

Chapter 4: Resampling Methods

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```
# Libraries
library(ISLR)

# The Validation Set Approach
attach(Auto)
set.seed(1)
train=sample(392,196)
lm.fit=lm(mpg~horsepower,data=Auto,subset=train)
mean((mpg-predict(lm.fit,Auto))[-train]^2)

## [1] 26.14142

lm.fit2=lm(mpg~poly(horsepower,2),data=Auto,subset=train)
mean((mpg-predict(lm.fit2,Auto))[-train]^2)

## [1] 19.82259

lm.fit3=lm(mpg~poly(horsepower,3),data=Auto,subset=train)
mean((mpg-predict(lm.fit3,Auto))[-train]^2)

## [1] 19.78252

set.seed(2)
train=sample(392,196)
lm.fit=lm(mpg~horsepower,subset=train)
mean((mpg-predict(lm.fit,Auto))[-train]^2)

## [1] 23.29559

lm.fit2=lm(mpg~poly(horsepower,2),data=Auto,subset=train)
mean((mpg-predict(lm.fit2,Auto))[-train]^2)

## [1] 18.90124

lm.fit3=lm(mpg~poly(horsepower,3),data=Auto,subset=train)
mean((mpg-predict(lm.fit3,Auto))[-train]^2)

## [1] 19.2574

# Leave-One-Out Cross-Validation

glm.fit=glm(mpg~horsepower,data=Auto)
coef(glm.fit)

## (Intercept) horsepower
## 39.9358610 -0.1578447

lm.fit=lm(mpg~horsepower,data=Auto)
coef(lm.fit)

## (Intercept) horsepower
## 39.9358610 -0.1578447
```

```

library(boot)
glm.fit=glm(mpg~horsepower,data=Auto)
cv.err=cv.glm(Auto,glm.fit)
cv.err$delta

## [1] 24.23151 24.23114

cv.error=rep(0,5)
for (i in 1:5){
  glm.fit=glm(mpg~poly(horsepower,i),data=Auto)
  cv.error[i]=cv.glm(Auto,glm.fit)$delta[1]
}
cv.error

## [1] 24.23151 19.24821 19.33498 19.42443 19.03321
# k-Fold Cross-Validation

set.seed(17)
cv.error.10=rep(0,10)
for (i in 1:10){
  glm.fit=glm(mpg~poly(horsepower,i),data=Auto)
  cv.error.10[i]=cv.glm(Auto,glm.fit,K=10)$delta[1]
}
cv.error.10

## [1] 24.20520 19.18924 19.30662 19.33799 18.87911 19.02103 18.89609
## [8] 19.71201 18.95140 19.50196
# The Bootstrap

alpha.fn=function(data,index){
  X=data$X[index]
  Y=data$Y[index]
  return((var(Y)-cov(X,Y))/(var(X)+var(Y)-2*cov(X,Y)))
}
alpha.fn(Portfolio,1:100)

## [1] 0.5758321

set.seed(1)
alpha.fn(Portfolio,sample(100,100,replace=T))

## [1] 0.5963833

boot(Portfolio,alpha.fn,R=1000)

##
## ORDINARY NONPARAMETRIC BOOTSTRAP
##
##
## Call:
## boot(data = Portfolio, statistic = alpha.fn, R = 1000)
##
##
## Bootstrap Statistics :
##      original      bias    std. error
## t1* 0.5758321 -7.315422e-05 0.08861826

```

```
# Estimating the Accuracy of a Linear Regression Model
```

```
boot.fn=function(data,index)
  return(coef(lm(mpg~horsepower,data=data,subset=index)))
boot.fn(Auto,1:392)
```

```
## (Intercept)  horsepower
## 39.9358610   -0.1578447
```

```
set.seed(1)
boot.fn(Auto,sample(392,392,replace=T))
```

```
## (Intercept)  horsepower
## 38.7387134   -0.1481952
```

```
boot.fn(Auto,sample(392,392,replace=T))
```

```
## (Intercept)  horsepower
## 40.0383086   -0.1596104
```

```
boot(Auto,boot.fn,1000)
```

```
##
## ORDINARY NONPARAMETRIC BOOTSTRAP
##
## Call:
## boot(data = Auto, statistic = boot.fn, R = 1000)
##
##
## Bootstrap Statistics :
##      original      bias    std. error
## t1* 39.9358610  0.02972191 0.860007896
## t2* -0.1578447 -0.00030823 0.007404467
```

```
summary(lm(mpg~horsepower,data=Auto))$coef
```

```
##           Estimate Std. Error  t value    Pr(>|t|)
## (Intercept) 39.9358610 0.717498656  55.65984 1.220362e-187
## horsepower  -0.1578447 0.006445501 -24.48914 7.031989e-81
```

```
boot.fn=function(data,index)
  coefficients(lm(mpg~horsepower+I(horsepower^2),data=data,subset=index))
set.seed(1)
boot(Auto,boot.fn,1000)
```

```
##
## ORDINARY NONPARAMETRIC BOOTSTRAP
##
## Call:
## boot(data = Auto, statistic = boot.fn, R = 1000)
##
##
## Bootstrap Statistics :
##      original      bias    std. error
## t1* 56.900099702 6.098115e-03 2.0944855842
## t2* -0.466189630 -1.777108e-04 0.0334123802
```

```
## t3* 0.001230536 1.324315e-06 0.0001208339
```

```
summary(lm(mpg~horsepower+I(horsepower^2),data=Auto))$coef
```

##	Estimate	Std. Error	t value	Pr(> t)
## (Intercept)	56.900099702	1.8004268063	31.60367	1.740911e-109
## horsepower	-0.466189630	0.0311246171	-14.97816	2.289429e-40
## I(horsepower^2)	0.001230536	0.0001220759	10.08009	2.196340e-21