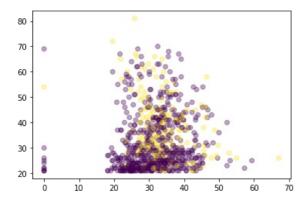
```
In [2]:
          #Given a dataset of various metrics can we predict if a patient has diabetes
          # Import libraries
          import pandas as pd
          import numpy as np
          import tensorflow as tf
          from tensorflow.keras.layers import Input, Dense
          from tensorflow.keras.models import Sequential
          from sklearn.model selection import train test split
          import matplotlib.pyplot as plt
          %matplotlib inline
In [3]:
          # Read the data
          data = pd.read_csv('./files/diabetes.csv')
          data.head()
            Number of times
                                               Diastolic blood
                                                              Triceps skin fold
                                                                              2-Hour serum
                                                                                                      Diabetes pedigree
                                                                                                                            Class
Out[3]:
                                Plasma glucose
                                                                                           Body mass
                                                                                                                      Age
                   pregnant
                                                                    thickness
                                                                                   insulin
                                                                                               index
                                                                                                                           variable
                                 concentration
                                                    pressure
                                                                                                              function
         0
                         6
                                                         72
                                                                          35
                                                                                                 33.6
                                                                                                                0.627
                                          148
                                                                                        0
                                                                                                                       50
                                                                                                                                1
                                                                          29
                                                                                        0
                                                                                                26.6
                                                                                                                0.351
                                                                                                                                0
         1
                         1
                                          85
                                                         66
                                                                                                                       31
         2
                                                                          0
                                                                                        0
                         8
                                          183
                                                         64
                                                                                                23.3
                                                                                                                0.672
                                                                                                                       32
                                                                                                                                1
         3
                                          89
                                                          66
                                                                          23
                                                                                       94
                                                                                                 28.1
                                                                                                                0.167
                                                                                                                       21
                                                                                                                                0
         4
                         0
                                                         40
                                                                          35
                                                                                      168
                                                                                                                2.288
                                          137
                                                                                                43.1
                                                                                                                       33
                                                                                                                                1
In [4]:
          #Check for data quality
          data.isna().sum()
        Number of times pregnant
                                            0
Out[4]:
         Plasma glucose concentration
                                            0
         Diastolic blood pressure
                                            0
         Triceps skin fold thickness
                                            0
         2-Hour serum insulin
                                            0
                                            0
         Body mass index
         Diabetes pedigree function
                                            0
                                            0
         Age
         Class variable
                                            0
         dtype: int64
In [5]:
          data.dtypes
                                              int64
         Number of times pregnant
         Plasma glucose concentration
                                              int64
         Diastolic blood pressure
                                              int64
         Triceps skin fold thickness
                                              int64
         2-Hour serum insulin
                                              int64
                                            float64
         Body mass index
         Diabetes pedigree function
                                            float64
         Age
                                              int64
         Class variable
                                              int64
         dtype: object
In [6]:
          # Create dataset
          X = data.iloc[:,:-1]
          y = data.iloc[:, -1]
In [7]:
          #Create training and test set
          X_train, X_test, y_train, y_test = train_test_split(X, y, random_state=42)
In [8]:
          #Calculate average accuracy for 10 runs
          accuracies = []
          for i in range(10):
            tf.random.set seed(i)
            model = Sequential()
            model.add(Dense(1, input dim=8, activation='sigmoid'))
            model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])
model.fit(X_train, y_train, epochs=1500, batch_size=100, verbose=0)
              accuracy = model.evaluate(X_test, y_test)
            accuracies append(accuracy*100)
```

```
from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero
2022-04-23 17:55:56.040391: I tensorflow/stream executor/cuda/cuda gpu executor.cc:939] successful NUMA node read
from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero
2022-04-23 17:55:56.040975: I tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:939] successful NUMA node read
from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero
2022-04-23 17:55:56.041420: I tensorflow/core/common runtime/gpu/gpu device.cc:1900] Ignoring visible gpu device
(device: 0, name: Quadro K1000M, pci bus id: 0000:01:00.0, compute capability: 3.0) with Cuda compute capability
3.0. The minimum required Cuda capability is 3.5.
2022-04-23 17:55:56.042651: I tensorflow/core/platform/cpu feature guard.cc:151] This TensorFlow binary is optimi
zed with oneAPI Deep Neural Network Library (oneDNN) to use the following CPU instructions in performance-critica
l operations: SSE4.1 SSE4.2 AVX
To enable them in other operations, rebuild TensorFlow with the appropriate compiler flags.
6/6 [====
                          ===] - 0s 1ms/step - loss: 0.5192 - accuracy: 0.7604
6/6 [============= ] - 0s 1ms/step - loss: 0.5285 - accuracy: 0.7552
           6/6 [===
6/6 [=
                           ==] - 0s 1ms/step - loss: 0.5206 - accuracy: 0.7448
6/6 [========== ] - 0s 2ms/step - loss: 0.5289 - accuracy: 0.7552
6/6 [============ ] - 0s 1ms/step - loss: 0.5219 - accuracy: 0.7448
6/6 [================== ] - 0s 1ms/step - loss: 0.5238 - accuracy: 0.7760
```

2022-04-23 17:55:55.990859: I tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:939] successful NUMA node read

```
sum(accuracies) / len(accuracies)
         75.15625
Out[9]:
In [10]:
          #Predict values
          y_pred = model.predict(X)
          y_pred = np.where(y_pred < .5, 0, 1)
In [11]:
          #Visualize correct vs incorrect predictions
          differ = np.abs(y.to_numpy() - y_pred.T)
          fig, ax = plt.subplots()
          ax.scatter(x=X['Body mass index'], y=X['Age'], c = differ, alpha = .35)
```

<matplotlib.collections.PathCollection at 0x7fea3d0d1b20> Out[11]:



In []:

In [9]:

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