

Lab6

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setwd ("C:/Users/usuario/OneDrive - University of East Anglia/PhD/First Semestre/Econometrics/Laboratories/Lab6") list.files()

```
library(data.table)

## Warning: package 'data.table' was built under R version 4.1.1

library(ggplot2)
library(stargazer)

## Warning: package 'stargazer' was built under R version 4.1.1

##
## Please cite as:

## Hlavac, Marek (2018). stargazer: Well-Formatted Regression and Summary
Statistics Tables.

## R package version 5.2.2. https://CRAN.R-project.org/package=stargazer

load("C:/Users/usuario/OneDrive - University of East Anglia/PhD/First
Semestre/Econometrics/Laboratories/Lab6/hprice1.RData")
dt.houseprice <- data.table(data)
rm(data)

#EXERCISE 1
# Estimate price =  $B_0 + B_1 \text{sqrft} + B_2 \text{bdrms} + u$ 

summary(dt.houseprice)
```

	price	assess	bdrms	lotsize	sqrft
## Min.	:111.0	Min. :198.7	Min. :2.000	Min. : 1000	Min. :1171
## 1st Qu.	:230.0	1st Qu.:253.9	1st Qu.:3.000	1st Qu.: 5733	1st Qu.:1660
## Median	:265.5	Median :290.2	Median :3.000	Median : 6430	Median :1845
## Mean	:293.5	Mean :315.7	Mean :3.568	Mean : 9020	Mean :2014
## 3rd Qu.	:326.2	3rd Qu.:352.1	3rd Qu.:4.000	3rd Qu.: 8583	3rd Qu.:2227
## Max.	:725.0	Max. :708.6	Max. :7.000	Max. :92681	Max. :3880

```
##      colonial      lprice      lassess      llotsize
## Min.   :0.0000   Min.   :4.710   Min.   :5.292   Min.    : 6.908
## 1st Qu.:0.0000   1st Qu.:5.438   1st Qu.:5.537   1st Qu.: 8.654
## Median :1.0000   Median :5.582   Median :5.671   Median : 8.769
## Mean   :0.6932   Mean    :5.633   Mean    :5.718   Mean    : 8.905
## 3rd Qu.:1.0000   3rd Qu.:5.788   3rd Qu.:5.864   3rd Qu.: 9.058
## Max.   :1.0000   Max.    :6.586   Max.    :6.563   Max.    :11.437
##      lsqrft
## Min.    :7.066
## 1st Qu.:7.415
## Median :7.520
## Mean    :7.573
## 3rd Qu.:7.708
## Max.    :8.264
```

```
lm.price <- lm(price~ sqrft + bdrms, data = dt.houseprice)
stargazer(lm.price, type = 'text')
```

```
##
## =====
##                      Dependent variable:
##                      -----
##                      price
## -----
## sqrft                0.128***
##                      (0.014)
##
## bdrms                15.198
##                      (9.484)
##
## Constant             -19.315
##                      (31.047)
##
## -----
## Observations          88
## R2                    0.632
## Adjusted R2           0.623
## Residual Std. Error   63.045 (df = 85)
## F Statistic           72.964*** (df = 2; 85)
## =====
## Note:                 *p<0.1; **p<0.05; ***p<0.01
```

#Answ i): Price = -19.315 + 0.128sqrft + 15.198bdrms

```
average.house <- data.table( sqrft = 2014, bdrms =3.568 )
average.house
```

```
##      sqrft bdrms
## 1:   2014 3.568
```

```

prediAverage <- predict(lm.price, newdata = average.house)
prediAverage

##          1
## 293.5827

new.house1 <- data.table( sqrft = 2014, bdrms =4.568 )
new.house1

##    sqrft bdrms
## 1:  2014 4.568

predinew1 <- predict(lm.price, newdata = new.house1)
predinew1

##          1
## 308.7809

#Answ ii): $15.198 $308.7809 (thousand of dollars)

new.house2 <- data.table( sqrft = 2014+140, bdrms =4.568 )
new.house2

##    sqrft bdrms
## 1:  2154 4.568

predinew2 <- predict(lm.price, newdata = new.house2)
predinew2

##          1
## 326.7619

Dif<-predinew2-predinew1
Dif

##          1
## 17.98107

#Answ iii): $326.7619 (thousand of dollars) is 17.98 (thousand of dollars)
greater 2 than 1

#Answ iv): 63.2%

new.house3 <- data.table( sqrft = 2483, bdrms =4 )
new.house3

##    sqrft bdrms
## 1:  2483     4

predinew3 <- predict(lm.price, newdata = new.house3)
predinew3

```

```
##          1
## 360.3849

#Answ v): $360.3849 (thousand of dollars)

360.3849-300

## [1] 60.3849

#Answ vi): 60.39 (thousand of dollars) wich suggest that the buyer overpaid
the house

lm.price2 <- lm(price~ sqrft + bdrms + colonial, data = dt.houseprice)
stargazer(lm.price2, type = 'text')

##
## =====
##                               Dependent variable:
##                               -----
##                               price
## -----
##  sqrft                        0.130***
##                               (0.014)
##
##  bdrms                        12.487
##                               (10.024)
##
##  colonial                     13.078
##                               (15.436)
##
##  Constant                     -21.552
##                               (31.210)
##
## -----
##  Observations                  88
##  R2                           0.635
##  Adjusted R2                   0.622
##  Residual Std. Error    63.150 (df = 84)
##  F Statistic             48.720*** (df = 3; 84)
##  =====
## Note:                *p<0.1; **p<0.05; ***p<0.01

dt.houseprice[colonial==0, mean(price)]

## [1] 272.3704

lm.price3 <- lm(price~ colonial, data = dt.houseprice)
stargazer(lm.price3, type = 'text')

##
## =====
```

```
##                               Dependent variable:
##                               -----
##                               price
## -----
## colonial                      30.548
##                               (23.652)
##
## Constant                      272.370***
##                               (19.692)
##
## -----
## Observations                  88
## R2                           0.019
## Adjusted R2                   0.008
## Residual Std. Error          102.321 (df = 86)
## F Statistic                   1.668 (df = 1; 86)
## =====
## Note:                        *p<0.1; **p<0.05; ***p<0.01
```

#Answ vii): In Average houses that are in colonial residences pay 13.078 (thousand of dollars)

#EXERCISE 2

```
load("C:/Users/usuario/OneDrive - University of East Anglia/PhD/First
Semestre/Econometrics/Laboratories/Lab6/ceosal2.RData")
dt.ceosal <- data.table(data)
rm(data)
```

#i) Estimate a model relating annual salary to firm sales and market value. Make the model of the constant elasticity variety for both independent variables. Write the results out in equation form.

```
lm.ceosalary <- lm(lsalary~ lsales + lmktval, data = dt.ceosal)
stargazer(lm.ceosalary, type = 'text')
```

```
##
## =====
##                               Dependent variable:
##                               -----
##                               lsalary
## -----
## lsales                      0.162***
##                               (0.040)
##
## lmktval                     0.107**
##                               (0.050)
##
## Constant                     4.621***
```

```
## (0.254)
##
## -----
## Observations          177
## R2                    0.299
## Adjusted R2           0.291
## Residual Std. Error   0.510 (df = 174)
## F Statistic           37.129*** (df = 2; 174)
## =====
## Note:                  *p<0.1; **p<0.05; ***p<0.01
```

#Answ i): $lsalary = 4.621 + 0.162lsales + 0.107lmktval$

```
lm.ceosalary1 <- lm(lsalary~ lsales + lmktval + profits, data = dt.ceosal)
stargazer(lm.ceosalary1, type = 'text')
```

```
##
## =====
##                      Dependent variable:
##                      -----
##                      lsalary
## -----
## lsales                0.161***
##                      (0.040)
##
## lmktval               0.098
##                      (0.064)
##
## profits               0.00004
##                      (0.0002)
##
## Constant              4.687***
##                      (0.380)
##
## -----
## Observations          177
## R2                    0.299
## Adjusted R2           0.287
## Residual Std. Error   0.512 (df = 173)
## F Statistic           24.636*** (df = 3; 173)
## =====
## Note:                  *p<0.1; **p<0.05; ***p<0.01
```

#Answ ii): Profits are not in Logarithmic form, ii) No because R2 is 0.29

#Answ iii) the coefficient is significant because the pvalue is < 0.05. An increment in 1% of the market value increases the CEO salary in 0.162%

```
exp(0.00004)
```

```
## [1] 1.00004
```

#Answ iv) the coefficient is significant because the pvalue is > 0.05. An increment in one unit of the profits increases the CEO salary in 0.004%

```
lm.ceosalary2 <- lm(lsalary~ lsales + lmktval + profits + ceoten + comten,  
data = dt.ceosal)  
stargazer(lm.ceosalary2, type = 'text')
```

```
##  
## =====  
##                               Dependent variable:  
##                               -----  
##                               lsalary  
## -----  
## lsales                        0.191***  
##                               (0.040)  
##  
## lmktval                      0.077  
##                               (0.062)  
##  
## profits                      0.0001  
##                               (0.0001)  
##  
## ceoten                      0.017***  
##                               (0.006)  
##  
## comten                      -0.010***  
##                               (0.003)  
##  
## Constant                    4.697***  
##                               (0.376)  
##  
## -----  
## Observations                177  
## R2                          0.349  
## Adjusted R2                 0.330  
## Residual Std. Error        0.496 (df = 171)  
## F Statistic                18.342*** (df = 5; 171)  
## =====  
## Note:                       *p<0.1; **p<0.05; ***p<0.01
```

```
stargazer(lm.ceosalary, lm.ceosalary1, lm.ceosalary2, type = 'text')
```

```
##  
##  
## =====  
##                               Dependent variable:  
##                               -----  
## -----
```

```

##                                     lsalary
##                                     (2)
##                                     (1)
## (3) -----
## -----
## lsales                0.162***          0.161***
## 0.191***
##                (0.040)          (0.040)
## (0.040)
##
## lmktval                0.107**          0.098
## 0.077
##                (0.050)          (0.064)
## (0.062)
##
## profits                0.00004
## 0.0001
##                (0.0002)
## (0.0001)
##
## ceoten                0.017***
##
## (0.006)
##
## comten                -0.010***
##
## (0.003)
##
## Constant                4.621***          4.687***
## 4.697***
##                (0.254)          (0.380)
## (0.376)
## -----
## -----
## Observations                177          177
## 177
## R2                0.299          0.299
## 0.349
## Adjusted R2                0.291          0.287
## 0.330
## Residual Std. Error    0.510 (df = 174)    0.512 (df = 173)
## 0.496 (df = 171)
## F Statistic        37.129*** (df = 2; 174) 24.636*** (df = 3; 173)
## 18.342*** (df = 5; 171)
##
=====
=====

```



```
## Note:
**p<0.05; ***p<0.01
```

*p<0.1;

#Answ v) 1.7%

*##Answ vi) An increment in 1 year of tenure increases the CEO salary in 1.7%,
An increment in 1 year of comten decreases the CEO salary in 1%. Both
variables comten and ceoten are significant*

```
summary(dt.ceosal)
```

```
##      salary      age      college      grad
## Min.   : 100.0   Min.   :33.00   Min.   :0.0000   Min.   :0.0000
## 1st Qu.: 471.0   1st Qu.:52.00   1st Qu.:1.0000   1st Qu.:0.0000
## Median : 707.0   Median :57.00   Median :1.0000   Median :1.0000
## Mean   : 865.9   Mean   :56.43   Mean   :0.9718   Mean   :0.5311
## 3rd Qu.:1119.0   3rd Qu.:62.00   3rd Qu.:1.0000   3rd Qu.:1.0000
## Max.   :5299.0   Max.   :86.00   Max.   :1.0000   Max.   :1.0000
##      comten      ceoten      sales      profits
## Min.   : 2.0     Min.   : 0.000   Min.   : 29     Min.   : -463.0
## 1st Qu.:12.0     1st Qu.: 3.000   1st Qu.: 561    1st Qu.: 34.0
## Median :23.0     Median : 6.000   Median :1400    Median : 63.0
## Mean   :22.5     Mean   : 7.955   Mean   :3529    Mean   :207.8
## 3rd Qu.:33.0     3rd Qu.:11.000   3rd Qu.:3500    3rd Qu.:208.0
## Max.   :58.0     Max.   :37.000   Max.   :51300   Max.   :2700.0
##      mktval      lsalary      lsales      lmktval
## Min.   : 387     Min.   :4.605   Min.   : 3.367   Min.   : 5.958
## 1st Qu.: 644     1st Qu.:6.155   1st Qu.: 6.330   1st Qu.: 6.468
## Median :1200     Median :6.561   Median : 7.244   Median : 7.090
## Mean   :3600     Mean   :6.583   Mean   : 7.231   Mean   : 7.399
## 3rd Qu.:3500     3rd Qu.:7.020   3rd Qu.: 8.161   3rd Qu.: 8.161
## Max.   :45400    Max.   :8.575   Max.   :10.845   Max.   :10.723
##      comtensq      ceotensq      profmarg
## Min.   : 4.0     Min.   : 0.0    Min.   : -203.077
## 1st Qu.:144.0     1st Qu.: 9.0    1st Qu.: 4.231
## Median :529.0     Median :36.0    Median : 6.834
## Mean   :656.7     Mean   :114.1   Mean   : 6.420
## 3rd Qu.:1089.0    3rd Qu.:121.0   3rd Qu.:10.947
## Max.   :3364.0    Max.   :1369.0   Max.   : 47.458
```

*#vii) The average of comten is 22.5 years. It could be that after a certain
amount of time productivity of CEO decreases a higher rate that increases in
the initial years.*

viii)Predict te values of CEO

```
fit.val <- fitted.values(lm.ceosalary2)
```

```
fit.val
```

```
##      1      2      3      4      5      6      7      8
## 7.151210 6.393152 6.185158 6.727215 6.303558 7.308279 6.231130 6.748995
##      9     10     11     12     13     14     15     16
## 6.424304 6.585552 6.408581 6.145786 6.752264 6.879309 6.845969 6.175660
##     17     18     19     20     21     22     23     24
## 6.295709 6.403426 6.671835 6.235924 6.693955 6.456440 6.504184 6.062382
##     25     26     27     28     29     30     31     32
## 7.187849 6.022501 6.737172 6.053943 6.745318 7.080347 7.106430 6.788409
##     33     34     35     36     37     38     39     40
## 6.696870 7.412483 6.645554 7.005641 6.414824 5.972263 7.110380 7.505438
##     41     42     43     44     45     46     47     48
## 6.589509 6.651134 7.254880 7.492498 6.567925 6.131221 7.361337 6.146312
##     49     50     51     52     53     54     55     56
## 6.479542 7.291735 6.698195 7.070538 6.314608 6.640608 6.321764 7.146681
##     57     58     59     60     61     62     63     64
## 6.450431 6.881708 6.147497 7.110562 6.500695 7.250963 6.593601 6.249849
##     65     66     67     68     69     70     71     72
## 6.819407 6.139394 6.320636 6.481749 6.149719 6.417198 6.656593 6.862629
##     73     74     75     76     77     78     79     80
## 5.978020 6.504380 6.308556 6.239654 7.205570 6.311297 6.739545 6.590062
##     81     82     83     84     85     86     87     88
## 6.743376 6.490247 6.668438 6.583575 6.377373 5.897672 6.461920 6.887926
##     89     90     91     92     93     94     95     96
## 6.306490 6.533651 6.655575 6.812945 7.237178 6.889879 6.324359 6.268027
##     97     98     99    100    101    102    103    104
## 6.798066 6.429280 6.108671 6.749980 7.067770 6.225946 6.575369 6.202740
##    105    106    107    108    109    110    111    112
## 6.675238 7.088067 7.197106 6.265588 6.321395 6.779425 6.510574 6.553468
##    113    114    115    116    117    118    119    120
## 7.135835 6.569187 6.370065 6.486983 6.014815 6.407470 6.904112 6.135279
##    121    122    123    124    125    126    127    128
## 6.880983 6.343876 6.541797 6.135624 6.677713 6.374578 7.262607 6.816456
##    129    130    131    132    133    134    135    136
## 6.783862 6.312388 6.569609 6.426906 6.570901 6.388564 7.347109 6.696450
##    137    138    139    140    141    142    143    144
## 6.218194 6.645024 6.609878 6.903268 6.586107 6.571357 6.948757 6.783637
##    145    146    147    148    149    150    151    152
## 6.849768 6.312563 6.305372 6.602141 6.216801 6.562929 6.067363 6.249218
##    153    154    155    156    157    158    159    160
## 7.121833 6.786744 6.749534 6.702540 6.753174 6.808535 6.159579 6.383498
##    161    162    163    164    165    166    167    168
## 6.158513 6.831989 6.582561 6.877790 6.518624 6.560536 6.638389 6.012189
##    169    170    171    172    173    174    175    176
## 6.628056 6.252875 6.371145 6.124556 5.936547 6.102017 6.249027 6.385066
##    177
## 6.052239
```

```
head(fit.val)
```

```

##           1           2           3           4           5           6
## 7.151210 6.393152 6.185158 6.727215 6.303558 7.308279

new.house4 <- data.table(lsales = 8.7323, lmktval=10.0519, profits=966,
ceoten=2, comten=9)
new.house4

##      lsales lmktval profits ceoten comten
## 1: 8.7323 10.0519     966      2      9

predinew4 <- predict(lm.ceosalary2, newdata = new.house4)
predinew4

##           1
## 7.151208

#ix) Estimate R2

summary(lm.ceosalary2)$r.squared

## [1] 0.3490874

summary(lm.ceosalary2)

##
## Call:
## lm(formula = lsalary ~ lsales + lmktval + profits + ceoten +
##      comten, data = dt.ceosal)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.5307 -0.2737 -0.0251  0.2910  1.9999
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  4.697e+00  3.759e-01  12.497 < 2e-16 ***
## lsales       1.909e-01  3.998e-02   4.774 3.86e-06 ***
## lmktval      7.726e-02  6.237e-02   1.239  0.21713
## profits      6.456e-05  1.479e-04   0.437  0.66298
## ceoten       1.694e-02  5.552e-03   3.052  0.00264 **
## comten      -9.539e-03  3.354e-03  -2.844  0.00500 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4961 on 171 degrees of freedom
## Multiple R-squared:  0.3491, Adjusted R-squared:  0.3301
## F-statistic: 18.34 on 5 and 171 DF,  p-value: 1.473e-14

#x) Calculate the residuals
residuals<-residuals(lm.ceosalary2)
residuals

```

##	1	2	3	4	5
6					
##	-0.094172931	0.003777270	-0.247622247	-0.248705293	-0.094967530 -
	0.335672471				
##	7	8	9	10	11
12					
##	0.620054466	0.390665454	-0.203713548	0.412043532	-0.009986326 -
	0.273667839				
##	13	14	15	16	17
18					
##	0.337812440	-0.332523520	0.101968184	-0.674401819	0.409929626
	1.020142090				
##	19	20	21	22	23
24					
##	0.206491280	0.175894249	-0.541222088	0.308599141	0.118552765 -
	0.557050907				
##	25	26	27	28	29
30					
##	-0.472465176	-0.141967576	0.320726309	-0.455520992	-0.025098407 -
	1.376564474				
##	31	32	33	34	35
36					
##	0.288062820	0.332035064	-0.405300736	0.081946817	-0.484346636
	0.191794369				
##	37	38	39	40	41
42					
##	-0.121404278	-1.112450520	0.328004060	-0.038067324	-0.153358342
	0.022163855				
##	43	44	45	46	47
48					
##	0.049635426	0.118849795	0.778084795	-0.137259210	-0.195071349 -
	0.039289284				
##	49	50	51	52	53
54					
##	-0.357049148	-0.251198272	-0.340352206	-0.673608656	0.160824865
	0.071132950				
##	55	56	57	58	59
60					
##	0.662952374	0.313809714	-0.085680572	-0.066067924	0.329475110
	0.585195406				
##	61	62	63	64	65
66					
##	-0.088876225	0.322568110	-0.280053207	-0.073981637	-0.554106143
	0.015463781				
##	67	68	69	70	71
72					
##	0.125083233	-0.048808517	0.757035957	-0.045586103	0.352816281
	0.139527065				
##	73	74	75	76	77
78					

0.074069206 1.430133545 -0.450622767 -0.345251508 0.519760010 -
0.379051946
79 80 81 82 83
84
0.039239503 -0.010811116 0.113086458 0.551164892 0.301352756
0.549721205
85 86 87 88 89
90
-0.241807799 -0.738616405 -0.300713147 0.241371765 0.697484367 -
0.681448384
91 92 93 94 95
96
-0.178602606 -0.038720980 0.140581332 0.423341800 -0.546706307 -
0.138976570
97 98 99 100 101
102
0.031728227 -0.502354298 -0.006112680 0.450444734 0.399028815 -
0.029502190
103 104 105 106 107
108
1.999904092 -0.136631803 -0.083564109 0.069668135 0.027647225
0.631105993
109 110 111 112 113
114
-0.077228626 0.391463022 0.215659871 0.190591571 -2.530665018 -
0.048565767
115 116 117 118 119
120
-0.029706181 -0.160833572 0.541963053 -0.677370304 0.334384730 -
0.171699858
121 122 123 124 125
126
-0.208950076 -0.362461541 -0.555344823 0.425406589 0.213912560 -
0.358420439
127 128 129 130 131
132
-0.264097194 -0.273983980 -0.057628127 1.083946878 -0.369099696
0.010845563
133 134 135 136 137
138
-0.390884109 0.208581421 0.303535516 0.052309489 -0.374649311
0.039588153
139 140 141 142 143
144
0.028690190 -0.211184138 -0.849534942 0.448834080 0.211312109
0.281122134
145 146 147 148 149
150
-0.069846244 0.682287041 0.697693754 -0.125168799 0.191727373
0.469694951

```
##          151          152          153          154          155
156
## -0.093553106  0.156010281  0.153339066  0.153478653  0.291002351 -
0.416542251
##          157          158          159          160          161
162
## -0.212143999 -0.724035261 -0.279045352  0.767987841  0.613422517 -
0.545991495
##          163          164          165          166          167
168
## -0.013080071  0.330069548  0.626572032 -0.418498226 -0.097358697 -
0.101392063
##          169          170          171          172          173
174
## -0.685256020 -0.106545805 -0.044995563 -0.740060933 -0.360597740 -
0.881660844
##          175          176          177
## -0.290602876  1.320196718  0.045835330
```

```
stargazer(residuals, type = "text")
```

```
head(residuals)
```

```
##          1          2          3          4          5          6
## -0.09417293  0.00377727 -0.24762225 -0.24870529 -0.09496753 -0.33567247
```

#xi) Correlation Coefficient between lmktval and profits 0.77689759

```
corremk_pro <- cor(dt.ceosal, y= NULL, use = "everything", method =
"pearson")
```

```
corremk_pro
```

```
##          salary          age          college          grad          comten
## salary  1.000000000  0.11538394 -0.06702522 -0.002999832  0.037698187
## age     0.115383944  1.000000000 -0.17806227 -0.123163323  0.479413536
## college -0.067025223 -0.17806227  1.000000000  0.181445273 -0.157109257
## grad    -0.002999832 -0.12316332  0.18144527  1.000000000 -0.228334613
## comten  0.037698187  0.47941354 -0.15710926 -0.228334613  1.000000000
## ceoten  0.142947678  0.33874170 -0.10628842 -0.102806453  0.315121243
## sales   0.380223875  0.12713402 -0.02149227  0.076326224  0.104399833
## profits 0.393927574  0.11474310 -0.04598209  0.097825529  0.143737237
## mktval  0.406307097  0.10717932 -0.02757797  0.122976062  0.136095997
## lsalary 0.886687117  0.09055855 -0.06504402  0.013136956 -0.002314525
## lsales  0.431751700  0.19390790 -0.07677381  0.083490527  0.237819855
## lmktval 0.441878721  0.13103224 -0.05730098  0.150120021  0.101931416
## comtensq 0.032751945  0.51327067 -0.15901123 -0.242566842  0.965136572
## ceotensq 0.069365886  0.32981691 -0.07776329 -0.120318867  0.324613068
## profmarg -0.028935381  0.01467793 -0.01753083 -0.015395225  0.047173910
##          ceoten          sales          profits          mktval          lsalary
## salary  0.142947678  0.38022387  0.39392757  0.406307097  0.886687117
```

## age	0.338741704	0.12713402	0.11474310	0.107179325	0.090558546
## college	-0.106288424	-0.02149227	-0.04598209	-0.027577967	-0.065044017
## grad	-0.102806453	0.07632622	0.09782553	0.122976062	0.013136956
## comten	0.315121243	0.10439983	0.14373724	0.136095997	-0.002314525
## ceoten	1.000000000	-0.06771469	-0.02160675	0.006609425	0.114728023
## sales	-0.067714685	1.000000000	0.79828723	0.754661598	0.409832114
## profits	-0.021606750	0.79828723	1.000000000	0.918127962	0.396694779
## mktval	0.006609425	0.75466160	0.91812796	1.000000000	0.403093726
## lsalary	0.114728023	0.40983211	0.39669478	0.403093726	1.000000000
## lsales	-0.037685375	0.71770779	0.60633246	0.578540457	0.529960173
## lmktval	-0.043469363	0.67805976	0.77689759	0.809066274	0.481490997
## comtensq	0.325551517	0.09411415	0.13347104	0.128006880	-0.024982111
## ceotensq	0.928526216	-0.05965380	-0.01011938	0.010011458	0.051806764
## profmarg	0.048804692	-0.01735348	0.12547925	0.067018760	-0.059272154
##	lsales	lmktval	comtensq	ceotensq	profmarg
## salary	0.43175170	0.44187872	0.03275195	0.06936589	-0.02893538
## age	0.19390790	0.13103224	0.51327067	0.32981691	0.01467793
## college	-0.07677381	-0.05730098	-0.15901123	-0.07776329	-0.01753083
## grad	0.08349053	0.15012002	-0.24256684	-0.12031887	-0.01539523
## comten	0.23781985	0.10193142	0.96513657	0.32461307	0.04717391
## ceoten	-0.03768537	-0.04346936	0.32555152	0.92852622	0.04880469
## sales	0.71770779	0.67805976	0.09411415	-0.05965380	-0.01735348
## profits	0.60633246	0.77689759	0.13347104	-0.01011938	0.12547925
## mktval	0.57854046	0.80906627	0.12800688	0.01001146	0.06701876
## lsalary	0.52996017	0.48149100	-0.02498211	0.05180676	-0.05927215
## lsales	1.000000000	0.73592316	0.19920093	-0.02741558	-0.01459387
## lmktval	0.73592316	1.000000000	0.08350722	-0.03623371	0.06077802
## comtensq	0.19920093	0.08350722	1.000000000	0.35366851	0.03649651
## ceotensq	-0.02741558	-0.03623371	0.35366851	1.000000000	0.02129125
## profmarg	-0.01459387	0.06077802	0.03649651	0.02129125	1.000000000