!pip install tensorflow

```
Requirement already satisfied: tensorflow in /usr/local/lib/python3.7/dist-packages (2.7
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Requirement already satisfied: wheel<1.0,>=0.32.0 in /usr/local/lib/python3.7/dist-pack@
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Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.7/dist-packa
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Requirement already satisfied: chardet<4,>=3.0.2 in /usr/local/lib/python3.7/dist-packas
Requirement already satisfied: oauthlib>=3.0.0 in /usr/local/lib/python3.7/dist-packages
```

```
from google.colab import drive
```

```
drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mour

```
import pandas as pd
import numpy as np
from sklearn.feature_extraction.text import CountVectorizer, TfidfVectorizer
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import GridSearchCV
from sklearn.metrics import classification_report
from sklearn.metrics import accuracy_score, confusion_matrix
from sklearn.model_selection import train_test_split, cross_val_score
```

forest=pd.read_csv('/content/drive/MyDrive/Colab Notebooks/forestfires.csv')
forest

month	day	FFMC	DMC	DC	ISI	temp	RH	wind	rain	area	dayfri	daymon	da
mar	fri	86.2	26.2	94.3	5.1	8.2	51	6.7	0.0	0.00	1	0	
oct	tue	90.6	35.4	669.1	6.7	18.0	33	0.9	0.0	0.00	0	0	
oct	sat	90.6	43.7	686.9	6.7	14.6	33	1.3	0.0	0.00	0	0	
mar	fri	91.7	33.3	77.5	9.0	8.3	97	4.0	0.2	0.00	1	0	
mar	sun	89.3	51.3	102.2	9.6	11.4	99	1.8	0.0	0.00	0	0	
aug	sun	81.6	56.7	665.6	1.9	27.8	32	2.7	0.0	6.44	0	0	
aug	sun	81.6	56.7	665.6	1.9	21.9	71	5.8	0.0	54.29	0	0	
aug	sun	81.6	56.7	665.6	1.9	21.2	70	6.7	0.0	11.16	0	0	
aug	sat	94.4	146.0	614.7	11.3	25.6	42	4.0	0.0	0.00	0	0	
nov	tue	79.5	3.0	106.7	1.1	11.8	31	4.5	0.0	0.00	0	0	
	mar oct oct mar aug aug aug	mar fri oct tue oct sat mar fri mar sun aug sun aug sun aug sun aug sat	mar fri 86.2 oct tue 90.6 oct sat 90.6 mar fri 91.7 mar sun 89.3 aug sun 81.6 aug sun 81.6 aug sun 81.6 aug sat 94.4	mar fri 86.2 26.2 oct tue 90.6 35.4 oct sat 90.6 43.7 mar fri 91.7 33.3 mar sun 89.3 51.3 aug sun 81.6 56.7	mar fri 86.2 26.2 94.3 oct tue 90.6 35.4 669.1 oct sat 90.6 43.7 686.9 mar fri 91.7 33.3 77.5 mar sun 89.3 51.3 102.2 aug sun 81.6 56.7 665.6 aug sun 81.6 56.7 665.6 aug sat 94.4 146.0 614.7	mar fri 86.2 26.2 94.3 5.1 oct tue 90.6 35.4 669.1 6.7 oct sat 90.6 43.7 686.9 6.7 mar fri 91.7 33.3 77.5 9.0 mar sun 89.3 51.3 102.2 9.6 aug sun 81.6 56.7 665.6 1.9 aug sun 81.6 56.7 665.6 1.9 aug sat 94.4 146.0 614.7 11.3	mar fri 86.2 26.2 94.3 5.1 8.2 oct tue 90.6 35.4 669.1 6.7 18.0 oct sat 90.6 43.7 686.9 6.7 14.6 mar fri 91.7 33.3 77.5 9.0 8.3 mar sun 89.3 51.3 102.2 9.6 11.4 aug sun 81.6 56.7 665.6 1.9 27.8 aug sun 81.6 56.7 665.6 1.9 21.2 aug sat 94.4 146.0 614.7 11.3 25.6	mar fri 86.2 26.2 94.3 5.1 8.2 51 oct tue 90.6 35.4 669.1 6.7 18.0 33 oct sat 90.6 43.7 686.9 6.7 14.6 33 mar fri 91.7 33.3 77.5 9.0 8.3 97 mar sun 89.3 51.3 102.2 9.6 11.4 99 aug sun 81.6 56.7 665.6 1.9 27.8 32 aug sun 81.6 56.7 665.6 1.9 21.9 71 aug sun 81.6 56.7 665.6 1.9 21.2 70 aug sat 94.4 146.0 614.7 11.3 25.6 42	mar fri 86.2 26.2 94.3 5.1 8.2 51 6.7 oct tue 90.6 35.4 669.1 6.7 18.0 33 0.9 oct sat 90.6 43.7 686.9 6.7 14.6 33 1.3 mar fri 91.7 33.3 77.5 9.0 8.3 97 4.0 mar sun 89.3 51.3 102.2 9.6 11.4 99 1.8 aug sun 81.6 56.7 665.6 1.9 27.8 32 2.7 aug sun 81.6 56.7 665.6 1.9 21.2 70 6.7 aug sat 94.4 146.0 614.7 11.3 25.6 42 4.0	mar fri 86.2 26.2 94.3 5.1 8.2 51 6.7 0.0 oct tue 90.6 35.4 669.1 6.7 18.0 33 0.9 0.0 oct sat 90.6 43.7 686.9 6.7 14.6 33 1.3 0.0 mar fri 91.7 33.3 77.5 9.0 8.3 97 4.0 0.2 mar sun 89.3 51.3 102.2 9.6 11.4 99 1.8 0.0 .	mar fri 86.2 26.2 94.3 5.1 8.2 51 6.7 0.0 0.00 oct tue 90.6 35.4 669.1 6.7 18.0 33 0.9 0.0 0.00 oct sat 90.6 43.7 686.9 6.7 14.6 33 1.3 0.0 0.00 mar fri 91.7 33.3 77.5 9.0 8.3 97 4.0 0.2 0.00 mar sun 89.3 51.3 102.2 9.6 11.4 99 1.8 0.0 0.00	mar fri 86.2 26.2 94.3 5.1 8.2 51 6.7 0.0 0.00 1 oct tue 90.6 35.4 669.1 6.7 18.0 33 0.9 0.0 0.00 0 oct sat 90.6 43.7 686.9 6.7 14.6 33 1.3 0.0 0.00 0 mar fri 91.7 33.3 77.5 9.0 8.3 97 4.0 0.2 0.00 1 mar sun 89.3 51.3 102.2 9.6 11.4 99 1.8 0.0 0.00 0 aug sun 81.6 56.7 665.6 1.9 27.8 32 2.7 0.0 6.44 0 aug sun 81.6 56.7 665.6 1.9	mar fri 86.2 26.2 94.3 5.1 8.2 51 6.7 0.0 0.00 1 0 oct tue 90.6 35.4 669.1 6.7 18.0 33 0.9 0.0 0.00 0 0 oct sat 90.6 43.7 686.9 6.7 14.6 33 1.3 0.0 0.00 0 0 mar fri 91.7 33.3 77.5 9.0 8.3 97 4.0 0.2 0.00 1 0 mar sun 89.3 51.3 102.2 9.6 11.4 99 1.8 0.0 0.00 0 0

517 rows × 31 columns

```
forest.shape
```

(517, 31)

forest.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 517 entries, 0 to 516
Data columns (total 31 columns):

#	Column	Non-Null Count	Dtype
0	month	517 non-null	object
1	dav	517 non-null	object

2	FFMC	517	non-null	float64
3	DMC	517	non-null	float64
4	DC	517	non-null	float64
5	ISI	517	non-null	float64
6	temp	517	non-null	float64
7	RH	517	non-null	int64
8	wind	517	non-null	float64
9	rain	517	non-null	float64
10	area	517	non-null	float64
11	dayfri	517	non-null	int64
12	daymon	517	non-null	int64
13	daysat	517	non-null	int64
14	daysun	517	non-null	int64
15	daythu	517	non-null	int64
16	daytue	517	non-null	int64
17	daywed	517	non-null	int64
18	monthapr	517	non-null	int64
19	monthaug	517	non-null	int64
20	monthdec	517	non-null	int64
21	monthfeb	517	non-null	int64
22	monthjan	517	non-null	int64
23	monthjul	517	non-null	int64
24	monthjun	517	non-null	int64
25	monthmar	517	non-null	int64
26	monthmay	517	non-null	int64
27	monthnov	517	non-null	int64
28	monthoct	517	non-null	int64
29	monthsep	517	non-null	int64
30	size_category	517		object
dtype	es: float64(8),	inte	54(20), objed	ct(3)
	125 2.	. IZD		

memory usage: 125.3+ KB

forest.describe()

₽		FFMC	DMC	DC	ISI	temp	RH	wind
	count	517.000000	517.000000	517.000000	517.000000	517.000000	517.000000	517.000000
	mean	90.644681	110.872340	547.940039	9.021663	18.889168	44.288201	4.017602
	std	5.520111	64.046482	248.066192	4.559477	5.806625	16.317469	1.791653
	min	18.700000	1.100000	7.900000	0.000000	2.200000	15.000000	0.400000
	25%	90.200000	68.600000	437.700000	6.500000	15.500000	33.000000	2.700000
	50%	91.600000	108.300000	664.200000	8.400000	19.300000	42.000000	4.000000
	75%	92.900000	142.400000	713.900000	10.800000	22.800000	53.000000	4.900000
	max	96.200000	291.300000	860.600000	56.100000	33.300000	100.000000	9.400000

forest.isna().sum()

6 PM	
month	0
day	0
FFMC	0
DMC	0
DC	0
ISI	0
temp	0
RH	0
wind	0
rain	0
area	0
dayfri	0
daymon	0
daysat	0
daysun	0
daythu	0
daytue	0
daywed	0
monthapr	0
monthaug	0
monthdec	0
monthfeb	0
monthjan	0
monthjul	0
monthjun	0
monthmar	0
monthmay	0
monthnov	0
monthoct	0
monthsep	0
size_category	0
dtype: int64	

forest2=forest.drop(['dayfri','daymon','daysat','daysun','daythu','daytue','daywed','monthapr
forest2

ize_category	area	rain	wind	RH	temp	ISI	DC	DMC	FFMC	day	month	
small	0.00	0.0	6.7	51	8.2	5.1	94.3	26.2	86.2	fri	mar	0
small	0.00	0.0	0.9	33	18.0	6.7	669.1	35.4	90.6	tue	oct	1

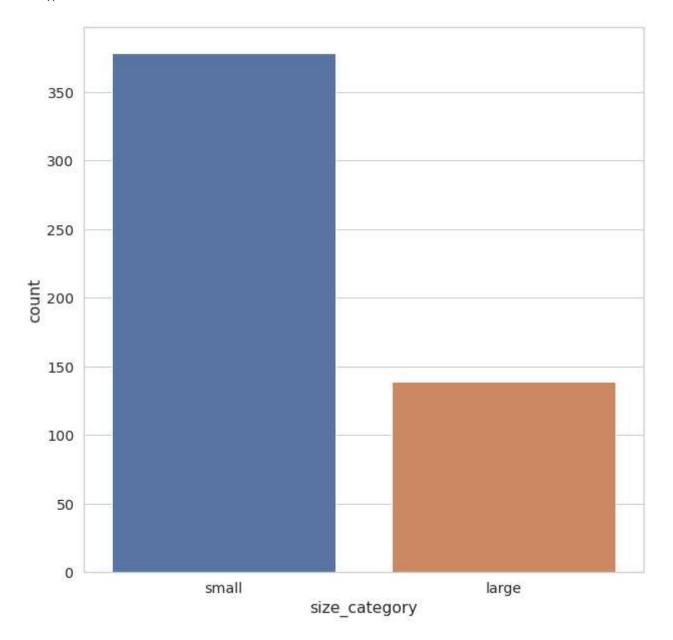
forest.size_category.value_counts()

small 378
large 139

Name: size_category, dtype: int64

import seaborn as sns
import matplotlib.pyplot as plt

sns.countplot(x='size_category', data=forest)
plt.show()



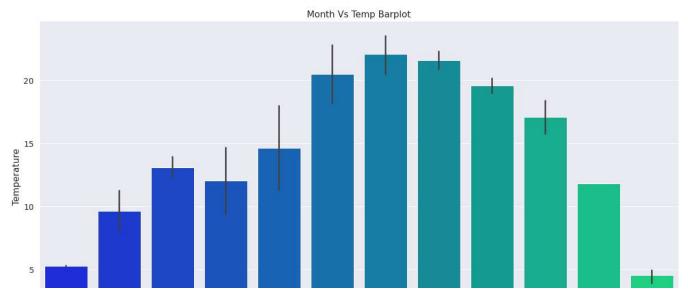
checking for which value of area is categorised into large and small by creating crosstab b pd.crosstab(forest.area, forest.size_category)

size_category large small

area		
0.00	0	247
0.09	0	1
0.17	0	1
0.21	0	1
0.24	0	1
200.94	1	0
212.88	1	0
278.53	1	0
746.28	1	0
1090.84	1	0

251 rows × 2 columns

```
plt.rcParams['figure.figsize']=[20,10]
sns.set(style="darkgrid", font_scale=1.3)
month_temp=sns.barplot(x='month', y='temp', data=forest, order =['jan','feb','mar','apr','may
month_temp.set(title='Month Vs Temp Barplot', xlabel='Month',ylabel='Temperature');
```



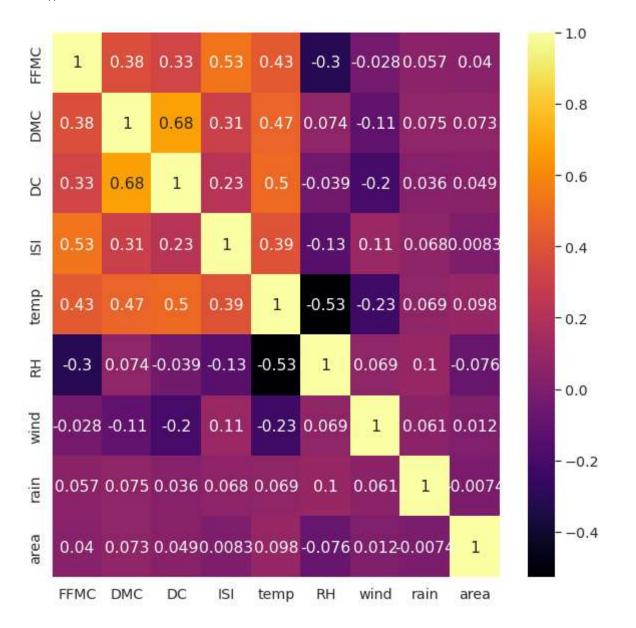
plt.rcParams['figure.figsize']=[10,10]
sns.set(style='whitegrid',font_scale=1.3)
day=sns.countplot(forest['day'], order=['sun','mon','tue','wed','thu','fri','sat'], palette='
day.set(title='countplot for the weekdays', xlabel ='days',ylabel='count');

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the FutureWarning

countplot for the weekdays



sns.heatmap(forest2.corr(), annot = True,cmap="inferno")
plt.show()



encoding month and day features

forest2.month.replace(('jan','feb','mar','apr','may','jun','jul','aug','sep','oct','nov','dec
forest2.day.replace(('mon','tue','wed','thu','fri','sat','sun'),(1,2,3,4,5,6,7), inplace=True
forest2.head(10)

	month	day	FFMC	DMC	DC	ISI	temp	RH	wind	rain	area	size_category
0	3	5	86.2	26.2	94.3	5.1	8.2	51	6.7	0.0	0.0	small
1	10	2	90.6	35.4	669.1	6.7	18.0	33	0.9	0.0	0.0	small
2	10	6	90.6	43.7	686.9	6.7	14.6	33	1.3	0.0	0.0	small
3	3	5	91.7	33.3	77.5	9.0	8.3	97	4.0	0.2	0.0	small
4	3	7	89.3	51.3	102.2	9.6	11.4	99	1.8	0.0	0.0	small
5	8	7	92.3	85.3	488.0	14.7	22.2	29	5.4	0.0	0.0	small
6	8	1	92.3	88.9	495.6	8.5	24.1	27	3.1	0.0	0.0	small
7	8	1	91.5	145.4	608.2	10.7	8.0	86	2.2	0.0	0.0	small

encoding target variable 'size category'
forest2.size_category.replace(('small','large'),(0,1), inplace=True)
forest2

	month	day	FFMC	DMC	DC	ISI	temp	RH	wind	rain	area	size_category
0	3	5	86.2	26.2	94.3	5.1	8.2	51	6.7	0.0	0.00	0
1	10	2	90.6	35.4	669.1	6.7	18.0	33	0.9	0.0	0.00	0
2	10	6	90.6	43.7	686.9	6.7	14.6	33	1.3	0.0	0.00	0
3	3	5	91.7	33.3	77.5	9.0	8.3	97	4.0	0.2	0.00	0
4	3	7	89.3	51.3	102.2	9.6	11.4	99	1.8	0.0	0.00	0
	•••		•••						•••			
512	8	7	81.6	56.7	665.6	1.9	27.8	32	2.7	0.0	6.44	1
513	8	7	81.6	56.7	665.6	1.9	21.9	71	5.8	0.0	54.29	1
514	8	7	81.6	56.7	665.6	1.9	21.2	70	6.7	0.0	11.16	1
515	8	6	94.4	146.0	614.7	11.3	25.6	42	4.0	0.0	0.00	0
516	11	2	79.5	3.0	106.7	1.1	11.8	31	4.5	0.0	0.00	0

517 rows × 12 columns

```
X = forest2.drop('size_category', axis=1)
y= forest2['size_category']
scaler=StandardScaler()
```

scaled_X = scaler.fit(X)
scaled_X

StandardScaler()

```
scaled_features=scaler.transform(X)
forest_head=pd.DataFrame(scaled_features, columns=forest2.columns[:-1])
forest_head
```

	month	day	FFMC	DMC	DC	ISI	temp	RH	
0	-1.968443	0.357721	-0.805959	-1.323326	-1.830477	-0.860946	-1.842640	0.411724	
1	1.110120	-1.090909	-0.008102	-1.179541	0.488891	-0.509688	-0.153278	-0.692456	
2	1.110120	0.840597	-0.008102	-1.049822	0.560715	-0.509688	-0.739383	-0.692456	
3	-1.968443	0.357721	0.191362	-1.212361	-1.898266	-0.004756	-1.825402	3.233519	
4	-1.968443	1.323474	-0.243833	-0.931043	-1.798600	0.126966	-1.291012	3.356206	
512	0.230531	1.323474	-1.640083	-0.846648	0.474768	-1.563460	1.536084	-0.753800	
513	0.230531	1.323474	-1.640083	-0.846648	0.474768	-1.563460	0.519019	1.638592	
514	0.230531	1.323474	-1.640083	-0.846648	0.474768	-1.563460	0.398350	1.577248	
515	0.230531	0.840597	0.680957	0.549003	0.269382	0.500176	1.156839	-0.140366	
516	1.549915	-1.090909	-2.020879	-1.685913	-1.780442	-1.739089	-1.222058	-0.815143	

517 rows × 11 columns

```
X_train,X_test,y_train,y_test = train_test_split(forest_head,y,test_size=0.20, random_state=1
```

model.add(Dense(units=1, activation= 'softmax'))

```
model.compile(optimizer='adam', loss = 'binary_crossentropy', metrics=['accuracy'])
```

```
model.fit(X_train, y_train, epochs= 10, batch_size=10)
   Epoch 1/10
   Epoch 2/10
   42/42 [============== ] - 0s 1ms/step - loss: 0.0115 - accuracy: 0.2518
   Epoch 3/10
   42/42 [============= ] - 0s 2ms/step - loss: 0.0105 - accuracy: 0.2518
   Epoch 4/10
   42/42 [============ ] - 0s 2ms/step - loss: 0.0108 - accuracy: 0.2518
   Epoch 5/10
   42/42 [============= ] - 0s 2ms/step - loss: 0.0100 - accuracy: 0.2518
   Epoch 6/10
   42/42 [============ ] - 0s 2ms/step - loss: 0.0104 - accuracy: 0.2518
   Epoch 7/10
   42/42 [============= ] - 0s 2ms/step - loss: 0.0095 - accuracy: 0.2518
   Epoch 8/10
   42/42 [============= ] - 0s 2ms/step - loss: 0.0094 - accuracy: 0.2518
   Epoch 9/10
   Epoch 10/10
   42/42 [============ ] - 0s 2ms/step - loss: 0.0089 - accuracy: 0.2518
   <keras.callbacks.History at 0x7f24a83afb10>
scores=model.evaluate(X test,y test)
scores
   4/4 [=========== ] - 0s 2ms/step - loss: 0.2233 - accuracy: 0.3365
   [0.2232731282711029, 0.3365384638309479]
print('Model Accuracy : ', round(scores[1],2))
                : ', round(scores[0],2))
print('Loss
   Model Accuracy: 0.34
```

: 0.22

Loss

✓ 2s completed at 7:25 PM