


```
import numpy as np
import pandas as pd

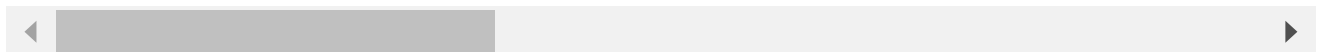
import matplotlib.pyplot as plt
from tensorflow.keras.utils import to_categorical
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from tensorflow.keras.layers import Dense, Dropout
from tensorflow.keras.models import Sequential
```

```
main_df = pd.read_csv('CarData.csv')
main_df
```



	car_ID	symboling	CarName	fueltype	aspiration	doornumber	carbody	driv
<b>0</b>	1	3	alfa-romero giulia	gas	std	two	convertible	
<b>1</b>	2	3	alfa-romero stelvio	gas	std	two	convertible	
<b>2</b>	3	1	alfa-romero Quadrifoglio	gas	std	two	hatchback	
<b>3</b>	4	2	audi 100 ls	gas	std	four	sedan	
<b>4</b>	5	2	audi 100ls	gas	std	four	sedan	
...	...	...	...	...	...	...	...	...
<b>200</b>	201	-1	volvo 145e (sw)	gas	std	four	sedan	
<b>201</b>	202	-1	volvo 144ea	gas	turbo	four	sedan	
<b>202</b>	203	-1	volvo 244dl	gas	std	four	sedan	
<b>203</b>	204	-1	volvo 246	diesel	turbo	four	sedan	
<b>204</b>	205	-1	volvo 264gl	gas	turbo	four	sedan	

205 rows × 26 columns



```
main_df.drop(columns=['CarName'], inplace=True)
main_df
```

	car_ID	symboling	fueltype	aspiration	doornumber	carbody	drivewheel	eng
0	1	3	gas	std	two	convertible	rwd	
1	2	3	gas	std	two	convertible	rwd	
2	3	1	gas	std	two	hatchback	rwd	
3	4	2	gas	std	four	sedan	fwd	
4	5	2	gas	std	four	sedan	4wd	
...	...	...	...	...	...	...	...	...
200	201	-1	gas	std	four	sedan	rwd	
201	202	-1	gas	turbo	four	sedan	rwd	
202	203	1	gas	std	four	sedan	rwd	

```
#convert columns using one-hot encoding
```

```
encoding_columns = ['symboling', 'fueltype', 'aspiration', 'doornumber', 'carbody', 'drive
```

```
main_array = np.array(main_df.car_ID).reshape(-1, 1)
```

```
for column in main_df.columns:
```

```
    if column in encoding_columns:
```

```
        temp = np.array(pd.get_dummies(main_df[column])) # If column is a categorical, pe
    else:
```

```
        temp = np.array(main_df[column]).reshape(-1, 1)
```

```
    main_array = np.hstack((main_array, temp)) # concatate the columns
```

```
main_array = main_array[:, 2:] # Remove car_ID column
```

```
pd.DataFrame(main_array) # Display array
```

	0	1	2	3	4	5	6	7	8	9	...	48	49	50	51	52	53
0	0.0	0.0	0.0	0.0	0.0	1.0	0.0	1.0	1.0	0.0	...	0.0	0.0	3.47	2.68	9.0	111.0
1	0.0	0.0	0.0	0.0	0.0	1.0	0.0	1.0	1.0	0.0	...	0.0	0.0	3.47	2.68	9.0	111.0
2	0.0	0.0	0.0	1.0	0.0	0.0	0.0	1.0	1.0	0.0	...	0.0	0.0	2.68	3.47	9.0	154.0
3	0.0	0.0	0.0	0.0	1.0	0.0	0.0	1.0	1.0	0.0	...	0.0	0.0	3.19	3.40	10.0	102.0
4	0.0	0.0	0.0	0.0	1.0	0.0	0.0	1.0	1.0	0.0	...	0.0	0.0	3.19	3.40	8.0	115.0
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
200	0.0	1.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	0.0	...	0.0	0.0	3.78	3.15	9.5	114.0
201	0.0	1.0	0.0	0.0	0.0	0.0	0.0	1.0	0.0	1.0	...	0.0	0.0	3.78	3.15	8.7	160.0
202	0.0	1.0	0.0	0.0	0.0	0.0	0.0	1.0	1.0	0.0	...	0.0	0.0	3.58	2.87	8.8	134.0
203	0.0	1.0	0.0	0.0	0.0	0.0	1.0	0.0	0.0	1.0	...	0.0	0.0	3.01	3.40	23.0	106.0
204	0.0	1.0	0.0	0.0	0.0	0.0	0.0	1.0	0.0	1.0	...	0.0	0.0	3.78	3.15	9.5	114.0

205 rows × 58 columns

```
X_data = main_array[:, :-1]
```

```
y_data = main_array[:, -1].reshape(-1, 1)
```

```
x_scaler = StandardScaler()
```

```

y_scaler = StandardScaler()
X_data_scaled = x_scaler.fit_transform(X_data)
y_data_scaled = y_scaler.fit_transform(y_data)
print("Shape of X_data: {}".format(X_data.shape))
print("Shape of y_data: {}".format(y_data.shape))
print("=====X_data after rescaling=====")
print(pd.DataFrame(X_data_scaled).head())
print("=====y_data after rescaling=====")
print(y_data_scaled.ravel())

```

Shape of X\_data: (205, 57)

Shape of y\_data: (205, 1)

=====X\_data after rescaling=====

	0	1	2	3	4	5	6	\
0	-0.121867	-0.346726	-0.696784	-0.598010	-0.430083	2.567604	-0.328798	
1	-0.121867	-0.346726	-0.696784	-0.598010	-0.430083	2.567604	-0.328798	
2	-0.121867	-0.346726	-0.696784	1.672213	-0.430083	-0.389468	-0.328798	
3	-0.121867	-0.346726	-0.696784	-0.598010	2.325134	-0.389468	-0.328798	
4	-0.121867	-0.346726	-0.696784	-0.598010	2.325134	-0.389468	-0.328798	
	7	8	9	...	47	48	49	50 \
0	0.328798	0.469295	-0.469295	...	1.08667	-0.214286	-0.070014	0.519071
1	0.328798	0.469295	-0.469295	...	1.08667	-0.214286	-0.070014	0.519071
2	0.328798	0.469295	-0.469295	...	1.08667	-0.214286	-0.070014	-2.404880
3	0.328798	0.469295	-0.469295	...	1.08667	-0.214286	-0.070014	-0.517266
4	0.328798	0.469295	-0.469295	...	1.08667	-0.214286	-0.070014	-0.517266
	51	52	53	54	55	56		
0	-1.839377	-0.288349	0.174483	-0.262960	-0.646553	-0.546059		
1	-1.839377	-0.288349	0.174483	-0.262960	-0.646553	-0.546059		
2	0.685946	-0.288349	1.264536	-0.262960	-0.953012	-0.691627		
3	0.462183	-0.035973	-0.053668	0.787855	-0.186865	-0.109354		
4	0.462183	-0.540725	0.275883	0.787855	-1.106241	-1.273900		

[5 rows x 57 columns]

=====y\_data after rescaling=====

```

[ 2.73911432e-02  4.04461099e-01  4.04461099e-01  8.44849301e-02
 5.23667906e-01  2.47610036e-01  5.56292928e-01  7.08124756e-01
 1.32988237e+00  5.75010530e-01  3.95677439e-01  4.57790460e-01
 9.65360500e-01  9.82300415e-01  1.41646416e+00  2.19381802e+00
 3.51826840e+00  2.96176083e+00 -1.01962107e+00 -8.76070979e-01
-8.40936341e-01 -9.66793634e-01 -8.65781550e-01 -6.67521806e-01
-8.84352716e-01 -8.26255082e-01 -7.11189142e-01 -5.92107815e-01
-5.46558266e-01 -3.92391883e-02 -8.52982503e-01 -8.05801703e-01
-9.88501821e-01 -8.46708460e-01 -7.71419950e-01 -7.50590129e-01
-7.50590129e-01 -6.75301619e-01 -5.24724598e-01 -5.56094811e-01
-3.74147578e-01 -4.16233245e-02 -3.67873535e-01 -8.14585363e-01
-5.47122930e-01 -5.47122930e-01 -2.79660498e-01  2.38078449e+00
 2.79487130e+00  2.85133768e+00 -1.01409991e+00 -9.01167150e-01
-8.13330554e-01 -8.25878639e-01 -7.38042044e-01 -2.92585025e-01
-1.79652260e-01  4.62132708e-02  2.97174972e-01 -5.56094811e-01
-6.00013109e-01 -3.36503323e-01 -3.80421620e-01 -3.11407153e-01
-2.54940770e-01  6.27817012e-01  6.35847787e-01  1.54031376e+00
 1.87861013e+00  1.86957551e+00  2.29922194e+00  2.62346446e+00
 2.73288376e+00  3.47372270e+00  4.03085767e+00  4.04837541e-01
-9.89756630e-01 -8.89371950e-01 -8.29141141e-01 -7.01150674e-01
-4.16309144e-01 -5.99511185e-01 -8.12752732e-02  1.99801832e-01
 1.52119108e-01 -7.88987269e-01 -6.38410249e-01 -5.01636122e-01
-5.01636122e-01 -9.75953736e-01 -7.75184376e-01 -8.31650758e-01]

```

```

-8.06554588e-01 -7.43814163e-01 -7.50088206e-01 -6.87347781e-01
-7.24992036e-01 -6.62251610e-01 -6.30881398e-01 -5.43044803e-01
-4.67756292e-01 2.78930666e-02 1.40825832e-01 2.78930666e-02
4.92172213e-01 8.05874339e-01 6.42749233e-01 -1.72750813e-01
-9.62570765e-03 -1.04991154e-01 7.31916536e-02 2.89018716e-01
4.54653439e-01 4.28929864e-01 4.76612588e-01 4.20773609e-01
5.86408332e-01 6.11504502e-01 -9.66793634e-01 -6.67521806e-01
-8.84352716e-01 -8.26255082e-01 -7.11189142e-01 -5.46558266e-01
-6.43353584e-02 1.09686443e+00 2.41566817e+00 2.60388944e+00

```

```

X_train, X_test, y_train, y_test = train_test_split(X_data_scaled, y_data_scaled, test_siz
print("Shape of X_train: {}".format(X_train.shape))
print("Shape of X_test: {}".format(X_test.shape))
print("Shape of y_train: {}".format(y_train.shape))
print("Shape of y_test: {}".format(y_test.shape))

```

```

Shape of X_train: (184, 57)
Shape of X_test: (21, 57)
Shape of y_train: (184, 1)
Shape of y_test: (21, 1)

```

```

model = Sequential()
model.add(Dense(8, activation='relu', input_shape=(None, 57)))
model.add(Dense(4, activation='relu'))
model.add(Dense(1, activation='linear'))
model.compile(optimizer='adam', loss='mse', metrics=['mae'])
model.summary()

```

Model: "sequential\_1"

Layer (type)	Output Shape	Param #
dense_1 (Dense)	(None, None, 8)	464
dense_2 (Dense)	(None, None, 4)	36
dense_3 (Dense)	(None, None, 1)	5

```

=====
Total params: 505
Trainable params: 505
Non-trainable params: 0
=====

```

