

## Class Notes

### Statistical Computing & Machine Learning

#### Class 15

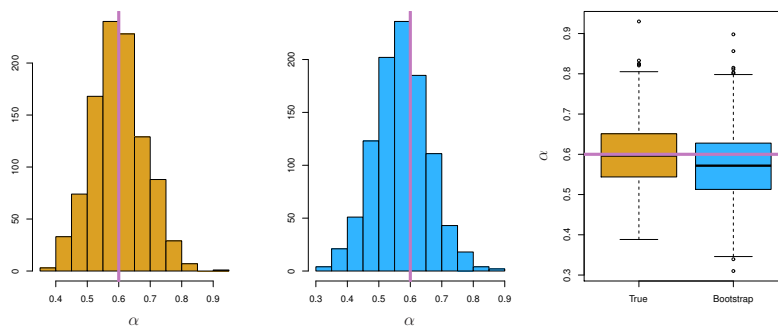
#### Programming

Functionals. How `optim()` allows you to pass the various other arguments, e.g. `data=` to the function being optimized.

The meaning of ...

Solvers

#### Bootstrapping



ISLR Figure 5.10

In-class demonstration. How many cases don't get used in a typical resample?

```
cases <- 1:10000
ntrials <- 50
result <- numeric(ntrials)
for (k in 1:50) {
  bootstrap_sample <- sample(cases, size=length(cases), replace = TRUE)
  result[k] <- length(setdiff(cases, unique(bootstrap_sample))) / length(cases)
}
result

## [1] 0.3678 0.3680 0.3675 0.3714 0.3649
## [6] 0.3615 0.3684 0.3686 0.3613 0.3669
## [11] 0.3648 0.3684 0.3705 0.3651 0.3687
## [16] 0.3694 0.3628 0.3674 0.3673 0.3647
## [21] 0.3694 0.3720 0.3681 0.3719 0.3622
## [26] 0.3666 0.3706 0.3648 0.3722 0.3737
## [31] 0.3700 0.3668 0.3666 0.3691 0.3696
## [36] 0.3666 0.3641 0.3700 0.3663 0.3715
```

```
## [41] 0.3646 0.3692 0.3662 0.3627 0.3712
## [46] 0.3656 0.3677 0.3639 0.3672 0.3689

## Or ...
(1 - 1/length(cases))^length(cases)

## [1] 0.367861
```

### Programming a bootstrap calculation

```
bootstrap <- function(formula, method, data, statistic, reps=10) {

}
```

### Mosaic software

Using `resample()` and `do()`

### Model Selection

Problem: Given a set of potential model terms, choose the best subset.

Two issues:

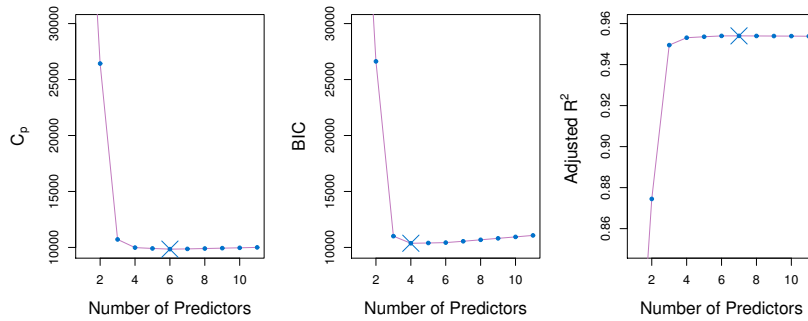
1. What does “best” mean?
2. How to optimize?

### Best

In-sample	Adjusted	Out-of-sample
$\frac{1}{n} \text{RSS}$	$C_p = \frac{1}{n}(\text{RSS} + 2d\hat{\sigma}^2)$	cross-validated prediction error
.	$\text{AIC} = -2 \ln \mathcal{L} - 2d$	.
.	$\text{AIC}_{ls} = \frac{C_p}{\hat{\sigma}^2}$	.
$\text{BIC} = \frac{1}{n}(\text{RSS} + \ln(n)d\hat{\sigma}^2)$	.	.
$R^2$	Adjusted $R^2$	???

$$\text{Adjusted } R^2 = 1 - \frac{\text{RSS}/(n-d-1)}{\text{TSS}/(n-1)}$$

ISLR Figure 6.2. Note that the values on the vertical axis are the best for that “number of predictors.”



### Uncertainty

Repeat the analysis for different test sets or using different folds in k-fold cross validation.

- At each value of “Number of Predictors”, there will be a distribution.
- *One-standard-error rule*: select the smallest model for which the estimated test error is within one standard error of the lowest point on the curve.

### Daily activity

Write a k-fold cross validator. Day-13-Programming-Tasks