

Lab7

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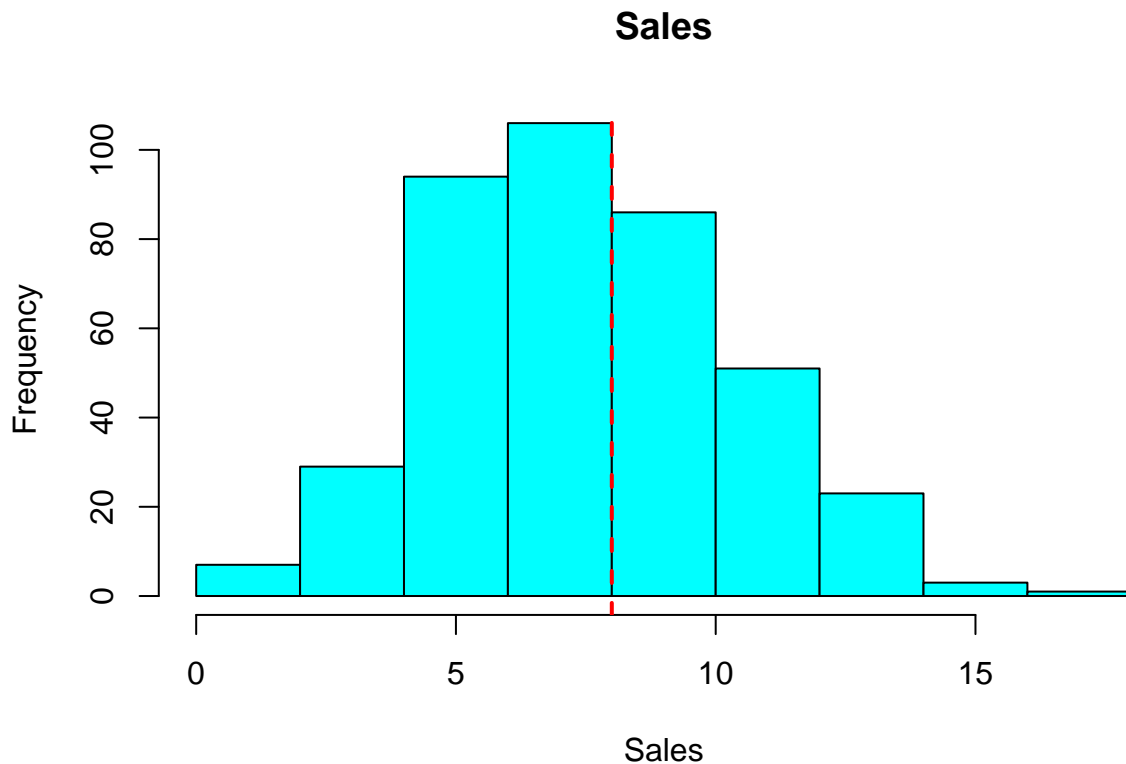
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8.3 Lab: Decision Trees Page - 323

8.3.1 Fitting classification Trees

We'll use Carseats data set. An additional variable “High” is created as a logical vector. High being “Yes”, when sales is > 8 .

```
data("Carseats")
#skimr::skim(Carseats)
hist(Carseats$Sales,xlab = "Sales", main = "Sales", col = c("#00FFFF"))
abline(v=8,col= "red", lty = 2, lwd = 2)
```



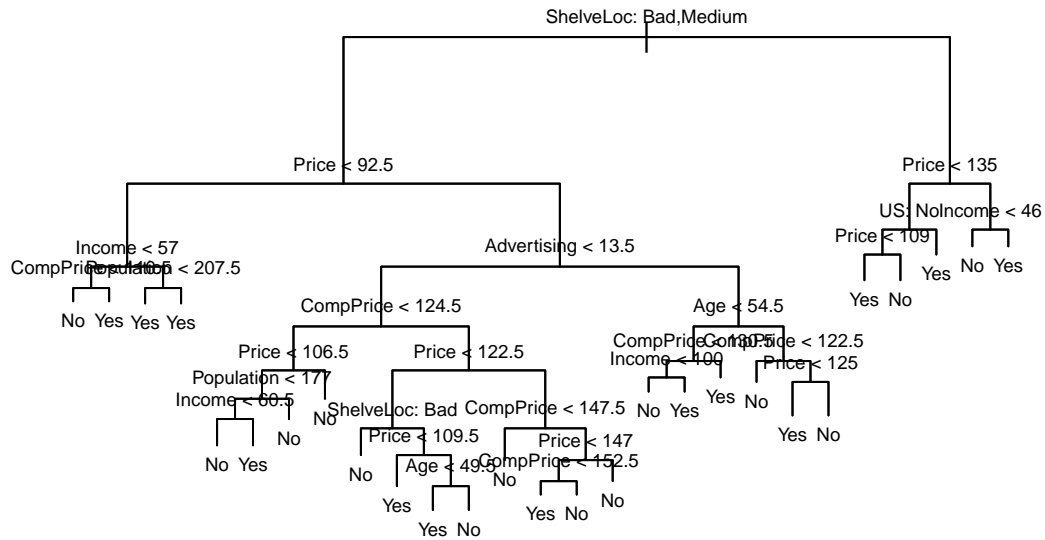
```
Carseats <- Carseats %>% dplyr::mutate(High = as.factor(if_else(Sales <= 8, "No", "Yes")))
```

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We'll use a `tree()` to fit a classification tree to predict if the “High” variable is “Yes” or “No”

```
set.seed(10)
tree.carseats <- tree(High ~ . , data = Carseats[, -1])
```

```
plot(tree.carseats)
text(tree.carseats,pretty=0,cex = .6)
```



```
summary(tree.carseats)

##
## Classification tree:
## tree(formula = High ~ ., data = Carseats[, -1])
## Variables actually used in tree construction:
## [1] "ShelveLoc" "Price" "Income" "CompPrice" "Population"
## [6] "Advertising" "Age" "US"
## Number of terminal nodes: 27
## Residual mean deviance: 0.4575 = 170.7 / 373
## Misclassification error rate: 0.09 = 36 / 400
```

What does rpart do?

Why is there a difference between misclassification between rpart and tree? Also notice the number of terminal nodes (of course the rpart's terminal node collapses the Yes and No into a stacked bar... but still look at the difference)

```
set.seed(10)
rpartTree <- rpart::rpart(High ~ ., data = Carseats[, -1])
plot(partykit::as.party(rpartTree), gp = gpar(fontsize = 6))
```



```
##
##      'Positive' Class : Yes
##
```

Let's predict outcome of `tree()` with `predict()`

The misclassifications are different between `tree()` and `rpart()`. What makes the difference?

```
tree.carseats.Pred <- predict(tree.carseats,newdata = Carseats[,-1])
tree.carseats.Pred <- as.factor(if_else(tree.carseats.Pred[,2] > 0.5,"Yes","No"))
caret::confusionMatrix(tree.carseats.Pred,Carseats$High, positive = "Yes")
```

```
## Confusion Matrix and Statistics
##
##      Reference
## Prediction  No Yes
##      No  213  13
##      Yes   23 151
##
##      Accuracy : 0.91
##      95% CI : (0.8776, 0.9362)
##      No Information Rate : 0.59
##      P-Value [Acc > NIR] : <2e-16
##
##      Kappa : 0.8157
##      Mcnemar's Test P-Value : 0.1336
##
##      Sensitivity : 0.9207
##      Specificity : 0.9025
##      Pos Pred Value : 0.8678
##      Neg Pred Value : 0.9425
##      Prevalence : 0.4100
##      Detection Rate : 0.3775
##      Detection Prevalence : 0.4350
##      Balanced Accuracy : 0.9116
##
##      'Positive' Class : Yes
##
```