

# **Predicting with trees**

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# **Key ideas**

- · Iteratively split variables into groups
- · Split where maximally predictive
- Evaluate "homogeneity" within each branch
- Fitting multiple trees often works better (forests)

#### Pros:

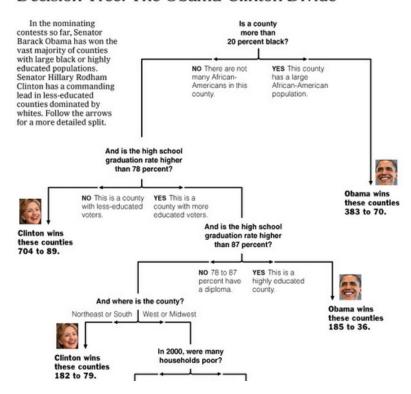
- · Easy to implement
- Easy to interpret
- Better performance in nonlinear settings

#### Cons:

- Without pruning/cross-validation can lead to overfitting
- Harder to estimate uncertainty
- · Results may be variable

# **Example Tree**

#### Decision Tree: The Obama-Clinton Divide



http://graphics8.nytimes.com/images/2008/04/16/us/0416-nat-subOBAMA.jpg

### **Basic algorithm**

- 1. Start with all variables in one group
- 2. Find the variable/split that best separates the outcomes
- 3. Divide the data into two groups ("leaves") on that split ("node")
- 4. Within each split, find the best variable/split that separates the outcomes
- 5. Continue until the groups are too small or sufficiently "pure"

### Measures of impurity

$$\hat{p}_{mk} = \frac{1}{N_m} \sum_{x_i \text{ in Leaf } m} \mathbb{1}(y_i = k)$$

**Misclassification Error**:

$$1 - \hat{p}_{mk(m)}$$

Gini index:

$$\sum_{k \neq k'} \hat{p}_{mk} \times \hat{p}_{mk'} = \sum_{k=1}^{K} \hat{p}_{mk} (1 - \hat{p}_{mk})$$

**Cross-entropy or deviance:** 

$$-\sum_{k=1}^K \hat{p}_{mk} \ln \hat{p}_{mk}$$

### **Example: Iris Data**

```
data(iris)
names(iris)
```

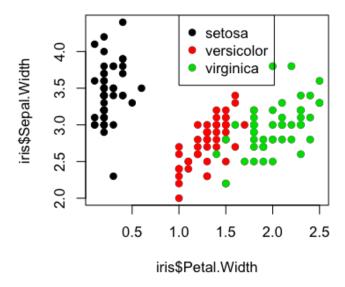
```
[1] "Sepal.Length" "Sepal.Width" "Petal.Length" "Petal.Width" "Species"
```

```
table(iris$Species)
```

```
setosa versicolor virginica
50 50 50
```

### Iris petal widths/sepal width

```
plot(iris$Petal.Width,iris$Sepal.Width,pch=19,col=as.numeric(iris$Species))
legend(1,4.5,legend=unique(iris$Species),col=unique(as.numeric(iris$Species)),pch=19)
```



# Iris petal widths/sepal width

```
# An alternative is library(rpart)
library(tree)
tree1 <- tree(Species ~ Sepal.Width + Petal.Width,data=iris)
summary(tree1)</pre>
```

```
Classification tree:

tree(formula = Species ~ Sepal.Width + Petal.Width, data = iris)

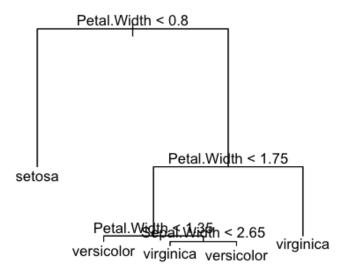
Number of terminal nodes: 5

Residual mean deviance: 0.204 = 29.6 / 145

Misclassification error rate: 0.0333 = 5 / 150
```

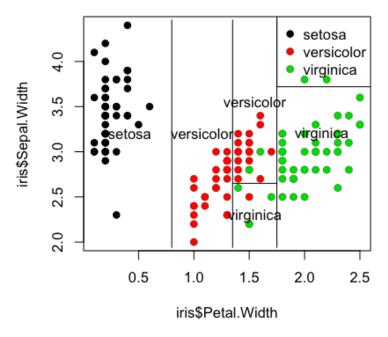
#### Plot tree

```
plot(tree1)
text(tree1)
```



# Another way of looking at a CART model

```
plot(iris$Petal.Width,iris$Sepal.Width,pch=19,col=as.numeric(iris$Species))
partition.tree(tree1,label="Species",add=TRUE)
legend(1.75,4.5,legend=unique(iris$Species),col=unique(as.numeric(iris$Species)),pch=19)
```



Predicting with trees

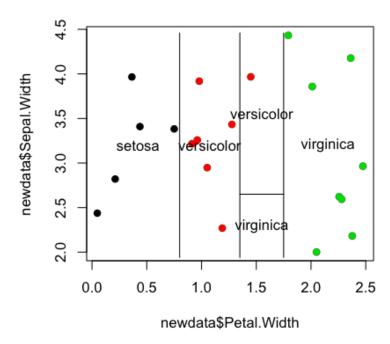
### **Predicting new values**

```
set.seed(32313)
newdata <- data.frame(Petal.Width = runif(20,0,2.5), Sepal.Width = runif(20,2,4.5))
pred1 <- predict(tree1, newdata)
pred1</pre>
```

```
setosa versicolor virginica
             0.02174
        0
1
                        0.97826
2
             0.02174
                        0.97826
        0
3
             0.00000
                        0.00000
        1
             1.00000
                        0.00000
4
        0
             0.02174
5
        0
                        0.97826
             0.02174
6
        0
                        0.97826
7
        0
             0.02174
                        0.97826
             0.90476
                        0.09524
8
        0
9
             1.00000
                        0.00000
        0
10
        0
             0.02174
                        0.97826
11
        0
             1.00000
                        0.00000
12
        1
             0.00000
                        0.00000
             0.00000
                        0.00000
13
        1
14
             0.00000
                        0.00000
        1
             0.02174
15
        0
                        0.97826
16
             0.02174
                        0.97826
        0
                                                                                                   11/18
17
             1.00000
                        0.00000
        0
```

### Overlaying new values

```
pred1 <- predict(tree1,newdata,type="class")
plot(newdata$Petal.Width,newdata$Sepal.Width,col=as.numeric(pred1),pch=19)
partition.tree(tree1,"Species",add=TRUE)</pre>
```

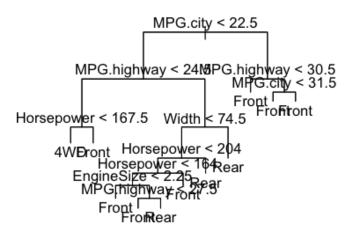


# Pruning trees example: Cars

data(Cars93,package="MASS")
head(Cars93)

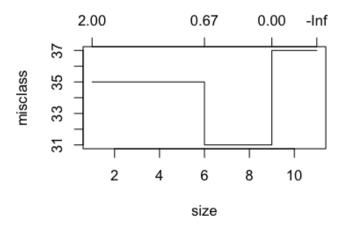
	Manufacturer	Model	Type	Min.Price	Price Ma	x.Price	MPG.ci	ty MPG.highw	ay	Aiı	rBags
1	Acura	Integra	Small	12.9	15.9	18.8		25	31		None
2	Acura	Legend	Midsize	29.2	33.9	38.7		18	25 Driv	er & Passe	enger
3	Audi	90	Compact	25.9	29.1	32.3	2	20	26	Driver	only
4	Audi	100	Midsize	30.8	37.7	44.6	:	19	26 Driv	er & Passe	enger
5	BMW	535i	Midsize	23.7	30.0	36.2		22	30	Driver	only
6	Buick	Century	Midsize	14.2	15.7	17.3		22	31	Driver	only
DriveTrain Cylinders EngineSize Horsepower RPM Rev.per.mile Man.trans.avail Fuel.tank.capacit											
1	Front	4	1	. 8	140 6300	)	2890		Yes		13.2
2	Front	6	3	3.2	200 5500	)	2335		Yes		18.0
3	Front	6	2	2.8	172 5500	)	2280		Yes		16.9
4	Front	6	2	2.8	172 5500	)	2535		Yes		21.
5	Rear	4	3	8.5	208 5700	)	2545		Yes		21.3
6	Front	4	2	2.2	110 5200	)	2565		No		16.4
	Passengers Le	ength Whe	eelbase W	didth Turn	.circle F	Rear.seat	t.room 1	Luggage. <mark>room</mark>	Weight	Origin	
1	5	177	102	68	37		26.5	11	2705	non-USA	
2	5	195	115	71	38		30.0	15	3560	non-USA	
3	5	180	102	67	37		28.0	14	3375	non-USA	
4	6	193	106	70	37		31.0	17	3405	non-USA	
5	4	186	109	69	39		27.0	13	3640	non-USA	13/18

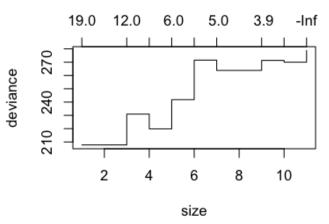
#### **Build a tree**



#### **Plot errors**

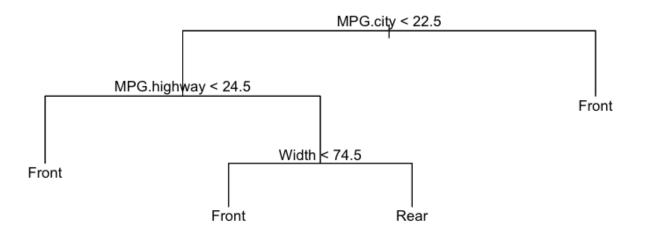
```
par(mfrow=c(1,2))
plot(cv.tree(treeCars,FUN=prune.tree,method="misclass"))
plot(cv.tree(treeCars))
```





#### Prune the tree

```
pruneTree <- prune.tree(treeCars,best=4)
plot(pruneTree)
text(pruneTree)</pre>
```



#### **Show resubstitution error**\*

```
table(Cars93$DriveTrain,predict(pruneTree,type="class"))
table(Cars93$DriveTrain,predict(treeCars,type="class"))
```

```
4WD Front Rear

4WD 5 5 0

Front 2 61 4

Rear 0 3 13
```

· Note that cross validation error is a better measure of test set accuracy

#### Notes and further resources

- · Hector Corrada Bravo's Notes, code
- · Cosma Shalizi's notes
- Elements of Statistical Learning
- Classification and regression trees
- Random forests