

02441 Applied Statistics and Statistical Software

Exercise 5A - Forbes

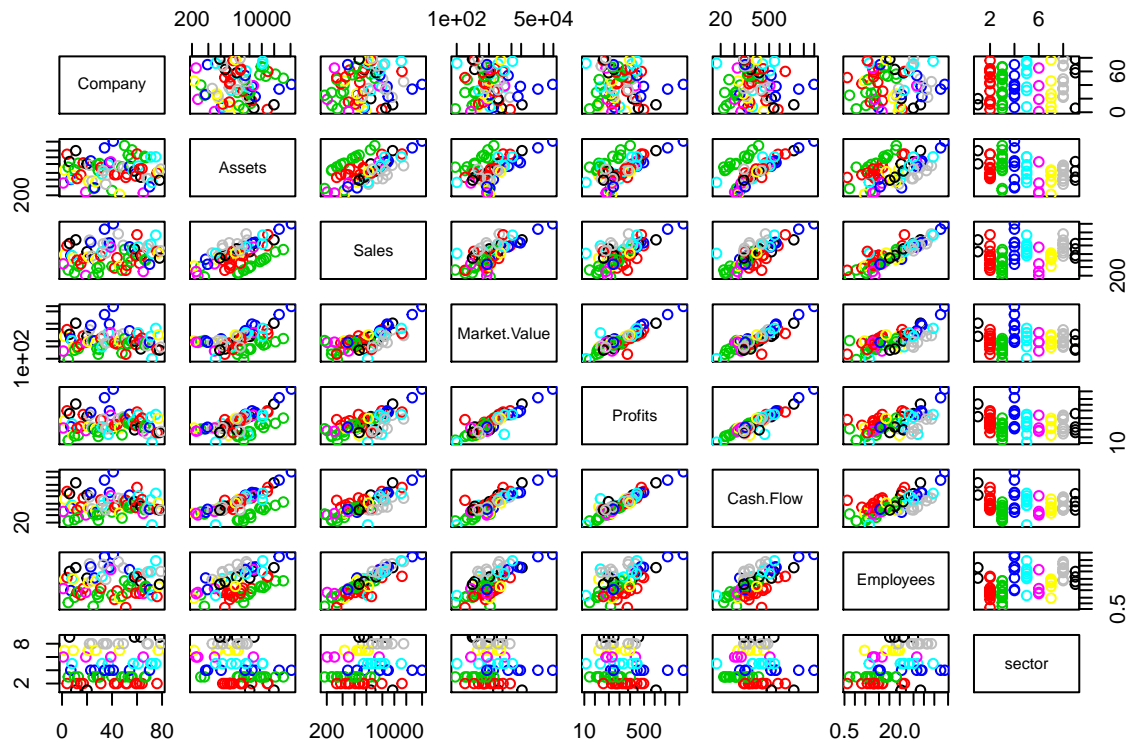
This dataset holds several facts about 77 companies selected from the Forbes 500 list for 1986. This is a 1/10 systematic sample from the alphabetical list of companies. The Forbes 500 includes all companies in the top 500 on any of the criteria, and thus has almost 800 companies in the list. Companies are often interested in how to increase sales. Many of the variables are skewed - a common occurrence with financial data - which suggests that much of the data are better analyzed after taking logarithms. For this study one should take the log of Sales and Assets.

Variable name	Description
Company	Company Name
Assets	Amount of assets (in millions)
Sales	Amount of sales (in millions)
Market.Value	Market Value of the company (in millions)
Profits	Profits (in millions)
Cash.Flow	Cash Flow (in millions)
Employees	Number of employees (in thousands)
Sector	Type of market the company is associated with

1. Investigate the relation between $\text{Log}(\text{Sales})$ predicted by $\text{Log}(\text{Assets})$

Let's start with some visualizations. Logarithmic transformations come in handy for all continuous variables. Cashflow and Profits contain negativ values. These are resulting in missing data due to the log transformation. The corresponding warnings are suppressed in the code below.

```
forbes <- read.table("forbes.txt", header = TRUE, sep = '\t')
# Remove rows with negativ CashFlow and Profits to make this exercise
# a bit easier (log transformation of negative values will result in NaNs).
forbes <- forbes[forbes$Cash.Flow>0 & forbes$Profits>0,]
pairs(forbes, col = forbes$sector, log = c(2:7))
```



There is a linear relationship between $\log(\text{Assets})$ and $\log(\text{Sales})$. However, it also seems that sector is significant (different intercepts; represented by color in pairs plot).

```
lm1 <- lm(log(Sales)~log(Assets), forbes)
summary(lm1)
```

```
##
## Call:
## lm(formula = log(Sales) ~ log(Assets), data = forbes)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.9484 -0.8641  0.2083  0.6392  1.9710
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   3.10267    0.79559   3.900 0.000221 ***
## log(Assets)    0.56067    0.09966   5.626 3.66e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.003 on 69 degrees of freedom
## Multiple R-squared:  0.3144, Adjusted R-squared:  0.3045
## F-statistic: 31.65 on 1 and 69 DF, p-value: 3.658e-07
```

$\log(\text{Assets})$ is significant at a 5% significance level. 31.4% of variance in $\log(\text{Sales})$ is explained by the model `lm1`.

2. Include Sector as a discrete factor in the model - what is your conclusion?

```
lm2a <- lm(log(Sales)~log(Assets)+sector, forbes)
lm2b <- lm(log(Sales)~sector+log(Assets), forbes)
anova(lm2a)

## Analysis of Variance Table
##
## Response: log(Sales)
##           Df Sum Sq Mean Sq F value    Pr(>F)
## log(Assets)  1 31.820   31.820 139.346 < 2.2e-16 ***
## sector       8 55.445    6.931  30.351 < 2.2e-16 ***
## Residuals   61 13.929    0.228
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

anova(lm2b)

## Analysis of Variance Table
##
## Response: log(Sales)
##           Df Sum Sq Mean Sq F value    Pr(>F)
## sector       8 46.684    5.835  25.555 < 2.2e-16 ***
## log(Assets)  1 40.582   40.582 177.716 < 2.2e-16 ***
## Residuals   61 13.929    0.228
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

First of all, there is an issue here. The standard `anova` function performs sequential *Type I* tests. Depending on the order of the main effects in the model formula the results may differ when data is unbalanced. The Sum of Squares (SS) of `lm2a` are calculated such that

$SS(\log(\text{Assets}))$ for Assets and $SS(\text{sector}|\log(\text{Assets}))$ for sector.

Using *Type II* tests you can make sure to control for other main effects when calculating Sum of Squares. In this particular example *Type II* Anova calculates the SS, such that

$SS(\log(\text{Assets})|\text{sector})$ for Assets and $SS(\text{sector}|\log(\text{Assets}))$ for sector.

The `Anova` function of the `car` package performs *Type II* tests. It yields similar results to `drop1` and the result does not depend on effect sequence in the model formula.

```
library(car)

## Loading required package: carData

Anova(lm2a)

## Anova Table (Type II tests)
##
## Response: log(Sales)
##           Sum Sq Df F value    Pr(>F)
## log(Assets) 40.582  1 177.716 < 2.2e-16 ***
## sector      55.445  8  30.351 < 2.2e-16 ***
## Residuals   13.929 61
```

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

Anova(lm2b)

## Anova Table (Type II tests)
##
## Response: log(Sales)
##           Sum Sq Df F value    Pr(>F)
## sector      55.445  8  30.351 < 2.2e-16 ***
## log(Assets) 40.582  1 177.716 < 2.2e-16 ***
## Residuals   13.929 61
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

drop1(lm2a, test = 'F')

## Single term deletions
##
## Model:
## log(Sales) ~ log(Assets) + sector
##           Df Sum of Sq    RSS      AIC F value    Pr(>F)
## <none>                 13.929 -95.636
## log(Assets)  1      40.582 54.511  -0.764 177.716 < 2.2e-16 ***
## sector       8      55.445 69.375   2.356  30.351 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

drop1(lm2b, test = 'F')

## Single term deletions
##
## Model:
## log(Sales) ~ sector + log(Assets)
##           Df Sum of Sq    RSS      AIC F value    Pr(>F)
## <none>                 13.929 -95.636
## sector       8      55.445 69.375   2.356  30.351 < 2.2e-16 ***
## log(Assets)  1      40.582 54.511  -0.764 177.716 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

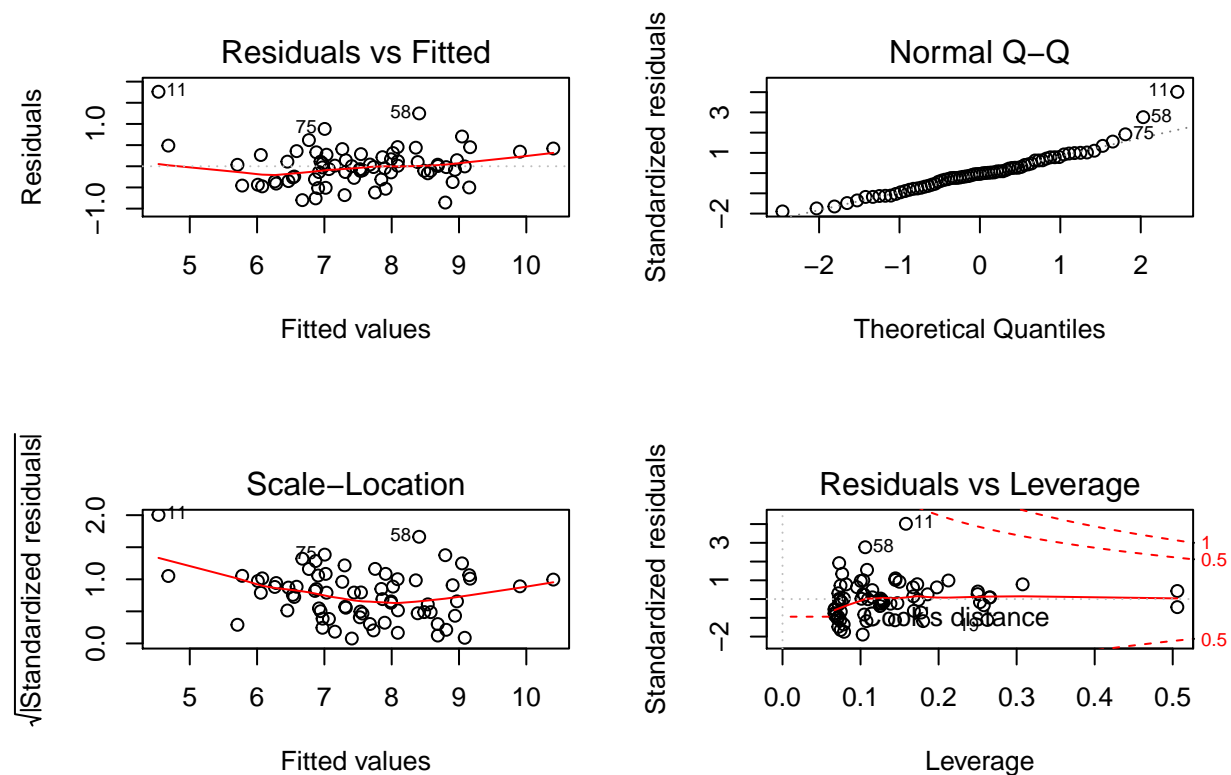
summary(lm2a)

##
## Call:
## lm(formula = log(Sales) ~ log(Assets) + sector, data = forbes)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.85688 -0.29117 -0.01791  0.24077  1.76024
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    1.85366    0.60106   3.084 0.003067 **
## log(Assets)     0.71934    0.05396  13.331 < 2e-16 ***
## sectorEnergy   -0.31491    0.36808  -0.856 0.395587
## sectorFinance  -1.48355    0.36195  -4.099 0.000125 ***
```

```
## sectorHiTech      0.72727    0.38134    1.907 0.061211 .
## sectorManufacturing 0.64875    0.38323    1.693 0.095586 .
## sectorMedical     -0.12279    0.43693   -0.281 0.779638
## sectorOther       0.31251    0.40851    0.765 0.447219
## sectorRetail      1.19386    0.38112    3.133 0.002662 **
## sectorTransportation 0.52312    0.42237    1.239 0.220263
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4779 on 61 degrees of freedom
## Multiple R-squared:  0.8624, Adjusted R-squared:  0.842
## F-statistic: 42.46 on 9 and 61 DF,  p-value: < 2.2e-16
```

Both, sector and $\log(\text{Assets})$ are significant. 86.2% of variance in $\log(\text{Sales})$ is explained by the model `lm2a`. Let's check if the model assumptions are fulfilled.

```
par(mfrow=c(2,2))
plot(lm2a)
```



Let's make a plot of the results.

```
plot(log(Sales)~log(Assets), forbes, col = forbes$sector, pch = 19)
legend("topleft", legend = levels(forbes$sector), col = 1:nlevels(forbes$sector),
      pch = 19, cex = 0.8)
new_data <- seq(min(forbes$Assets), max(forbes$Assets), length.out = 100)
```

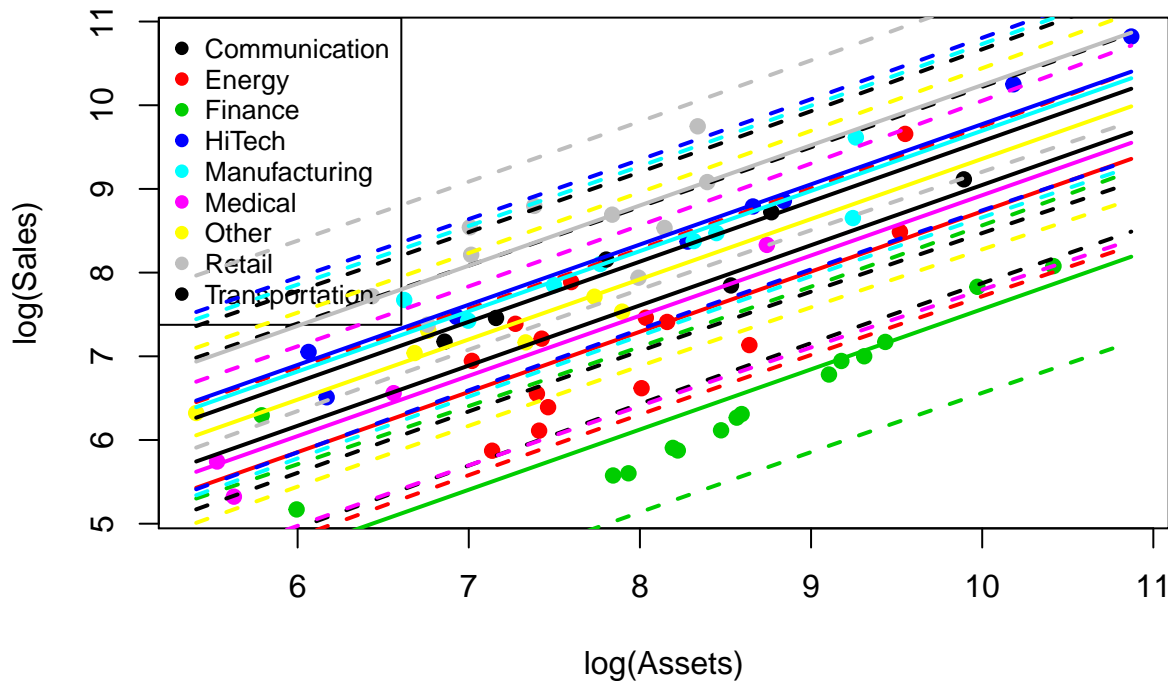
```

new_data_grid <- expand.grid(Assets = new_data, sector = levels(forbes$sector))
color_codes <- as.character(levels(forbes$sector))

plot_confidence <- function(sector){
  matlines(log(new_data), predict(lm2a, new_data_grid[new_data_grid$sector==color_codes[sector],],
                                interval = "prediction"), lty = c(1,2,2), lw = 2,
          col = c(sector, sector, sector))
}

for (sector in c(1:9)) {
  plot_confidence(sector)
}

```

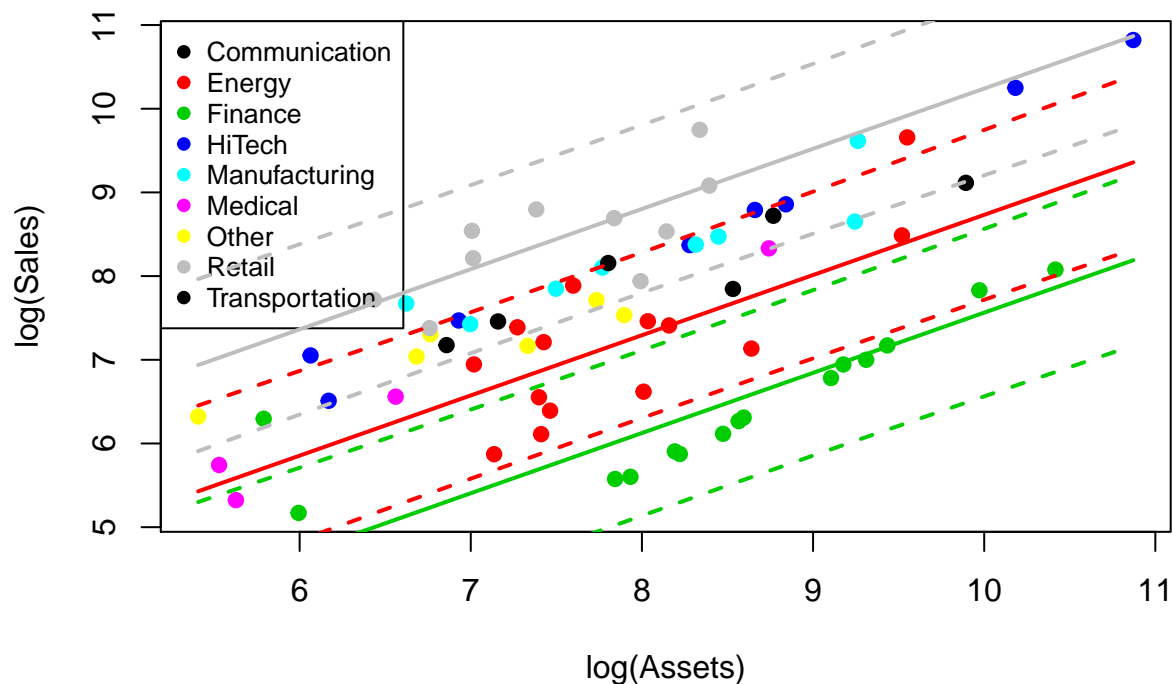


The plot above shows 95% prediction intervals for all sectors, but it is not very informative due to information overflow. Let's only plot the prediction intervals for three sectors. In the R chunk below we use the `sapply` command. `apply` type functions can be viewed as a substitute to the loop and are often computationally more efficient.

```

plot(log(Sales)~log(Assets), forbes, col = forbes$sector, pch = 19)
legend("topleft", legend = levels(forbes$sector), col = 1:nlevels(forbes$sector),
      pch = 19, cex = 0.8)
lapply(c(2,3,8), plot_confidence)

```



```
## [[1]]
## NULL
##
## [[2]]
## NULL
##
## [[3]]
## NULL
```

3. Are other variables significant?

```
lm3a <- lm(log(Sales)~(log(Assets)+log(Employees)+log(Cash.Flow)+
                      log(Profits)+log(Market.Value))*sector, forbes, na.action = )
Anova(lm3a)
```

```
## Note: model has aliased coefficients
##      sums of squares computed by model comparison

## Anova Table (Type II tests)
##
## Response: log(Sales)
##
##      Sum Sq Df F value    Pr(>F)
## log(Assets)      0.1327  1  1.8326 0.1879311
## log(Employees)    0.9360  1 12.9309 0.0013869 **
## log(Cash.Flow)    0.0904  1  1.2486 0.2744506
## log(Profits)      0.0365  1  0.5036 0.4844883
## log(Market.Value) 0.0210  1  0.2896 0.5952301
## sector           3.4021  8  5.8751 0.0002869 ***
## log(Assets):sector 0.7837  5  2.1654 0.0904546 .
## log(Employees):sector 0.7707  5  2.1294 0.0950223 .
## log(Cash.Flow):sector 1.1097  5  3.0662 0.0271135 *
## log(Profits):sector 0.8852  5  2.4459 0.0617695 .
```

```
## log(Market.Value):sector 0.1728 5 0.4775 0.7895251
## Residuals 1.8096 25
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

summary(lm3a)

##
## Call:
## lm(formula = log(Sales) ~ (log(Assets) + log(Employees) + log(Cash.Flow) +
## log(Profits) + log(Market.Value)) * sector, data = forbes)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.36129 -0.05652  0.00000  0.06643  0.51219
##
## Coefficients: (8 not defined because of singularities)
##              Estimate Std. Error t value
## (Intercept)    -10.65222    21.31733  -0.500
## log(Assets)       1.18820     3.42516   0.347
## log(Employees)   -3.82151     7.24437  -0.528
## log(Cash.Flow)    4.02263     6.75290   0.596
## log(Profits)    -0.74641     0.72544  -1.029
## log(Market.Value) -0.17727     0.80261  -0.221
## sectorEnergy     16.50212    21.37247   0.772
## sectorFinance     13.97121    21.41017   0.653
## sectorHiTech      11.73447    21.39026   0.549
## sectorManufacturing 15.04533    21.36807   0.704
## sectorMedical       8.39602    24.74028   0.339
## sectorOther       -3.70877    31.28291  -0.119
## sectorRetail      16.03921    19.90827   0.806
## sectorTransportation 6.49592     9.63048   0.675
## log(Assets):sectorEnergy -0.82495     3.45429  -0.239
## log(Assets):sectorFinance -1.16113     3.42851  -0.339
## log(Assets):sectorHiTech  0.29838     3.53057   0.085
## log(Assets):sectorManufacturing -0.68998     3.43245  -0.201
## log(Assets):sectorMedical -0.14030     5.25602  -0.027
## log(Assets):sectorOther -0.23308     3.49189  -0.067
## log(Assets):sectorRetail -1.61685     3.46798  -0.466
## log(Assets):sectorTransportation -0.35736     0.62002  -0.576
## log(Employees):sectorEnergy 4.70485     7.24765   0.649
## log(Employees):sectorFinance 4.19797     7.24835   0.579
## log(Employees):sectorHiTech 3.82270     7.24655   0.528
## log(Employees):sectorManufacturing 4.29748     7.25189   0.593
## log(Employees):sectorMedical 2.78894     6.99533   0.399
## log(Employees):sectorOther 5.03365     7.39198   0.681
## log(Employees):sectorRetail 4.49161     7.13917   0.629
## log(Employees):sectorTransportation NA         NA         NA
## log(Cash.Flow):sectorEnergy -2.82369     6.77741  -0.417
## log(Cash.Flow):sectorFinance -2.18709     6.79354  -0.322
## log(Cash.Flow):sectorHiTech -6.69059     7.04549  -0.950
## log(Cash.Flow):sectorManufacturing -4.75398     6.77000  -0.702
## log(Cash.Flow):sectorMedical -2.21363     7.26272  -0.305
## log(Cash.Flow):sectorOther -6.54255     7.27923  -0.899
## log(Cash.Flow):sectorRetail -2.43601     6.60501  -0.369
```


## log(Cash.Flow):sectorTransportation	NA	NA	NA
## log(Profits):sectorEnergy	-0.59094	0.87133	-0.678
## log(Profits):sectorFinance	-0.27130	1.04558	-0.259
## log(Profits):sectorHiTech	2.77955	1.95598	1.421
## log(Profits):sectorManufacturing	0.85105	0.74433	1.143
## log(Profits):sectorMedical	NA	NA	NA
## log(Profits):sectorOther	-2.61723	4.70904	-0.556
## log(Profits):sectorRetail	NA	NA	NA
## log(Profits):sectorTransportation	NA	NA	NA
## log(Market.Value):sectorEnergy	-0.23077	0.88875	-0.260
## log(Market.Value):sectorFinance	-0.03707	0.87224	-0.042
## log(Market.Value):sectorHiTech	0.19214	1.10948	0.173
## log(Market.Value):sectorManufacturing	0.41673	0.87463	0.476
## log(Market.Value):sectorMedical	NA	NA	NA
## log(Market.Value):sectorOther	5.79333	7.19698	0.805
## log(Market.Value):sectorRetail	NA	NA	NA
## log(Market.Value):sectorTransportation	NA	NA	NA
##	Pr(> t)		
## (Intercept)	0.622		
## log(Assets)	0.732		
## log(Employees)	0.602		
## log(Cash.Flow)	0.557		
## log(Profits)	0.313		
## log(Market.Value)	0.827		
## sectorEnergy	0.447		
## sectorFinance	0.520		
## sectorHiTech	0.588		
## sectorManufacturing	0.488		
## sectorMedical	0.737		
## sectorOther	0.907		
## sectorRetail	0.428		
## sectorTransportation	0.506		
## log(Assets):sectorEnergy	0.813		
## log(Assets):sectorFinance	0.738		
## log(Assets):sectorHiTech	0.933		
## log(Assets):sectorManufacturing	0.842		
## log(Assets):sectorMedical	0.979		
## log(Assets):sectorOther	0.947		
## log(Assets):sectorRetail	0.645		
## log(Assets):sectorTransportation	0.570		
## log(Employees):sectorEnergy	0.522		
## log(Employees):sectorFinance	0.568		
## log(Employees):sectorHiTech	0.602		
## log(Employees):sectorManufacturing	0.559		
## log(Employees):sectorMedical	0.694		
## log(Employees):sectorOther	0.502		
## log(Employees):sectorRetail	0.535		
## log(Employees):sectorTransportation	NA		
## log(Cash.Flow):sectorEnergy	0.681		
## log(Cash.Flow):sectorFinance	0.750		
## log(Cash.Flow):sectorHiTech	0.351		
## log(Cash.Flow):sectorManufacturing	0.489		
## log(Cash.Flow):sectorMedical	0.763		
## log(Cash.Flow):sectorOther	0.377		

```
## log(Cash.Flow):sectorRetail          0.715
## log(Cash.Flow):sectorTransportation   NA
## log(Profits):sectorEnergy             0.504
## log(Profits):sectorFinance            0.797
## log(Profits):sectorHiTech             0.168
## log(Profits):sectorManufacturing      0.264
## log(Profits):sectorMedical            NA
## log(Profits):sectorOther              0.583
## log(Profits):sectorRetail             NA
## log(Profits):sectorTransportation     NA
## log(Market.Value):sectorEnergy        0.797
## log(Market.Value):sectorFinance       0.966
## log(Market.Value):sectorHiTech        0.864
## log(Market.Value):sectorManufacturing 0.638
## log(Market.Value):sectorMedical       NA
## log(Market.Value):sectorOther         0.428
## log(Market.Value):sectorRetail        NA
## log(Market.Value):sectorTransportation NA
##
## Residual standard error: 0.269 on 25 degrees of freedom
## Multiple R-squared:  0.9821, Adjusted R-squared:  0.9499
## F-statistic: 30.51 on 45 and 25 DF,  p-value: 7.238e-14
```

When looking at the summary of model `lm3a` it is clear that some coefficients could not be estimated (Coefficients: (8 not defined because of singularities)). This is due to perfect multicollinearity or said differently, some of the coefficients can be expressed as a linear combination of others. When this happens you might also get a warning stating that coefficients are aliased.

To fix this problem use the `alias` function to find out which coefficients are affected.

```
alias(lm3a)

## Model :
## log(Sales) ~ (log(Assets) + log(Employees) + log(Cash.Flow) +
##   log(Profits) + log(Market.Value)) * sector
##
## Complete :
##                                     (Intercept)
## log(Employees):sectorTransportation      843/169
## log(Cash.Flow):sectorTransportation    36374081/12095508
## log(Profits):sectorMedical              0
## log(Profits):sectorRetail              72875/3097
## log(Profits):sectorTransportation     -2608497/133736
## log(Market.Value):sectorMedical         0
## log(Market.Value):sectorRetail        22326/87919
## log(Market.Value):sectorTransportation 352317227/116471077
##                                     log(Assets)
## log(Employees):sectorTransportation    -11214/11849
## log(Cash.Flow):sectorTransportation    -101954/92861
## log(Profits):sectorMedical              0
## log(Profits):sectorRetail              -2462/2327
## log(Profits):sectorTransportation     -37381/665302
```

```

## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -42341615/21079183
## log(Market.Value):sectorTransportation 158741/214512
## log(Employees)
## log(Employees):sectorTransportation 1
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 371813/45315
## log(Profits):sectorTransportation -163921/19978
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -12105/5741
## log(Market.Value):sectorTransportation 17254/8183
## log(Cash.Flow)
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation 1
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -1090707/153104
## log(Profits):sectorTransportation 2526976/354715
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 568117/228017
## log(Market.Value):sectorTransportation -240465/96512
## log(Profits)
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 1
## log(Profits):sectorTransportation 0
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 0
## log(Market.Value):sectorTransportation 0
## log(Market.Value)
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 0
## log(Profits):sectorTransportation 0
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 1
## log(Market.Value):sectorTransportation 0
## sectorEnergy
## log(Employees):sectorTransportation -843/169
## log(Cash.Flow):sectorTransportation -9704934/3227191
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -156433/6648
## log(Profits):sectorTransportation 91616462715/4697118401
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -13315/52434
## log(Market.Value):sectorTransportation -16365182/5410097
## sectorFinance
## log(Employees):sectorTransportation -843/169
## log(Cash.Flow):sectorTransportation -16964213/5641126
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -156433/6648
## log(Profits):sectorTransportation 204128547449/10465542196

```

```

## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -13315/52434
## log(Market.Value):sectorTransportation -19824544/6553713
## sectorHiTech
## log(Employees):sectorTransportation -843/169
## log(Cash.Flow):sectorTransportation -9704934/3227191
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -156433/6648
## log(Profits):sectorTransportation 2454017359/125815926
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -13315/52434
## log(Market.Value):sectorTransportation -19824544/6553713
## sectorManufacturing
## log(Employees):sectorTransportation -843/169
## log(Cash.Flow):sectorTransportation -9704934/3227191
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -156433/6648
## log(Profits):sectorTransportation 967463228/49601231
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -13315/52434
## log(Market.Value):sectorTransportation -19824544/6553713
## sectorMedical
## log(Employees):sectorTransportation -843/169
## log(Cash.Flow):sectorTransportation -99494995/33085166
## log(Profits):sectorMedical 7185/2641
## log(Profits):sectorRetail -262536255/10000849
## log(Profits):sectorTransportation 1908835420/97864791
## log(Market.Value):sectorMedical 71053997/4069157
## log(Market.Value):sectorRetail -157438/8887
## log(Market.Value):sectorTransportation -68786515/22739846
## sectorOther
## log(Employees):sectorTransportation -843/169
## log(Cash.Flow):sectorTransportation -21622465/7190139
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -156433/6648
## log(Profits):sectorTransportation 6102732410/312883251
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -13315/52434
## log(Market.Value):sectorTransportation -128928022/42621775
## sectorRetail
## log(Employees):sectorTransportation -843/169
## log(Cash.Flow):sectorTransportation -9704934/3227191
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -156433/6648
## log(Profits):sectorTransportation 10738314672/550546637
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -13315/52434
## log(Market.Value):sectorTransportation -18094863/5981905
## sectorTransportation
## log(Employees):sectorTransportation -843/169
## log(Cash.Flow):sectorTransportation -9704934/3227191
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -971614/91981
## log(Profits):sectorTransportation 19115593/2924130

```

```

## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 207809/198959
## log(Market.Value):sectorTransportation -6101734/1411343
## log(Assets):sectorEnergy
## log(Employees):sectorTransportation 4309/4553
## log(Cash.Flow):sectorTransportation 210983/192166
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 2462/2327
## log(Profits):sectorTransportation 76261/1357283
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 66206473/32959970
## log(Market.Value):sectorTransportation -498686/673891
## log(Assets):sectorFinance
## log(Employees):sectorTransportation 4309/4553
## log(Cash.Flow):sectorTransportation 210983/192166
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 2462/2327
## log(Profits):sectorTransportation 76261/1357283
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 60818372/30277579
## log(Market.Value):sectorTransportation -498686/673891
## log(Assets):sectorHiTech
## log(Employees):sectorTransportation 4309/4553
## log(Cash.Flow):sectorTransportation 210983/192166
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 2462/2327
## log(Profits):sectorTransportation 113642/2022585
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 373633461/186008212
## log(Market.Value):sectorTransportation -498686/673891
## log(Assets):sectorManufacturing
## log(Employees):sectorTransportation 4309/4553
## log(Cash.Flow):sectorTransportation 210983/192166
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 2462/2327
## log(Profits):sectorTransportation 76261/1357283
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 204267001/101691480
## log(Market.Value):sectorTransportation -498686/673891
## log(Assets):sectorMedical
## log(Employees):sectorTransportation 4309/4553
## log(Cash.Flow):sectorTransportation 210983/192166
## log(Profits):sectorMedical -43764/86764247
## log(Profits):sectorRetail 1560546/1474273
## log(Profits):sectorTransportation 113642/2022585
## log(Market.Value):sectorMedical -55982/16383
## log(Market.Value):sectorRetail 9137/1684
## log(Market.Value):sectorTransportation -498686/673891
## log(Assets):sectorOther
## log(Employees):sectorTransportation 4309/4553
## log(Cash.Flow):sectorTransportation 210983/192166
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 2462/2327
## log(Profits):sectorTransportation 76261/1357283

```

```

## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 1005964395/500805356
## log(Market.Value):sectorTransportation -498686/673891
## log(Assets):sectorRetail
## log(Employees):sectorTransportation 4309/4553
## log(Cash.Flow):sectorTransportation 210983/192166
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 2462/2327
## log(Profits):sectorTransportation 76261/1357283
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 373633461/186008212
## log(Market.Value):sectorTransportation -498686/673891
## log(Assets):sectorTransportation
## log(Employees):sectorTransportation 4309/4553
## log(Cash.Flow):sectorTransportation 210983/192166
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 782311/1389520
## log(Profits):sectorTransportation 393/713
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 1698/31097
## log(Market.Value):sectorTransportation 1857817/1530226
## log(Employees):sectorEnergy
## log(Employees):sectorTransportation -1
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -163921/19978
## log(Profits):sectorTransportation 371813/45315
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 17254/8183
## log(Market.Value):sectorTransportation -12105/5741
## log(Employees):sectorFinance
## log(Employees):sectorTransportation -1
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -163921/19978
## log(Profits):sectorTransportation 371813/45315
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 17254/8183
## log(Market.Value):sectorTransportation -12105/5741
## log(Employees):sectorHiTech
## log(Employees):sectorTransportation -1
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -163921/19978
## log(Profits):sectorTransportation 371813/45315
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 17254/8183
## log(Market.Value):sectorTransportation -12105/5741
## log(Employees):sectorManufacturing
## log(Employees):sectorTransportation -1
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -163921/19978
## log(Profits):sectorTransportation 371813/45315

```

```

## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 17254/8183
## log(Market.Value):sectorTransportation -12105/5741
## log(Employees):sectorMedical
## log(Employees):sectorTransportation -1
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 1613/2682
## log(Profits):sectorRetail -1326566/150635
## log(Profits):sectorTransportation 371813/45315
## log(Market.Value):sectorMedical 392260/114339
## log(Market.Value):sectorRetail -5245/3967
## log(Market.Value):sectorTransportation -12105/5741
## log(Employees):sectorOther
## log(Employees):sectorTransportation -1
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -163921/19978
## log(Profits):sectorTransportation 371813/45315
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 17254/8183
## log(Market.Value):sectorTransportation -12105/5741
## log(Employees):sectorRetail
## log(Employees):sectorTransportation -1
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -163921/19978
## log(Profits):sectorTransportation 371813/45315
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 17254/8183
## log(Market.Value):sectorTransportation -12105/5741
## log(Cash.Flow):sectorEnergy
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation -1
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 2526976/354715
## log(Profits):sectorTransportation -1090707/153104
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -240465/96512
## log(Market.Value):sectorTransportation 568117/228017
## log(Cash.Flow):sectorFinance
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation -1
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 2526976/354715
## log(Profits):sectorTransportation -1090707/153104
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -240465/96512
## log(Market.Value):sectorTransportation 568117/228017
## log(Cash.Flow):sectorHiTech
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation -1
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 2526976/354715
## log(Profits):sectorTransportation -1090707/153104

```

```

## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -240465/96512
## log(Market.Value):sectorTransportation 568117/228017
## log(Cash.Flow):sectorManufacturing
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation -1
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 2526976/354715
## log(Profits):sectorTransportation -1090707/153104
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -240465/96512
## log(Market.Value):sectorTransportation 568117/228017
## log(Cash.Flow):sectorMedical
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation -1
## log(Profits):sectorMedical 153757/2141490
## log(Profits):sectorRetail 10230029/1450623
## log(Profits):sectorTransportation -1090707/153104
## log(Market.Value):sectorMedical 4336617/4319816
## log(Market.Value):sectorRetail -21102/6037
## log(Market.Value):sectorTransportation 568117/228017
## log(Cash.Flow):sectorOther
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation -1
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 2526976/354715
## log(Profits):sectorTransportation -1090707/153104
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -240465/96512
## log(Market.Value):sectorTransportation 568117/228017
## log(Cash.Flow):sectorRetail
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation -1
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 2526976/354715
## log(Profits):sectorTransportation -1090707/153104
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -240465/96512
## log(Market.Value):sectorTransportation 568117/228017
## log(Profits):sectorEnergy
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -1
## log(Profits):sectorTransportation 0
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 0
## log(Market.Value):sectorTransportation 0
## log(Profits):sectorFinance
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -1
## log(Profits):sectorTransportation 0

```



```

## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 0
## log(Market.Value):sectorTransportation 0
## log(Profits):sectorHiTech
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -1
## log(Profits):sectorTransportation 0
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 0
## log(Market.Value):sectorTransportation 0
## log(Profits):sectorManufacturing
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -1
## log(Profits):sectorTransportation 0
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 0
## log(Market.Value):sectorTransportation 0
## log(Profits):sectorOther
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail -1
## log(Profits):sectorTransportation 0
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail 0
## log(Market.Value):sectorTransportation 0
## log(Market.Value):sectorEnergy
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 0
## log(Profits):sectorTransportation 0
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -1
## log(Market.Value):sectorTransportation 0
## log(Market.Value):sectorFinance
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 0
## log(Profits):sectorTransportation 0
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -1
## log(Market.Value):sectorTransportation 0
## log(Market.Value):sectorHiTech
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 0
## log(Profits):sectorTransportation 0

```

```
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -1
## log(Market.Value):sectorTransportation 0
## log(Market.Value):sectorManufacturing
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 0
## log(Profits):sectorTransportation 0
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -1
## log(Market.Value):sectorTransportation 0
## log(Market.Value):sectorOther
## log(Employees):sectorTransportation 0
## log(Cash.Flow):sectorTransportation 0
## log(Profits):sectorMedical 0
## log(Profits):sectorRetail 0
## log(Profits):sectorTransportation 0
## log(Market.Value):sectorMedical 0
## log(Market.Value):sectorRetail -1
## log(Market.Value):sectorTransportation 0
```

Let's remove all the affected interaction effects from our model (as shown in the alias table above). Afterwards we can reduce the model using backwards selection. Again `drop1` helps us not to make mistakes during this process.

```
lm3b <- update(lm3a, ~.-log(Market.Value):sector-log(Profits):sector-
               log(Cash.Flow):sector-log(Employees):sector)
drop1(lm3b, test = "F")
```

```
## Single term deletions
##
## Model:
## log(Sales) ~ log(Assets) + log(Employees) + log(Cash.Flow) +
## log(Profits) + log(Market.Value) + sector + log(Assets):sector
## Df Sum of Sq RSS AIC F value Pr(>F)
## <none> 8.2636 -108.708
## log(Employees) 1 1.93403 10.1976 -95.777 11.4681 0.001403 **
## log(Cash.Flow) 1 0.12489 8.3885 -109.643 0.7405 0.393683
## log(Profits) 1 0.04449 8.3081 -110.327 0.2638 0.609816
## log(Market.Value) 1 0.00000 8.2636 -110.708 0.0000 0.997347
## log(Assets):sector 8 0.92658 9.1902 -117.163 0.6868 0.701003
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
drop1(lm3c <- update(lm3b, ~.-log(Market.Value)), test = "F")
```

```
## Single term deletions
##
## Model:
## log(Sales) ~ log(Assets) + log(Employees) + log(Cash.Flow) +
## log(Profits) + sector + log(Assets):sector
## Df Sum of Sq RSS AIC F value Pr(>F)
## <none> 8.2636 -110.708
## log(Employees) 1 2.15058 10.4142 -96.285 13.0124 0.000714 ***
```

```
## log(Cash.Flow)      1  0.13596  8.3996 -111.550  0.8227 0.368750
## log(Profits)        1  0.05035  8.3139 -112.277  0.3047 0.583428
## log(Assets):sector  8  0.92700  9.1906 -119.159  0.7011 0.688913
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

drop1(lm3d <- update(lm3c, ~.-log(Assets):sector), test = "F")

## Single term deletions
##
## Model:
## log(Sales) ~ log(Assets) + log(Employees) + log(Cash.Flow) +
##   log(Profits) + sector
##           Df Sum of Sq    RSS      AIC F value    Pr(>F)
## <none>                9.1906 -119.16
## log(Assets)      1    1.1712 10.3618 -112.64   7.3912 0.0086308 **
## log(Employees)   1    2.1742 11.3648 -106.08  13.7208 0.0004749 ***
## log(Cash.Flow)   1    0.1950  9.3856 -119.67   1.2308 0.2718254
## log(Profits)     1    0.0214  9.2120 -121.00   0.1349 0.7147754
## sector           8    3.7795 12.9701 -110.70   2.9815 0.0072313 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

drop1(lm3e <- update(lm3d, ~.-log(Profits)), test = "F")

## Single term deletions
##
## Model:
## log(Sales) ~ log(Assets) + log(Employees) + log(Cash.Flow) +
##   sector
##           Df Sum of Sq    RSS      AIC F value    Pr(>F)
## <none>                9.212 -121.00
## log(Assets)      1    1.1655 10.377 -114.54   7.4646 0.008288 **
## log(Employees)   1    2.4092 11.621 -106.50  15.4300 0.000227 ***
## log(Cash.Flow)   1    0.3360  9.548 -120.45   2.1520 0.147691
## sector           8    4.1974 13.409 -110.34   3.3604 0.003084 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

drop1(lm3f <- update(lm3e, ~.-log(Cash.Flow)), test = "F")

## Single term deletions
##
## Model:
## log(Sales) ~ log(Assets) + log(Employees) + sector
##           Df Sum of Sq    RSS      AIC F value    Pr(>F)
## <none>                9.548 -120.451
## log(Assets)      1    1.8946 11.443 -109.599 11.9057 0.001031 **
## log(Employees)   1    4.3814 13.929  -95.636 27.5328 2.126e-06 ***
## sector           8    6.7310 16.279  -98.569  5.2872 4.960e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Everything else is significant. Let's look at our final model lm3f.

```
Anova(lm3f)
```

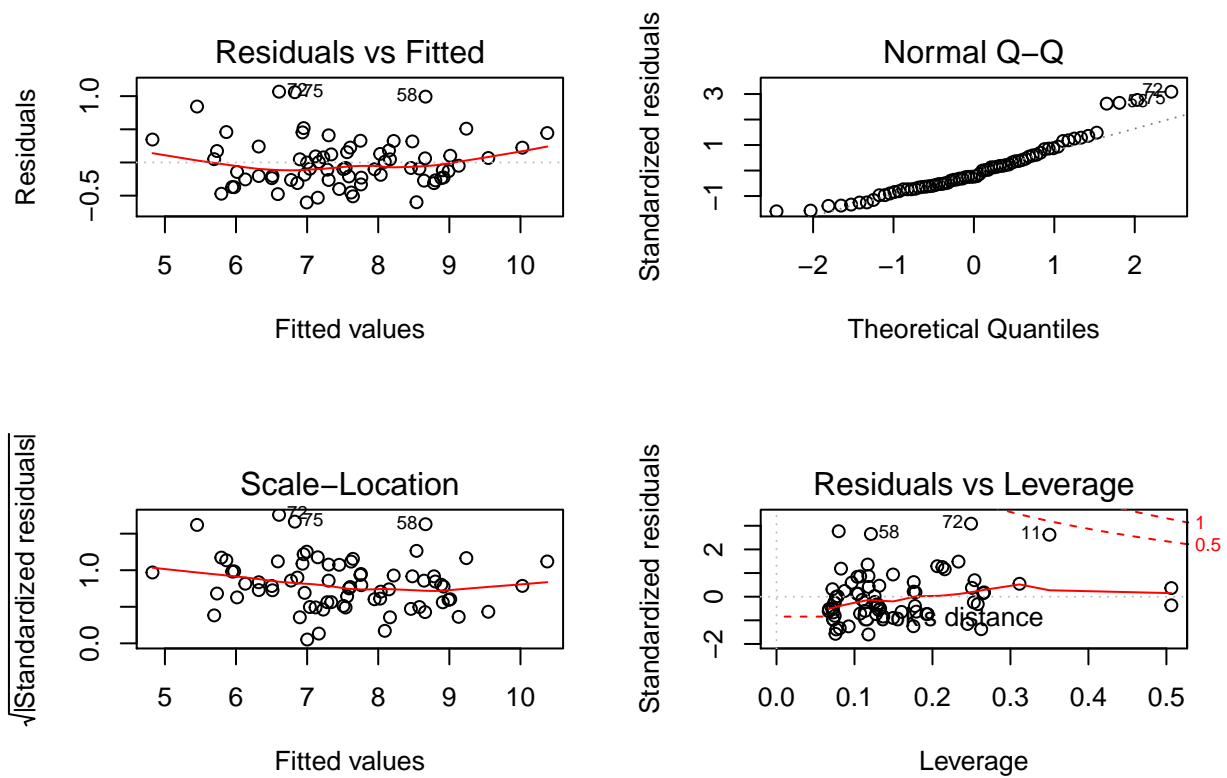
```
## Anova Table (Type II tests)
##
## Response: log(Sales)
##           Sum Sq Df F value    Pr(>F)
## log(Assets)  1.8946  1 11.9057  0.001031 **
## log(Employees) 4.3814  1 27.5328 2.126e-06 ***
## sector       6.7310  8  5.2872 4.960e-05 ***
## Residuals    9.5480 60
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
summary(lm3f)
```

```
##
## Call:
## lm(formula = log(Sales) ~ log(Assets) + log(Employees) + sector,
##     data = forbes)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.60248 -0.24602 -0.08277  0.17644  1.06718
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    3.76251    0.61976   6.071 9.35e-08 ***
## log(Assets)     0.31052    0.09000   3.450 0.00103 **
## log(Employees)  0.49786    0.09488   5.247 2.13e-06 ***
## sectorEnergy    0.29063    0.32823   0.885 0.37945
## sectorFinance  -0.62102    0.34398  -1.805 0.07603 .
## sectorHiTech    0.25597    0.33077   0.774 0.44205
## sectorManufacturing 0.31084    0.32633   0.953 0.34466
## sectorMedical  -0.38089    0.36805  -1.035 0.30487
## sectorOther     0.10317    0.34335   0.300 0.76484
## sectorRetail    0.34654    0.35679   0.971 0.33531
## sectorTransportation 0.23088    0.35696   0.647 0.52024
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.3989 on 60 degrees of freedom
## Multiple R-squared:  0.9056, Adjusted R-squared:  0.8899
## F-statistic: 57.59 on 10 and 60 DF, p-value: < 2.2e-16
```

90.6% Variance in log(Sales) is explained by model lm3f. Let's look at our model diagnostics.

```
par(mfrow=c(2,2))
plot(lm3f)
```



One could argue that the outliers 72 and 11 should be removed due to their high Cook's distances. Proceed with re-modelling thereafter and check model assumptions again...