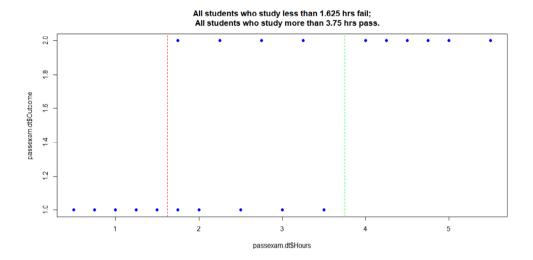
Exercise 8.2: CART Analysis for passexam.csv and passexam2.csv

1. In previous unit on Logistic Regression, we analyzed passexam.csv. Plotting the dataset shows at least two clear cut-offs for failing and passing the exam. Can CART detect the two cut-offs?



Solution:

```
Root node error: 10/20 = 0.5

n= 20

CP nsplit rel error xerror xstd
1 0.600000 0 1.0 1.6 0.17889
2 0.033333 1 0.4 0.6 0.20494
3 0.000000 8 0.1 0.9 0.22249
```

CP table reveals either 1 split or 8 splits (maximal tree).

Printing the maximal tree shows that node 3 and node 4 correctly identifies the two cutoffs.

```
n= 20
node), split, n, loss, yval, (yprob)
         * denotes terminal node
  1) root 20 10 0 (0.5000000 0.5000000)
2) Hours< 3.75 14 4 0 (0.7142857 0.2857143)
         4) Hours< 1.625 5 0 0 (1.0000000 0.0000000) *
5) Hours>=1.625 9 4 0 (0.5555556 0.4444444) 10) Hours>=3.375 1 0 0 (1.0000000 0.0000000)
                                      4 0 (0.5000000 0.5000000)
          11) Hours< 3.375 8
             22) Hours< 3.125
                                          3
                                             0 (0.5714286 0.4285714)
                44) Hours>=2.875 1 0 0 (1.0000000 0.0000000)
45) Hours< 2.875 6 3 0 (0.5000000 0.5000000)
                    (0.3000000 0.5000000)

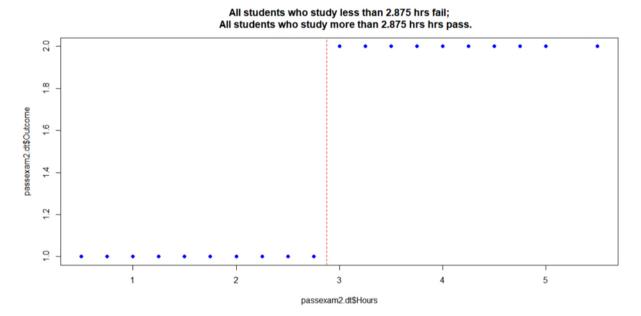
180) Hours>=2.375 1 0 0 (1.0000000 0.4000000)

181) Hours
                   90) Hours< 2.625 5
                                                   0 0 (1.0000000 0.0000000)
                     181) Hours< 2.375 4
                                                  2 0 (0.5000000 0.5000000)
                                                     1 0 (0.6666667 0.3333333)
0 1 (0.0000000 1.0000000)
                       362) Hours< 2.125 3
                       363) Hours>=2.125 1
             91) Hours>=2.625 1 0 1 (0.0000000 1.0000000)
23) Hours>=3.125 1 0 1 (0.0000000 1.0000000) *
      3) Hours>=3.75 6 0 1 (0.0000000 1.0000000) *
```

The rules found can be reconciled and simplified by rpart.rules() function.

```
Outcome
                 0
 nn
 4
          0
             Γ1.00
                     .001
                          when Hours
                                                           25%
             [1.00
[1.00
[1.00
10
          0
                                          3.4
                                                             5%
                     .00]
                          when Hours
                                               to 3.8
44
          0
                     .00]
                          when Hours
                                       is
                                          2.9
                                               to
                                                  3.1
                                       is 2.4
180
          0
                     .001
                          when Hours
                                               to
                                                  2.6
                                               to 2.1
          0
               .67
362
                     .33]
                          when Hours
                                          1.6
               .00
                                                  2.4
363
                    1.00]
                          when Hours
                                       is 2.1
                                               to
                                                             5%
               .00
                                                  2.9
                                                             5%
91
          1
                   1.001
                                       is 2.6
                          when Hours
                                               to
               .00
                   1.00]
                          when Hours is 3.1
                                                             5%
               .00
                   1.00]
                                                  3.8
                                                           30%
                          when Hours
```

2. Previously, we learnt that Logistic Regression fails on perfectly separable data passexam2.csv. Plotting the dataset shows at least one clear cut-off for failing and passing the exam. Can CART succeed in creating the model and detect the single cut-off?



Solution:

CART succeeds in finding the cut-off 2.875 hrs.

```
Root node error: 10/20 = 0.5

n= 20

CP nsplit rel error xerror xstd
1 1 0 1 1.6 0.178885
2 0 1 0 0.1 0.097468
```

```
> print(pass.cart2)
n= 20

node), split, n, loss, yval, (yprob)
   * denotes terminal node

1) root 20 10 0 (0.5000000 0.5000000)
   2) Hours< 2.875 10 0 0 (1.0000000 0.0000000) *
   3) Hours>=2.875 10 0 1 (0.0000000 1.0000000) *
```