

In [1]:

```
1 import pandas as pd
2 import numpy as np
3 import matplotlib.pyplot as plt
4 import seaborn as sns
5
6 # Make numpy printouts easier to read.
7 np.set_printoptions(precision=3, suppress=True)
```

In [2]:

```
1 import tensorflow as tf
2
3 from tensorflow import keras
4 from tensorflow.keras import layers
5 from tensorflow.keras.layers.experimental import preprocessing
6
7 print(tf.__version__)
```

2.11.0

## The Auto MPG dataset

In [3]:

```
1 url = 'http://archive.ics.uci.edu/ml/machine-learning-databases/auto-mpg/auto-mpg.data'
2 column_names = ['MPG', 'Cylinders', 'Displacement', 'Horsepower', 'Weight',
3                 'Acceleration', 'Model Year', 'Origin']
4
5 raw_dataset = pd.read_csv(url, names=column_names,
6                           na_values='?', comment='\t',
7                           sep=' ', skipinitialspace=True)
```

In [5]:

```
1 carsdata = raw_dataset.copy()
2 carsdata.tail()
```

Out[5]:

	MPG	Cylinders	Displacement	Horsepower	Weight	Acceleration	Model Year	Origin
393	27.0	4	140.0	86.0	2790.0	15.6	82	1
394	44.0	4	97.0	52.0	2130.0	24.6	82	2
395	32.0	4	135.0	84.0	2295.0	11.6	82	1
396	28.0	4	120.0	79.0	2625.0	18.6	82	1
397	31.0	4	119.0	82.0	2720.0	19.4	82	1

## Clean the Data

In [7]:

```
1 carsdata.isna().sum()
```

Out[7]:

```
MPG          0
Cylinders     0
Displacement  0
Horsepower    6
Weight        0
Acceleration  0
Model Year    0
Origin        0
dtype: int64
```

In [10]:

```
1 carsdata = carsdata.replace("?", np.nan)
```

In [12]:

```
1 carsdata.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 398 entries, 0 to 397
Data columns (total 8 columns):
#   Column          Non-Null Count  Dtype  
---  -
0   MPG              398 non-null   float64
1   Cylinders         398 non-null   int64   
2   Displacement      398 non-null   float64
3   Horsepower        392 non-null   float64
4   Weight            398 non-null   float64
5   Acceleration      398 non-null   float64
6   Model Year        398 non-null   int64   
7   Origin            398 non-null   int64   
dtypes: float64(5), int64(3)
memory usage: 25.0 KB
```

In [14]:

```
1 carsdata = carsdata.fillna(carsdata.median())
```

In [16]:

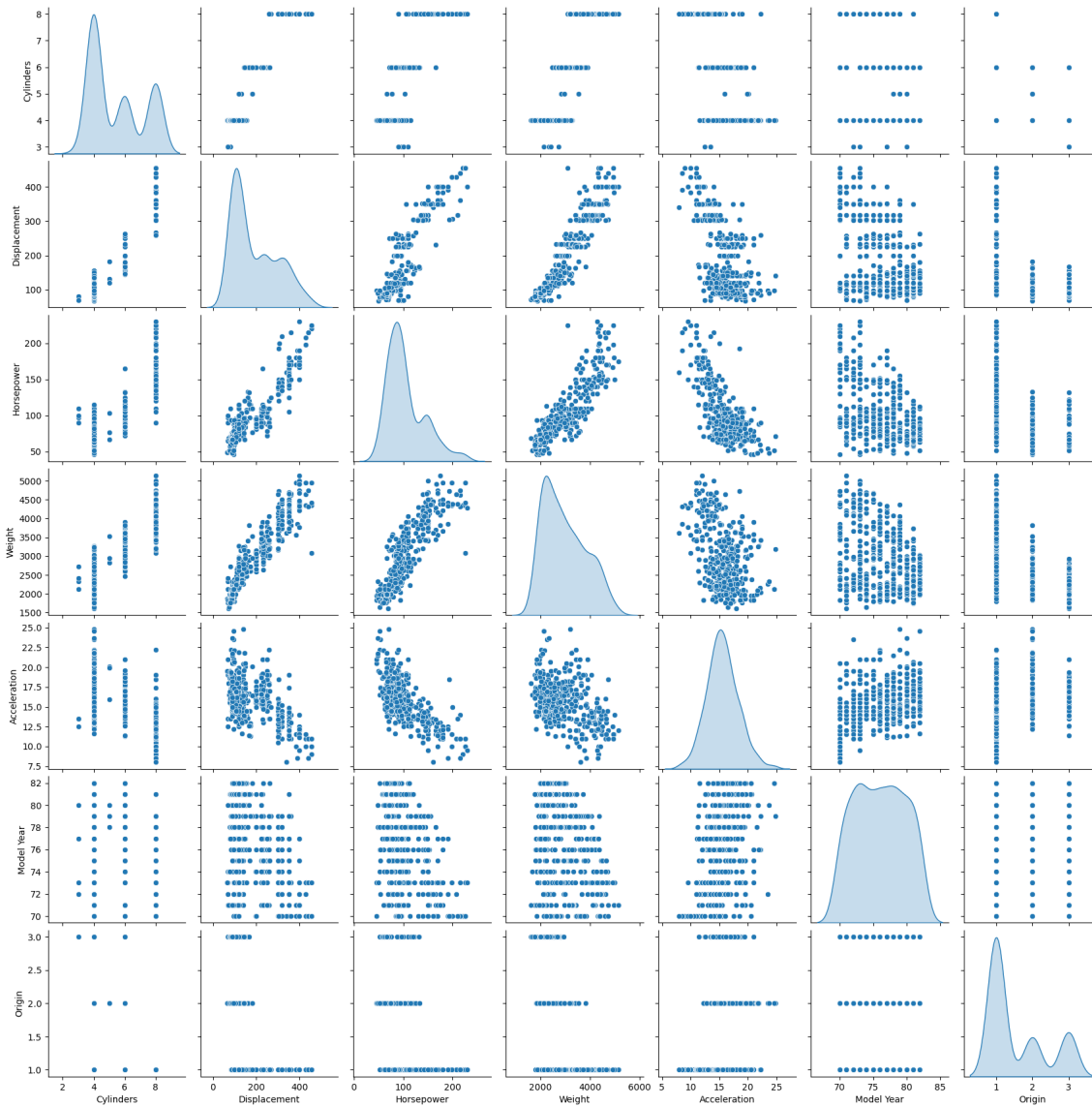
```
1 carsdata["Horsepower"] = carsdata["Horsepower"].astype('float64')
```

In [19]:

```
1 X = carsdata.drop(['MPG'], axis=1)
2 y = carsdata[['MPG']]
3
4 sns.pairplot(X, diag_kind= 'kde')
```

Out[19]:

<seaborn.axisgrid.PairGrid at 0x17c658efb50>



## Scaling

In [23]:

```
1 x = carsdata.iloc[:, :-1].values
2 y = carsdata.iloc[:, -1].values
```

In [25]:

```
1 from sklearn.preprocessing import StandardScaler
2 sc = StandardScaler()
3 scaler = sc.fit_transform(x)
```

## splitting

In [26]:

```
1 from sklearn.model_selection import train_test_split
2 x_train,x_test,y_train,y_test = train_test_split(scaler,y,random_state=10,test_size=
```

## ANN

In [33]:

```
1 from keras.models import Sequential
2 from keras.layers import Dense, Flatten, Dropout
```

In [34]:

```
1 carsANN = Sequential()
```

In [35]:

```
1 carsANN.add(Dense(units=8, activation= "relu"))
2 carsANN.add(Dense(units=3, activation= "relu"))
3 carsANN.add(Dense(units=1, activation= "sigmoid"))
```

In [36]:

```
1 from tensorflow.keras.optimizers.schedules import ExponentialDecay
2 from keras.optimizers import Adam
3
4 carsANN.compile(loss='binary_crossentropy',
5                 metrics=['accuracy'],
6                 optimizer='Adam'
7                 )
```

In [37]:

```
1 carsANN.fit(x_train,y_train, batch_size=40, epochs=50)
```

Epoch 1/50  
8/8 [=====] - 2s 7ms/step - loss: 0.5830 - accuracy: 0.2673  
Epoch 2/50  
8/8 [=====] - 0s 5ms/step - loss: 0.4828 - accuracy: 0.3176  
Epoch 3/50  
8/8 [=====] - 0s 3ms/step - loss: 0.3862 - accuracy: 0.3522  
Epoch 4/50  
8/8 [=====] - 0s 3ms/step - loss: 0.2855 - accuracy: 0.3742  
Epoch 5/50  
8/8 [=====] - 0s 4ms/step - loss: 0.1922 - accuracy: 0.3868  
Epoch 6/50  
8/8 [=====] - 0s 5ms/step - loss: 0.0983 - accuracy: 0.4151  
Epoch 7/50  
8/8 [=====] - 0s 4ms/step - loss: 0.0087 - accuracy: 0.4843  
Epoch 8/50  
8/8 [=====] - 0s 3ms/step - loss: -0.0787 - accuracy: 0.5472  
Epoch 9/50  
8/8 [=====] - 0s 5ms/step - loss: -0.1664 - accuracy: 0.6069  
Epoch 10/50  
8/8 [=====] - 0s 4ms/step - loss: -0.2501 - accuracy: 0.6258  
Epoch 11/50  
8/8 [=====] - 0s 4ms/step - loss: -0.3402 - accuracy: 0.6289  
Epoch 12/50  
8/8 [=====] - 0s 4ms/step - loss: -0.4393 - accuracy: 0.6289  
Epoch 13/50  
8/8 [=====] - 0s 5ms/step - loss: -0.5443 - accuracy: 0.6289  
Epoch 14/50  
8/8 [=====] - 0s 5ms/step - loss: -0.6489 - accuracy: 0.6289  
Epoch 15/50  
8/8 [=====] - 0s 4ms/step - loss: -0.7647 - accuracy: 0.6289  
Epoch 16/50  
8/8 [=====] - 0s 5ms/step - loss: -0.8836 - accuracy: 0.6289  
Epoch 17/50  
8/8 [=====] - 0s 5ms/step - loss: -1.0093 - accuracy: 0.6289  
Epoch 18/50  
8/8 [=====] - 0s 5ms/step - loss: -1.1331 - accuracy: 0.6289  
Epoch 19/50  
8/8 [=====] - 0s 4ms/step - loss: -1.2698 - accuracy: 0.6289  
Epoch 20/50  
8/8 [=====] - 0s 4ms/step - loss: -1.4019 - accuracy: 0.6289  
Epoch 21/50

8/8 [=====] - 0s 5ms/step - loss: -1.5453 - accur  
acy: 0.6289  
Epoch 22/50  
8/8 [=====] - 0s 5ms/step - loss: -1.7009 - accur  
acy: 0.6289  
Epoch 23/50  
8/8 [=====] - 0s 6ms/step - loss: -1.8512 - accur  
acy: 0.6289  
Epoch 24/50  
8/8 [=====] - 0s 6ms/step - loss: -2.0151 - accur  
acy: 0.6289  
Epoch 25/50  
8/8 [=====] - 0s 5ms/step - loss: -2.1933 - accur  
acy: 0.6289  
Epoch 26/50  
8/8 [=====] - 0s 7ms/step - loss: -2.3852 - accur  
acy: 0.6289  
Epoch 27/50  
8/8 [=====] - 0s 5ms/step - loss: -2.5835 - accur  
acy: 0.6289  
Epoch 28/50  
8/8 [=====] - 0s 7ms/step - loss: -2.8123 - accur  
acy: 0.6289  
Epoch 29/50  
8/8 [=====] - 0s 4ms/step - loss: -3.0373 - accur  
acy: 0.6289  
Epoch 30/50  
8/8 [=====] - 0s 5ms/step - loss: -3.2917 - accur  
acy: 0.6289  
Epoch 31/50  
8/8 [=====] - 0s 7ms/step - loss: -3.5575 - accur  
acy: 0.6289  
Epoch 32/50  
8/8 [=====] - 0s 5ms/step - loss: -3.8469 - accur  
acy: 0.6289  
Epoch 33/50  
8/8 [=====] - 0s 5ms/step - loss: -4.1449 - accur  
acy: 0.6289  
Epoch 34/50  
8/8 [=====] - 0s 5ms/step - loss: -4.4840 - accur  
acy: 0.6289  
Epoch 35/50  
8/8 [=====] - 0s 4ms/step - loss: -4.8299 - accur  
acy: 0.6289  
Epoch 36/50  
8/8 [=====] - 0s 5ms/step - loss: -5.1848 - accur  
acy: 0.6289  
Epoch 37/50  
8/8 [=====] - 0s 5ms/step - loss: -5.6059 - accur  
acy: 0.6289  
Epoch 38/50  
8/8 [=====] - 0s 4ms/step - loss: -6.0411 - accur  
acy: 0.6289  
Epoch 39/50  
8/8 [=====] - 0s 7ms/step - loss: -6.4765 - accur  
acy: 0.6289  
Epoch 40/50  
8/8 [=====] - 0s 6ms/step - loss: -6.9852 - accur  
acy: 0.6289  
Epoch 41/50  
8/8 [=====] - 0s 5ms/step - loss: -7.4907 - accur

```
acy: 0.6289
Epoch 42/50
8/8 [=====] - 0s 5ms/step - loss: -8.0696 - accur
acy: 0.6289
Epoch 43/50
8/8 [=====] - 0s 4ms/step - loss: -8.6585 - accur
acy: 0.6289
Epoch 44/50
8/8 [=====] - 0s 5ms/step - loss: -9.3187 - accur
acy: 0.6289
Epoch 45/50
8/8 [=====] - 0s 5ms/step - loss: -9.9816 - accur
acy: 0.6289
Epoch 46/50
8/8 [=====] - 0s 5ms/step - loss: -10.7609 - accur
acy: 0.6289
Epoch 47/50
8/8 [=====] - 0s 4ms/step - loss: -11.5090 - accur
acy: 0.6289
Epoch 48/50
8/8 [=====] - 0s 3ms/step - loss: -12.2879 - accur
acy: 0.6289
Epoch 49/50
8/8 [=====] - 0s 5ms/step - loss: -13.1968 - accur
acy: 0.6289
Epoch 50/50
8/8 [=====] - 0s 5ms/step - loss: -14.0783 - accur
acy: 0.6289
```

Out[37]:

```
<keras.callbacks.History at 0x17c6931ef70>
```



In [38]:

```
1 pred= carsANN.predict(x_test)
2 print(pred)
```

3/3 [=====] - 0s 2ms/step

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```
[1.   ]
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[0.989]
[1.   ]
[0.997]
[0.999]
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[0.997]
[1.   ]
[1.   ]
[0.998]
[0.995]
[0.996]
[0.996]]
```

In [42]:

```
1 carsANN.evaluate(x_test, y_test)
```

```
3/3 [=====] - 0s 0s/step - loss: -14.8295 - accur
acy: 0.6125
```

Out[42]:

```
[-14.829455375671387, 0.612500011920929]
```

In [ ]:

```
1
```