BUILDING THE TRAINED MODEL

DATA PREPROCESSING

```
In [1]:
           import pandas as pd
  In [2]:
            dataset = pd.read csv('../dataset/train dataset.csv', index col = 0)
  In [3]:
            dataset.head()
 Out[3]:
                                              Sudden
                                                                            Genital
                                                                                                              Delayed Partial
                                                                                                                               Muscle
              Age Gender Polyuria Polydipsia weight Weakness Polyphagia
                                                                                            Itching Irritability
                                                                                                               healing paresis stiffness
                                                                            thrush blurring
                                                 loss
                                 0
                                                                                          0
               40
                                           1
                                                                                                                                    1
                        1
                                 0
                                           0
                                                   0
                                                                         0
                                                                                 0
                                                                                         1
                                                                                                 0
                                                                                                                    0
                                                                                                                                     0
               58
                                 1
                                                   0
                                                                                                                                    1
               45
                                 1
                                           1
                                                   1
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                                                                         1
                                                                                 0
                                                                                         1
                                                                                                 1
                                                                                                                            1
                                                                                                                                    1
  In [4]:
            import numpy as np
  In [5]:
           x = dataset.iloc[ : , : 16].values
           y = dataset.iloc[ : , 16].values
 In [6]:
           from sklearn.model selection import train test split
            <u>v tosin v tost v tosin v tost - t</u>rain test_split(x, y, test_size = 0.15, random_state = 0)
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```

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x train.shape
  In [8]:
  Out[8]: (544, 16)
  In [9]:
           x test.shape
 Out[9]: (96, 16)
 In [10]:
           class 1 = np.count nonzero(y train == 1)
           class 0 = np.count nonzero(y train == 0)
 In [11]:
           print('Positive outcomes in training data =', class 1)
           print('Negative outcomes in training data =', class 0)
          Positive outcomes in training data = 271
          Negative outcomes in training data = 273
 In [12]:
           from sklearn.preprocessing import StandardScaler
           scaler = StandardScaler()
 In [13]:
           x train = scaler.fit transform(x train)
           x test = scaler.transform(x test)
 In [14]:
           x train[: 5]
 Out[14]: array([[-0.60659035, 0.66268653, -0.83666003, -0.78679579, 1.29099445,
                   -1.12127669, -0.80218734, -0.52345457, -0.80218734, -0.94629297,
                   -0.50917508, -0.8753478 , -0.7685332 , -0.74450367, -0.73557901,
                    2.26105275],
                  [-0.94322402, 0.66268653, -0.83666003, -0.78679579, -0.77459667,
                   -1.12127669, -0.80218734, -0.52345457, -0.80218734, -0.94629297,
                   -0.50917508, -0.8753478 , -0.7685332 , -0.74450367, -0.73557901,
                   -0.44227186],
                  [ 1.41321165, 0.66268653, -0.83666003, -0.78679579, -0.77459667,
                    0.89184053, 1.2465916, -0.52345457, 1.2465916, 1.05675518,
                    1.96396101, 1.14240306, -0.7685332, 1.34317672, 1.35947326,
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-1.12127669, -0.80218734, -0.52345457, -0.80218734, 1.05675518,
          -0.50917508, -0.8753478 , -0.7685332 , -0.74450367, 1.35947326,
          -0.44227186]])
     USING ARTIFICIAL NEURAL NETWORK (ANN)
In [15]:
      import tensorflow as tf
In [16]:
      model = tf.keras.models.Sequential()
In [17]:
      model.add(tf.keras.layers.Dense(16, input dim = 16, activation = 'relu'))
      model.add(tf.keras.layers.Dense(32, activation = 'relu'))
      model.add(tf.keras.layers.Dense(64, activation = 'relu'))
      model.add(tf.keras.layers.Dense(1, activation = 'sigmoid'))
In [18]:
      model.compile(optimizer = 'adam', loss = 'binary crossentropy', metrics = ['accuracy'])
In [19]:
      model.fit(x train, y train, batch size = 10, epochs = 100)
     Epoch 1/100
     Epoch 2/100
     Epoch 3/100
     Epoch 4/100
     Epoch 5/100
     Epoch 6/100
     Epoch 7/100
     Epoch 8/100
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                            - 0s 923us/step - loss: 0.0655 - accuracy: 0.9737
```

-1.12127669, -0.80218734, -0.52345457, -0.80218734, 1.05675518, -0.50917508, -0.8753478, -0.7685332, -0.74450367, 1.35947326,

[-0.01748144, 0.66268653, -0.83666003, -0.78679579, -0.77459667,

-0.44227186],

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	11/100
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	12/100
	[=====================================
•	13/100 [===================================
	14/100
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	15/100
	[=====================================
	16/100
	[=====================================
	17/100
	[=====================================
	18/100
55/55	[=====================================
Epoch	19/100
55/55	[=====================================
	20/100
55/55	[=====================================
	21/100
	[=====================================
	22/100
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	23/100
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	25/100 Last 96845/ston last 9 0113 postupative 9 0051
	[=====================================
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	27/100
	[=====================================
	28/100
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	29/100
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	30/100
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	31/100
•	[=====================================
Epoch	32/100
55/55	[=====================================
Epoch	33/100
	output/CommonHTML/fonts/TeX/fontdata.js] - 0s 889us/step - loss: 0.0094 - accuracy: 0.9948
Epoch	34/100

```
Epoch 35/100
 Epoch 36/100
 Epoch 37/100
 Epoch 38/100
 Epoch 39/100
 Epoch 40/100
 Epoch 41/100
 Epoch 42/100
 Epoch 43/100
 Epoch 44/100
 Epoch 45/100
 Epoch 46/100
 Epoch 47/100
 Epoch 48/100
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 Epoch 53/100
 Epoch 54/100
 Epoch 55/100
 Epoch 56/100
 Epoch 57/100
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Epoch	68/100
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Epoch	69/100
55/55	[=====================================
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	82/100
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Epoch	83/100

```
Epoch 84/100
 Epoch 85/100
 Epoch 86/100
 Epoch 87/100
 Epoch 88/100
 Epoch 89/100
 Epoch 90/100
 Epoch 91/100
 Epoch 92/100
 Epoch 93/100
 Epoch 94/100
 Epoch 95/100
 Epoch 96/100
 Epoch 97/100
 Epoch 98/100
 Epoch 99/100
 Epoch 100/100
 Out[19]: <tensorflow.python.keras.callbacks.History at 0x17a8b46c040>
In [20]:
  model.summary()
 Model: "sequential"
 Layer (type)
       Output Shape
            Param #
 ______
 dense (Dense)
       (None, 16)
            272
```

544

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```
dense_2 (Dense)
                                    (None, 64)
                                                            2112
         dense 3 (Dense)
                                    (None, 1)
         ______
         Total params: 2,993
         Trainable params: 2,993
         Non-trainable params: 0
In [21]:
         loss, accuracy = model.evaluate(x_test, y_test, verbose = 0)
In [22]:
         print("Model accuracy:", round(accuracy * 100, 2), "%")
         Model accuracy: 96.88 %
In [23]:
         y_pred = model.predict(x_test)
         y pred = y pred \Rightarrow= 0.5
In [24]:
         from sklearn.metrics import accuracy score
         print("Accuracy =", accuracy score(y test, y pred))
         Accuracy = 0.96875
        SAVING THE TRAINED MODEL
In [25]:
         import os.path
In [26]:
         if os.path.isfile('../model/project model.h5') is False:
             model.save('../model/project model.h5')
 In [ ]:
```