

importing libraries

```
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import os
from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

```
import seaborn as sns
import matplotlib.pyplot as plt
import plotly.express as px
import warnings
warnings.filterwarnings('ignore')
```

reading data

```
df1 = pd.read_csv('/content/drive/MyDrive/mental fitness/mental-
disease.csv')
df2 = pd.read_csv('/content/drive/MyDrive/mental fitness/prevalence-by-
mental-and-substance-use-disorder.csv')
```

```
df1.head()
```

	Entity	Code	Year	\
0	Afghanistan	AFG	1990	
1	Afghanistan	AFG	1991	
2	Afghanistan	AFG	1992	
3	Afghanistan	AFG	1993	
4	Afghanistan	AFG	1994	

DALYs (Disability-Adjusted Life Years) - Mental disorders - Sex:
Both - Age: All Ages (Percent)

0	1.696670
1	1.734281
2	1.791189
3	1.776779
4	1.712986

```
df2.head()
```

	Entity	Code	Year	\
0	Afghanistan	AFG	1990	
1	Afghanistan	AFG	1991	
2	Afghanistan	AFG	1992	
3	Afghanistan	AFG	1993	
4	Afghanistan	AFG	1994	

Prevalence - Schizophrenia - Sex: Both - Age: Age-standardized
(Percent) \

0	0.228979
1	0.228120
2	0.227328
3	0.226468
4	0.225567

Prevalence - Bipolar disorder - Sex: Both - Age: Age-standardized
(Percent) \

0	0.721207
1	0.719952
2	0.718418
3	0.717452
4	0.717012

Prevalence - Eating disorders - Sex: Both - Age: Age-standardized
(Percent) \

0	0.131001
1	0.126395
2	0.121832
3	0.117942
4	0.114547

Prevalence - Anxiety disorders - Sex: Both - Age: Age-standardized
(Percent) \

0	4.835127
1	4.821765

2	4.801434
3	4.789363
4	4.784923

Prevalence - Drug use disorders - Sex: Both - Age: Age-standardized (Percent) \

0	0.454202
1	0.447112
2	0.441190
3	0.435581
4	0.431822

Prevalence - Depressive disorders - Sex: Both - Age: Age-standardized (Percent) \

0	5.125291
1	5.116306
2	5.106558
3	5.100328
4	5.099424

Prevalence - Alcohol use disorders - Sex: Both - Age: Age-standardized (Percent)

0	0.444036
1	0.444250
2	0.445501
3	0.445958
4	0.445779

merging two datasets

```
data = pd.merge(df1, df2)
data.head()
```

	Entity	Code	Year	\
0	Afghanistan	AFG	1990	
1	Afghanistan	AFG	1991	
2	Afghanistan	AFG	1992	
3	Afghanistan	AFG	1993	
4	Afghanistan	AFG	1994	

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1	5.116306
2	5.106558
3	5.100328
4	5.099424

Prevalence - Alcohol use disorders - Sex: Both - Age: Age-standardized (Percent)

0	0.444036
---	----------

1	0.444250
2	0.445501
3	0.445958
4	0.445779

data cleaning

```
data.isnull().sum()
```

```
Entity
```

```
0
```

```
Code
```

```
690
```

```
Year
```

```
0
```

```
DALYs (Disability-Adjusted Life Years) - Mental disorders - Sex: Both  
- Age: All Ages (Percent) 0
```

```
Prevalence - Schizophrenia - Sex: Both - Age: Age-standardized  
(Percent) 0
```

```
Prevalence - Bipolar disorder - Sex: Both - Age: Age-standardized  
(Percent) 0
```

```
Prevalence - Eating disorders - Sex: Both - Age: Age-standardized  
(Percent) 0
```

```
Prevalence - Anxiety disorders - Sex: Both - Age: Age-standardized  
(Percent) 0
```

```
Prevalence - Drug use disorders - Sex: Both - Age: Age-standardized  
(Percent) 0
```

```
Prevalence - Depressive disorders - Sex: Both - Age: Age-standardized  
(Percent) 0
```

```
Prevalence - Alcohol use disorders - Sex: Both - Age: Age-standardized  
(Percent) 0
```

```
dtype: int64
```

```
data.drop('Code',axis=1,inplace=True)
```

```
data.head()
```

```

      Entity  Year  \
0  Afghanistan  1990
1  Afghanistan  1991
2  Afghanistan  1992
3  Afghanistan  1993
4  Afghanistan  1994
```

```

      DALYs (Disability-Adjusted Life Years) - Mental disorders - Sex:
Both - Age: All Ages (Percent)  \
```

0	1.696670
1	1.734281
2	1.791189
3	1.776779
4	1.712986

Prevalence - Schizophrenia - Sex: Both - Age: Age-standardized
(Percent) \

0	0.228979
1	0.228120
2	0.227328
3	0.226468
4	0.225567

Prevalence - Bipolar disorder - Sex: Both - Age: Age-standardized
(Percent) \

0	0.721207
1	0.719952
2	0.718418
3	0.717452
4	0.717012

Prevalence - Eating disorders - Sex: Both - Age: Age-standardized
(Percent) \

0	0.131001
1	0.126395
2	0.121832
3	0.117942
4	0.114547

Prevalence - Anxiety disorders - Sex: Both - Age: Age-standardized

(Percent) \	
0	4.835127
1	4.821765
2	4.801434
3	4.789363
4	4.784923

Prevalence - Drug use disorders - Sex: Both - Age: Age-standardized (Percent) \	
0	0.454202
1	0.447112
2	0.441190
3	0.435581
4	0.431822

Prevalence - Depressive disorders - Sex: Both - Age: Age-standardized (Percent) \	
0	5.125291
1	5.116306
2	5.106558
3	5.100328
4	5.099424

Prevalence - Alcohol use disorders - Sex: Both - Age: Age-standardized (Percent)	
0	0.444036
1	0.444250
2	0.445501
3	0.445958
4	0.445779

data.size,data.shape


```
(68400, (6840, 10))
```

```
data.set_axis(['Country', 'Year', 'Schizophrenia', 'Bipolar_disorder',  
'Eating_disorder', 'Anxiety', 'drug_usage', 'depression', 'alcohol', 'mental_fitness'], axis='columns', inplace=True)
```

```
data.head()
```

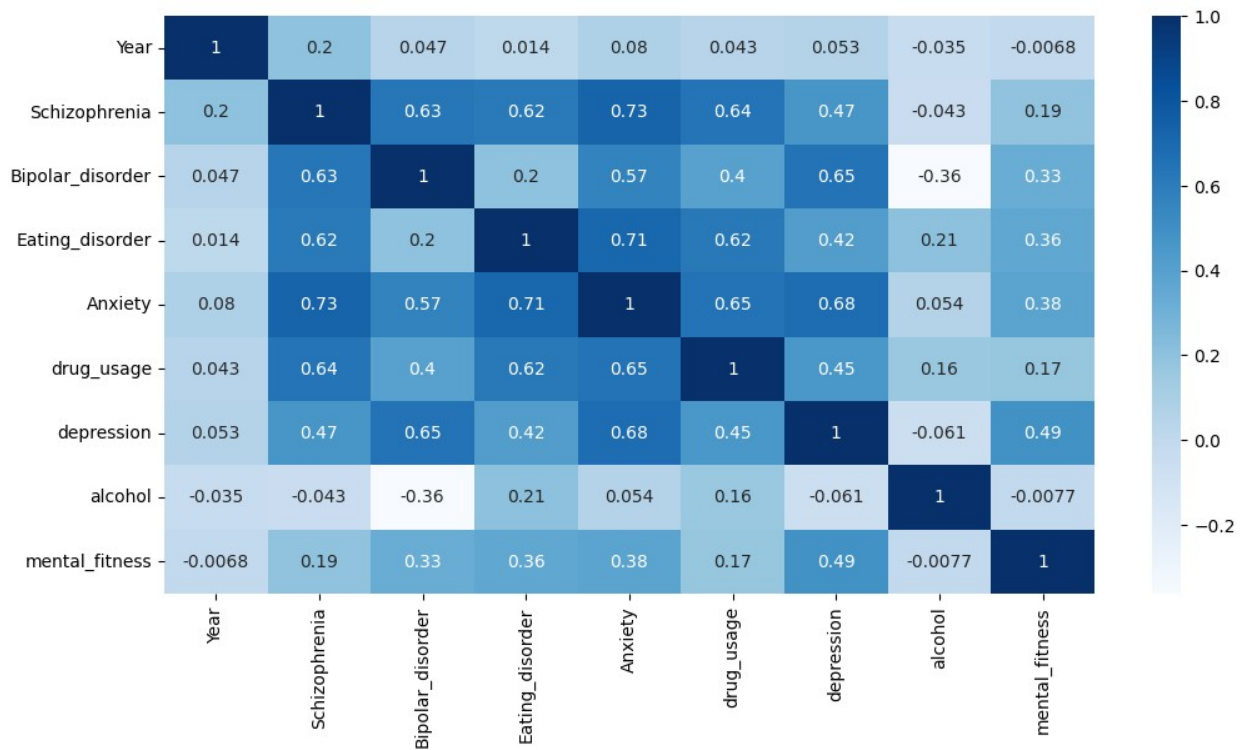
	Country	Year	Schizophrenia	Bipolar_disorder	Eating_disorder
0	Afghanistan	1990	1.696670	0.228979	0.721207
1	Afghanistan	1991	1.734281	0.228120	0.719952
2	Afghanistan	1992	1.791189	0.227328	0.718418
3	Afghanistan	1993	1.776779	0.226468	0.717452
4	Afghanistan	1994	1.712986	0.225567	0.717012

	Anxiety	drug_usage	depression	alcohol	mental_fitness
0	0.131001	4.835127	0.454202	5.125291	0.444036
1	0.126395	4.821765	0.447112	5.116306	0.444250
2	0.121832	4.801434	0.441190	5.106558	0.445501
3	0.117942	4.789363	0.435581	5.100328	0.445958
4	0.114547	4.784923	0.431822	5.099424	0.445779

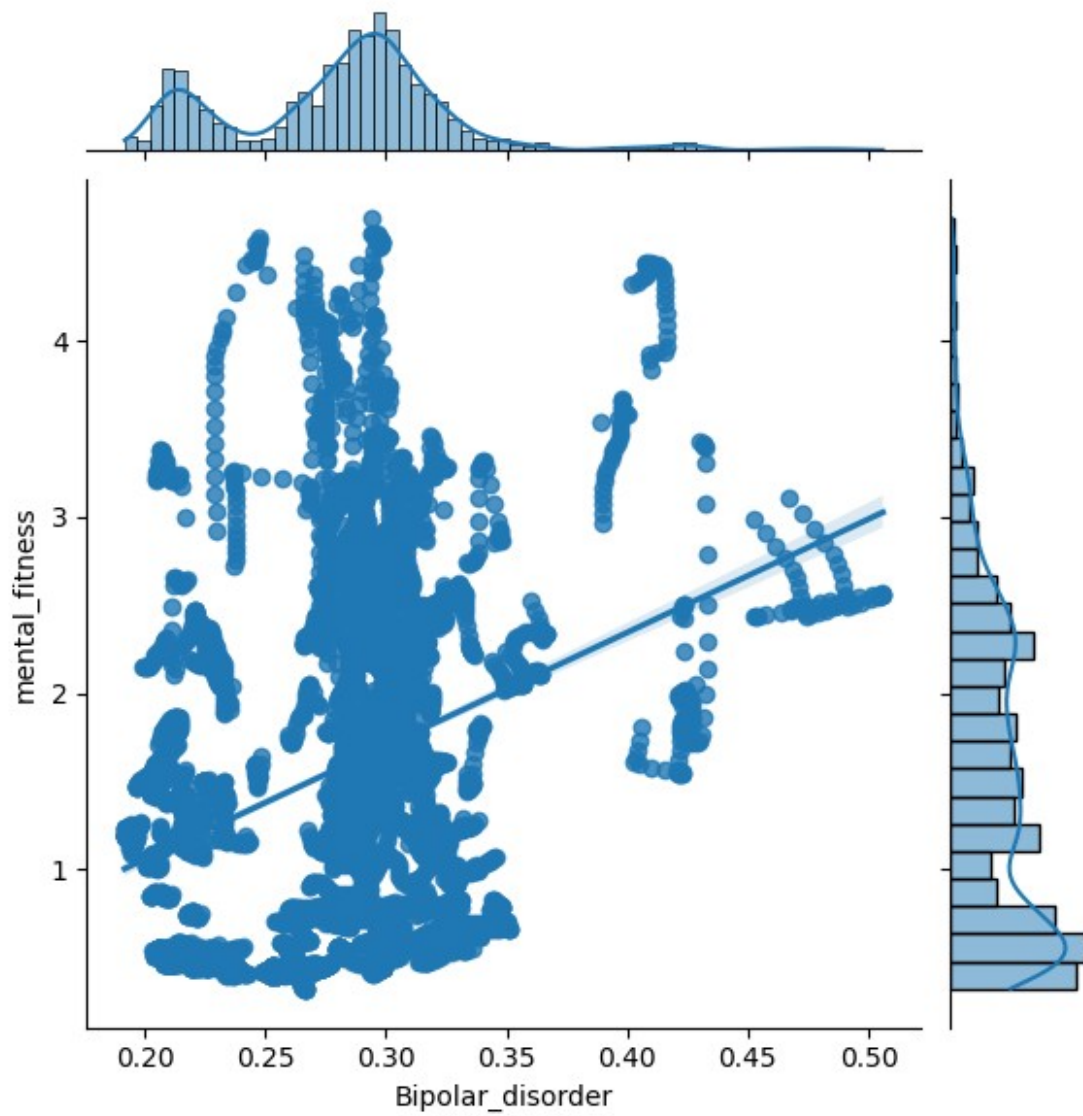
explaratory data analysis(EDA)

```
plt.figure(figsize=(12,6))  
sns.heatmap(data.corr(),annot=True,cmap='Blues')  
plt.plot()
```

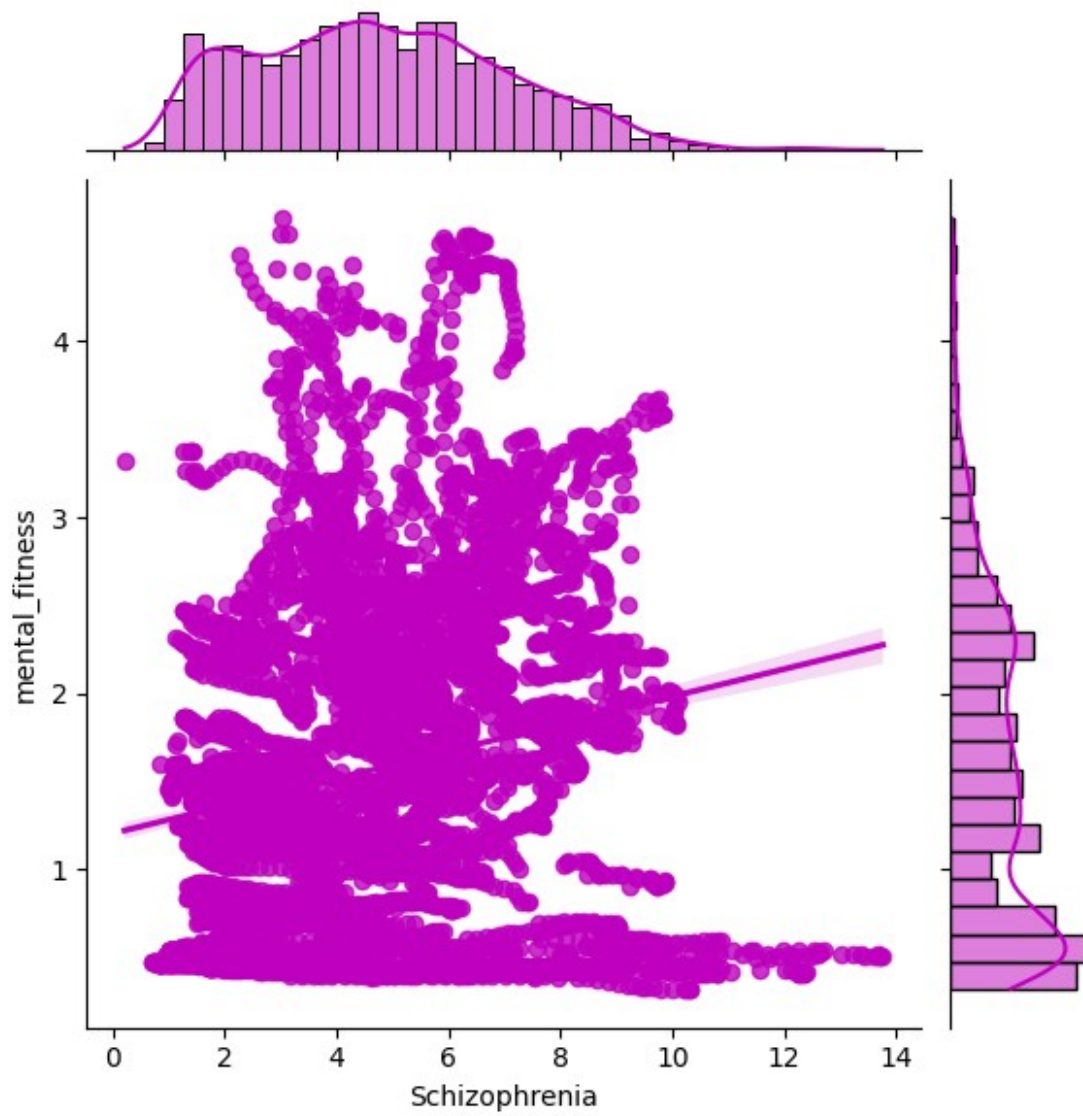
```
[]
```



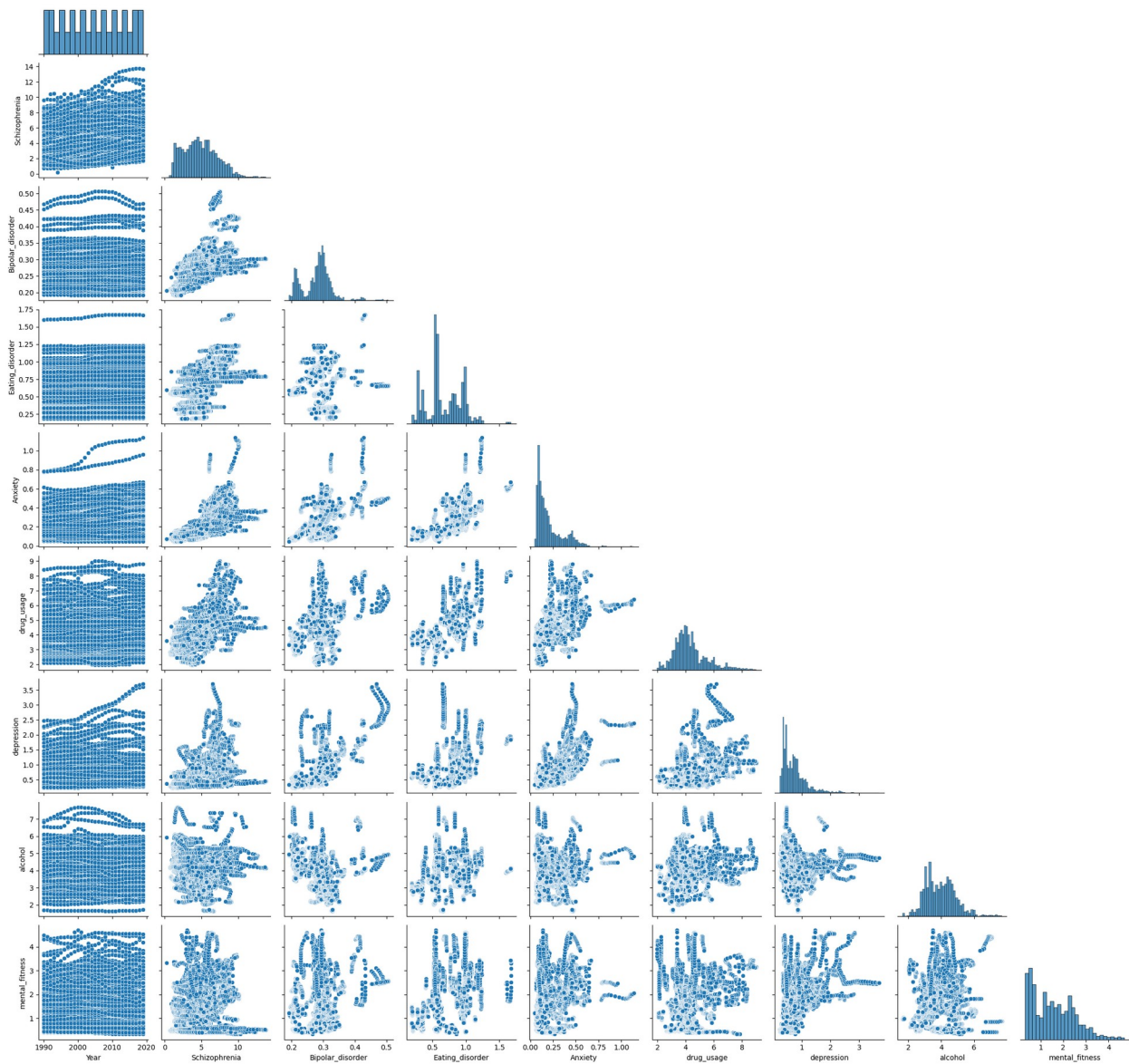
```
sns.jointplot(data=data,x='Bipolar_disorder',y='mental_fitness',kind='
reg')
plt.show()
```



```
sns.jointplot(x='Schizophrenia',y='mental_fitness',data=data,kind='reg',color='m')  
plt.show()
```



```
sns.pairplot(data,corner=True)  
plt.show()
```



```
mean = data['mental_fitness'].mean()
```

```
mean
```

```
1.5788071625377194
```

```
fig = px.pie(data, values='mental_fitness', names='Year')
```

```
fig.show()
```

year wise variation in mental fitness in different countries

```
fig = px.line(data, x="Year", y="mental_fitness",  
color='Country', markers=True, color_discrete_sequence=['red', 'blue'], te  
mplate='plotly_dark')  
fig.show()
```

```
df = data.copy()
```

```
df.head()
```

	Country	Year	Schizophrenia	Bipolar_disorder	Eating_disorder
0	Afghanistan	1990	1.696670	0.228979	0.721207
1	Afghanistan	1991	1.734281	0.228120	0.719952
2	Afghanistan	1992	1.791189	0.227328	0.718418
3	Afghanistan	1993	1.776779	0.226468	0.717452
4	Afghanistan	1994	1.712986	0.225567	0.717012

	Anxiety	drug_usage	depression	alcohol	mental_fitness
0	0.131001	4.835127	0.454202	5.125291	0.444036
1	0.126395	4.821765	0.447112	5.116306	0.444250
2	0.121832	4.801434	0.441190	5.106558	0.445501
3	0.117942	4.789363	0.435581	5.100328	0.445958
4	0.114547	4.784923	0.431822	5.099424	0.445779

```
df.info()
```

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

```
Int64Index: 6840 entries, 0 to 6839
```

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

#	Column	Non-Null	Count	Dtype
---	-----	-----	-----	-----
0	Country	6840	non-null	object
1	Year	6840	non-null	int64
2	Schizophrenia	6840	non-null	float64
3	Bipolar_disorder	6840	non-null	float64
4	Eating_disorder	6840	non-null	float64
5	Anxiety	6840	non-null	float64
6	drug_usage	6840	non-null	float64
7	depression	6840	non-null	float64
8	alcohol	6840	non-null	float64
9	mental_fitness	6840	non-null	float64

```

dtypes: float64(8), int64(1), object(1)
memory usage: 587.8+ KB

from sklearn.preprocessing import LabelEncoder
l=LabelEncoder()
for i in df.columns:
    if df[i].dtype == 'object':
        df[i]=l.fit_transform(df[i])

X = df.drop('mental_fitness',axis=1)
y = df['mental_fitness']

from sklearn.model_selection import train_test_split
xtrain, xtest, ytrain, ytest = train_test_split(X, y, test_size=0.2,
random_state=2)

X = df.drop('mental_fitness',axis=1)
y = df['mental_fitness']

from sklearn.model_selection import train_test_split
xtrain, xtest, ytrain, ytest = train_test_split(X, y, test_size=0.2,
random_state=2)

```

applying classification algorithm

linear regression

```

from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score
lr = LinearRegression()
lr.fit(xtrain,ytrain)

# model evaluation for training set
ytrain_pred = lr.predict(xtrain)
mse = mean_squared_error(ytrain, ytrain_pred)
rmse = (np.sqrt(mean_squared_error(ytrain, ytrain_pred)))
r2 = r2_score(ytrain, ytrain_pred)

print("The model performance for training set")
print("-----")
print('MSE is {}'.format(mse))
print('RMSE is {}'.format(rmse))
print('R2 score is {}'.format(r2))
print("\n")

# model evaluation for testing set
ytest_pred = lr.predict(xtest)

```

```
mse = mean_squared_error(ytest, ytest_pred)
rmse = (np.sqrt(mean_squared_error(ytest, ytest_pred)))
r2 = r2_score(ytest, ytest_pred)
```

```
print("The model performance for testing set")
print("-----")
print('MSE is {}'.format(mse))
print('RMSE is {}'.format(rmse))
print('R2 score is {}'.format(r2))
```

The model performance for training set

```
MSE is 0.576867540006079
RMSE is 0.7595179655584712
R2 score is 0.33581211668682887
```

The model performance for testing set

```
MSE is 0.5792230514362919
RMSE is 0.7610670479243546
R2 score is 0.3513086902731888
```

random forest regression

```
from sklearn.ensemble import RandomForestRegressor
rf = RandomForestRegressor()
rf.fit(xtrain, ytrain)
```

model evaluation for training set

```
ytrain_pred = rf.predict(xtrain)
mse = mean_squared_error(ytrain, ytrain_pred)
rmse = (np.sqrt(mean_squared_error(ytrain, ytrain_pred)))
r2 = r2_score(ytrain, ytrain_pred)
```

```
print("The model performance for training set")
print("-----")
print('MSE is {}'.format(mse))
print('RMSE is {}'.format(rmse))
print('R2 score is {}'.format(r2))
print("\n")
```

model evaluation for testing set

```
ytest_pred = rf.predict(xtest)
mse = mean_squared_error(ytest, ytest_pred)
rmse = (np.sqrt(mean_squared_error(ytest, ytest_pred)))
r2 = r2_score(ytest, ytest_pred)
```



```

print("The model performance for testing set")
print("-----")
print('MSE is {}'.format(mse))
print('RMSE is {}'.format(rmse))
print('R2 score is {}'.format(r2))

```

```

The model performance for training set
-----
MSE is 0.0006193541437514021
RMSE is 0.02488682671116191
R2 score is 0.9992868943228195

```

```

The model performance for testing set
-----
MSE is 0.0036496873709222374
RMSE is 0.06041264247591093
R2 score is 0.9959125927829249

```

decision tree regression

```

from sklearn.tree import DecisionTreeRegressor
dtr=DecisionTreeRegressor(random_state=0)
dtr.fit(xtrain,ytrain)

#predicting the value

ytest_pred=dtr.predict(xtest)

#evaluating the model

from sklearn.metrics import r2_score,mean_squared_error
print("Results for Decision Tree Regression:\n1)Mean Square Error={}\n2)R-Square Score={}").format(mean_squared_error(ytest,ytest_pred),r2_score(ytest,ytest_pred))

```

```

Results for Decision Tree Regression:
1)Mean Square Error=0.012478922917737525
2)R-Square Score=0.9860244359553473

```

svm regression

```

from sklearn.svm import SVR
svr=SVR()
svr.fit(xtrain,ytrain)

```

```

#predicting the value
ytest_pred=svr.predict(xtest)

#evaluating the model

from sklearn.metrics import r2_score,mean_squared_error
print("Results for SVM Regression:\n1)Mean Square Error={} \n2)R-Square Score={}".format(mean_squared_error(ytest,ytest_pred),r2_score(ytest,ytest_pred)))

```

Results for SVM Regression:
1)Mean Square Error=0.902874531063811
2)R-Square Score=-0.011159449925972176

conclusion

Random Forest Regression works well on both train and test sets with r2 score of 0.99. As well as Decision Tree Regression also works well on both train and test set with r2 score of 0.98.

```

np.random.seed(range(0,100))
print("Welcome to Mental Fitness Tracker!\nFill the detail to check your mental fitness!")
country=l.fit_transform([input('Enter Your country Name:')])
year=int(input("Enter the Year:"))
schi=(float(input("Enter your Schizophrenia rate in % (it not enter 0):")))*100
bipo_dis=(float(input("Enter your Bipolar disorder rate in % (it not enter 0):")))*100
eat_dis=(float(input("Enter your Eating disorder rate in % (it not enter 0):")))*100
anx=(float(input("Enter your Anxiety rate in % (it not enter 0):")))*100
drug_use=(float(input("Enter your Drug Usage rate in per year % (it not enter 0):")))*100
depr=(float(input("Enter your Depression rate in % (it not enter 0):")))*100
alch=(float(input("Enter your Alcohol Consuming rate per year in % (it not enter 0):")))*100

prediction=rf.predict([[country,year,schi,bipo_dis,eat_dis,anx,drug_use,depr,alch]])
print("Your Mental Fitness is {}".format(prediction))
print("Bye...!")

```

```
Welcome to Mental Fitness Tracker!
Fill the detail to check your mental fitness!
Enter Your country Name:pakistan
Enter the Year:87
Enter your Schizophrenia rate in % (it not enter 0):98
Enter your Bipolar disorder rate in % (it not enter 0):78
Enter your Eating disorder rate in % (it not enter 0):76
Enter your Anxiety rate in % (it not enter 0):99
Enter your Drug Usage rate in per year % (it not enter 0):78
Enter your Depression rate in % (it not enter 0):85
Enter your Alcohol Consuming rate per year in % (it not enter 0):67
Your Mental Fitness is [4.02270229]%
Bye...!
```