CS105 Lab 10: Data Mining II

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R₀

Say I'm trying to classify whether people are coffee drinkers or not

| ID | State | Color | Coffee? |
|----|-------|---------|---------|
| 1 | MA | Red | Yes |
| 2 | NH | Red | No |
| 3 | MA | Magenta | Yes |

 R0 just ignores all input attributes and just looks at the output

R₀

Say I'm trying to classify whether people are coffee drinkers or not

| ID | State | Color | Coffee? |
|----|-------|-------|---------|
| _ | - | _ | Yes |
| - | - | _ | No |
| - | - | _ | Yes |

- R0 just ignores all input attributes and just looks at the output
- Which is the most common output?
- Always predict the most common output

We can do better than R0 ... R1!

| ID | State | Color | Coffee? |
|----|-------|---------|---------|
| 1 | MA | Red | Yes |
| 2 | NH | Red | No |
| 3 | MA | Magenta | Yes |

- For each value of each input attribute, find the most frequent class and create a rule
- Choose the rules with the higest accuracy

We can do better than R0 ... R1!

| ID | State | Color | Coffee? |
|----|-------|---------|---------|
| 1 | MA | Red | Yes |
| 2 | NH | Red | No |
| 3 | MA | Magenta | Yes |

• State: $MA \rightarrow Yes (2/2)$

 $NH \rightarrow No (1/1)$

Overall Accuracy = 3/3

We can do better than R0 ... R1!

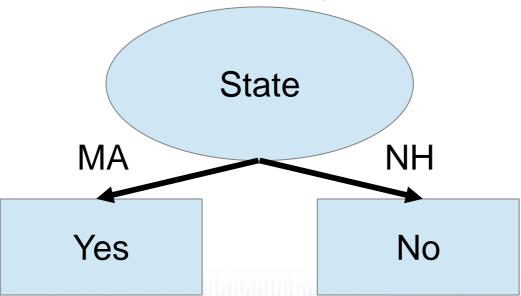
| ID | State | Color | Coffee? |
|----|-------|---------|---------|
| 1 | MA | Red | Yes |
| 2 | NH | Red | No |
| 3 | MA | Magenta | Yes |

Color: Magenta → Yes (1/1)
Red → No (1/2)

Overall Accuracy = 2/3

- State had an overall accuracy of 100%
- Color had an overall accuracy of 66.66%

So we create the following classifier:



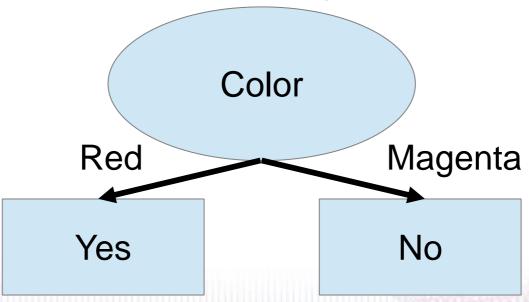
Building a Tree

| IC | | State | Color | | Coffee? | |
|--------|---------|--------------------|--------|------------|-------------------|--|
| 1 | | MA | Red | | Yes | |
| 2 | | NH | Red | | No | |
| 3 | | MA | Magent | ta | No | |
| 4 | | ME | Red | | Yes | |
| 5 | | ME | Magent | ta | No | |
| Color: | Red | \rightarrow Yes | (2/3) | Ov | erall Acc = 4/5 | |
| | Magenta | $a \rightarrow No$ | (2/2) | Go | odness = $4/5$ | |
| State: | MA | \rightarrow Yes | (1/2) | \bigcirc | Overall Acc = 3/5 | |
| N | NH | \rightarrow No | (1/1) | | odness = $3/10$ | |
| | ME | \rightarrow Yes | (1/2) | | | |

Building a Trees

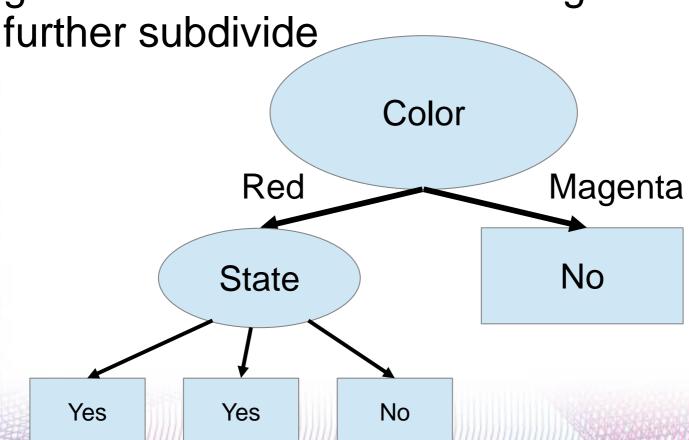
- State had an overall goodness of 30%
- Color had an overall goodness of 90%

So we create the following classifier:



Building a Trees

 The "Red" subgroup still had some inaccuracies, can go back, recalculate goodness scores for remaining attributes, and further subdivide



Wait ... what?

- We're just trained a model which now determines whether someone is a coffee drinker based on their favorite color....
- How does this make sense?

 There isn't enough training data and there aren't enough input attributes to 'filter' out the color attribute

Awesome Segue

- Choosing a data set is very important
- You need to have enough information to filter out noise (irrelevant information/anomalies)
- You need a problem that fits your dataset and a dataset that fits your problem
 - I can't model a generic person if I only have data on people from a specific place
 - If I only have data on people from a specific place, I shouldn't choose to try and model a generic person

Lab

 Nothing to submit, but I recommend you work through the questions and ask for help

You may also use lab time to work on your proposals

• Questions?