

CODE TO PREDICT CAR PURCHASING DOLLAR AMOUNT USING ANNS(REGRESSION TASK)

PROBLEM STATEMENT

You are working as a car salesman and you would like to develop a model to predict the total amount in dollars that customers are willing to pay given the following attributes:

- Customer Name
- Customer e-mail
- Country
- Gender
- Age
- Annual Salary
- Credit Card Debt
- Net Worth

The model should predict:

- Car Purchase Amount

STEP #0: LIBRARIES IMPORT

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

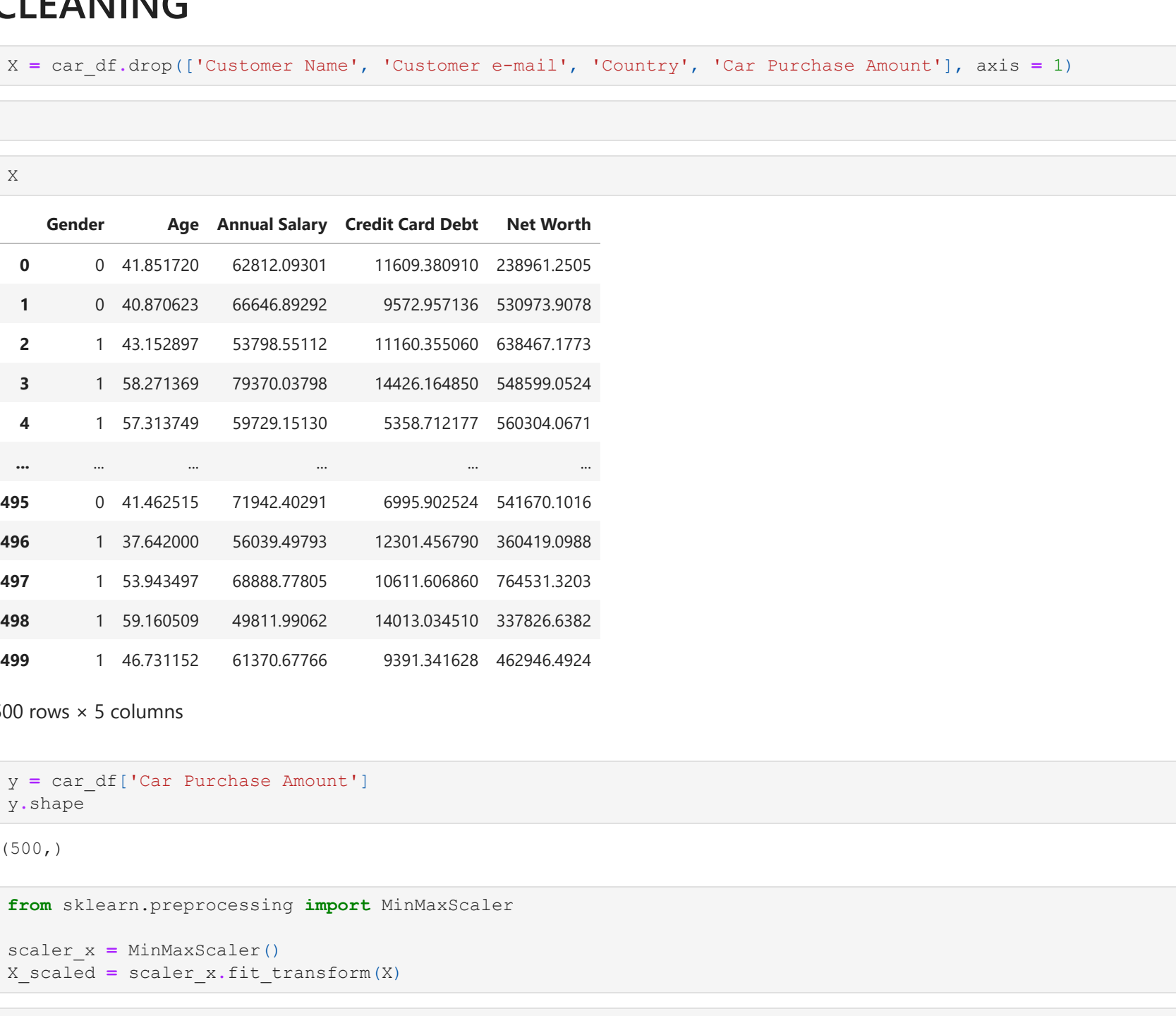
STEP #1: IMPORT DATASET

```
In [2]: car_df = pd.read_csv('Car_Purchasing_Data.csv', encoding='ISO-8859-1')
In [3]: car_df
```

	Customer Name	Customer e-mail	Country	Gender	Age	Annual Salary	Credit Card Debt	Net Worth	Pu Ar	
0	Martina Rivas	rcubilla@CuraePhasellus@quissacanconsecteturadip...	Bulgaria		0	41.851720	62812.09301	11609.380910	238961.2505	53531
1	Harlan Barnes	edolor@diam.co.uk	Belize		0	40.870623	66646.89292	9572.957136	530973.9078	45115
2	Naim Rodriguez	vulpulatae@ametusconsecteturadip...	Algeria		1	43.152897	53798.55112	11160.355060	638467.1773	42945
3	Jade Cunningham	malesuada@dignissim.com	Cook Islands		1	58.271369	79370.03798	14426.164850	548599.0524	67922
4	Cedric Leach	felis.ullamcorper.viverra@egemollislectus.net	Brazil		1	57.313749	59729.15130	5358.712177	560304.0671	55915
...
495	Wilhem	ligula@Cumsocia.ca	Nepal		0	41.462515	71942.40291	6995.902524	541670.1016	48901
496	Vanna	Cumsocis.natoque@Sedmolestie.edu	Zimbabwe		1	37.642000	56039.49793	12301.456790	360419.0988	31491
497	Pearl	penatibus.en@mossanonante.com	Philippines		1	53.943497	68888.77805	10611.606860	764531.6303	61447
498	Neel	Quitiquevarius@arcuVivamusnisi.net	Botswana		1	59.160509	49811.99062	14013.034510	337826.6382	45442
499	Maria	Camaron.maria@hotmail.com	malta		1	46.731152	61370.67766	9391.341628	462946.4924	45107

500 rows x 9 columns

STEP #2: VISUALIZE DATASET



STEP #3: CREATE TESTING AND TRAINING DATASET/DATA SPLITTING

```
In [5]: X = car_df.drop(['Customer Name', 'Customer e-mail', 'Country', 'Car Purchase Amount'], axis = 1)
In [6]: y = car_df['Car Purchase Amount']
```

```
Out [7]: (500,)
```

```
In [8]: from sklearn.preprocessing importMinMaxScaler
scaler_x=MinMaxScaler()
X_scaled=scaler_x.fit_transform(X)
```

```
In [9]: scaler_x.data_max_
Out [9]: array([1.e+00, 7.e+04, 1.e+05, 2.e+04, 1.e+06])
```

```
In [10]: scaler_x.data_min_
Out [10]: array([ 0., 20., 20000., 100., 20000.])
```

```
In [11]: print(X_scaled)
[[0. 0.4370344 0.53515116 0.57836085 0.22342985]
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...
[0. 0.67886994 0.61110973 0.52822145 0.75972594]
[1. -0.78321017 0.37264988 0.63914746 0.3243129 ]
[1. -0.53462305 0.51713347 0.46690159 0.45198622]]
```

```
In [12]: X_scaled.shape
Out [12]: (500, 5)
```

```
In [13]: y.shape
Out [13]: (500,)
```

```
In [14]: y = y.values.reshape(-1,1)
In [15]: y.shape
Out [15]: (500, 1)
```

```
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```

STEP#4: TRAINING THE MODEL

```
In [19]: from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X_scaled, y_scaled, test_size = 0.25)

In [20]: import tensorflow.keras
from keras.models import Sequential
from keras.layers import Dense
from sklearn.preprocessing import MinMaxScaler

model = Sequential()
model.add(Dense(25, input_dim=5, activation='relu'))
model.add(Dense(25, activation='relu'))
model.add(Dense(1, activation='linear'))
model.summary()

Model: "sequential"
Layer (type) Output Shape Param #
-----
dense_1 (Dense) (None, 25) 150
dense_2 (Dense) (None, 1) 26
Total params: 826
Trainable params: 826
Non-trainable params: 0

In [21]: model.compile(optimizer='adam', loss='mean_squared_error')

In [22]: epochs_hist = model.fit(X_train, y_train, epochs=20, batch_size=25, verbose=1, validation_split=0.2)

Epoch 1/20 12/12 ===== - ls 16ms/step - loss: 0.1013 - val_loss: 0.0337
Epoch 2/20 12/12 ===== - ls 16ms/step - loss: 0.0253 - val_loss: 0.0158
Epoch 3/20 12/12 ===== - ls 16ms/step - loss: 0.0189 - val_loss: 0.0167
Epoch 4/20 12/12 ===== - ls 16ms/step - loss: 0.0162 - val_loss: 0.0117
Epoch 5/20 12/12 ===== - ls 16ms/step - loss: 0.0136 - val_loss: 0.0098
Epoch 6/20 12/12 ===== - ls 16ms/step - loss: 0.0118 - val_loss: 0.0080
Epoch 7/20 12/12 ===== - ls 16ms/step - loss: 0.0094 - val_loss: 0.0064
Epoch 8/20 12/12 ===== - ls 16ms/step - loss: 0.0075 - val_loss: 0.0048
Epoch 9/20 12/12 ===== - ls 16ms/step - loss: 0.0063 - val_loss: 0.0039
Epoch 10/20 12/12 ===== - ls 16ms/step - loss: 0.0052 - val_loss: 0.0033
Epoch 11/20 12/12 ===== - ls 16ms/step - loss: 0.0043 - val_loss: 0.0029
Epoch 12/20 12/12 ===== - ls 16ms/step - loss: 0.0037 - val_loss: 0.0025
Epoch 13/20 12/12 ===== - ls 16ms/step - loss: 0.0032 - val_loss: 0.0022
Epoch 14/20 12/12 ===== - ls 16ms/step - loss: 0.0028 - val_loss: 0.0020
Epoch 15/20 12/12 ===== - ls 16ms/step - loss: 0.0026 - val_loss: 0.0019
Epoch 16/20 12/12 ===== - ls 16ms/step - loss: 0.0023 - val_loss: 0.0018
Epoch 17/20 12/12 ===== - ls 16ms/step - loss: 0.0021 - val_loss: 0.0016
Epoch 18/20 12/12 ===== - ls 16ms/step - loss: 0.0019 - val_loss: 0.0015
Epoch 19/20 12/12 ===== - ls 16ms/step - loss: 0.0018 - val_loss: 0.0015
Epoch 20/20 12/12 ===== - ls 16ms/step - loss: 0.0016 - val_loss: 0.0014
```

STEP#5: EVALUATING THE MODEL

```
In [23]: print(epochs_hist.history.keys())
dict_keys(['loss', 'val_loss'])

In [24]: plt.plot(epochs_hist.history['loss'])
plt.plot(epochs_hist.history['val_loss'])
plt.title('Model Loss Progression During Training/Validation')
plt.ylabel('Training and Validation Losses')
plt.xlabel('Epoch Number')
plt.legend(['Training Loss', 'Validation Loss'])

Out[24]: <matplotlib.legend.Legend at 0x25050d00280>
```



```
In [25]: # Gender, Age, Annual Salary, Credit Card Debt, Net Worth
# *** (Note that input data must be normalized) ***
X_test_sample = np.array([[0, 0.4370344, 0.53515116, 0.57836085, 0.22342985]])
#X_test_sample = np.array([[1, 0.53462305, 0.51713347, 0.46690159, 0.45198622]])

y_predict_sample = model.predict(X_test_sample)

print('Expected Purchase Amount=', y_predict_sample)
y_predict_sample_orig = scaler.y.inverse_transform(y_predict_sample)
print('Expected Purchase Amount=', y_predict_sample_orig)

Expected Purchase Amount= [[0.42265666]]
Expected Purchase Amount= [[39008.62]]
```

Prediction Using ANNs FOR REGRESSION TASKS!

```
In [26]: numerator = y_predict_sample_orig
denominator = y_predict_sample * 100000
percentage_accuracy = (numerator/denominator) * 100
print(percentage_accuracy)

[[92.29387637]]
```

```
In [ ]:
```