

```
In [3]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
import seaborn as sns
from sklearn.preprocessing import OrdinalEncoder
from tensorflow.keras.models import Sequential
from keras.layers import Dense
import tensorflow as tf
from sklearn.model_selection import train_test_split

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

```
In [7]: data = pd.read_csv("forestfires (1).csv")
data
```

Out[7]:

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

517 rows × 31 columns



```
In [8]: data.shape
```

Out[8]: (517, 31)

```
In [9]: data.dtypes
```

```
Out[9]: month          object
        day            object
        FPMC           float64
        DMC            float64
        DC             float64
        ISI            float64
        temp           float64
        RH             int64
        wind           float64
        rain           float64
        area           float64
        dayfri         int64
        daymon         int64
        daysat         int64
        daysun         int64
        daythu         int64
        daytue         int64
        daywed         int64
        monthapr       int64
        monthaug       int64
        monthdec       int64
        monthfeb       int64
        monthjan       int64
        monthjul       int64
        monthjun       int64
        monthmar       int64
        monthmay       int64
        monthnov       int64
        monthoct       int64
        monthsep       int64
        size_category  object
dtype: object
```

```
In [10]: data.isna().sum()
```

```
Out[10]: month                0  
         day                  0  
         FFMCI                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
```

In [16]: `data.describe(include = 'all')`

Out[16]:

	month	day	FFMC	DMC	DC	ISI	temp	RH
<b>count</b>	517	517	517.000000	517.000000	517.000000	517.000000	517.000000	517.000000
<b>unique</b>	12	7	NaN	NaN	NaN	NaN	NaN	NaN
<b>top</b>	aug	sun	NaN	NaN	NaN	NaN	NaN	NaN
<b>freq</b>	184	95	NaN	NaN	NaN	NaN	NaN	NaN
<b>mean</b>	NaN	NaN	90.644681	110.872340	547.940039	9.021663	18.889168	44.288201
<b>std</b>	NaN	NaN	5.520111	64.046482	248.066192	4.559477	5.806625	16.317469
<b>min</b>	NaN	NaN	18.700000	1.100000	7.900000	0.000000	2.200000	15.000000
<b>25%</b>	NaN	NaN	90.200000	68.600000	437.700000	6.500000	15.500000	33.000000
<b>50%</b>	NaN	NaN	91.600000	108.300000	664.200000	8.400000	19.300000	42.000000
<b>75%</b>	NaN	NaN	92.900000	142.400000	713.900000	10.800000	22.800000	53.000000
<b>max</b>	NaN	NaN	96.200000	291.300000	860.600000	56.100000	33.300000	100.000000

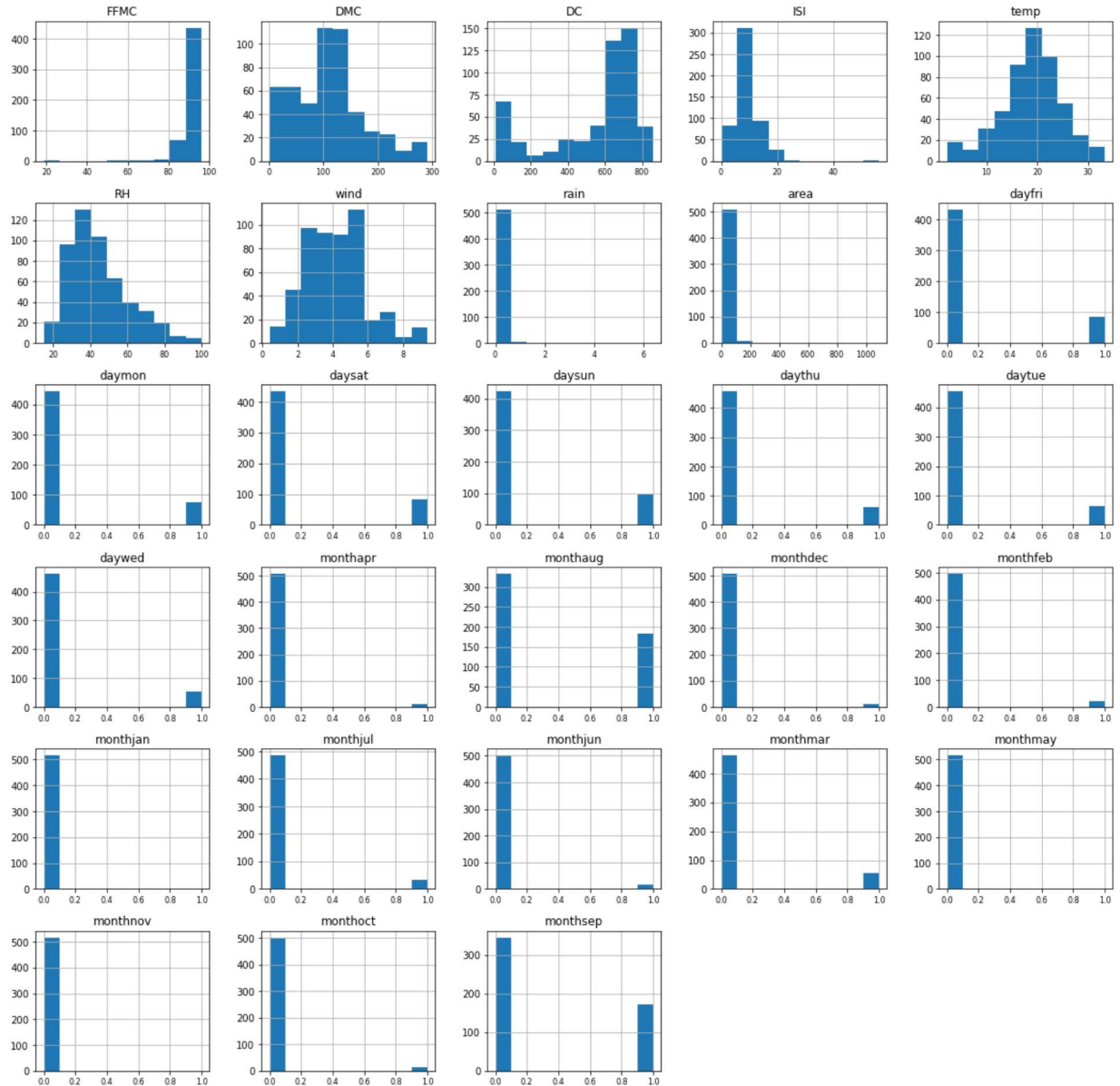
11 rows × 31 columns

In [11]: `data.describe(include='object')`

Out[11]:

	month	day	size_category
<b>count</b>	517	517	517
<b>unique</b>	12	7	2
<b>top</b>	aug	sun	small
<b>freq</b>	184	95	378

```
In [18]: data.hist(figsize = (20,20),xlabelsize = 8 , ylabelsize=10)
plt.show()
```



```
In [19]: data = data.drop(['month','day'],axis =1 )
```

```
In [20]: mapping = {'small':0 , 'large' :1}
```

```
In [21]: data = data.replace(mapping)
```

```
In [22]: data
```

Out[22]:

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

517 rows × 29 columns



```
In [23]: x = np.array(data.iloc[:,0:28])
y = np.array(data.iloc[:,28])
```

```
In [26]: def norm_func(i):
x = (i-i.min())/(i.max()-i.min())
return(x)
```

```
In [27]: x_norm = norm_func(x)
```

```
In [31]: x_train,x_test,y_train,y_test= train_test_split(x_norm,y, test_size=0.2,stratify = y)
```



```
In [32]: model = Sequential()
model.add(Dense(8, input_dim=28, activation='linear'))
model.add(Dense(4, activation='tanh'))
model.add(Dense(1, activation='sigmoid'))
```

```
In [33]: model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])
```

```
In [34]: history = model.fit(x_train,y_train,validation_split=0.3,epochs = 120 ,batch_size=
```

```
Epoch 2/120
29/29 [=====] - 0s 4ms/step - loss: 0.6674 - accuracy: 0.7336 - val_loss: 0.6514 - val_accuracy: 0.7258
Epoch 3/120
29/29 [=====] - 0s 4ms/step - loss: 0.6346 - accuracy: 0.7336 - val_loss: 0.6241 - val_accuracy: 0.7258
Epoch 4/120
29/29 [=====] - 0s 5ms/step - loss: 0.6099 - accuracy: 0.7336 - val_loss: 0.6067 - val_accuracy: 0.7258
Epoch 5/120
29/29 [=====] - 0s 5ms/step - loss: 0.5946 - accuracy: 0.7336 - val_loss: 0.5977 - val_accuracy: 0.7258
Epoch 6/120
29/29 [=====] - 0s 5ms/step - loss: 0.5857 - accuracy: 0.7336 - val_loss: 0.5940 - val_accuracy: 0.7258
Epoch 7/120
29/29 [=====] - 0s 5ms/step - loss: 0.5817 - accuracy: 0.7336 - val_loss: 0.5927 - val_accuracy: 0.7258
Epoch 8/120
29/29 [=====] - 0s 5ms/step - loss: 0.5817 - accuracy: 0.7336 - val_loss: 0.5927 - val_accuracy: 0.7258
```

```
In [36]: scores = model.evaluate(x_train,y_train)
print("%s: %.2f%%" % (model.metrics_names[1], scores[1]*100))
```

```
13/13 [=====] - 0s 2ms/step - loss: 0.2388 - accuracy: 0.8910
accuracy: 89.10%
```

```
In [37]: scores = model.evaluate(x_test, y_test)
print("%s: %.2f%%" % (model.metrics_names[1], scores[1]*100))
```

```
4/4 [=====] - 0s 3ms/step - loss: 0.2422 - accuracy: 0.9038
accuracy: 90.38%
```

```
In [ ]:
```