



# Chapter 3: ANN

## Ex1: Cars - Linear activation function

- Cho dữ liệu cars.csv
- Áp dụng ANN để dự đoán giá trị của một "potential car sale"

```
In [1]: # from google.colab import drive
# drive.mount("/content/gdrive", force_remount=True)
```

```
In [2]: # %cd '/content/gdrive/My Drive/LDS8_DeepLearning/Practice/Chapter3/'
```

```
In [3]: import warnings
warnings.filterwarnings('ignore')
```

```
In [4]: import tensorflow as tf
from tensorflow import keras
```

```
In [5]: import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.model_selection import KFold
from sklearn.pipeline import Pipeline
from sklearn.preprocessing import MinMaxScaler
from tensorflow.keras import Sequential
from tensorflow.keras.layers import Dense
from tensorflow.keras.wrappers.scikit_learn import KerasRegressor
```

```
In [6]: print(tf.__version__)
print(keras.__version__)
```

```
2.5.0
2.5.0
```

```
In [7]: # Load Dataset
#Variables
dataset=pd.read_csv("cars.csv", delimiter=";", header=None)
dataset.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 963 entries, 0 to 962
Data columns (total 6 columns):
 #   Column  Non-Null Count  Dtype
---  -
 0    0      963 non-null    int64
 1    1      963 non-null    int64
 2    2      963 non-null    int64
 3    3      963 non-null    int64
 4    4      963 non-null    int64
 5    5      963 non-null    int64
dtypes: int64(6)
memory usage: 45.3 KB
```

```
In [8]: dataset.head()
```

Out[8]:

	0	1	2	3	4	5
0	28	0	23	0	4099	620
1	26	0	27	0	2677	1792
2	30	1	58	41576	6215	27754
3	26	1	25	43172	7626	28256
4	20	1	17	6979	8071	4438

```
In [9]: dataset.describe()
```

Out[9]:

	0	1	2	3	4	5
<b>count</b>	963.000000	963.000000	963.000000	963.000000	963.000000	963.000000
<b>mean</b>	37.971963	0.512980	27.704050	14109.004154	6176.047767	11689.860852
<b>std</b>	12.290838	0.500091	13.378181	18273.702481	3260.670142	8986.896921
<b>min</b>	19.000000	0.000000	10.000000	0.000000	0.000000	500.000000
<b>25%</b>	27.000000	0.000000	20.000000	1475.000000	3506.500000	3554.000000
<b>50%</b>	37.000000	1.000000	25.000000	6236.000000	6360.000000	9130.000000
<b>75%</b>	49.000000	1.000000	32.000000	16686.000000	8649.500000	19245.000000
<b>max</b>	60.000000	1.000000	97.000000	59770.000000	11970.000000	29926.000000

```
In [10]: x=dataset.iloc[:,0:5]
y=dataset.iloc[:,5]
print(type(x))
print(type(y))

<class 'pandas.core.frame.DataFrame'>
<class 'pandas.core.series.Series'>
```

```
In [11]: y=np.reshape(np.array(y), (-1,1))

scaler_x = MinMaxScaler()
scaler_y = MinMaxScaler()
# print(scaler.fit(x))
# print(scaler.fit(y))
```

```
In [12]: xscale=scaler_x.fit_transform(x)
yscale=scaler_y.fit_transform(y)
#print(xscale)
#print(yscale)
```

```
In [13]: # Keras Model Configuration: Neural Network API
model = Sequential()
model.add(Dense(3, input_dim=5, kernel_initializer='normal', activation='relu'))
# model.add(Dense(12, input_dim=5, kernel_initializer='normal', activation='relu'))
# tham số đầu tiên là units: là số nguyên, nodes của đầu ra
model.add(Dense(3, activation='relu'))
model.add(Dense(1, activation='linear'))
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 3)	18
dense_1 (Dense)	(None, 3)	12
dense_2 (Dense)	(None, 1)	4

=====

Total params: 34  
Trainable params: 34  
Non-trainable params: 0

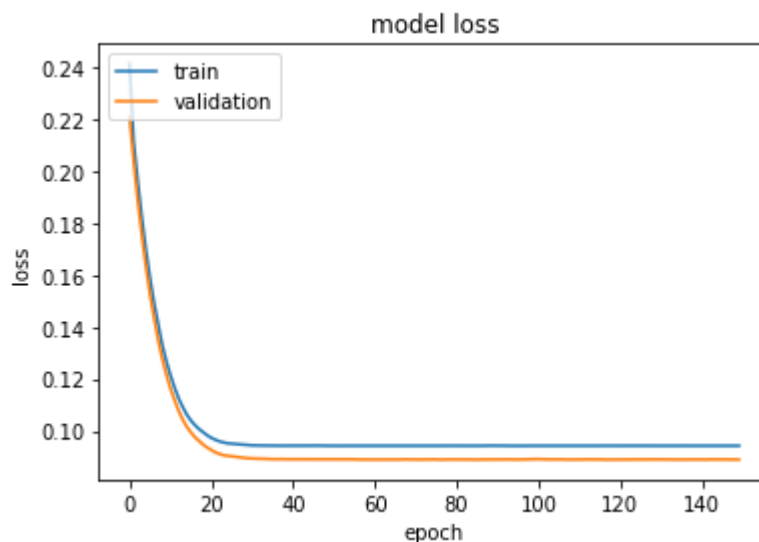
```
In [14]: model.compile(loss='mse', optimizer='adam', metrics=['mse', 'mae'])
```

```
In [15]: # fit model
history = model.fit(xscale, yscale, epochs=150,
                    batch_size=32,
                    verbose=1,
                    validation_split=0.2)
```

```
Epoch 1/150
25/25 [=====] - 0s 6ms/step - loss: 0.2415 - mse: 0.
2415 - mae: 0.3786 - val_loss: 0.2211 - val_mse: 0.2211 - val_mae: 0.3642
Epoch 2/150
25/25 [=====] - 0s 2ms/step - loss: 0.2118 - mse: 0.
2118 - mae: 0.3469 - val_loss: 0.2044 - val_mse: 0.2044 - val_mae: 0.3430
Epoch 3/150
25/25 [=====] - 0s 2ms/step - loss: 0.1965 - mse: 0.
1965 - mae: 0.3292 - val_loss: 0.1890 - val_mse: 0.1890 - val_mae: 0.3246
Epoch 4/150
25/25 [=====] - 0s 2ms/step - loss: 0.1825 - mse: 0.
1825 - mae: 0.3144 - val_loss: 0.1756 - val_mse: 0.1756 - val_mae: 0.3099
Epoch 5/150
25/25 [=====] - 0s 2ms/step - loss: 0.1700 - mse: 0.
1700 - mae: 0.3028 - val_loss: 0.1635 - val_mse: 0.1635 - val_mae: 0.2969
Epoch 6/150
25/25 [=====] - 0s 2ms/step - loss: 0.1591 - mse: 0.
1591 - mae: 0.2938 - val_loss: 0.1527 - val_mse: 0.1527 - val_mae: 0.2856
Epoch 7/150
25/25 [=====] - 0s 2ms/step - loss: 0.1491 - mse: 0.
1491 - mae: 0.2856 - val_loss: 0.1427 - val_mse: 0.1427 - val_mae: 0.2756
```

```
In [16]: print(history.history.keys())
# "Loss"
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['train', 'validation'], loc='upper left')
plt.show()
```

```
dict_keys(['loss', 'mse', 'mae', 'val_loss', 'val_mse', 'val_mae'])
```



```
In [17]: # Evaluate the model on the test data using `evaluate`  
print('\n# Evaluate on test data')  
results = model.evaluate(xscale, yscale)  
print('loss, mse, mae:', results)
```

```
# Evaluate on test data  
31/31 [=====] - 0s 831us/step - loss: 0.0932 - mse: 0.  
0932 - mae: 0.2699  
loss, mse, mae: [0.09317748993635178, 0.09317748993635178, 0.2698577344417572]
```

```
In [20]: # Predictions  
Xnew = np.array([[40, 0, 26, 9000, 8000]])  
X_new = scaler_x.transform(Xnew)  
ynew=model.predict(X_new)  
y_new = scaler_y.inverse_transform(ynew)  
print("X=%s, Predicted=%s" % (Xnew[0], y_new[0]))
```

```
X=[ 40    0   26 9000 8000], Predicted=[11720.976]
```

```
In [ ]:
```