=None,
               interaction constraints=None, learning rate=None, max bin=None,
               max cat to onehot=None, max delta step=None, max depth=None,
               max_leaves=None, min_child_weight=None, missing=nan,
               monotone constraints=None, n estimators=100, n jobs=None,
               num parallel tree=None, predictor=None, random state=0,
               reg alpha=None, reg lambda=None, ...)}
```

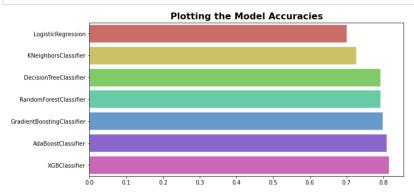
In [63]:

```
predicted =[]
for name,algo in models.items():
    model=algo
    model.fit(X_train,y_train)
    predict = model.predict(X_test)
    acc = accuracy_score(y_test, predict)
    predicted.append(acc)
    print(name,acc)
```

LogisticRegression 0.7929936305732485 KNeighborsClassifier 0.7006369426751592 DecisionTreeClassifier 0.7261146496815286 RandomForestClassifier 0.8152866242038217 GradientBoostingClassifier 0.8089171974522293 AdaBoostClassifier 0.7929936305732485 XGBClassifier 0.7993630573248408

In [64]:

```
plt.figure(figsize = (10,5))
ax = sns.barplot(x = predicted, y = key, palette='hls', order=predicted.sort())
plt.title("Plotting the Model Accuracies", fontsize=16, fontweight="bold")
plt.show()
```

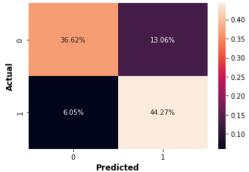


In [65]:

```
import numpy as np
```

In [66]:

Confusion Matrix of Gradient Boost Classifier



In [67]:

```
fpr, tpr, thresholds = metrics.roc_curve(y_test, pred)
plt.figure(figsize=(8,8))
roc_auc = metrics.auc(fpr, tpr)
plt.plot(fpr, tpr, color='darkorange', label='ROC curve (area = %0.2f)' % roc_auc)
plt.plot([0, 1], [0, 1], color='navy', linestyle='--')
plt.xlim([0.0, 1.0])
plt.xlim([0.0, 1.0])
plt.rcParams['font.size'] = 12
plt.title('ROC Curve', fontweight = 'bold', fontsize=16)
plt.xlabel('False Positive Rate (1 - Specificity)', fontweight = 'bold', fontsize=14)
plt.ylabel('True Positive Rate (Sensitivity)', fontweight = 'bold', fontsize=14)
plt.legend(loc="lower right")
plt.show()
metrics.roc_curve(y_test, pred)
plt.show()
```

