

# COST FOR US AIRLINES

Greene's book on Econometric Analysis (7th edition).

The data can be found in

<http://people.stern.nyu.edu/wgreene/Text/Edition7/TableF6-1.txt>

90 observations on 6 firms for 15 years (1970-1984)

Columns are:

I : Airline

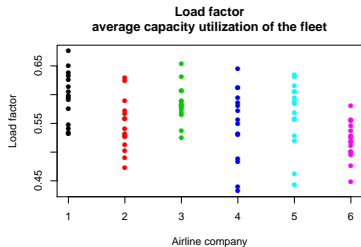
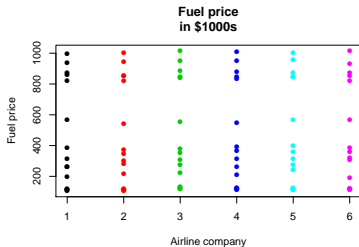
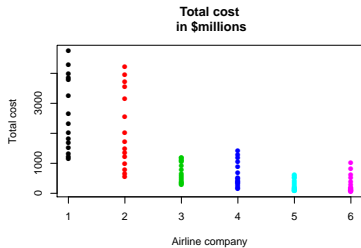
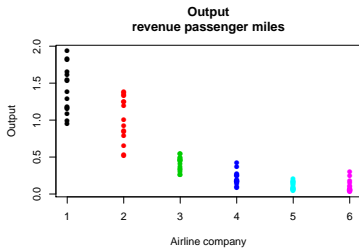
T : Year

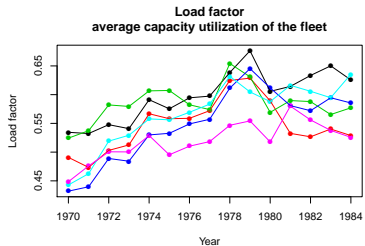
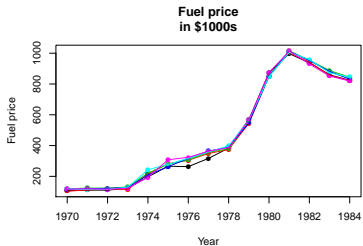
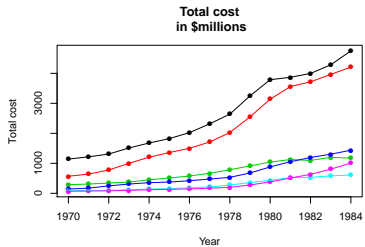
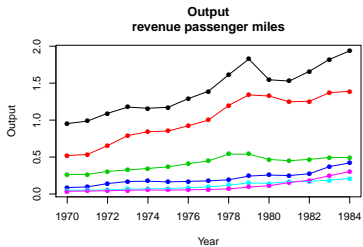
Q : Output, in revenue passenger miles, index number,

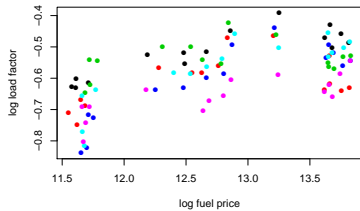
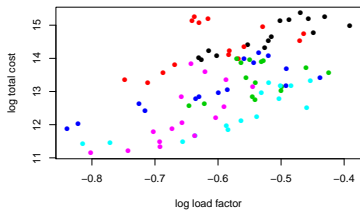
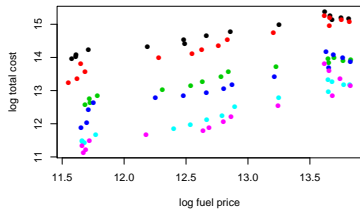
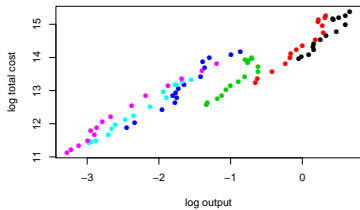
C : Total cost, in \$1000,

FP : Fuel price,

LF : Load factor, the average capacity utilization of the fleet.







Let us start by fitting the following model

$$\log C = \beta_0 + \beta_1 \log Q + \beta_2 \log FP + \beta_3 \log LF + \varepsilon$$

via OLS.

```
data = read.table("http://people.stern.nyu.edu/wgreene/Text/  
Edition7/TableF6-1.txt",header=TRUE)
```

```
attach(data)
```

```
n = nrow(data)
```

```
lc = log(C)
```

```
lq = log(Q)
```

```
lf = log(PF)
```

```
ll = log(LF)
```

```
reg = lm(lc~lq+lf+ll)
coef.ols = reg$coef
se.ols = sqrt(diag(summary(reg)$cov.unscaled))*summary(reg)$sigma
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	8.07565	0.33420	24.164	< 2e-16	***
lq	0.88285	0.01330	66.369	< 2e-16	***
lf	0.45469	0.02046	22.223	< 2e-16	***
ll	-0.89146	0.19065	-4.676	1.08e-05	***

---

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

Residual standard error: 0.1248 on 86 degrees of freedom

Multiple R-squared: 0.9883, Adjusted R-squared: 0.9878

F-statistic: 2411 on 3 and 86 DF, p-value: < 2.2e-16



## HR STANDARD ERRORS

```
k = 3
p = k + 1
x = cbind(lq,lf,ll)
X = cbind(1,x)
se.hr = rep(0,ncol(X))
for (i in 1:p){
  reg1 = lm(X[,i]~X[,-i]-1)
  se.hr[i] = sqrt(sum((reg1$res^2)*(reg$res^2))/(sum(reg1$res^2))^2)
}
round(cbind(coef.ols,se.ols,se.hr),3)
```

	coef.ols	se.ols	se.hr
(Intercept)	8.076	0.334	0.327
lq	0.883	0.013	0.009
lf	0.455	0.020	0.020
ll	-0.891	0.191	0.169



## HR $F$ TEST

```
# Hetereskedasticity-robust F test
esq      = reg$res^2
R2.e     = summary(lm(esq~x))$r.sq
Ftest    = R2.e/(1-R2.e)*(n-p)/k
pvalue   = 1-pf(Ftest,k,n-p)
```

```
> R2.e
```

```
[1] 0.04748007
```

```
> Ftest
```

```
[1] 1.428942
```

```
> pvalue
```

```
[1] 0.2399109
```

# GENERALIZED LS

```
# Generalized least squares
lesq = log(esq)
reg1 = lm(lesq~x)
g     = reg1$fit
hhat = exp(g)
reg.gls = lm(lc~x,weights=1/hhat)
sig = summary(reg.gls)$sigma
se.gls = sqrt(diag(summary(reg.gls)$cov.unscaled))*sig
coef.gls = reg.gls$coef

round(cbind(coef.ols,se.ols,se.hr,coef.gls,se.gls),3)
```

	coef.ols	se.ols	se.hr	coef.gls	se.gls
(Intercept)	8.076	0.334	0.327	8.144	0.341
lq	0.883	0.013	0.009	0.882	0.014
lf	0.455	0.020	0.020	0.452	0.021
ll	-0.891	0.191	0.169	-0.843	0.189

# MAXIMIZING THE LIKELIHOOD

```
# Minus the likelihood function
like = function(beta){
  m = beta[1]+beta[2]*lq+beta[3]*lp+beta[4]*ll
  s = exp((beta[5]+beta[6]*ll)/2)
  return(-sum(dnorm(lc,m,s,log=TRUE)))
}

# initial values
beta = c(lm(lc~lq+lp+ll)$coef,lm(lesq~ll)$coef)

# Nonlinear minimization.
nlm(like,beta)

$minimum
[1] -64.56045

$estimate
[1] 8.2403727 0.8790963 0.4478516 -0.7548482 -1.4087092 4.9078414

$gradient
[1] -4.408520e-06 4.859189e-06 -5.569056e-05 2.446399e-06 1.361860e-08 2.923048e-08

$code
[1] 1

$iterations
[1] 45
```

	coef.ols	coef.gls	coef.mle
(Intercept)	8.076	8.144	8.240
lq	0.883	0.882	0.879
lf	0.455	0.452	0.448
ll	-0.891	-0.843	-0.755

Therefore, by MLE, the estimated model is

$$\begin{aligned}
 E(\log C \mid Q, FP, LF) &= 8.24 + 0.88 \log Q + 0.45 \log FP - 0.76 \log LF \\
 V(\log C \mid LF) &= \exp\{-1.41 + 4.91 \log LF\}
 \end{aligned}$$