

## ISLR Lab 4.6.1-4.6.2

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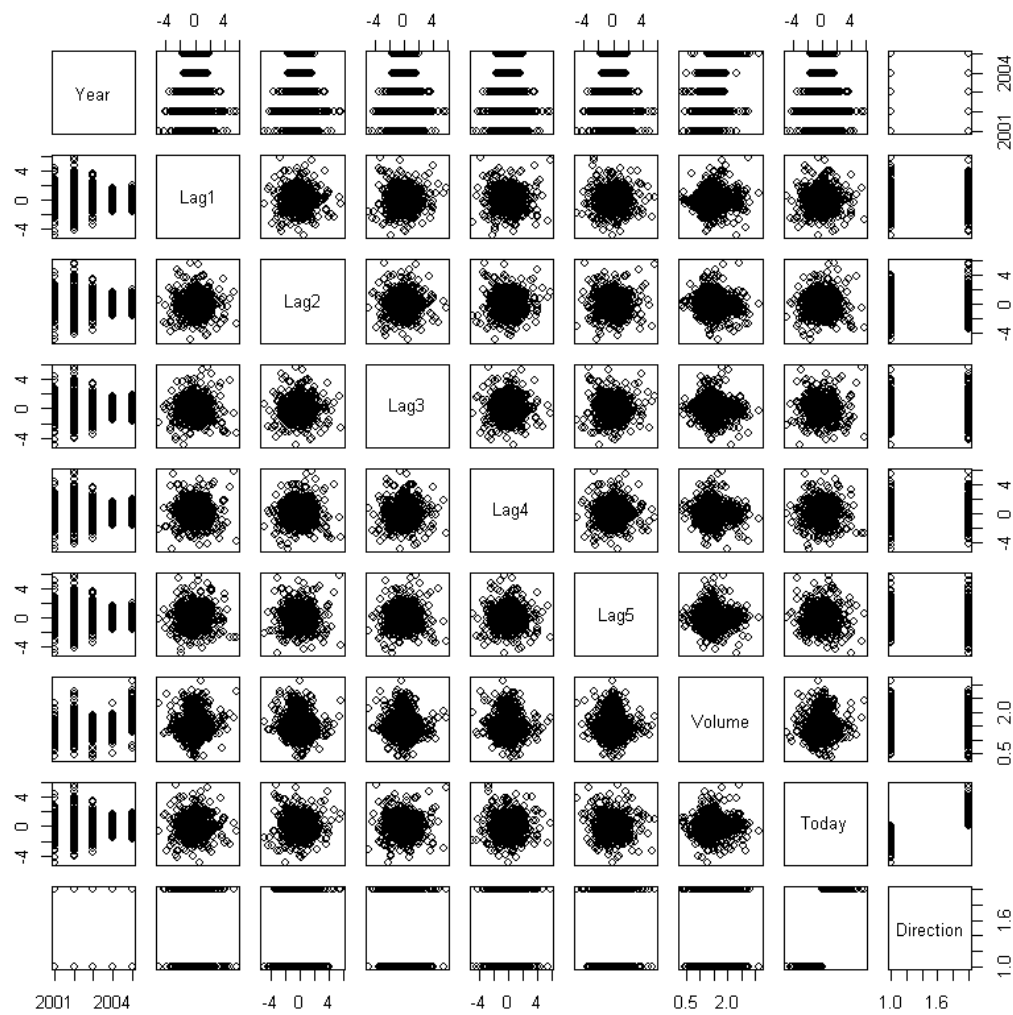
### Section 4.6.1

```
[2]: library(ISLR)
```

```
[3]: print(names(Smarket))  
      print(dim(Smarket))  
      summary(Smarket)
```

```
[1] "Year"      "Lag1"      "Lag2"      "Lag3"      "Lag4"      "Lag5"  
[7] "Volume"    "Today"     "Direction"  
[1] 1250      9  
      Year      Lag1      Lag2      Lag3  
Min.   :2001   Min.    :-4.922000   Min.    :-4.922000   Min.    :-4.922000  
1st Qu.:2002   1st Qu. :-0.639500   1st Qu. :-0.639500   1st Qu. :-0.640000  
Median :2003   Median  : 0.039000   Median  : 0.039000   Median  : 0.038500  
Mean   :2003   Mean    : 0.003834   Mean    : 0.003919   Mean    : 0.001716  
3rd Qu.:2004   3rd Qu. : 0.596750   3rd Qu. : 0.596750   3rd Qu. : 0.596750  
Max.   :2005   Max.    : 5.733000   Max.    : 5.733000   Max.    : 5.733000  
      Lag4      Lag5      Volume      Today  
Min.    :-4.922000   Min.    :-4.92200   Min.     :0.3561   Min.    :-4.922000  
1st Qu. :-0.640000   1st Qu. :-0.64000   1st Qu. :1.2574   1st Qu. :-0.639500  
Median : 0.038500   Median  : 0.03850   Median  :1.4229   Median  : 0.038500  
Mean   : 0.001636   Mean    : 0.00561   Mean    :1.4783   Mean    : 0.003138  
3rd Qu.: 0.596750   3rd Qu. : 0.59700   3rd Qu. :1.6417   3rd Qu. : 0.596750  
Max.    : 5.733000   Max.    : 5.73300   Max.    :3.1525   Max.    : 5.733000  
Direction  
Down:602  
Up   :648
```

```
[4]: pairs(Smarket)
```

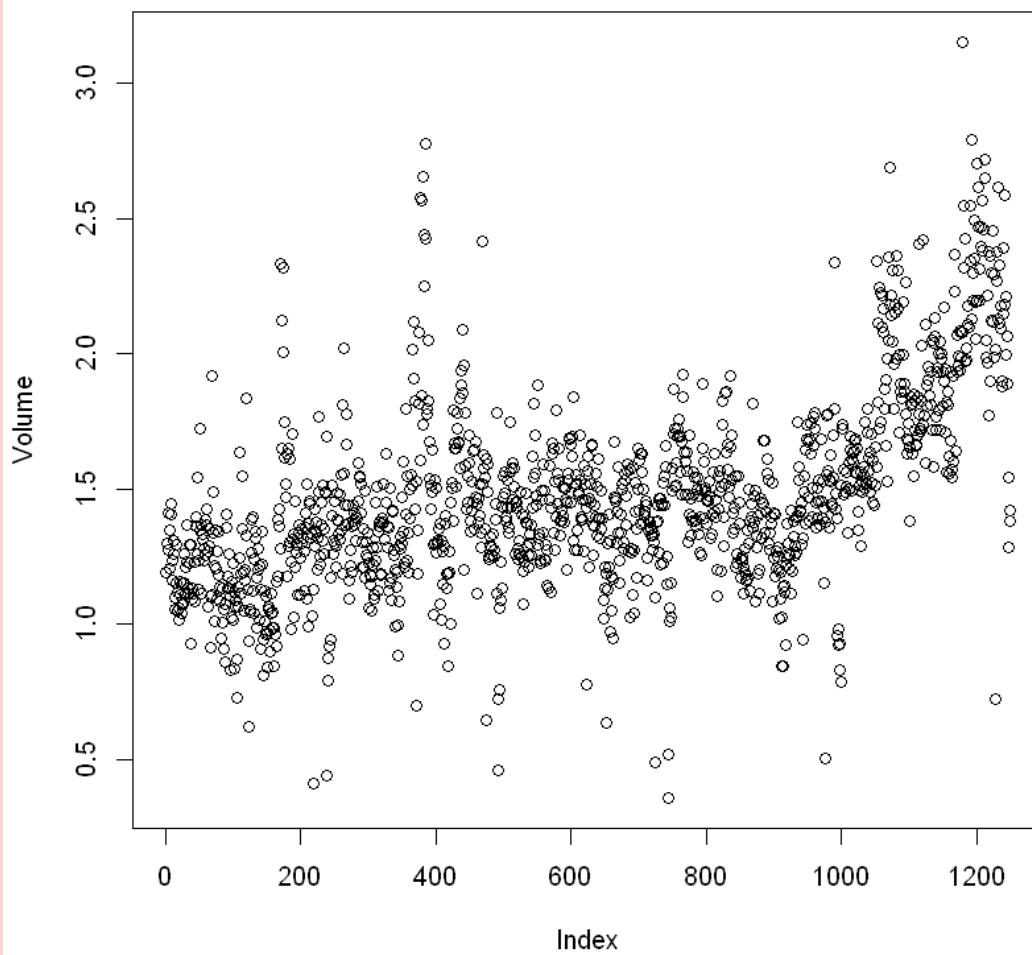


```
[19]: print(cor(Smarket[, -9]))
```

	Year	Lag1	Lag2	Lag3	Lag4
Year	1.00000000	0.029699649	0.030596422	0.033194581	0.035688718
Lag1	0.02969965	1.000000000	-0.026294328	-0.010803402	-0.002985911
Lag2	0.03059642	-0.026294328	1.000000000	-0.025896670	-0.010853533
Lag3	0.03319458	-0.010803402	-0.025896670	1.000000000	-0.024051036
Lag4	0.03568872	-0.002985911	-0.010853533	-0.024051036	1.000000000
Lag5	0.02978799	-0.005674606	-0.003557949	-0.018808338	-0.027083641
Volume	0.53900647	0.040909908	-0.043383215	-0.041823686	-0.048414246
Today	0.03009523	-0.026155045	-0.010250033	-0.002447647	-0.006899527
	Lag5	Volume	Today		
Year	0.029787995	0.53900647	0.030095229		
Lag1	-0.005674606	0.04090991	-0.026155045		

```
Lag2 -0.003557949 -0.04338321 -0.010250033
Lag3 -0.018808338 -0.04182369 -0.002447647
Lag4 -0.027083641 -0.04841425 -0.006899527
Lag5  1.000000000 -0.02200231 -0.034860083
Volume -0.022002315 1.00000000 0.014591823
Today -0.034860083 0.01459182 1.000000000
```

```
[6]: attach(Smarket)
      plot(Volume)
```



## Section 4.6.2

```
[9]: glm.fits=glm(Direction~Lag1+Lag2+Lag3+Lag4+Lag5+ Volume, data=Smarket
, family=binomial )
summary(glm.fits)
```

Call:

```
glm(formula = Direction ~ Lag1 + Lag2 + Lag3 + Lag4 + Lag5 +
    Volume, family = binomial, data = Smarket)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z )
(Intercept)	-0.126000	0.240736	-0.523	0.601
Lag1	-0.073074	0.050167	-1.457	0.145
Lag2	-0.042301	0.050086	-0.845	0.398
Lag3	0.011085	0.049939	0.222	0.824
Lag4	0.009359	0.049974	0.187	0.851
Lag5	0.010313	0.049511	0.208	0.835
Volume	0.135441	0.158360	0.855	0.392

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 1731.2 on 1249 degrees of freedom  
 Residual deviance: 1727.6 on 1243 degrees of freedom  
 AIC: 1741.6

Number of Fisher Scoring iterations: 3

```
[11]: print(coef(glm.fits))
```

(Intercept)	Lag1	Lag2	Lag3	Lag4	Lag5
-0.126000257	-0.073073746	-0.042301344	0.011085108	0.009358938	0.010313068
Volume					
0.135440659					

```
[12]: summary(glm.fits)$coef
```

	Estimate	Std. Error	z value	Pr(> z )
(Intercept)	-0.126000257	0.24073574	-0.5233966	0.6006983
Lag1	-0.073073746	0.05016739	-1.4565986	0.1452272
Lag2	-0.042301344	0.05008605	-0.8445733	0.3983491
Lag3	0.011085108	0.04993854	0.2219750	0.8243333
Lag4	0.009358938	0.04997413	0.1872757	0.8514445
Lag5	0.010313068	0.04951146	0.2082966	0.8349974
Volume	0.135440659	0.15835970	0.8552723	0.3924004

```
[14]: print(summary(glm.fits)$coef[,4])
```

```
(Intercept)      Lag1      Lag2      Lag3      Lag4      Lag5
  0.6006983  0.1452272  0.3983491  0.8243333  0.8514445  0.8349974
      Volume
  0.3924004
```

```
[18]: glm.probs=predict(glm.fits,type="response")
print(glm.probs[1:10])
```

```
      1      2      3      4      5      6      7      8
0.5070841 0.4814679 0.4811388 0.5152224 0.5107812 0.5069565 0.4926509 0.5092292
      9     10
0.5176135 0.4888378
```

```
[20]: contrasts(Direction)
```

	Up
Down	0
Up	1

```
[25]: glm.pred = rep("Down" ,1250)
glm.pred[glm.probs >.5]="Up"
```

```
[26]: table(glm.pred ,Direction )
```

```
      Direction
glm.pred Down  Up
      Down  145 141
      Up    457 507
```

```
[27]: (507+145) /1250
mean(glm.pred==Direction )
```

```
0.5216
0.5216
```

```
[30]: train=(Year <2005)
Smarket.2005= Smarket[!train ,]
print(dim(Smarket.2005))
Direction.2005= Direction [!train]
```

```
[1] 252  9
```

```
[34]: glm.fits=glm(Direction~Lag1+Lag2+Lag3+Lag4+Lag5+ Volume ,data=Smarket
,family=binomial ,subset=train)
glm.probs=predict (glm.fits,Smarket.2005, type="response")
```

```
[37]: glm.pred=rep("Down",252)
      glm.pred[glm.probs >.5]="Up"
      table(glm.pred ,Direction.2005)
```

```
      Direction.2005
glm.pred Down Up
      Down   77 97
      Up    34 44
```

```
[38]: mean(glm.pred==Direction.2005)
      mean(glm.pred!=Direction.2005)
```

```
0.48015873015873
0.51984126984127
```

```
[39]: glm.fits=glm(Direction~Lag1+Lag2 ,data=Smarket ,family=binomial ,subset=train)
      glm.probs=predict (glm.fits,Smarket.2005, type="response")
```

```
[40]: glm.pred=rep("Down",252)
      glm.pred[glm.probs >.5]="Up"
      table(glm.pred ,Direction.2005)
```

```
      Direction.2005
glm.pred Down  Up
      Down   35 35
      Up    76 106
```

```
[42]: mean(glm.pred==Direction.2005)
      106/(106+76)
```

```
0.55952380952381
0.582417582417582
```

```
[44]: print(predict(glm.fits,newdata =data.frame(Lag1=c(1.2 ,1.5),Lag2=c(1.1,-0.8)
      ),type="response"))
```

```
      1      2
0.4791462 0.4960939
```