

Ch2 Regression

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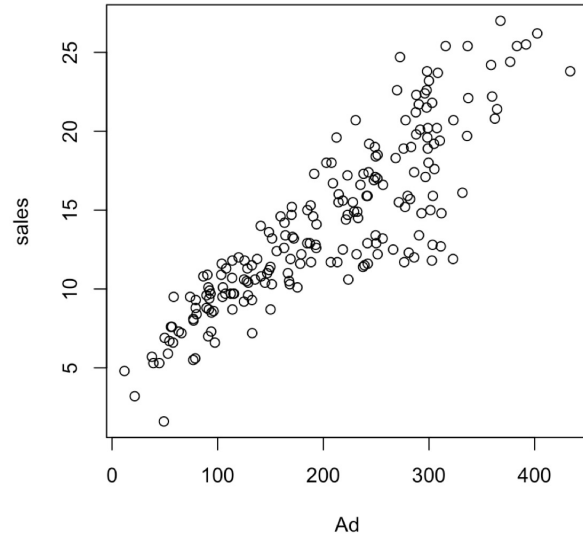
A Subsection

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A.1 Simple Linear Regression (Advertising Data)

Advertising.csv from ISLR web site.

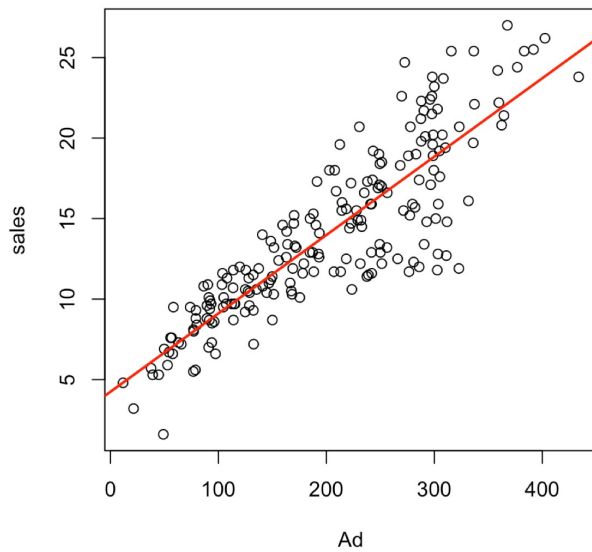
$\text{Ad} = \text{TV} + \text{radio} + \text{newspaper}$ (Total Spending)



A.2 Questions

1. Is there a relationship between advertising budget and sales?
2. If so, what is the form of the relationship?
3. How strong is the relationship between advertising budget and sales?
4. How accurately can we estimate the effect?
5. How accurately can we predict future sales?

A.3 Regression



A.4 Regression Approach 1

$$Y = f(X) + \epsilon$$

- Assumes that $f()$ is linear
- Estimate parameters β_0 and β_1 based on

$$\text{RSS} = \sum_{i=1}^n (Y_i - \hat{Y})^2$$

- Formula for $\hat{\beta}_1$ and $\hat{\beta}_0$:

$$\hat{\beta}_1 = rs_y/s_x \qquad \hat{\beta}_0 = \bar{y} - \hat{\beta}_1\bar{x}$$

A.5 Regression Approach 2

- Assumes that

$$Y = \beta_0 + \beta_1 X + \epsilon, \quad \epsilon \sim N(0, \sigma^2)$$

- Estimate parameters β_0 and β_1 with best estimators possible. (unbiased, minimum variance)
- Formula for $\hat{\beta}_1$ and $\hat{\beta}_0$:

$$\hat{\beta}_1 = rs_y/s_x \quad \hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$$

```

Call:
lm(formula = sales ~ Ad)

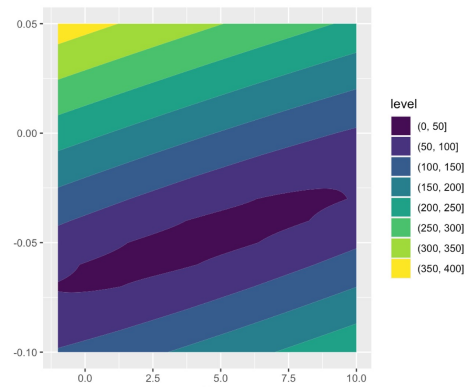
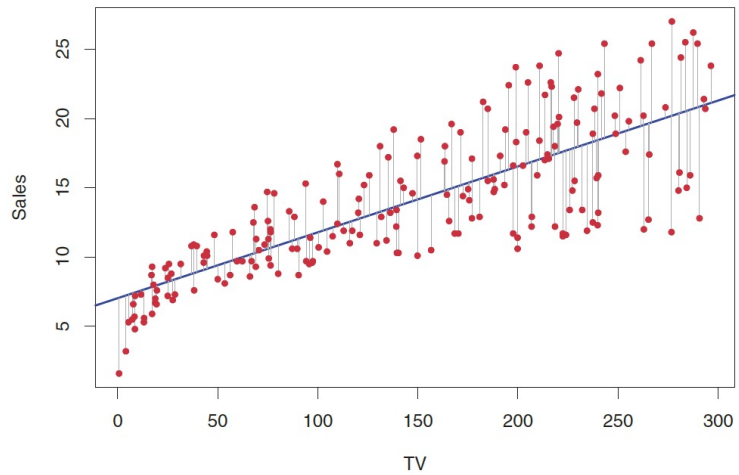
Residuals:
    Min       1Q   Median       3Q      Max
-8.0546 -1.3071  0.1173  1.5961  7.1895

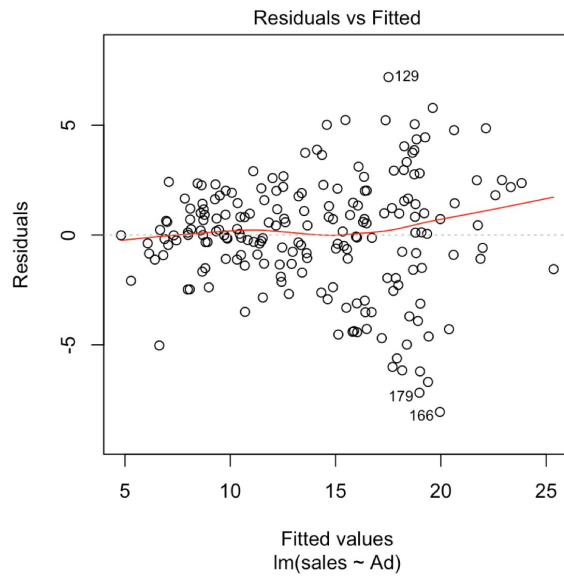
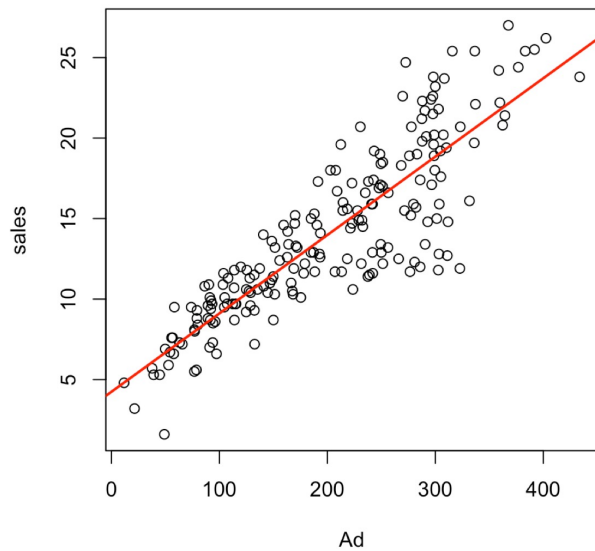
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 4.243028   0.438525   9.676  <2e-16 ***
Ad          0.048688   0.001982  24.564  <2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.6 on 198 degrees of freedom
Multiple R-squared:  0.7529,    Adjusted R-squared:  0.7517
F-statistic: 603.4 on 1 and 198 DF,  p-value: < 2.2e-16

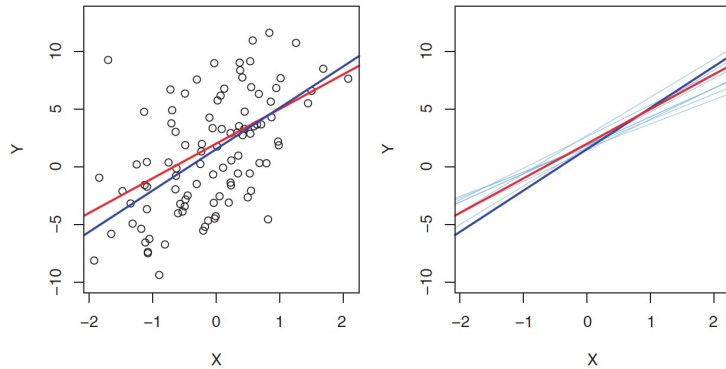
```

- is assumption true?
- how off is the estimator?
- how accurate is the prediction?





A.6 Simulation Under perfect case



```
X = rnorm(30, 3, 5)
Y = 4+1.5* X + rnorm(30, 0, 5)

plot(X, Y, xlim=c(-6, 11), ylim=c(-10, 25))
abline(a=4, b=1.5, col="blue", lwd=2)
```

```
m1 <- lm(Y~X)
abline(m1, col="red")
```

A.7 Multiple Linear Regression (Adv data)

Multiple Linear Regression (Adv data)

A.8 MLR

- Want to guess the next Y as accurate as possible

$$\text{sales} = \beta_0 + \beta_1 \text{TV} + \beta_2 \text{radio} + \beta_3 \text{newspaper} + \epsilon, \quad \epsilon \sim N(0, \sigma^2)$$

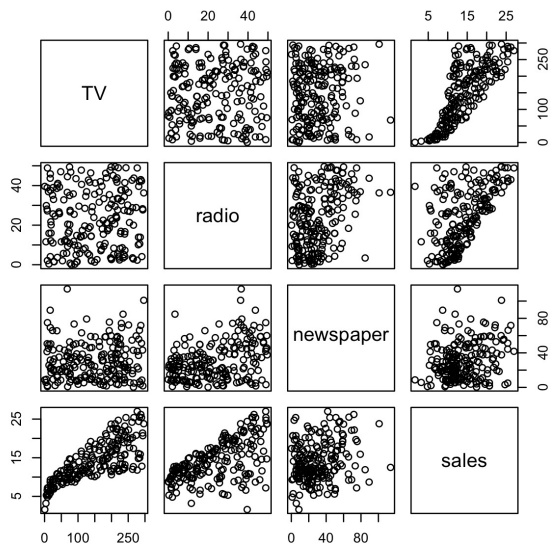
- Estimate parameters by minimizing

$$\text{RSS} = \sum_{i=1}^n (Y_i - \hat{Y})^2$$

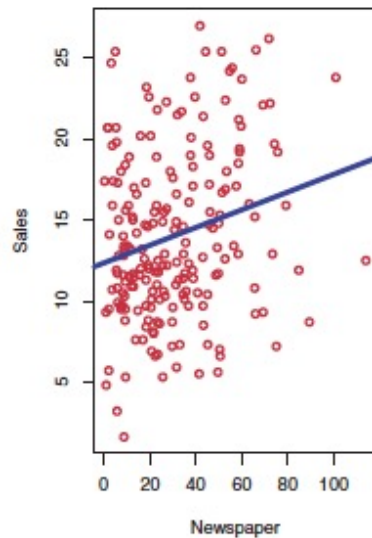
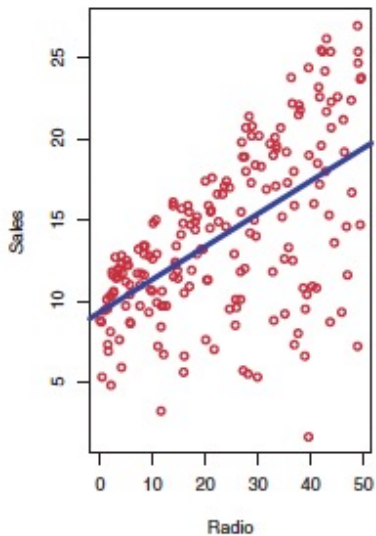
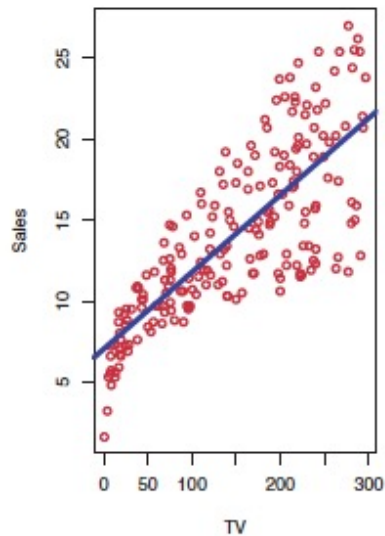
- Formula for $\hat{\boldsymbol{\beta}} = (\beta_0, \beta_1, \beta_2, \beta_3)'$:

$$\hat{\boldsymbol{\beta}} = (\mathbf{X}'\mathbf{X})^{-1}\mathbf{X}'\mathbf{Y}$$

A.9 Pairs()



A.10 Last row



```
Model2 <- lm(sales ~ TV + radio + newspaper)
summary(Model2)
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	2.938889	0.311908	9.422	<2e-16	***
TV	0.045765	0.001395	32.809	<2e-16	***
radio	0.188530	0.008611	21.893	<2e-16	***
newspaper	-0.001037	0.005871	-0.177	0.86	

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 1.686 on 196 degrees of freedom

Multiple R-squared: 0.8972, Adjusted R-squared: 0.8956

F-statistic: 570.3 on 3 and 196 DF, p-value: < 2.2e-16

A.11 Without Newspaper

```
lm(formula = sales ~ TV + radio)
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	2.92110	0.29449	9.919	<2e-16 ***
TV	0.04575	0.00139	32.909	<2e-16 ***
radio	0.18799	0.00804	23.382	<2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

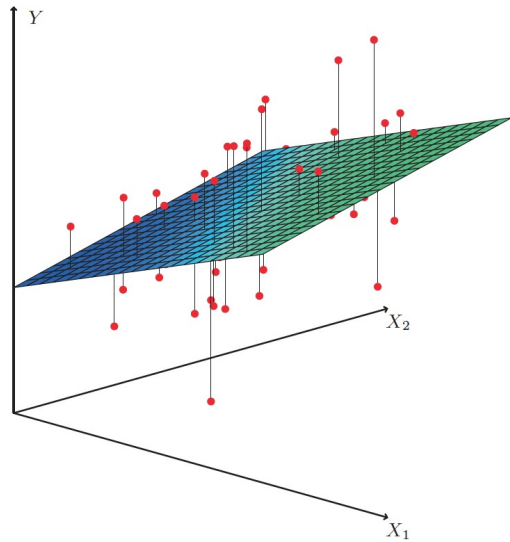
Residual standard error: 1.681 on 197 degrees of freedom

Multiple R-squared: 0.8972, Adjusted R-squared: 0.8962

F-statistic: 859.6 on 2 and 197 DF, p-value: < 2.2e-16

A.12 With only TV and Radio as predictor

It's like fitting plane in 3-d space



A.13 The Marketing Questions (3.4)

1. Is there a relationship between advertising sales and budget?
2. How strong is the relationship?
3. Is it important to advertise in newspaper?
4. Is all predictor important, or just a subset?
5. Which media contribute to most to the sales? How much?
6. How accurately can we predict future sales?
7. Is the relationship linear?
8. Is there synergy among the advertising media?

A.14 1 At least one X useful?

- In SLR, we only need to test $\beta_1 = 0$.
- Now we have to test $\beta_1 = \beta_2 = \beta_3 = 0$.
- Use F -statistic

$$F = \frac{(TSS - RSS)/p}{RSS/(n - p - 1)}$$

```
summary(Model2)
```

```
# Residual standard error: 1.686 on 196 degrees of freedom  
# Multiple R-squared:  0.8972, Adjusted R-squared:  0.8956  
# F-statistic: 570.3 on 3 and 196 DF,  p-value: < 2.2e-16
```


- We can test SUBSET of parameters ($\beta_2 = \beta_3 = 0$) by

$$F = \frac{(RSS_0 - RSS)/q}{RSS/(n - p - 1)}$$

Where RSS_0 is the RSS from the model using $\beta_2 = \beta_3 = 0$, and q is the number of suppressed parameters.

- Why test as a whole when you can do the t-test individually? (important when p is large)

A.15 2 How good is Model Fit?

- Coefficient of Determination

$$\text{TSS} = \sum_{i=1}^n (Y_i - \bar{Y})^2, \quad \text{RSS} = \sum_{i=1}^n (Y_i - \hat{Y})^2, \quad R^2 = 1 - \frac{\text{RSS}}{\text{TSS}}$$

- $R^2 = .89719$ without newspaper
- $R^2 = .8972$ with newspaper
- MSE (RSE in ISLR) estimates σ^2 and represents irreducible error.
- With p predictors,

$$MSE = \sqrt{\frac{1}{n - p - 1} \text{RSS}}$$

A.16 3 newspaper? (Confounding Effect)

```
Model3 <- lm(sales ~ newspaper)
summary(Model3)
```

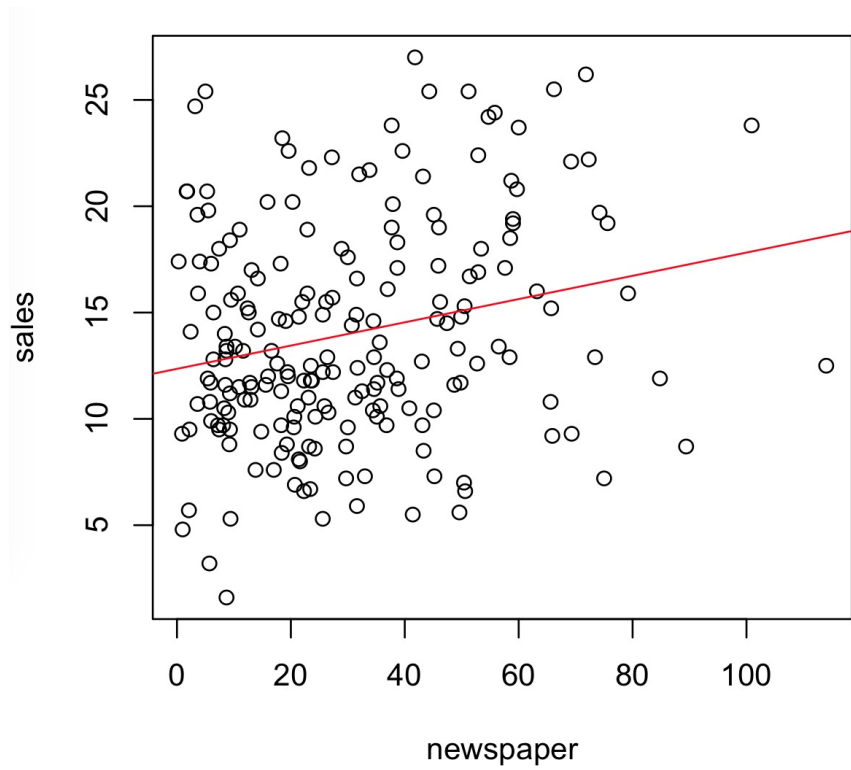
Coefficients:

```
#           Estimate Std. Error t value Pr(>|t|)
#(Intercept) 12.35141    0.62142   19.88 < 2e-16 ***
#newspaper    0.05469    0.01658    3.30 0.00115 **
#---
#Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
#
#Residual standard error: 5.092 on 198 degrees of freedom
#Multiple R-squared:  0.05212, Adjusted R-squared:  0.04733
#F-statistic: 10.89 on 1 and 198 DF,  p-value: 0.001148
```

```
plot(newspaper, sales)
abline(Model3, col="red")    # plot reg line from Model3
```

```
cor(Adv)    # correlation matrix of each column
```

```
#           TV           radio      newspaper  sales
# TV         1.00000000  0.05480866  0.05664787  0.7822244
# radio      0.05480866  1.00000000  0.35410375  0.5762226
# newspaper  0.05664787  0.35410375  1.00000000  0.2282990
# sales      0.78222442  0.57622257  0.22829903  1.0000000
```



- Multiple Reg suggests **newspaper** has no effect
- In simple regression, **newspaper** gets credit through **radio** because of the correlation.
- Many examples of confounding variables (lurking variables) (shark attack vs ice cream sales, num of cavity vs vocabulary score)

A.17 4 All predictors or just a few?

- Have to try out many models, and use some kind of criteria to pick the best
- Mallows's C_p , AIC, BIC, Adjusted R^2 . (more in Ch6)
- There's 2^p models with p predictors. $2^3 = 8$, $2^{10} = 1024$, $2^{30} = 1,073,741,824$.
- Forward, Backward, Mixed selection

A.18 5 Effect of each medium?

We can construct CI for parameters.

For the Advertising data, the 95% CI

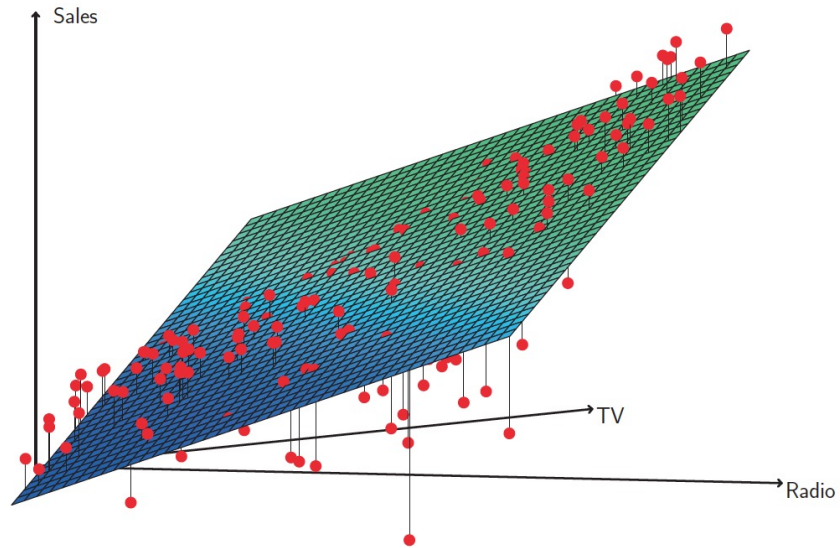
(0.043, 0.049) for TV,

(0.172, 0.206) for radio,

(-0.013, 0.011) for newspaper.

A.19 6 Prediction Accuracy?

- We can get $\hat{f}(X)$ using estimated β_i .
- There could be model bias
- Get CI for parameters, and PI for predictions



```
newAdv <- data.frame(TV=c(50, 60), radio=c(20, 10), newspaper=c(0, 0))
newAdv
```

```
# TV radio newspaper
#1 50    20          0
#2 60    10          0
```

```
predict(Model2, newdata=newAdv, interval="confidence")
#      fit      lwr      upr
#1 8.997722 8.515752 9.479692
#2 7.570068 7.099337 8.040800
```

```
predict(Model2, newdata=newAdv, interval="prediction")

#      fit      lwr      upr
#1 8.997722 5.638898 12.35655
#2 7.570068 4.212838 10.92730
```

A.20 7 Is the relationship linear?

If the relationships are linear, then the residual plots should display no pattern. Needs transformation?

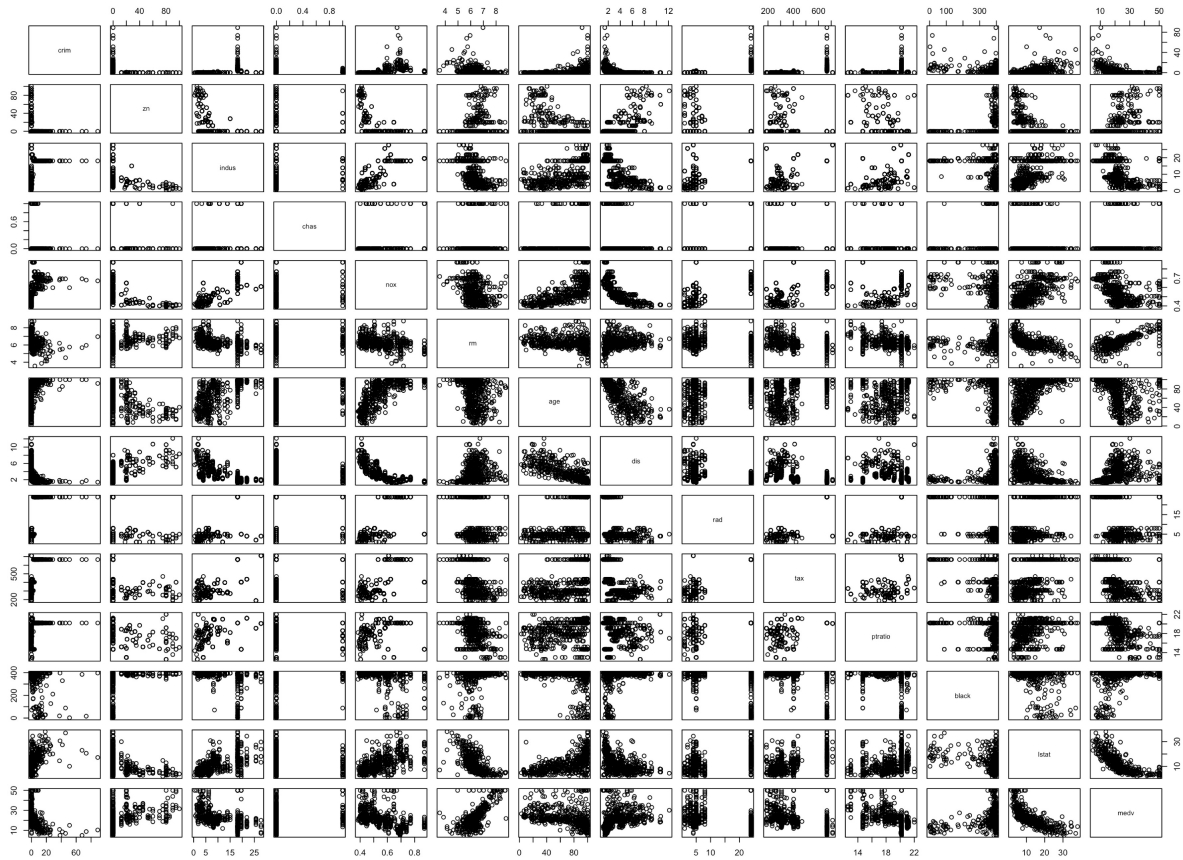
A.21 8 Is there synergy among the advertising media?

The standard linear regression model assumes an additive relationship between the predictors and the response. Including an interaction term in the model results in a substantial increase in R^2 , from around 90% to almost 97

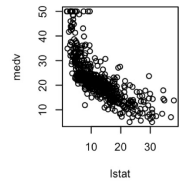
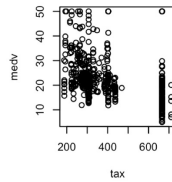
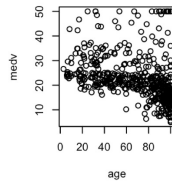
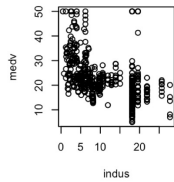
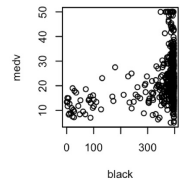
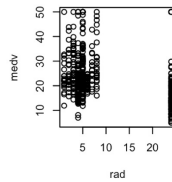
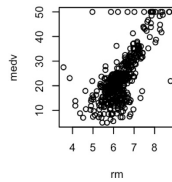
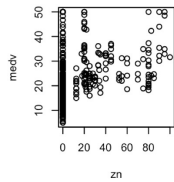
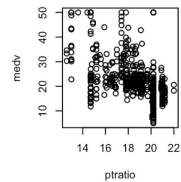
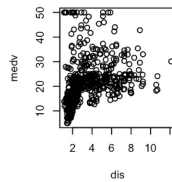
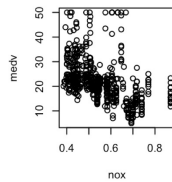
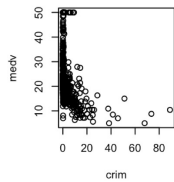
A.22 Boston House Price 1978

Boston House Price 1978

A.23 Boston Data



A.24 Median Home Value vs predictors



A.25 Inference vs Prediction

Inference vs Prediction

A.26 Advertisement

- Prediction

Who is more likely to be next customer?

- Inference

Which media contribute to sales?

Which media generate the biggest boost in sales?

How much increase in sales is associated with a given increase in TV advertising?

A.27 Housing

- Prediction

Is THIS house overpriced, or underpriced?

- Inference

What contributes to the value of a house? How can I increase them? If I have X amount of dollars, where should I spend them?

A.28 Flexibility and Interpretability trade-off

