

Tracking Returns

[Code ▾](#)

Casey Tirshfield

The file `d_nasdaq_82stocks.txt` contains the daily log returns of the NASDAQ Composite Index and 82 stocks from January 3, 1990 to December 29, 2006. We want to track the returns of NASDAQ by using a small number of stocks from the given 82 stocks.

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```
# import .txt file as data frame
df <- read.table('d_nasdaq_82stocks.txt', header=TRUE)
# in order to avoid issues with naming, we change the ticker for Ford Motor Company from F to FORD, and the ticker for Brown-Forman Corporation Class B from BF-B to BFB
names(df)[23] <- 'BFB'
names(df)[51] <- 'FORD'
# we make the headers of our data frame callable objects
attach(df)
```

(a) Construct a full regression model.

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```
# this question's code is inspired by the sample code found at the following link https://web.stanford.edu/~xing/statfinbook/_BookFun/chap1_ana_data.txt
fit_full <- lm(Nasdaq ~ ., data=df[, -1])
summary(fit_full)
```

Call:

```
lm(formula = Nasdaq ~ ., data = df[, -1])
```

Residuals:

Min	1Q	Median	3Q	Max
-0.0288872	-0.0014119	0.0000462	0.0015263	0.0173120

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-8.597e-05	4.454e-05	-1.930	0.053679

AA	1.824e-03	2.745e-03	0.665	0.506407	
AAPL	1.474e-02	1.540e-03	9.570	< 2e-16	***
AET	2.254e-03	2.371e-03	0.951	0.341774	
ALTR	2.266e-02	1.455e-03	15.570	< 2e-16	***
AMAT	2.231e-02	1.726e-03	12.929	< 2e-16	***
AMD	7.580e-03	1.318e-03	5.753	9.40e-09	***
AMGN	2.978e-02	2.027e-03	14.689	< 2e-16	***
AOC	5.162e-03	2.478e-03	2.083	0.037294	*
APA	2.686e-03	2.493e-03	1.078	0.281210	
APD	-8.368e-03	3.233e-03	-2.588	0.009681	**
ASH	6.134e-03	3.328e-03	1.843	0.065348	.
AT	2.123e-02	3.170e-03	6.697	2.41e-11	***
AVY	-1.905e-04	2.989e-03	-0.064	0.949174	
AXP	7.856e-03	3.004e-03	2.615	0.008950	**
AYE	1.275e-04	2.315e-03	0.055	0.956061	
BA	5.296e-03	2.725e-03	1.943	0.052075	.
BAC	-4.788e-03	3.297e-03	-1.452	0.146551	
BAX	-3.604e-03	2.745e-03	-1.313	0.189264	
BBY	7.881e-03	1.338e-03	5.889	4.19e-09	***
BC	5.916e-03	2.332e-03	2.537	0.011208	*
BFB	-3.074e-03	3.644e-03	-0.844	0.398978	
BLL	4.434e-03	2.824e-03	1.570	0.116560	
BMJ	1.811e-03	3.069e-03	0.590	0.555232	
BNI	3.887e-04	3.173e-03	0.122	0.902524	
C	1.082e-02	3.044e-03	3.555	0.000382	***
CAH	3.707e-03	2.390e-03	1.551	0.120954	
CAT	-4.740e-03	2.837e-03	-1.671	0.094887	.
CCE	3.300e-03	2.322e-03	1.421	0.155258	
CCL	7.846e-03	2.267e-03	3.461	0.000544	***
CEG	-8.797e-03	4.161e-03	-2.114	0.034558	*
CFC	6.246e-03	2.115e-03	2.952	0.003170	**
CLX	1.114e-03	3.002e-03	0.371	0.710563	
CMI	2.473e-03	2.533e-03	0.976	0.328942	
COP	-3.952e-03	3.548e-03	-1.114	0.265414	
COST	1.202e-02	2.235e-03	5.379	7.89e-08	***
CSX	-1.959e-03	3.120e-03	-0.628	0.530244	
CVX	-1.141e-03	4.255e-03	-0.268	0.788613	
DD	-1.688e-02	3.387e-03	-4.983	6.51e-07	***
DELL	1.511e-02	1.517e-03	9.964	< 2e-16	***
DIS	1.248e-02	2.744e-03	4.549	5.55e-06	***
DOV	5.097e-03	3.388e-03	1.504	0.132596	
DUK	-2.919e-03	4.001e-03	-0.730	0.465736	
ED	-1.407e-02	5.246e-03	-2.683	0.007336	**
EDS	6.362e-03	1.883e-03	3.379	0.000734	***
EIX	-4.434e-03	2.348e-03	-1.888	0.059040	.
EK	2.272e-03	2.532e-03	0.897	0.369664	
EOG	2.406e-03	2.435e-03	0.988	0.323128	
EXC	-3.029e-03	3.872e-03	-0.782	0.434071	
FORD	-5.682e-03	2.752e-03	-2.065	0.038997	*
FPL	-4.521e-04	5.072e-03	-0.089	0.928981	

FRX	8.976e-03	2.183e-03	4.111	4.01e-05	***
GE	3.303e-02	3.949e-03	8.365	< 2e-16	***
GM	4.229e-03	2.829e-03	1.495	0.135017	
GPS	3.683e-03	1.935e-03	1.903	0.057133	.
GT	-4.592e-04	2.135e-03	-0.215	0.829681	
HD	-1.197e-03	2.813e-03	-0.425	0.670507	
HLT	3.403e-04	2.250e-03	0.151	0.879795	
HON	1.036e-02	2.667e-03	3.885	0.000104	***
HPC	4.338e-03	2.625e-03	1.653	0.098506	.
HPQ	2.027e-02	2.152e-03	9.417	< 2e-16	***
HRB	2.853e-03	2.566e-03	1.112	0.266295	
HSY	-1.075e-02	3.469e-03	-3.100	0.001945	**
HUM	3.524e-03	1.730e-03	2.037	0.041757	*
IBM	1.947e-02	2.833e-03	6.872	7.24e-12	***
INTC	4.606e-02	2.336e-03	19.717	< 2e-16	***
IPG	1.324e-02	2.339e-03	5.659	1.62e-08	***
JNJ	-1.286e-02	3.724e-03	-3.454	0.000558	***
JPM	1.274e-02	2.913e-03	4.374	1.25e-05	***
JWN	2.014e-03	1.949e-03	1.033	0.301561	
K	-4.565e-03	3.274e-03	-1.394	0.163361	
KBH	4.928e-03	2.065e-03	2.387	0.017041	*
KMI	7.191e-03	2.459e-03	2.924	0.003471	**
KO	-1.361e-02	3.609e-03	-3.770	0.000165	***
KR	-5.139e-03	2.270e-03	-2.264	0.023637	*
LEG	9.504e-03	2.865e-03	3.317	0.000917	***
LEN	9.544e-03	2.287e-03	4.173	3.07e-05	***
LLY	-4.022e-03	2.931e-03	-1.372	0.170060	
LM	2.092e-02	2.592e-03	8.072	8.93e-16	***
LMT	-1.871e-03	2.705e-03	-0.692	0.489179	
LOW	3.257e-03	2.318e-03	1.405	0.160068	
LUV	5.386e-03	2.128e-03	2.531	0.011411	*
MAS	-6.793e-03	2.719e-03	-2.498	0.012535	*

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

Residual standard error: 0.0029 on 4203 degrees of freedom

Multiple R-squared: 0.8065, Adjusted R-squared: 0.8027

F-statistic: 213.6 on 82 and 4203 DF, p-value: < 2.2e-16

(b) Use partial F-statistics and backward elimination to select variables from the full regression model in (a). Write down the selected model.

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```
# tries all one-term deletions from the model and removes the least significant explanatory variable of each iteration until all explanatory variables have a p-value of less than alpha
backward_elimination <- function(data, alpha) {
  while (1) {
    model <- lm(Nasdaq ~ ., data=data)
    temp <- dropterm(model, test='F')
    stock_removed = which.min(temp$F)
    p_value <- temp$`Pr(F)`[stock_removed]
    if (p_value < alpha) {
      break;
    }
    data = data[, -stock_removed];
  }
  data
}
selected <- backward_elimination(df[, -1], 0.05);
summary(selected)
```

Nasdaq	AAPL	ALTR	AMAT
AMD	AMGN		
Min. : -0.0441609	Min. : -0.7308588	Min. : -0.3157970	Min. : -0.1880522
Min. : -0.4769241	Min. : -0.2240090		
1st Qu.: -0.0026999	1st Qu.: -0.0166505	1st Qu.: -0.0202560	1st Qu.: -0.0201772
1st Qu.: -0.0198026	1st Qu.: -0.0125642		
Median : 0.0005034	Median : 0.0000000	Median : 0.0000000	Median : 0.0000000
Median : 0.0000000	Median : 0.0000000		
Mean : 0.0001682	Mean : 0.0005314	Mean : 0.0008815	Mean : 0.0008658
Mean : 0.0003727	Mean : 0.0009698		
3rd Qu.: 0.0032769	3rd Qu.: 0.0171961	3rd Qu.: 0.0220693	3rd Qu.: 0.0213710
3rd Qu.: 0.0205771	3rd Qu.: 0.0142315		
Max. : 0.0575642	Max. : 0.2866694	Max. : 0.2247073	Max. : 0.2281850
Max. : 0.2381447	Max. : 0.1406494		
AOC	APD	ASH	AT
AXP	BA		
Min. : -0.3611045	Min. : -0.1118734	Min. : -0.1854172	Min. : -0.1251912
Min. : -0.1461158	Min. : -0.1938306		
1st Qu.: -0.0078238	1st Qu.: -0.0090580	1st Qu.: -0.0079640	1st Qu.: -0.0080953
1st Qu.: -0.0103951	1st Qu.: -0.0100884		
Median : 0.0000000	Median : 0.0000000	Median : 0.0000000	Median : 0.0000000
Median : 0.0000000	Median : 0.0000000		
Mean : 0.0003561	Mean : 0.0004858	Mean : 0.0003012	Mean : 0.0004307
Mean : 0.0005206	Mean : 0.0004106		
3rd Qu.: 0.0091244	3rd Qu.: 0.0099811	3rd Qu.: 0.0092309	3rd Qu.: 0.0086336
3rd Qu.: 0.0113347	3rd Qu.: 0.0107299		
Max. : 0.1997797	Max. : 0.1122120	Max. : 0.0821425	Max. : 0.1177031
Max. : 0.1198440	Max. : 0.1098945		
BBY	BC	C	CCL
CEG	CFC		

Min. : -0.490829	Min. : -0.1563542	Min. : -0.1708447	Min. : -0.3838365	M
in. : -0.2394297	Min. : -0.1271005			
1st Qu.: -0.016000	1st Qu.: -0.0109104	1st Qu.: -0.0102296	1st Qu.: -0.0110978	1
st Qu.: -0.0067715	1st Qu.: -0.0126284			
Median : 0.000000	Median : 0.0000000	Median : 0.0000000	Median : 0.0000000	M
edian : 0.0000000	Median : 0.0000000			
Mean : 0.001262	Mean : 0.0002761	Mean : 0.0008202	Mean : 0.0005948	M
ean : 0.0004482	Mean : 0.0009504			
3rd Qu.: 0.018479	3rd Qu.: 0.0113409	3rd Qu.: 0.0117187	3rd Qu.: 0.0116921	3
rd Qu.: 0.0080498	3rd Qu.: 0.0139302			
Max. : 0.208311	Max. : 0.1753633	Max. : 0.1680757	Max. : 0.1527422	M
ax. : 0.1228797	Max. : 0.1978257			
COST	DD	DELL	DIS	
ED	EDS			
Min. : -0.2677794	Min. : -0.1093209	Min. : -0.277632	Min. : -0.2031775	M
in. : -0.0695699	Min. : -0.7510589			
1st Qu.: -0.0127526	1st Qu.: -0.0095107	1st Qu.: -0.015614	1st Qu.: -0.0104395	1
st Qu.: -0.0067014	1st Qu.: -0.0104599			
Median : 0.0000000	Median : 0.0000000	Median : 0.000000	Median : 0.0000000	M
edian : 0.0000000	Median : 0.0000000			
Mean : 0.0003597	Mean : 0.0003229	Mean : 0.001408	Mean : 0.0003275	M
ean : 0.0003557	Mean : 0.0002211			
3rd Qu.: 0.0132804	3rd Qu.: 0.0097128	3rd Qu.: 0.018962	3rd Qu.: 0.0107027	3
rd Qu.: 0.0072938	3rd Qu.: 0.0115512			
Max. : 0.1774138	Max. : 0.0939340	Max. : 0.187742	Max. : 0.1423899	M
ax. : 0.0904406	Max. : 0.1491984			
EIX	FRX	GE	GPS	
HON	HPQ			
Min. : -0.4273862	Min. : -0.2767231	Min. : -0.1128923	Min. : -0.2366094	
Min. : -0.1909042	Min. : -0.200151			
1st Qu.: -0.0076689	1st Qu.: -0.0110103	1st Qu.: -0.0080678	1st Qu.: -0.0134817	
1st Qu.: -0.0098796	1st Qu.: -0.012980			
Median : 0.0000000	Median : 0.0000000	Median : 0.0000000	Median : 0.0000000	
Median : 0.0000000	Median : 0.000000			
Mean : 0.0003543	Mean : 0.0006935	Mean : 0.0005367	Mean : 0.0005706	
Mean : 0.0004709	Mean : 0.000547			
3rd Qu.: 0.0086233	3rd Qu.: 0.0122489	3rd Qu.: 0.0093230	3rd Qu.: 0.0146702	
3rd Qu.: 0.0106013	3rd Qu.: 0.013908			
Max. : 0.3024843	Max. : 0.1603890	Max. : 0.1174122	Max. : 0.1751452	
Max. : 0.2484743	Max. : 0.159253			
HSY	HUM	IBM	INTC	
IPG	JNJ			
Min. : -0.1283812	Min. : -0.2976997	Min. : -0.1688993	Min. : -0.2487877	
Min. : -0.3534155	Min. : -0.1725264			
1st Qu.: -0.0071985	1st Qu.: -0.0134747	1st Qu.: -0.0095253	1st Qu.: -0.0144195	
1st Qu.: -0.0096853	1st Qu.: -0.0080216			
Median : 0.0000000	Median : 0.0000000	Median : 0.0000000	Median : 0.0000000	
Median : 0.0000000	Median : 0.0000000			
Mean : 0.0004786	Mean : 0.0004477	Mean : 0.0003897	Mean : 0.0006914	
Mean : 0.0002314	Mean : 0.0005772			

3rd Qu.: 0.0079745	3rd Qu.: 0.0144927	3rd Qu.: 0.0097648	3rd Qu.: 0.0158764	
3rd Qu.: 0.0101508	3rd Qu.: 0.0086888			
Max. : 0.2255167	Max. : 0.1469095	Max. : 0.1236629	Max. : 0.1832673	
Max. : 0.1867144	Max. : 0.0789695			
JPM	KBH	KMI	KO	
KR	LEG			
Min. : -0.199319	Min. : -0.1916917	Min. : -0.3323115	Min. : -0.1108306	M
1st Qu.: -0.2949261	1st Qu.: -0.216096			
1st Qu.: -0.010987	1st Qu.: -0.0148953	1st Qu.: -0.0077043	1st Qu.: -0.0077254	1
1st Qu.: -0.0106225	1st Qu.: -0.008459			
Median : 0.000000	Median : 0.0000000	Median : 0.0000000	Median : 0.0000000	M
Median : 0.0000000	Median : 0.000000			
Mean : 0.000517	Mean : 0.0005156	Mean : 0.0006464	Mean : 0.0004365	M
Mean : 0.0004202	Mean : 0.000514			
3rd Qu.: 0.011055	3rd Qu.: 0.0152638	3rd Qu.: 0.0088541	3rd Qu.: 0.0088047	3
3rd Qu.: 0.0109494	3rd Qu.: 0.009559			
Max. : 0.148359	Max. : 0.1745596	Max. : 0.4372542	Max. : 0.0935826	M
Max. : 0.1307342	Max. : 0.159386			
LEN	LM	LUV	MAS	
Min. : -0.2809803	Min. : -0.1892698	Min. : -0.2752111	Min. : -0.1740038	
1st Qu.: -0.0121708	1st Qu.: -0.0096546	1st Qu.: -0.0132730	1st Qu.: -0.0096001	
Median : 0.0000000	Median : 0.0000000	Median : 0.0000000	Median : 0.0000000	
Mean : 0.0009141	Mean : 0.0008879	Mean : 0.0006299	Mean : 0.0002951	
3rd Qu.: 0.0128193	3rd Qu.: 0.0112263	3rd Qu.: 0.0139167	3rd Qu.: 0.0101298	
Max. : 0.1495949	Max. : 0.1425047	Max. : 0.1436243	Max. : 0.1128037	

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```
# now we write down the selected model, which contains only significant explanatory variables
fit_reduce <- lm(Nasdaq ~ AAPL + ALTR + AMAT + AMD + AMGN + AOC + APD + ASH + AT + AXP + BA + BBY + BC + C + CCL + CEG + CFC + COST + DD + DELL + DIS + ED + EDS + EIX + FRX + GE + GPS + HON + HPQ + HSY + HUM + IBM + INTC + IPG + JNJ + JPM + KBH + KMI + KO + KR + LEG + LEN + LM + LUV + MAS)
summary(fit_reduce)
```

Call:

```
lm(formula = Nasdaq ~ AAPL + ALTR + AMAT + AMD + AMGN + AOC + APD + ASH + AT + AXP + BA + BBY + BC + C + CCL + CEG + CFC + COST + DD + DELL + DIS + ED + EDS + EIX + FRX + GE + GPS + HON + HPQ + HSY + HUM + IBM + INTC + IPG + JNJ + JPM + KBH + KMI + KO + KR + LEG + LEN + LM + LUV + MAS)
```

Residuals:

Min	1Q	Median	3Q	Max
-0.0289356	-0.0014230	0.0000468	0.0015105	0.0177293

Coefficients:

Estimate	Std. Error	t value	Pr(> t)
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(Intercept)	-8.623e-05	4.450e-05	-1.938	0.052718	.
AAPL	1.471e-02	1.527e-03	9.632	< 2e-16	***
ALTR	2.273e-02	1.448e-03	15.698	< 2e-16	***
AMAT	2.242e-02	1.719e-03	13.041	< 2e-16	***
AMD	7.755e-03	1.310e-03	5.920	3.48e-09	***
AMGN	2.948e-02	2.001e-03	14.735	< 2e-16	***
AOC	5.209e-03	2.451e-03	2.125	0.033624	*
APD	-6.824e-03	3.110e-03	-2.194	0.028265	*
ASH	7.142e-03	3.183e-03	2.244	0.024900	*
AT	2.156e-02	3.144e-03	6.858	7.98e-12	***
AXP	7.837e-03	2.970e-03	2.639	0.008354	**
BA	5.389e-03	2.674e-03	2.015	0.043965	*
BBY	8.543e-03	1.312e-03	6.510	8.40e-11	***
BC	6.401e-03	2.296e-03	2.788	0.005325	**
C	9.739e-03	2.968e-03	3.281	0.001041	**
CCL	8.636e-03	2.232e-03	3.870	0.000111	***
CEG	-1.043e-02	3.827e-03	-2.724	0.006470	**
CFC	5.482e-03	2.082e-03	2.633	0.008497	**
COST	1.271e-02	2.174e-03	5.848	5.36e-09	***
DD	-1.631e-02	3.217e-03	-5.070	4.14e-07	***
DELL	1.511e-02	1.508e-03	10.023	< 2e-16	***
DIS	1.287e-02	2.711e-03	4.748	2.12e-06	***
ED	-1.700e-02	4.693e-03	-3.622	0.000296	***
EDS	6.869e-03	1.860e-03	3.693	0.000225	***
EIX	-4.514e-03	2.283e-03	-1.977	0.048073	*
FRX	8.639e-03	2.102e-03	4.109	4.05e-05	***
GE	3.180e-02	3.897e-03	8.162	4.32e-16	***
GPS	4.265e-03	1.868e-03	2.283	0.022454	*
HON	1.124e-02	2.608e-03	4.307	1.69e-05	***
HPQ	2.064e-02	2.139e-03	9.652	< 2e-16	***
HSY	-1.201e-02	3.341e-03	-3.593	0.000330	***
HUM	4.184e-03	1.667e-03	2.510	0.012103	*
IBM	1.965e-02	2.816e-03	6.979	3.42e-12	***
INTC	4.640e-02	2.314e-03	20.049	< 2e-16	***
IPG	1.363e-02	2.322e-03	5.873	4.60e-09	***
JNJ	-1.448e-02	3.438e-03	-4.212	2.58e-05	***
JPM	1.159e-02	2.725e-03	4.253	2.15e-05	***
KBH	5.539e-03	2.049e-03	2.703	0.006897	**
KMI	7.713e-03	2.414e-03	3.196	0.001406	**
KO	-1.408e-02	3.420e-03	-4.118	3.90e-05	***
KR	-5.119e-03	2.245e-03	-2.280	0.022665	*
LEG	1.113e-02	2.824e-03	3.940	8.29e-05	***
LEN	1.007e-02	2.269e-03	4.439	9.28e-06	***
LM	2.214e-02	2.566e-03	8.627	< 2e-16	***
LUV	5.576e-03	2.084e-03	2.675	0.007496	**
MAS	-6.272e-03	2.661e-03	-2.357	0.018472	*

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

Residual standard error: 0.002902 on 4240 degrees of freedom

Multiple R-squared: 0.8045, Adjusted R-squared: 0.8024
F-statistic: 387.6 on 45 and 4240 DF, p-value: < 2.2e-16

(c) Compare the full and selected models. Summarize your comparison in an ANOVA table.

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```
anova(fit_reduce, fit_full)
```

Analysis of Variance Table

Model 1: Nasdaq ~ AAPL + ALTR + AMAT + AMD + AMGN + AOC + APD + ASH +
AT + AXP + BA + BBY + BC + C + CCL + CEG + CFC + COST + DD +
DELL + DIS + ED + EDS + EIX + FRX + GE + GPS + HON + HPQ +
HSY + HUM + IBM + INTC + IPG + JNJ + JPM + KBH + KMI + KO +
KR + LEG + LEN + LM + LUV + MAS

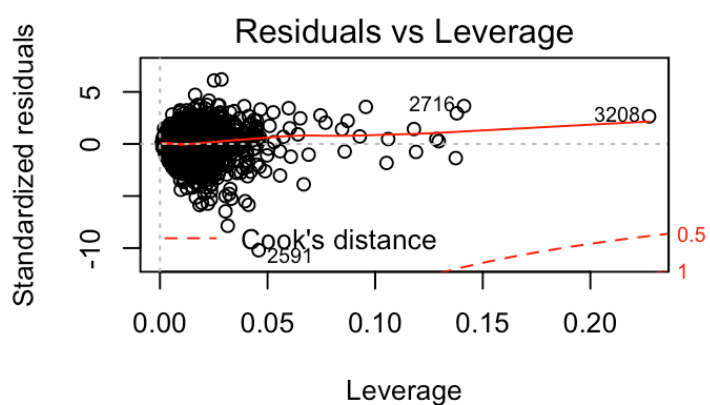
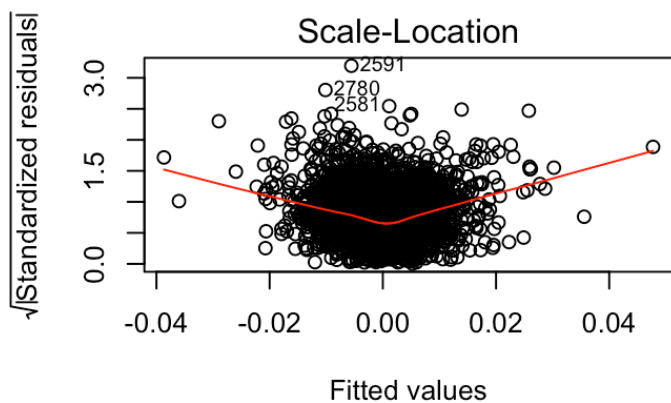
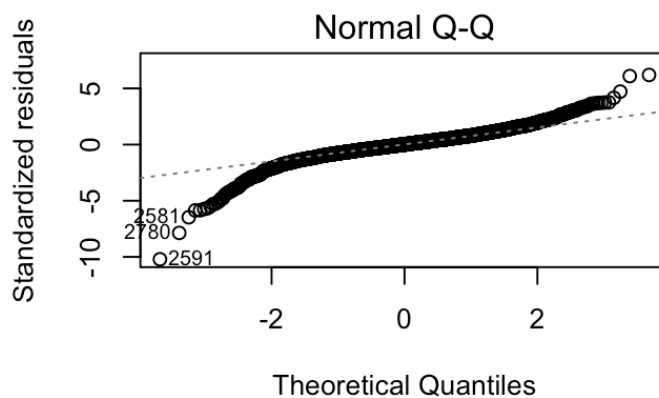
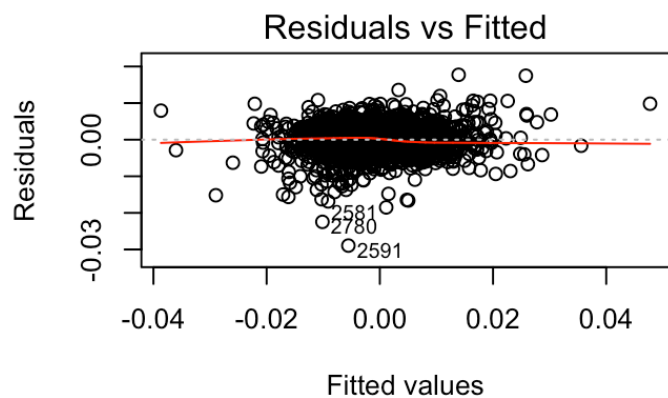
Model 2: Nasdaq ~ AA + AAPL + AET + ALTR + AMAT + AMD + AMGN + AOC + APA +
APD + ASH + AT + AVY + AXP + AYE + BA + BAC + BAX + BBY +
BC + BFB + BLL + BMY + BNI + C + CAH + CAT + CCE + CCL +
CEG + CFC + CLX + CMI + COP + COST + CSX + CVX + DD + DELL +
DIS + DOV + DUK + ED + EDS + EIX + EK + EOG + EXC + FORD +
FPL + FRX + GE + GM + GPS + GT + HD + HLT + HON + HPC + HPQ +
HRB + HSY + HUM + IBM + INTC + IPG + JNJ + JPM + JWN + K +
KBH + KMI + KO + KR + LEG + LEN + LLY + LM + LMT + LOW +
LUV + MAS

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	4240	0.035718				
2	4203	0.035353	37	0.00036456	1.1714	0.2204

(d) For the selected regression model in (b), perform residual diagnostics.

Hide

```
par(mfrow=c(2, 2))  
plot(fit_reduce)  
# plot studentized residuals  
par(mfrow=c(1, 2))
```

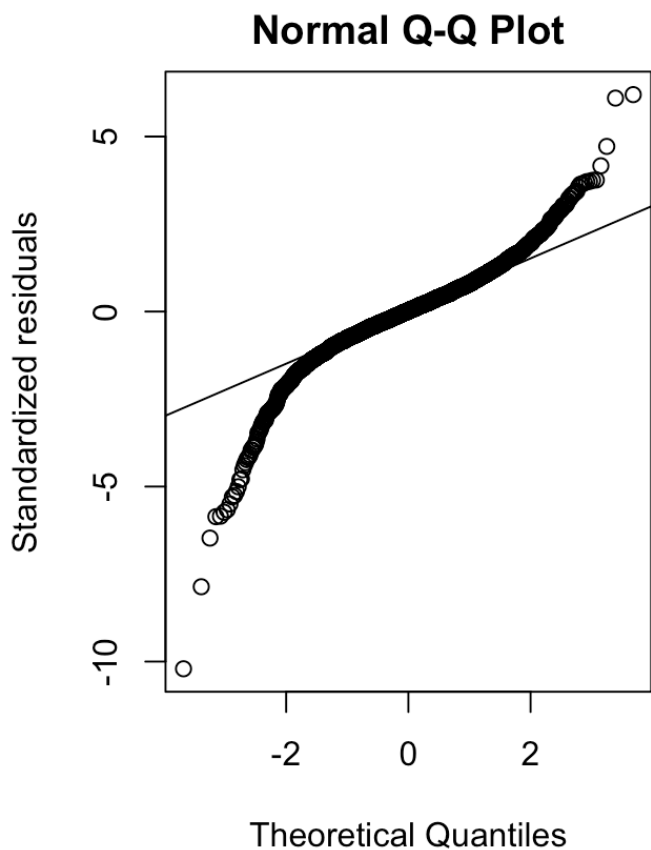
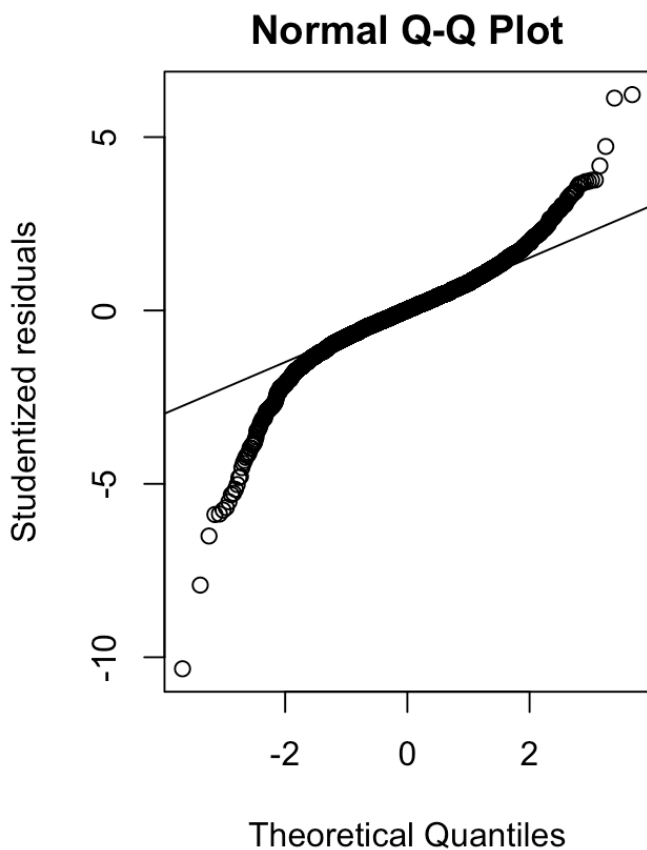



Hide

```
studresid <- as.numeric(studres(fit_reduce))
qqnorm(studresid, ylab='Studentized residuals')
qqline(studresid)
```

Hide

```
# plot standardized residuals
standresid <- as.numeric(stdres(fit_reduce))
qqnorm(standresid, ylab='Standardized residuals')
qqline(standresid)
```



(e) If you can only use at most five stocks to track the daily NASDAQ log returns, describe your model selection procedure and your constructed model.

Hide

```
# we employ forward selection
# first we initialize the null model consisting of no predictors, just the intercept
null <- lm(Nasdaq ~ 1, data=df[, -1])
# we then add terms one by one
stepAIC(null, direction='forward', steps=5, trace=0,
        scope = list(upper=fit_full, lower=null))
```

```
Call:
lm(formula = Nasdaq ~ INTC + GE + ALTR + LM + AMAT, data = df[,
  -1])
```

```
Coefficients:
(Intercept)      INTC          GE      ALTR          LM          AMAT
-5.105e-05    7.688e-02    9.438e-02    3.544e-02    5.636e-02    3.944e-02
```

Hide

```
# now we write down the constructed model
fit_five <- lm(Nasdaq ~ INTC + GE + ALTR + LM + AMAT)
summary(fit_five)
```

Call:

```
lm(formula = Nasdaq ~ INTC + GE + ALTR + LM + AMAT)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-0.034947	-0.001797	0.000065	0.001923	0.022637

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-5.105e-05	5.463e-05	-0.935	0.35
INTC	7.688e-02	2.634e-03	29.193	<2e-16 ***
GE	9.438e-02	3.858e-03	24.465	<2e-16 ***
ALTR	3.544e-02	1.735e-03	20.429	<2e-16 ***
LM	5.636e-02	2.871e-03	19.629	<2e-16 ***
AMAT	3.944e-02	2.016e-03	19.559	<2e-16 ***

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

Residual standard error: 0.003572 on 4280 degrees of freedom

Multiple R-squared: 0.701, Adjusted R-squared: 0.7006

F-statistic: 2007 on 5 and 4280 DF, p-value: < 2.2e-16

Hide

```
# we compare the full and constructed models
anova(fit_five, fit_full)
```

Analysis of Variance Table

Model 1: Nasdaq ~ INTC + GE + ALTR + LM + AMAT

Model 2: Nasdaq ~ AA + AAPL + AET + ALTR + AMAT + AMD + AMGN + AOC + APA +
APD + ASH + AT + AVY + AXP + AYE + BA + BAC + BAX + BBY +
BC + BFB + BLL + BMY + BNI + C + CAH + CAT + CCE + CCL +
CEG + CFC + CLX + CMI + COP + COST + CSX + CVX + DD + DELL +
DIS + DOV + DUK + ED + EDS + EIX + EK + EOG + EXC + FORD +
FPL + FRX + GE + GM + GPS + GT + HD + HLT + HON + HPC + HPQ +
HRB + HSY + HUM + IBM + INTC + IPG + JNJ + JPM + JWN + K +
KBH + KMI + KO + KR + LEG + LEN + LLY + LM + LMT + LOW +
LUV + MAS

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	4280	0.054619				
2	4203	0.035353	77	0.019265	29.745	< 2.2e-16 ***

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