

```
In [50]: library(MASS)
library(lmtest)
library(gridExtra)
library(tidyverse)
library(readxl)
```

```
In [7]: library(ggcorrplot)
```

```
In [8]: getwd()
list.files()
```

'/Users/xuren/pydev/p1_udj_assignments/s11_forecasting'

'S11 UDJ SALES.xls' 's11_sales.ipynb'

```
In [9]: sales_tibble = read_excel("S11 UDJ SALES.xls")
```

```
In [10]: head(sales_tibble)
```

A tibble: 6 × 9

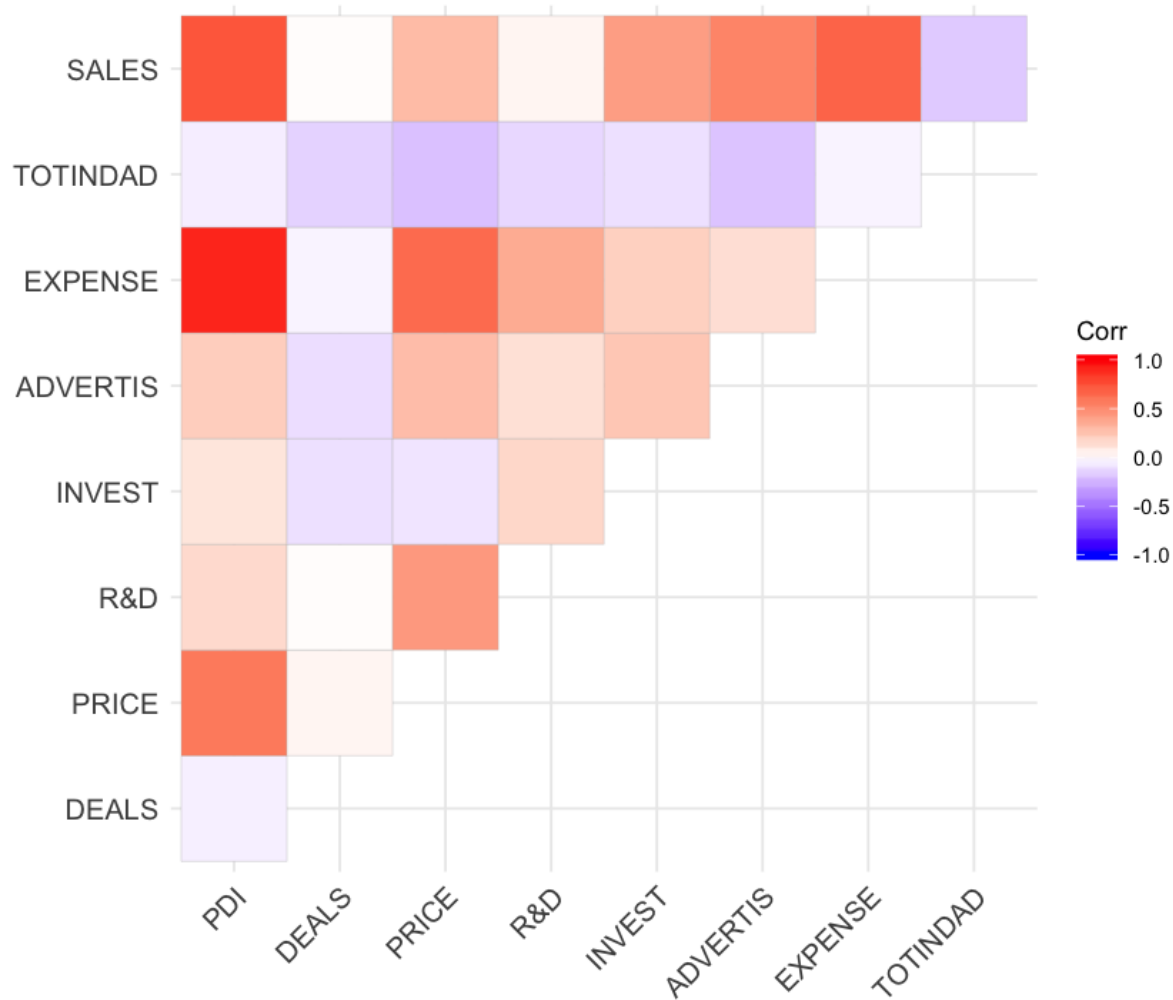
PDI	DEALS	PRICE	R&D	INVEST	ADVERTIS	EXPENSE	TOTINDAD	SALES
<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
398	138	56.20	12.11	49.89	76.86	228.80	98.20	5540.39
369	118	59.04	9.33	16.59	88.81	177.45	224.95	5439.04
268	129	56.72	28.75	89.18	51.30	166.40	263.03	4290.00
484	111	57.86	12.89	106.73	39.65	258.05	320.93	5502.34
394	146	59.11	13.38	142.55	51.65	209.30	406.99	4871.77
332	140	60.11	11.09	61.28	20.55	180.05	247.00	4708.08

```
In [11]: cor(sales_tibble %>% dplyr::select(-SALES))
```

A matrix: 8 × 8 of type dbl

	PDI	DEALS	PRICE	R&D	INVEST	ADVERTIS	EXPI
PDI	1.00000000	-0.051975827	0.58178908	0.162751135	0.10655595	0.2057018	0.9017
DEALS	-0.05197583	1.000000000	0.04420068	0.008085699	-0.09995092	-0.1095744	-0.0409
PRICE	0.58178908	0.044200685	1.00000000	0.437099447	-0.08937882	0.2763723	0.6386
R&D	0.16275114	0.008085699	0.43709945	1.000000000	0.16733956	0.1268537	0.3627
INVEST	0.10655595	-0.099950919	-0.08937882	0.167339564	1.00000000	0.2380239	0.1956
ADVERTIS	0.20570182	-0.109574424	0.27637231	0.126853716	0.23802387	1.0000000	0.1439
EXPENSE	0.90171491	-0.040904276	0.63869001	0.362759847	0.19561952	0.1439814	1.0000
TOTINDAD	-0.05611052	-0.152765420	-0.21352408	-0.129457379	-0.10084968	-0.1986011	-0.0409

```
In [12]: ggcorrplot(cor(sales_tibble), type="upper")
```



```
In [13]: model1 <- lm(SALES ~ ., data = sales_tibble)
```

In [14]: `summary(model1)`

Call:

```
lm(formula = SALES ~ ., data = sales_tibble)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-385.9	-118.0	-13.0	138.2	462.4

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	3027.6336	572.3599	5.290	9.36e-06 ***
PDI	3.3723	1.4137	2.385	0.0234 *
DEALS	4.6953	3.0103	1.560	0.1290
PRICE	-18.1112	7.4927	-2.417	0.0217 *
`R&D`	-9.9033	6.0021	-1.650	0.1090
INVEST	1.6895	0.7157	2.361	0.0247 *
ADVERTIS	8.2907	1.6445	5.042	1.91e-05 ***
EXPENSE	4.4434	2.5076	1.772	0.0862 .
TOTINDAD	-0.4427	0.3646	-1.214	0.2339

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

Residual standard error: 234.6 on 31 degrees of freedom

Multiple R-squared: 0.8338, Adjusted R-squared: 0.791

F-statistic: 19.44 on 8 and 31 DF, p-value: 4.437e-10

In [15]: `step.model <- stepAIC(model1, direction = "both",
trace = FALSE)
summary(step.model)`

Call:

```
lm(formula = SALES ~ PDI + DEALS + PRICE + `R&D` + INVEST + ADVERTIS +  
EXPENSE, data = sales_tibble)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-466.83	-138.75	4.12	118.98	417.08

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	2732.9641	522.1804	5.234	1.01e-05 ***
PDI	3.4996	1.4202	2.464	0.0193 *
DEALS	5.3063	2.9898	1.775	0.0855 .
PRICE	-16.2325	7.3852	-2.198	0.0353 *
`R&D`	-9.5020	6.0372	-1.574	0.1253
INVEST	1.8193	0.7129	2.552	0.0157 *
ADVERTIS	8.4837	1.6489	5.145	1.30e-05 ***
EXPENSE	4.0191	2.5015	1.607	0.1179

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

Residual standard error: 236.4 on 32 degrees of freedom

Multiple R-squared: 0.8259, Adjusted R-squared: 0.7879

F-statistic: 21.69 on 7 and 32 DF, p-value: 1.812e-10

In [19]: `# ?stepAIC`

In [20]: `final.model <- lm(SALES ~ PDI + PRICE + ADVERTIS + INVEST, data=sales_tibble)`

In [21]: `summary(final.model)`

Call:

`lm(formula = SALES ~ PDI + PRICE + ADVERTIS + INVEST, data = sales_tibble)`

Residuals:

	Min	1Q	Median	3Q	Max
	-571.71	-138.10	16.65	190.58	455.81

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	3333.3884	363.2868	9.176	7.66e-11	***
PDI	5.5343	0.7056	7.844	3.22e-09	***
PRICE	-15.3901	6.5801	-2.339	0.02518	*
ADVERTIS	7.5453	1.6519	4.568	5.88e-05	***
INVEST	1.9211	0.6879	2.793	0.00842	**

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

Residual standard error: 248.7 on 35 degrees of freedom

Multiple R-squared: 0.7892, Adjusted R-squared: 0.7651

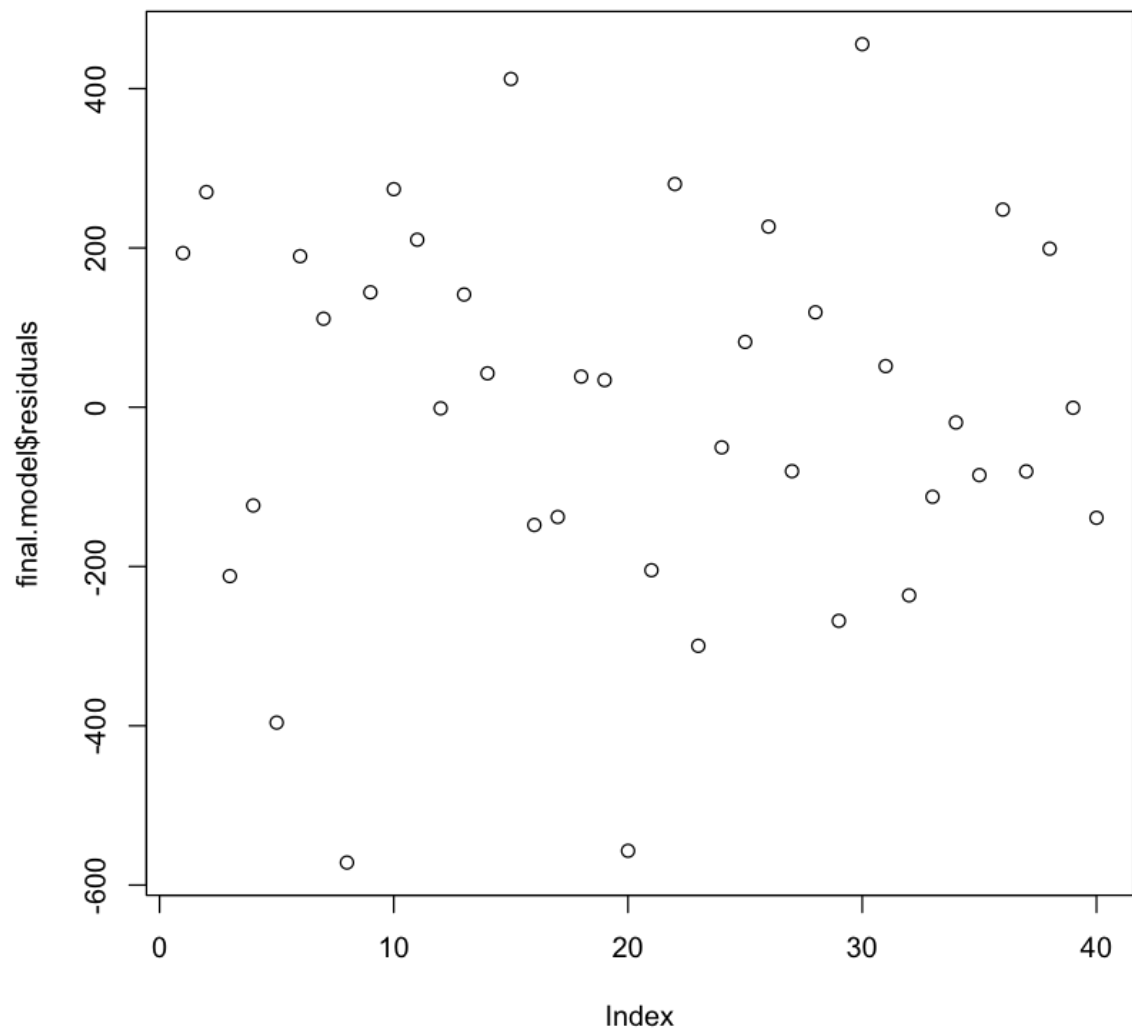
F-statistic: 32.76 on 4 and 35 DF, p-value: 2.176e-11

In []:

```
In [22]: final.model$residuals
```

```
1  193.482955103847
2  270.142481480832
3 -212.061562254153
4 -123.405856405797
5 -396.005373164189
6  189.610711808098
7  111.093253475075
8 -571.714499394502
9  144.234478379027
10 273.768203403839
11 210.282836640206
12 -1.39599268870773
13 141.517770970126
14 42.5378192248762
15 412.130946614613
16 -147.884253976745
17 -137.846844843227
18 38.4711286979419
19 34.0144479685638
20 -556.952779260608
21 -204.67736459852
22 280.208426696856
23 -299.616681963429
24 -50.4365603766034
25 81.9734260855476
26 226.747524619609
27 -80.3240474703674
28 119.14843034955
29 -268.237432438695
30 455.814682025909
31 51.5936621005788
32 -236.230804849129
33 -112.491499560325
34 -19.2001703998441
35 -85.2775708596983
36 248.107383309242
37 -80.5578291120477
38 199.02167802253
39 -0.720546114289327
40 -138.864577245988
```

```
In [23]: plot(final.model$residuals)
```

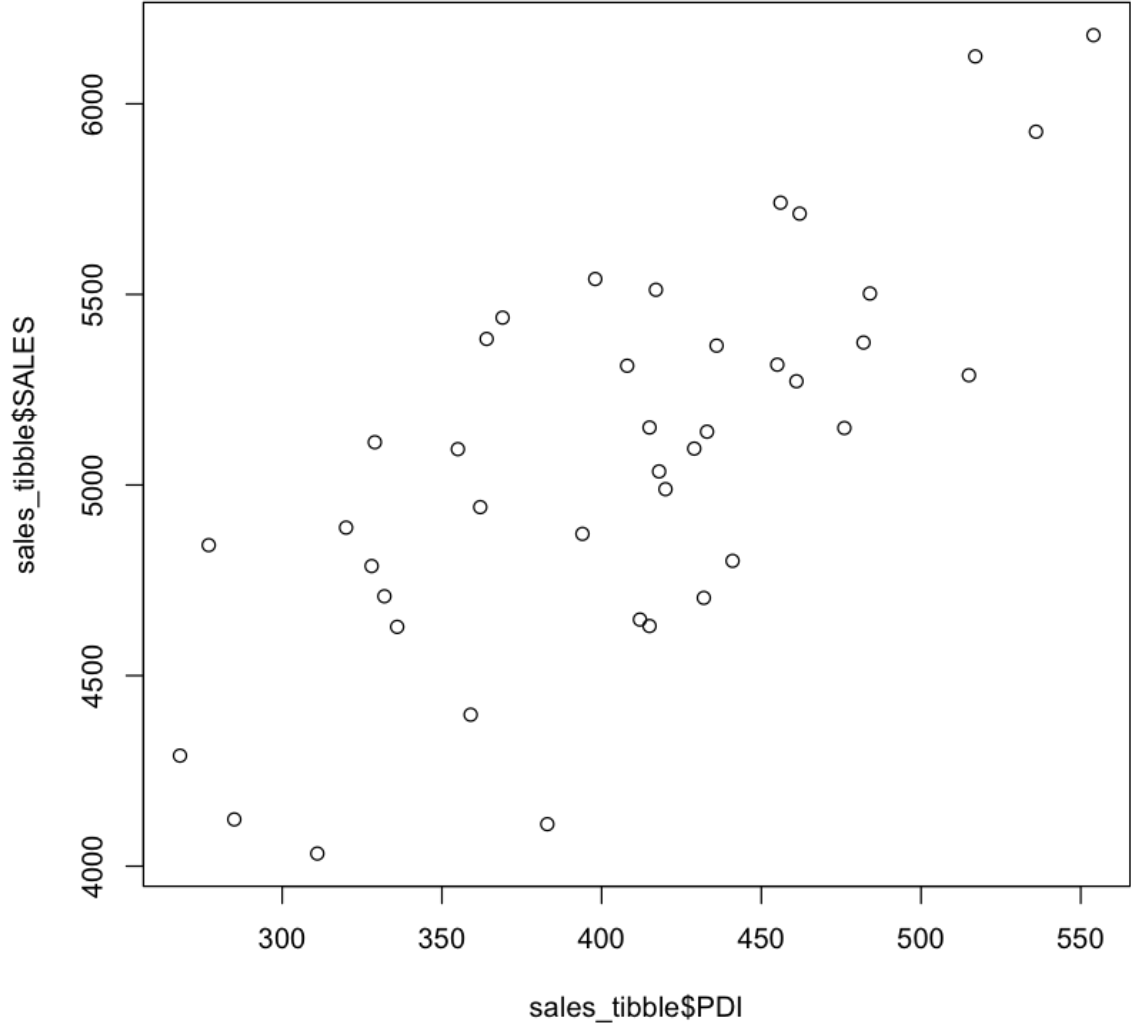


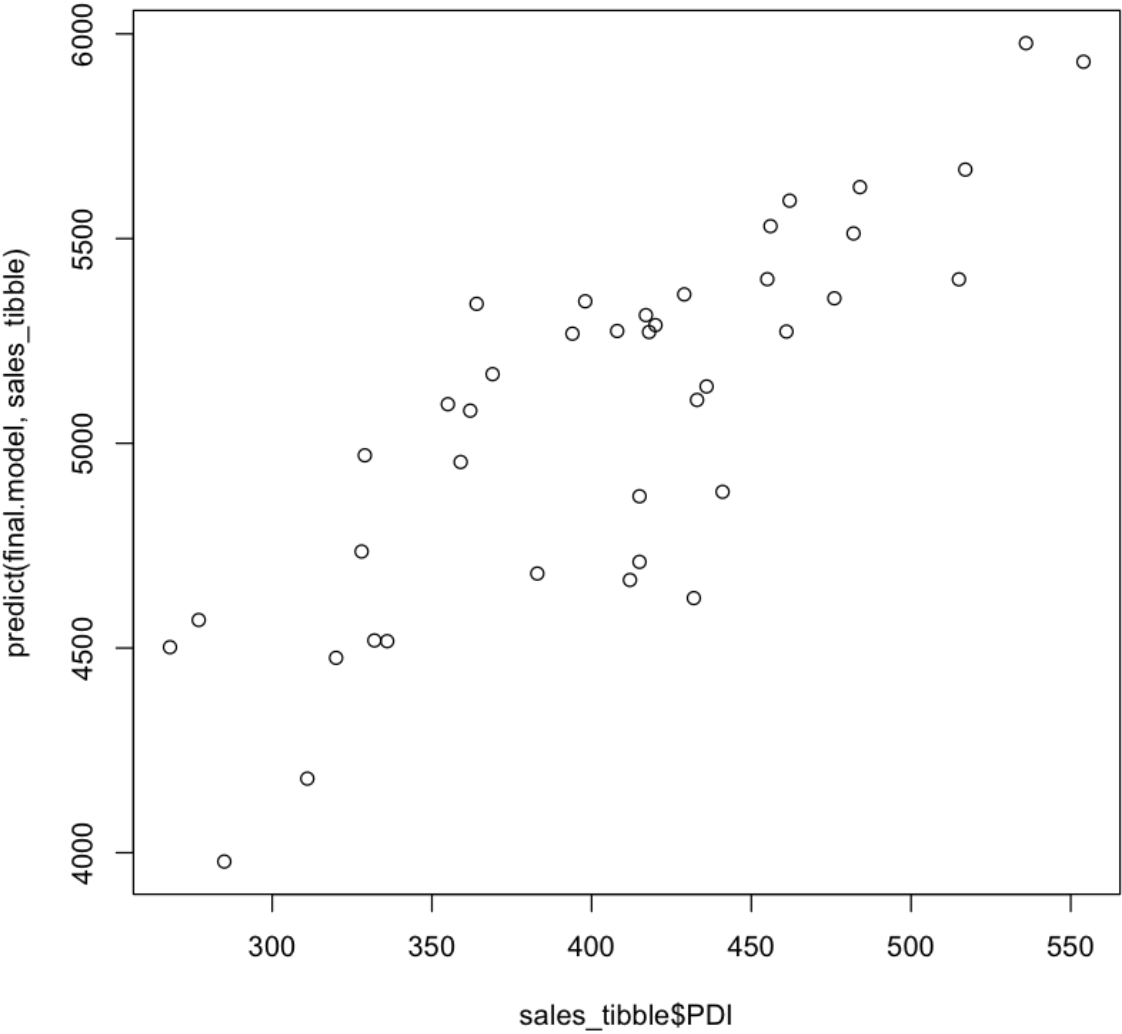
```
In [24]: predict(final.model, sales_tibble)
```

```
1  5346.90704489616
2  5168.89751851917
3  4502.06156225416
4  5625.7458564058
5  5267.77537316419
6  4518.4692881919
7  4516.71674652493
8  4681.95449939451
9  3978.45552162097
10 4568.48179659616
11 5530.3671633598
12 5095.49599268871
13 4970.70222902988
14 5340.66218077513
15 4476.03905338539
16 4181.01425397675
17 5079.80684484323
18 5274.32887130206
19 5105.85555203144
20 4954.31277926061
21 5354.14736459852
22 4870.62157330315
23 5288.63668196343
24 5977.29656037661
25 4621.90657391445
26 5138.84247538039
27 4710.41404747037
28 5592.71156965045
29 5363.7174324387
30 5668.55531797409
31 4735.74633789942
32 5271.85080484913
33 5400.50149956033
34 4666.21017039985
35 5400.9075708597
36 5931.95261669076
37 4881.52782911205
38 5313.10832197747
39 5272.93054611429
40 5512.63457724599
```

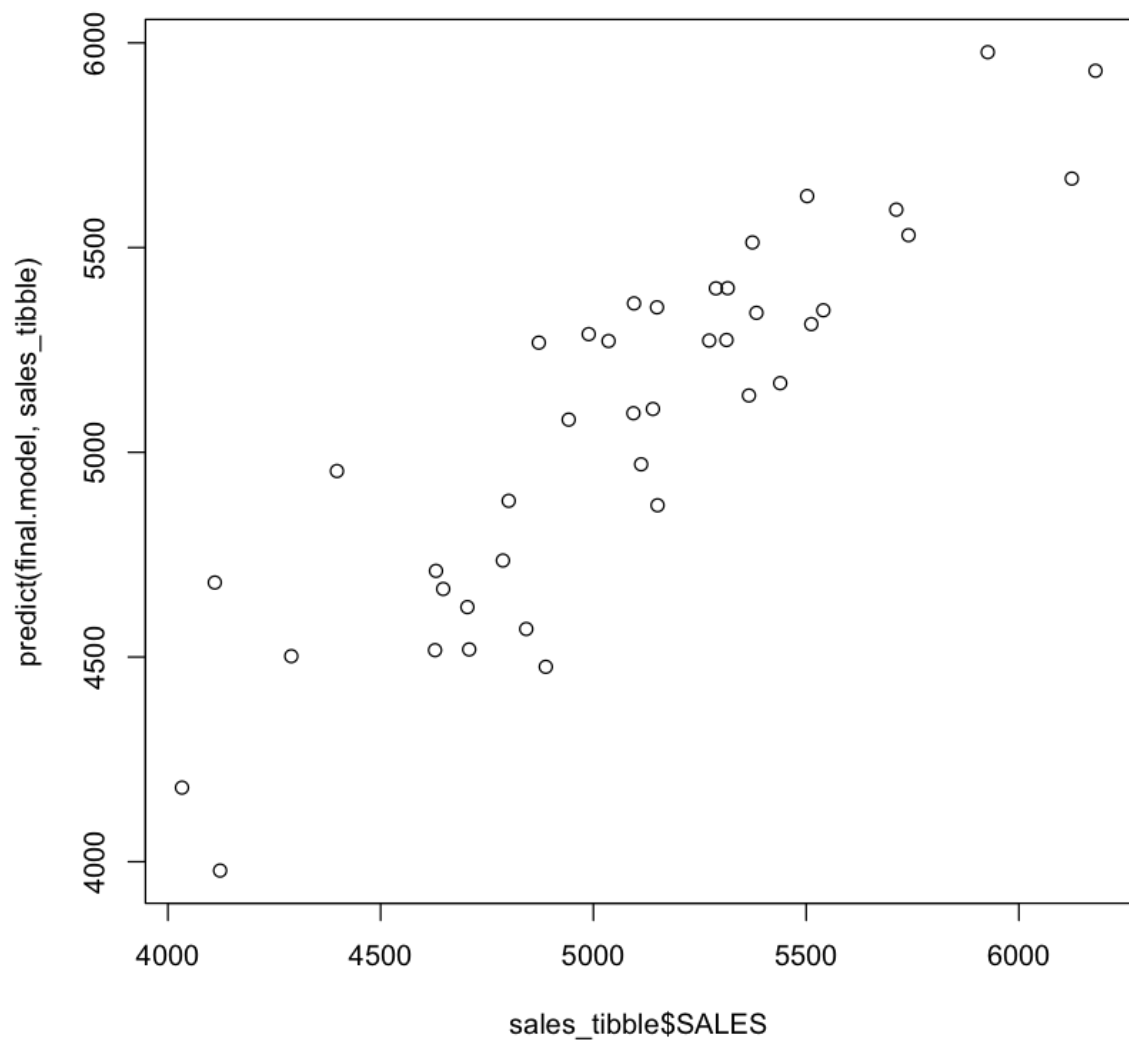


```
In [25]: plot(x=sales_tibble$PDI, y=sales_tibble$SALES)
plot(x=sales_tibble$PDI, y=predict(final.model, sales_tibble))
```





```
In [26]: plot(sales_tibble$SALES, predict(final.model, sales_tibble))
```



```
In [27]: dwtest(final.model)
```

Durbin-Watson test

data: final.model

DW = 2.222, p-value = 0.6875

alternative hypothesis: true autocorrelation is greater than 0

```
In [28]: dw_stat <- dwtest(final.model)
```

```
In [29]: str(dw_stat)
```

```
List of 5
 $ statistic : Named num 2.22
  ..- attr(*, "names")= chr "DW"
 $ method    : chr "Durbin-Watson test"
 $ alternative: chr "true autocorrelation is greater than 0"
 $ p.value    : num 0.687
 $ data.name  : chr "final.model"
 - attr(*, "class")= chr "htest"
```

```
In [30]: summary(dw_stat)
```

	Length	Class	Mode
statistic	1	-none-	numeric
method	1	-none-	character
alternative	1	-none-	character
p.value	1	-none-	numeric
data.name	1	-none-	character

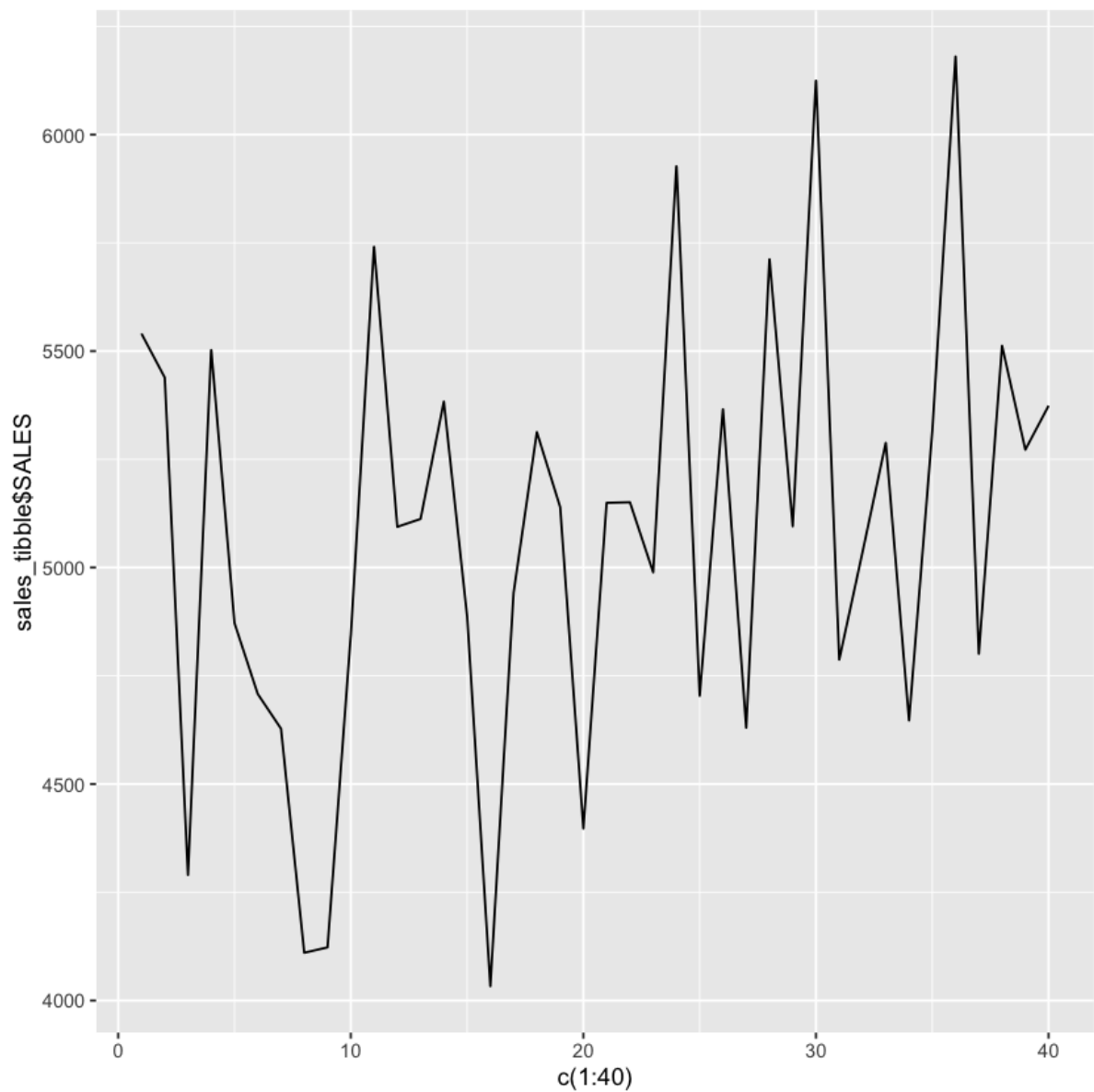
```
In [33]: length(c(1971:1991))
```

```
21
```

```
In [36]: nrow(sales_tibble)
```

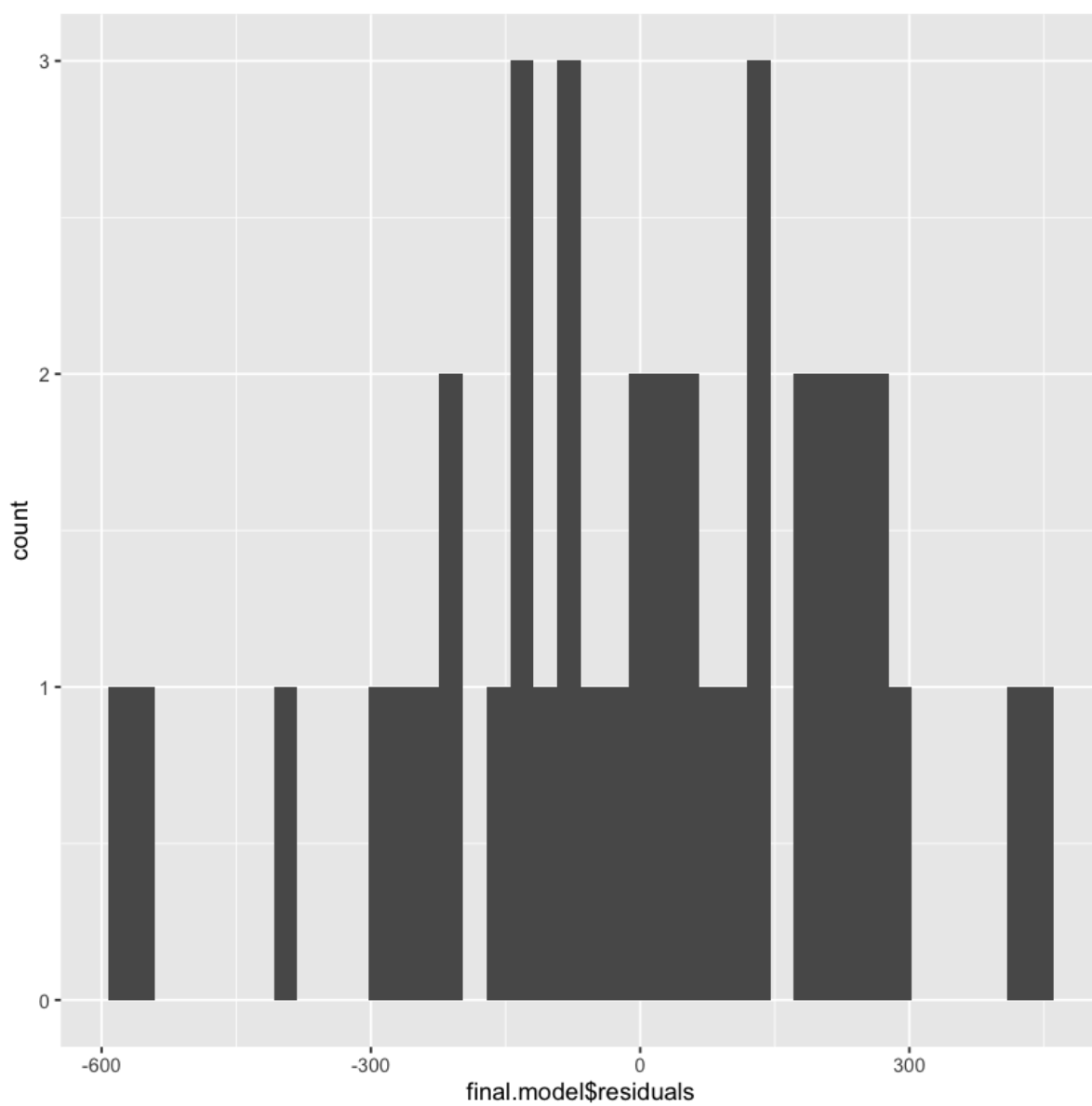
```
40
```

```
In [40]: ggplot(data=sales_tibble, aes(x=c(1:40), y=sales_tibble$SALES)) + geom_line()
```



Histograms of Residuals against Normal KDE

```
In [45]: df <- data.frame(PF = 10*rnorm(1000))  
ggplot(data=sales_tibble, aes(final.model$residuals)) +  
geom_histogram(bins=40)
```

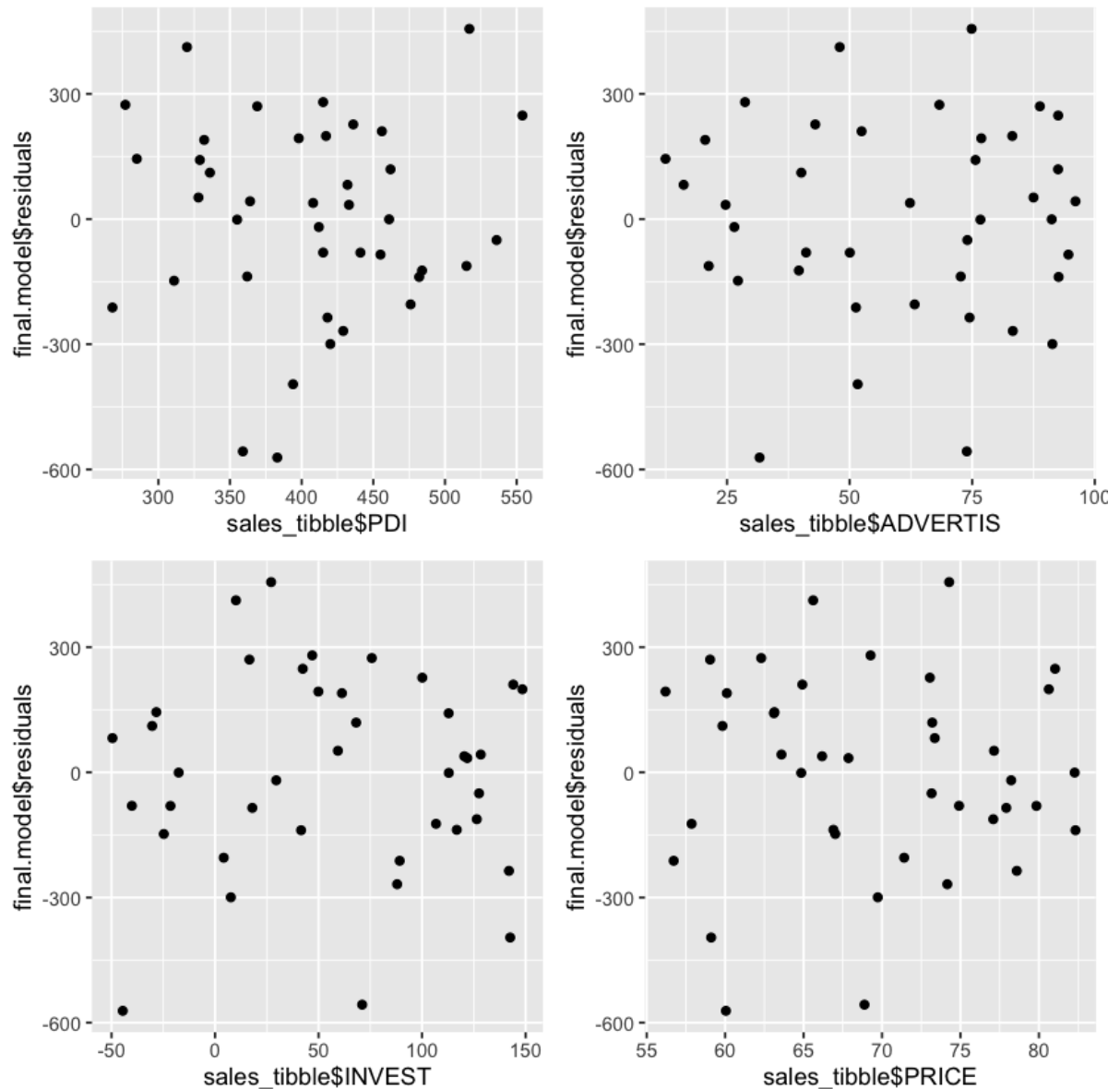


Plot residuals against:

- PDI
- Advertising
- Investment
- Price

```
In [52]: g1 <- ggplot(data=sales_tibble, aes(x=sales_tibble$PDI, y=final.model$residuals)) + geom_point()
g2 <- ggplot(data=sales_tibble, aes(x=sales_tibble$ADVERTIS, y=final.model$residuals)) + geom_point()
g3 <- ggplot(data=sales_tibble, aes(x=sales_tibble$INVEST, y=final.model$residuals)) + geom_point()
g4 <- ggplot(data=sales_tibble, aes(x=sales_tibble$PRICE, y=final.model$residuals)) + geom_point()
```

```
In [53]: grid.arrange(g1, g2, g3, g4)
```



```
In [ ]:
```