

ANN

Keras Classification Code Along

```

In [1]: 1 import pandas as pd
        2 import numpy as np
        3 import matplotlib.pyplot as plt
        4 import seaborn as sns
        5 import plotly.express as px

In [2]: 1 # df = pd.read_csv(r"D:\LIUZHICHENG\Udemy\Deep Learning\Jose\TF_2_Notebooks_and_Data\DATA\cancer_classification.csv")
        2 df = pd.read_csv("/Volumes/GoogleDrive/My Drive/Udemy/Deep Learning/Jose/TF_2_Notebooks_and_Data/DATA/cancer_classification.csv")
        3 df.head()

```

```

Out[2]:
   mean radius  mean texture  mean perimeter  mean area  mean smoothness  mean compactness  mean concavity  mean concave points  mean symmetry  mean fractal dimension  ...  worst texture  worst perimeter  worst area  worst smoothness  c
0      17.99      10.38      122.80      1001.0         0.11840         0.27760         0.3001         0.14710         0.2419         0.07871  ...        17.33        184.60       2019.0         0.1622
1      20.57      17.77      132.90      1326.0         0.08474         0.07864         0.0869         0.07017         0.1812         0.05667  ...        23.41        158.80       1956.0         0.1238
2      19.69      21.25      130.00      1203.0         0.10960         0.15990         0.1974         0.12790         0.2069         0.05999  ...        25.53        152.50       1709.0         0.1444
3      11.42      20.38       77.58       386.1         0.14250         0.28390         0.2414         0.10520         0.2597         0.09744  ...        26.50         98.87        567.7         0.2098
4      20.29      14.34      135.10      1297.0         0.10030         0.13280         0.1980         0.10430         0.1809         0.05883  ...        16.67        152.20       1575.0         0.1374

```

5 rows × 31 columns

```

In [3]: 1 df.shape

```

```

Out[3]: (569, 31)

```

```

In [4]: 1 df.isna().sum()

```

```

Out[4]: mean radius      0
        mean texture    0
        mean perimeter  0
        mean area       0
        mean smoothness 0
        mean compactness 0
        mean concavity  0
        mean concave points 0
        mean symmetry    0
        mean fractal dimension 0
        radius error     0
        texture error    0
        perimeter error  0
        area error       0
        smoothness error 0
        compactness error 0
        concavity error  0
        concave points error 0
        symmetry error   0
        fractal dimension error 0
        worst radius     0
        worst texture    0
        worst perimeter  0
        worst area       0
        worst smoothness 0
        worst compactness 0
        worst concavity  0
        worst concave points 0
        worst symmetry    0
        worst fractal dimension 0
        benign_0__mal_1    0
        dtype: int64

```

EDA for new df

[...]

Building a Model

```

In [7]: 1 X = df.drop(columns="benign_0__mal_1").to_numpy()
        2 y = df["benign_0__mal_1"].to_numpy()

```

```

In [8]: 1 from sklearn.model_selection import train_test_split
        2
        3 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.25, random_state=101)
        4
        5 from sklearn.preprocessing import MinMaxScaler, StandardScaler
        6
        7 sc = MinMaxScaler()
        8 X_train = sc.fit_transform(X_train)
        9 X_test = sc.transform(X_test)

```

```
In [9]: 1 from tensorflow.keras.models import Sequential
2 from tensorflow.keras.layers import Dense, Dropout
3
4 model = Sequential()
5
6 model.add(layer=Dense(units=30, activation="relu"))
7 model.add(layer=Dense(units=15, activation="relu"))
8
9 # Binary classification
10 model.add(layer=Dense(units=1, activation="sigmoid")) # Last Layer
11
12 model.compile(optimizer="adam", loss="binary_crossentropy")
```

```
In [10]: 1 model.fit(x=X_train, y=y_train, epochs=600,
2 validation_data=(X_test, y_test),
3 verbose=1)
```

```
Epoch 1/600
14/14 [=====] - 1s 10ms/step - loss: 0.6611 - val_loss: 0.6341
Epoch 2/600
14/14 [=====] - 0s 2ms/step - loss: 0.6171 - val_loss: 0.5904
Epoch 3/600
14/14 [=====] - 0s 2ms/step - loss: 0.5717 - val_loss: 0.5439
Epoch 4/600
14/14 [=====] - 0s 2ms/step - loss: 0.5250 - val_loss: 0.4943
Epoch 5/600
14/14 [=====] - 0s 2ms/step - loss: 0.4745 - val_loss: 0.4428
Epoch 6/600
14/14 [=====] - 0s 2ms/step - loss: 0.4295 - val_loss: 0.3957
Epoch 7/600
14/14 [=====] - 0s 2ms/step - loss: 0.3842 - val_loss: 0.3532
Epoch 8/600
14/14 [=====] - 0s 2ms/step - loss: 0.3470 - val_loss: 0.3163
Epoch 9/600
14/14 [=====] - 0s 2ms/step - loss: 0.3129 - val_loss: 0.2832
Epoch 10/600
14/14 [=====] - 0s 2ms/step - loss: 0.2886 - val_loss: 0.2613
```

```
In [11]: 1 model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 30)	930
dense_1 (Dense)	(None, 15)	465
dense_2 (Dense)	(None, 1)	16
Total params: 1,411		
Trainable params: 1,411		
Non-trainable params: 0		

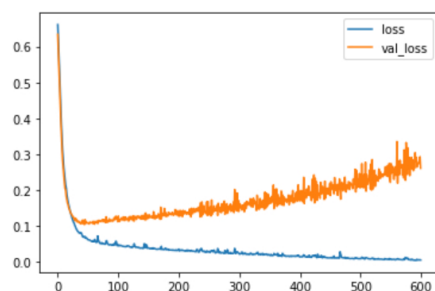
```
In [12]: 1 loss_df = pd.DataFrame(data=model.history.history)
2 loss_df
3 # loss, validation Loss
```

```
Out[12]:
```

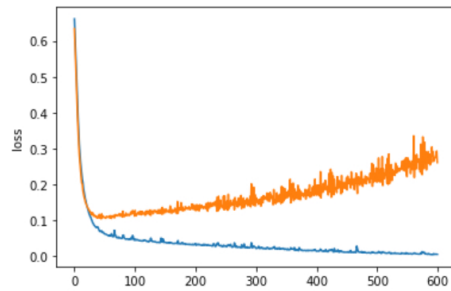
	loss	val_loss
0	0.661094	0.634076
1	0.617055	0.590358
2	0.571726	0.543933
3	0.525043	0.494313
4	0.474459	0.442837
...
595	0.005870	0.272632
596	0.006248	0.278575
597	0.006599	0.283509
598	0.006501	0.293628
599	0.006030	0.261371

600 rows × 2 columns

```
In [13]: 1 # overfitting
2 loss_df.plot();
```



```
In [14]: 1 sns.lineplot(data=loss_df, x=loss_df.index, y="loss")
2 sns.lineplot(data=loss_df, x=loss_df.index, y="val_loss");
```



EarlyStopping

```
In [15]: 1 from tensorflow.keras.models import Sequential
2 from tensorflow.keras.layers import Dense, Dropout
3
4 model = Sequential()
5
6 model.add(layer=Dense(units=30, activation="relu"))
7 model.add(layer=Dense(units=15, activation="relu"))
8
9 # Binary classification
10 model.add(layer=Dense(units=1, activation="sigmoid")) # Last Layer
11
12 model.compile(optimizer="adam", loss="binary_crossentropy")
13
14 from tensorflow.keras.callbacks import EarlyStopping
15
16 early_stop = EarlyStopping(monitor="val_loss", mode="min", verbose=1, patience=25)
17
18 model.fit(x=X_train, y=y_train, epochs=600,
19         validation_data=(X_test, y_test),
20         verbose=1, callbacks=[early_stop])
```

```
Epoch 1/600
14/14 [=====] - 0s 8ms/step - loss: 0.6681 - val_loss: 0.6541
Epoch 2/600
14/14 [=====] - 0s 2ms/step - loss: 0.6345 - val_loss: 0.6214
Epoch 3/600
14/14 [=====] - 0s 2ms/step - loss: 0.5998 - val_loss: 0.5871
Epoch 4/600
14/14 [=====] - 0s 2ms/step - loss: 0.5647 - val_loss: 0.5477
Epoch 5/600
14/14 [=====] - 0s 2ms/step - loss: 0.5232 - val_loss: 0.5060
Epoch 6/600
14/14 [=====] - 0s 2ms/step - loss: 0.4811 - val_loss: 0.4590
Epoch 7/600
14/14 [=====] - 0s 2ms/step - loss: 0.4305 - val_loss: 0.4067
Epoch 8/600
14/14 [=====] - 0s 2ms/step - loss: 0.3823 - val_loss: 0.3600
Epoch 9/600
14/14 [=====] - 0s 2ms/step - loss: 0.3415 - val_loss: 0.3193
Epoch 10/600
14/14 [=====] - 0s 2ms/step - loss: 0.3062 - val loss: 0.2860
```

```
In [16]: 1 model.history.history
```

```
Out[16]: {'loss': [0.6681123971939087,
0.6344802975654602,
0.5997874140739441,
0.564723014831543,
0.5232142210006714,
0.48110640048980713,
0.4304735064506531,
0.38230839371681213,
0.34154751896858215,
0.306238055229187,
0.2775701880455017,
0.25517910718917847,
0.2378966063261032,
0.22003865242004395,
0.2091105580329895,
0.19965435564517975,
0.18735867738723755,
0.17597214877605438,
0.16939526796340942,
0.15864303708076477,
```

```
In [17]: 1 loss_df = pd.DataFrame(data=model.history.history)
2 loss_df
```

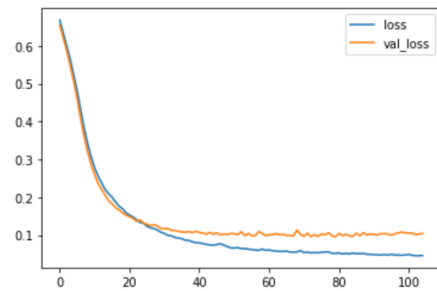
```
Out[17]:
```

	loss	val_loss
0	0.668112	0.654135
1	0.634480	0.621362
2	0.599787	0.587133
3	0.564723	0.547748
4	0.523214	0.506039
...

100	0.048395	0.105124
101	0.046471	0.105373
102	0.045163	0.100747
103	0.045341	0.102875
104	0.045771	0.103905

105 rows × 2 columns

In [18]: `loss_df.plot();`



Dropout

```
In [19]: 1 from tensorflow.keras.models import Sequential
2         from tensorflow.keras.layers import Dense, Dropout
3
4         model = Sequential()
5
6         model.add(layer=Dense(units=30, activation="relu"))
7         model.add(layer=Dropout(rate=0.5)) # 50% neurons be turned off
8
9         model.add(layer=Dense(units=15, activation="relu"))
10        model.add(layer=Dropout(rate=0.5)) # 50% neurons be turned off
11
12        # Binary classification
13        model.add(layer=Dense(units=1, activation="sigmoid")) # Last Layer
14
15        model.compile(optimizer="adam", loss="binary_crossentropy")
16
17        from tensorflow.keras.callbacks import EarlyStopping
18
19        early_stop = EarlyStopping(monitor="val_loss", mode="min", verbose=1, patience=25)
20
21        model.fit(x=X_train, y=y_train, epochs=600,
22                  validation_data=(X_test, y_test),
23                  verbose=1, callbacks=[early_stop])
```

```
Epoch 1/600
14/14 [=====] - 0s 8ms/step - loss: 0.6976 - val_loss: 0.6665
Epoch 2/600
14/14 [=====] - 0s 2ms/step - loss: 0.6719 - val_loss: 0.6579
Epoch 3/600
14/14 [=====] - 0s 2ms/step - loss: 0.6706 - val_loss: 0.6471
Epoch 4/600
14/14 [=====] - 0s 2ms/step - loss: 0.6563 - val_loss: 0.6329
Epoch 5/600
14/14 [=====] - 0s 2ms/step - loss: 0.6432 - val_loss: 0.6101
Epoch 6/600
14/14 [=====] - 0s 2ms/step - loss: 0.6226 - val_loss: 0.5866
Epoch 7/600
14/14 [=====] - 0s 2ms/step - loss: 0.5918 - val_loss: 0.5583
Epoch 8/600
14/14 [=====] - 0s 2ms/step - loss: 0.5893 - val_loss: 0.5337
Epoch 9/600
14/14 [=====] - 0s 2ms/step - loss: 0.5681 - val_loss: 0.5093
Epoch 10/600
14/14 [=====] - 0s 2ms/step - loss: 0.5608 - val_loss: 0.4897
```

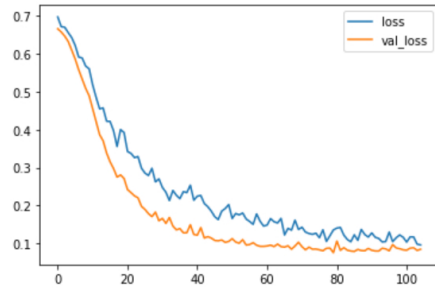
In [20]: `1 loss_df = pd.DataFrame(data=model.history.history)
2 loss_df`

Out[20]:

	loss	val_loss
0	0.697633	0.666492
1	0.671942	0.657884
2	0.670568	0.647124
3	0.656304	0.632946
4	0.643187	0.610075
...
100	0.103479	0.082177
101	0.117253	0.087561
102	0.117101	0.088553
103	0.098061	0.081630
104	0.096125	0.084610

105 rows × 2 columns

```
1 loss_df.plot();
```



- ▼ *model.predict_classes* instead of *model.predict*

```
1 model.predict_classes(X_test)
```

```

/opt/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/engine/sequential.py:455: UserWarning: `model.predict_classes()` is deprecated and will be removed after 2021-01-01. Please use instead: * `np.argmax(model.predict(x), axis=-1)`, if your model does multi-class classification (e.g. if it uses a `softmax` last-layer activation). * `model.predict(x) > 0.5).astype("int32")`, if your model does binary classification (e.g. if it uses a `sigmoid` last-layer activation).
warnings.warn("`model.predict_classes()` is deprecated and ")

```

```
Out[22]: array([[1],  
                [1],  
                [1],  
                [0],  
                [1],  
                [1],  
                [1],  
                [0],  
                [1],  
                [1],  
                [0],  
                [1],  
                [1]])
```

```
1 from sklearn.metrics import classification_report, confusion_matrix, accuracy_score
```

```
3 print(classification_report(y_test, model.predict_classes(X_test)))
4 print(confusion_matrix(y_test, model.predict_classes(X_test)))
5 print(accuracy_score(y_test, model.predict_classes(X_test)))
```

```

/opt/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/engine/sequential.py:455: UserWarning: `model.predict_classes()` is deprecated and will be removed after 2021-01-01. Please use instead: * `np.argmax(model.predict(x), axis=-1)`, if your model does multi-class classification (e.g. if it uses a `softmax` last-layer activation). * `(model.predict(x) > 0.5).astype("int32")`, if your model does binary classification (e.g. if it uses a `sigmoid` last-layer activation).
warnings.warn("`model.predict_classes()` is deprecated and ")
/opt/anaconda3/lib/python3.8/site-packages/tensorflow/python/keras/engine/sequential.py:455: UserWarning: `model.predict_classes()` is deprecated and will be removed after 2021-01-01. Please use instead: * `np.argmax(model.predict(x), axis=-1)`, if your model does multi-class classification (e.g. if it uses a `softmax` last-layer activation). * `(model.predict(x) > 0.5).astype("int32")`, if your model does binary classification (e.g. if it uses a `sigmoid` last-layer activation).
warnings.warn("`model.predict_classes()` is deprecated and ")

```

	precision	recall	f1-score	support
0	0.96	0.98	0.97	55
1	0.99	0.98	0.98	88
accuracy			0.98	143
macro avg	0.98	0.98	0.98	143
weighted avg	0.98	0.98	0.98	143

```
[[54  1]
 [ 2 86]]
0.9790209790209791
```