1 Logistic Survey

Write down the following information:

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Degree: BSc Year of study: 3

Registered/auditting/crashing. NA

2.1 Summary Statistics

1. max_age: 71 min_age: 14

2. median_female: 25 median_male: 26

3. 10% 25% 50% 75% 90% 20 23 26 29 33

- 2.2 Data Visualization(Attached file at the end)
- 1.
- 2.
- 3.
- 3 Decision Trees and Cross-Validation
- 3.1

Plot attached at the end.

Accuracy when depth is 10: 1-0.38 = 0.62

3.2

Given that all the features are binary data, we claim that once a stump uses a feature, this feature will no longer appearing in it's children. For the first pass, we iterate through all the the D features, under each of the feature, we iterate through the n examples. Our goal is to find the feature that 'best' split the data. Even thought not every stump will need to go through every example, but the stump on the same level would go through N examples in total. Under worst case, there could be in total of N^*D operations at first stump. Then, at the second level of our decision tree, we iterate through the rest of (D-1) features and N examples, the total operations would be $N^*(D-1)$.

Follow this fashion, we could conclude that the cost of operations at each level has relationship with the depth.

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\begin{split} \text{depth} &= 0, \ cost = N^*(D\text{-}0) \\ \text{depth} &= 1, \ cost = N^* \ (D\text{-}1) \\ \text{depth} &= 2, \ cost = N^*(D\text{-}2) \\ \dots \\ \text{depth} &= m, \ cost = N^*(D\text{-}m) \\ \text{To sum up, } \ N[(D) + (D\text{-}1) + (D\text{-}2) + \dots + (D\text{-}m)] = (2D\text{-}M)(M\text{+}1)N/2 \\ \text{That is } O(NMD) \end{split}
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Attached file at the end

3.3 From cross-validation, we would choose depth 5.

Depth	Average_Error
1	0.1280
2	0.1278
3	0.1278
4	0.1320
5	0.1188
6	0.1258
7	0.1216
8	0.1218
9	0.1236
10	0.1262
11	0.1262
12	0.1262
13	0.1262
14	0.1262
15	0.1262

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4. Probability excersice
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4.1

Apply Bayes Rule:

P(D=1|T=1) = P(D=1)P(T=1|D=1)/P(T=1)

- = P(D=1)P(T=1|D=1)/P(T=1|D=1)P(D=1)+P(T=1|D=0)P(D=0)
- = 0.001*0.99/0.99*0.001+(1-0.99)*(1-0.001)
- = 0.09

4.2

Two sons

case1: Because the two results have same probability, we could list all the possible results.

- 1. head, tail
- 2. head, head

Then the probability is 1/2

case2: Same as before, list all the possibilities.

- 1. head tail
- 2. head head
- 3. tail head

Then the probability is 1/3

4.3

Prosecutor's fallacy

It is not possible to determine the probability that the person is guilty. We will reason from the info provided in the description.

Let us use G to represent the probability that a random choose person is guilty;

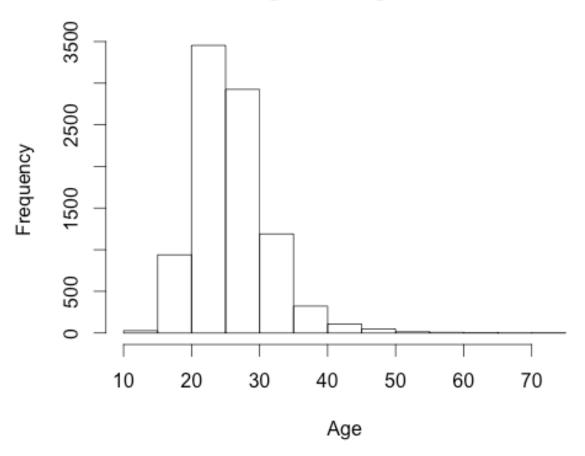
Let us use F to represent the probability that a person out of the innocent people match the footprint of the guilty person;

This question ask us to find the probability that given a person's footprint matches with the one at the crime scene, what is the probability that he is guilty?

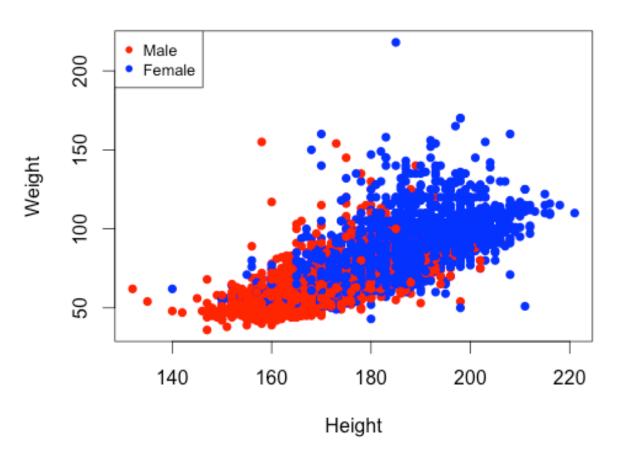
P(G|F) = P(F|G)*P(G)/P(F)

However, we do not know the probability that a random chosen people is guilty so therefore we could bot determine.

Histogram of age values



Height VS Weight



Boxplot of weight values for each age value

