ANN Theory

In [1]: pip install torch

Artifical Netural Network

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Requirement already satisfied: torch in c:\users\roy62\anaconda3\envs\tensorflow_
        env\lib\site-packages (2.5.1)
        Requirement already satisfied: filelock in c:\users\roy62\anaconda3\envs\tensorfl
        ow_env\lib\site-packages (from torch) (3.16.1)
        Requirement already satisfied: typing-extensions>=4.8.0 in c:\users\roy62\anacond
        a3\envs\tensorflow_env\lib\site-packages (from torch) (4.11.0)
        Requirement already satisfied: networkx in c:\users\roy62\anaconda3\envs\tensorfl
        ow env\lib\site-packages (from torch) (3.4.2)
        Requirement already satisfied: jinja2 in c:\users\roy62\anaconda3\envs\tensorflow
        env\lib\site-packages (from torch) (3.1.4)
        Requirement already satisfied: fsspec in c:\users\roy62\anaconda3\envs\tensorflow
        _env\lib\site-packages (from torch) (2024.10.0)
        Requirement already satisfied: sympy==1.13.1 in c:\users\roy62\anaconda3\envs\ten
        sorflow_env\lib\site-packages (from torch) (1.13.1)
        Requirement already satisfied: mpmath<1.4,>=1.1.0 in c:\users\roy62\anaconda3\env
        s\tensorflow_env\lib\site-packages (from sympy==1.13.1->torch) (1.3.0)
        Requirement already satisfied: MarkupSafe>=2.0 in c:\users\roy62\anaconda3\envs\t
        ensorflow_env\lib\site-packages (from jinja2->torch) (2.1.3)
        Note: you may need to restart the kernel to use updated packages.
In [2]: import numpy as np
In [3]: import pandas as pd
In [4]: import tensorflow as tf
In [8]: tf.__version__
Out[8]: '2.10.0'
In [9]: # Part-1 Data Preprocessing
         dataset = pd.read_csv(r"E:\Data Science & AI\Dataset files\Churn_Modelling.csv")
         X = dataset.iloc[:, 3:-1].values
         y = dataset.iloc[:, -1].values
In [10]: print(X)
        [[619 'delhi' 'Female' ... 1 1 101348.88]
         [608 'bangalore' 'Female' ... 0 1 112542.58]
         [502 'delhi' 'Female' ... 1 0 113931.57]
         [709 'delhi' 'Female' ... 0 1 42085.58]
         [772 'mumbai' 'Male' ... 1 0 92888.52]
         [792 'delhi' 'Female' ... 1 0 38190.78]]
In [11]: | print(y)
```

```
[1 0 1 ... 1 1 0]
```

```
In [12]: # Encoding Categorical Dat
             ## Label Encoding Gender Column
         from sklearn.preprocessing import LabelEncoder
         le = LabelEncoder()
         X[:, 2] = le.fit_transform(X[:, 2])
In [13]: print(X)
        [[619 'delhi' 0 ... 1 1 101348.88]
         [608 'bangalore' 0 ... 0 1 112542.58]
         [502 'delhi' 0 ... 1 0 113931.57]
         . . .
         [709 'delhi' 0 ... 0 1 42085.58]
         [772 'mumbai' 1 ... 1 0 92888.52]
         [792 'delhi' 0 ... 1 0 38190.78]]
In [14]: # Geography Column
         from sklearn.compose import ColumnTransformer
         from sklearn.preprocessing import OneHotEncoder
         ct = ColumnTransformer(transformers=[('encoder', OneHotEncoder(), [1])], remaind
         X = np.array(ct.fit_transform(X))
In [15]: print(X)
        [[0.0 1.0 0.0 ... 1 1 101348.88]
         [1.0 0.0 0.0 ... 0 1 112542.58]
         [0.0 1.0 0.0 ... 1 0 113931.57]
         [0.0 1.0 0.0 ... 0 1 42085.58]
         [0.0 0.0 1.0 ... 1 0 92888.52]
         [0.0 1.0 0.0 ... 1 0 38190.78]]
In [16]: # Feature Scaling
         from sklearn.preprocessing import StandardScaler
         sc = StandardScaler()
         X = sc.fit_transform(X)
In [17]: | print(X)
        [[-0.57380915  0.99720391 -0.57873591 ...  0.64609167  0.97024255
           0.02188649]
         [ 1.74273971 -1.00280393 -0.57873591 ... -1.54776799 0.97024255 ]
           0.21653375]
         [-0.57380915 0.99720391 -0.57873591 ... 0.64609167 -1.03067011
           0.2406869 ]
         [-0.57380915 \quad 0.99720391 \quad -0.57873591 \quad \dots \quad -1.54776799 \quad 0.97024255
          -1.00864308]
         [-0.57380915 -1.00280393 1.72790383 ... 0.64609167 -1.03067011
         -0.12523071]
         [-0.57380915 0.99720391 -0.57873591 ... 0.64609167 -1.03067011
          -1.07636976]]
In [18]: # Splitting the dataset into the Training set and Test set
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```
from sklearn.model_selection import train_test_split
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, rando
In [19]: # Part 2 - Building the ANN
            # Initializing the ANN
         ann = tf.keras.models.Sequential()
In [22]: # Adding the input layer and the first hidden layer
         ann.add(tf.keras.layers.Dense(units=6, activation='relu'))
In [23]: # Adding the second hidden Layer
         ann.add(tf.keras.layers.Dense(units=6, activation='relu'))
In [24]: # Adding the output layer
         ann.add(tf.keras.layers.Dense(units=1, activation='sigmoid'))
In [25]: # part-3 Training
             ## Compiling the ANN
         ann.compile(optimizer = 'adam', loss = 'binary_crossentropy', metrics = ['accura
In [26]: # Training the ANN on the Training set
         ann.fit(X_train, y_train, batch_size = 32, epochs = 100)
```

```
Epoch 1/100
0.5660
Epoch 2/100
0.7930
Epoch 3/100
0.8021
Epoch 4/100
0.8173
Epoch 5/100
0.8201
Epoch 6/100
0.8211
Epoch 7/100
0.8221
Epoch 8/100
0.8280
Epoch 9/100
0.8295
Epoch 10/100
0.8369
Epoch 11/100
0.8432
Epoch 12/100
0.8526
Epoch 13/100
0.8551
Epoch 14/100
0.8569
Epoch 15/100
0.8572
Epoch 16/100
0.8575
Epoch 17/100
0.8593
Epoch 18/100
0.8571
Epoch 19/100
0.8586
Epoch 20/100
0.8595
```

```
Epoch 21/100
0.8601
Epoch 22/100
0.8595
Epoch 23/100
0.8605
Epoch 24/100
0.8605
Epoch 25/100
0.8619
Epoch 26/100
0.8602
Epoch 27/100
0.8609
Epoch 28/100
0.8614
Epoch 29/100
0.8609
Epoch 30/100
0.8610
Epoch 31/100
0.8614
Epoch 32/100
0.8605
Epoch 33/100
0.8609
Epoch 34/100
0.8611
Epoch 35/100
0.8621
Epoch 36/100
0.8615
Epoch 37/100
0.8606
Epoch 38/100
0.8621
Epoch 39/100
0.8624
Epoch 40/100
0.8611
```

```
Epoch 41/100
0.8618
Epoch 42/100
0.8631
Epoch 43/100
0.8610
Epoch 44/100
0.8640
Epoch 45/100
0.8625
Epoch 46/100
0.8636
Epoch 47/100
0.8615
Epoch 48/100
0.8616
Epoch 49/100
0.8629
Epoch 50/100
0.8619
Epoch 51/100
0.8639
Epoch 52/100
0.8626
Epoch 53/100
0.8626
Epoch 54/100
0.8636
Epoch 55/100
0.8640
Epoch 56/100
0.8611
Epoch 57/100
0.8646
Epoch 58/100
0.8627
Epoch 59/100
0.8649
Epoch 60/100
0.8634
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```
Epoch 61/100
0.8655
Epoch 62/100
0.8641
Epoch 63/100
0.8645
Epoch 64/100
0.8629
Epoch 65/100
0.8634
Epoch 66/100
0.8645
Epoch 67/100
0.8625
Epoch 68/100
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Epoch 69/100
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Epoch 70/100
0.8636
Epoch 71/100
0.8630
Epoch 72/100
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Epoch 73/100
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Epoch 74/100
0.8635
Epoch 75/100
0.8624
Epoch 76/100
0.8636
Epoch 77/100
0.8644
Epoch 78/100
0.8651
Epoch 79/100
0.8644
Epoch 80/100
0.8626
```

```
Epoch 81/100
0.8640
Epoch 82/100
0.8640
Epoch 83/100
0.8646
Epoch 84/100
0.8643
Epoch 85/100
0.8648
Epoch 86/100
0.8634
Epoch 87/100
0.8640
Epoch 88/100
0.8633
Epoch 89/100
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Epoch 90/100
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Epoch 91/100
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Epoch 92/100
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Epoch 93/100
0.8645
Epoch 94/100
0.8640
Epoch 95/100
0.8646
Epoch 96/100
0.8641
Epoch 97/100
0.8641
Epoch 98/100
0.8648
Epoch 99/100
0.8640
Epoch 100/100
0.8627
```

```
Out[26]: <keras.callbacks.History at 0x239574cd750>
In [27]: # Part 4 - Making the predictions and evaluating the model
                 ## Predicting the Test set results
         y_pred = ann.predict(X_test)
         y_pred = (y_pred > 0.5)
         print(np.concatenate((y_pred.reshape(len(y_pred),1), y_test.reshape(len(y_test),
        63/63 [========= ] - 0s 2ms/step
        [[0 0]]
        [0 1]
        [0 0]
         [0 0]
         [0 0]
        [0 0]]
In [28]: # Making the Confusion Matrix
         from sklearn.metrics import confusion_matrix
         cm = confusion_matrix(y_test, y_pred)
         print(cm)
        [[1518 77]
         [ 197 208]]
In [ ]:
In [ ]:
In [ ]:
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