

## R Output for Linear Models using functions lm(), gls() & glm()

Different kinds of output related to linear models can be obtained in R using function `lm()` {stats} in the base installation as well as `gls()` & `glm()` in package {nlme} among others. The latter functions are primarily designed to extend linear models in various ways, but default to more-or-less the same analysis with differences in the format of the output. The objective of this worksheet is to show output from each function on the same data set, along with basic definitions. Since linear models may include a combination of numeric covariates and factors, simple ANCOVA data is utilized here from Kuter et al. (KNNL) *Applied Linear Statistical Models* 5th Edition. Specifics of the t & F tests can be found in Worksheet LM 03. Matrix algebra calculations are described in *Biostatistics* Worksheet 380.

### Example in R:

**#COMPARISON OF OUTPUT FROM lm(), gls() & glm():**

`library(nlme)`

`op=options(digits=15)`

`setwd("c:/DATA/Models")`

**#KNNL TABLE 22.1**

`K=read.table("CH22TA01.txt",header=TRUE)`

`K$T=factor(K$treatment)`

`K`

**> K**

	Y	X	treatment	store	T
1	38	21	1	1	1
2	39	26	1	2	1
3	36	22	1	3	1
4	45	28	1	4	1
5	33	19	1	5	1
6	43	34	2	1	2
7	38	26	2	2	2
8	38	29	2	3	2
9	27	18	2	4	2
10	34	25	2	5	2
11	24	23	3	1	3
12	32	29	3	2	3
13	31	30	3	3	3
14	21	16	3	4	3
15	28	29	3	5	3

**Variables:**

**Y = dependent variable**

**X = numeric covariate**

**T = factor**

### Using lm():

**#LINEAR MODEL lm():**

`FM=lm(Y~X+T,data=K) #FULL MODEL`

`summary(FM)`

`anova(FM)`

`drop1(FM)`

**> summary(FM)**

Call:

`lm(formula = Y ~ X + T, data = K)`

Residuals:

	Min	1Q	Median	3Q	Max
	-2.434813925570	-1.273889555822	-0.336254501801	1.671008403361	2.486914765906

**Marginal output for  
covariate & each level  
of the factor**

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	17.353421368547	2.523004120596	6.87808	2.6635e-05 ***
X	0.898559423770	0.102584878233	8.75918	2.7310e-06 ***
T2	-5.075390156062	1.228965128624	-4.12981	0.0016727 **
T3	-12.976830732293	1.205623298658	-10.76359	3.5272e-07 ***

---

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

Residual standard error: 1.87256043775 on 11 degrees of freedom

Multiple R-squared: 0.940329040032, Adjusted R-squared: 0.924055141

F-statistic: 57.7814258162 on 3 and 11 DF, p-value: 5.08171867379e-07

**> anova(FM)**

Analysis of Variance Table

**Serial output for  
covariate & combined  
factor levels in order  
specified by the formula**

Response: Y

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
X	1	190.677777777778	190.677777777778	54.37865	1.4048e-05 ***
T	2	417.1509136988	208.57545684941	59.48282	1.2636e-06 ***
Residuals	11	38.5713085234	3.50648259304		

---

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

### Marginal output for covariate & factor

```
> drop1(FM)
```

```
Single term deletions
```

```
Model:
Y ~ X + T
Df      Sum of Sq      RSS      AIC
<none>                38.5713085234 22.16687744720
X        1 269.0286914766 307.6000000000 51.31125054794
T        2 417.1509136988 455.7222222222 55.20749891507
```

### Using gls():

```
#GAUSSIAN GENERALIZED LEAST SQUARES gls():
```

```
FM2=gls(Y~X+T,data=K) #FULL MODEL
```

```
summary(FM2)
```

```
anova(FM2)
```

```
drop1(FM2)
```

```
> summary(FM2)
```

```
Generalized least squares fit by REML
```

```
Model: Y ~ X + T
```

```
Data: K
```

```
              AIC              BIC              logLik
65.6544520507195 67.6439284147113 -27.8272260253597
```

```
Coefficients:
```

	Value	Std. Error	t-value
(Intercept)	17.35342136854739	2.523004120596327	6.87807888496266
X	0.89855942376951	0.102584878232654	8.75918009798337
T2	-5.07539015606242	1.228965128623691	-4.12980811078530
T3	-12.97683073229291	1.205623298657670	-10.76358655870469

```
Correlation:
```

	(Intr)	X	T2
X	-0.943		
T2	0.026	-0.267	
T3	-0.054	-0.187	0.523

```
Standardized residuals:
```

	Min	Q1	Med
Q3	-1.300259194032781	-0.680292892095387	-0.179569371979373

```
> anova(FM2)
```

```
Denom. DF: 11 Max
1.328082509791125 F-value p-value
(Intercept) 1 4887.11965490090 <.0001
Residual standard error: 3.7864651684375001
Degrees of freedom: 18.48281541838 Residual
```

```
> drop1(FM2)
```

function not defined for gls()

```
Error in terms.default(object) : no terms component nor attribute
```

### Using glm():

```
#GAUSSIAN GENERALIZED LINEAR MODEL glm():
```

```
FM3=glm(Y~X+T,data=K) #FULL MODEL
```

```
summary(FM3)
```

```
anova(FM3)
```

```
drop1(FM3)
```

**> summary(FM3)**

```
Call:
glm(formula = Y ~ X + T, data = K)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-2.434813925570 -1.273889555822 -0.336254501801  1.671008403361
 2.486914765906

Coefficients:
            Estimate      Std. Error  t value Pr(>|t|)
(Intercept) 17.353421368547    2.523004120596   6.87808 2.6635e-05 ***
X             0.898559423770    0.102584878233   8.75918 2.7310e-06 ***
T2          -5.075390156062    1.228965128624  -4.12981 0.0016727 **
T3          -12.976830732293    1.205623298658 -10.76359 3.5272e-07 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 3.50648259303722)

Null deviance: 646.4000000000 on 14 degrees of freedom
Residual deviance: 38.57130852341 on 11 degrees of freedom
AIC: 66.73503344334
```

Number of Fisher Scoring iterations: 2

**> anova(FM3)**

Analysis of Deviance Table

Model: gaussian, link: identity

Response: Y

Terms added sequentially (first to last)

	Df	Deviance	Resid. Df	Resid. Dev
NULL			14	646.4000000000
X	1	190.67777777778	13	455.72222222222
T	2	417.1509136988	11	38.5713085234

**> drop1(FM3)**

Single term deletions

```
Model:
Y ~ X + T
Df      Deviance      AIC
<none> 38.5713085234 66.73503344334
X       1 307.6000000000 95.87940654408
T       2 455.7222222222 99.77565491121
```

**> X #INDEPENDENT VARIABLES IN MODEL FORM****Model Matrix, Hat Matrix & Fitted Values:****#MODEL MATRIX, HAT MATRIX & FITTED VALUES:**

**X=model.matrix(FM)**

**X #INDEPENDENT VARIABLES IN MODEL FORM**

**H=X%\*%solve(t(X)%\*%X)%\*%t(X)**

**H # HAT MATRIX**

**^ NxN matrix**

```
(Intercept) X T2 T3
1           1 21  0  0
2           1 26  0  0
3           1 22  0  0
4           1 28  0  0
5           1 19  0  0
6           1 34  1  0
7           1 26  1  0
8           1 29  1  0
9           1 18  1  0
10          1 25  1  0
11          1 23  0  1
12          1 29  0  1
13          1 30  0  1
14          1 16  0  1
15          1 29  0  1
attr(,"assign")
[1] 0 1 2 2
attr(,"contrasts")
attr(,"contrasts")$T
[1] "contr treatment"
```

**H%\*%K\$Y #FITTED VALUES CALCULATED USING HAT MATRIX****fitted(FM)**

$$H = X \cdot (X^T X)^{-1} \cdot X^T$$

$$\text{Fitted} = H \cdot Y$$

**> H%\*%K\$Y #FITTED VALUES CALCULATED USING HAT MATRIX**

```
[ ,1]
1 36.2231692677070
2 40.7159663865545
3 37.1217286914766
4 42.5130852340937
5 34.4260504201680
6 42.8290516206482
7 35.6405762304921
8 38.3362545018006
9 28.4521008403361
10 34.7420168067227
11 25.0434573829531
12 30.4348139255701
13 31.3333733493397
14 18.7535414165666
15 30.4348139255701
```

**> fitted(FM)**

```
      1      2      3      4
36.2231692677071 40.7159663865546 37.1217286914766 42.5130852340936
      5      6      7      8
34.4260504201681 42.8290516206483 35.6405762304922 38.3362545018007
      9     10     11     12
28.4521008403361 34.7420168067227 25.0434573829532 30.4348139255702
     13     14     15
31.3333733493397 18.7535414165666 30.4348139255702
```

**Regression Parameter Estimates:****#REGRESSION PARAMETER ESTIMATES:****b=solve(t(X)%\*%X)%\*%t(X)%\*%K\$Y #Coefficients****b**

$$b = X^T X \cdot (X^T X)^{-1}$$

**> b**

```
[ ,1]
(Intercept) 17.353421368547416
X            0.898559423769506
T2          -5.075390156062418
T3          -12.976830732292926
```

**Sum of Squares Error & Mean Squares Error:****#SUM OF SQUARES ERROR(SSE) = RESIDUAL SUM OF SQUARES (RSS):****SSE=sum(residuals(FM)^2)****SSE****RSS=deviance(FM) # deviance is the same as RSS=SSE in Linear Models****RSS****deviance() calculates****SSE = RSS for Linear Models****#RESIDUAL DEGREES OF FREEDOM:****n=length(K\$T) # number of objects****n****k=4 # number of parameters/factor levels including intercept****dfR=df.residual(FM)****dfR # = n-k****df.residual() gives  
residual degrees of freedom****> SSE**

```
[1] 38.5713085234092
```

**> RSS**

```
[1] 38.5713085234092
```

**> n**

```
[1] 15
```

**#MEAN SQUARES ERROR:****MSE=SSE/dfR****MSE****> dfR # = n-k**

```
[1] 11
```

**> MSE**

```
[1] 3.5064825930372
```

## Residual Standard Error = Error Standard Deviation:

**#RESIDUAL STANDARD ERROR:**

**RSE=sqrt(MSE)**

**RSE**

**> RSE**

[1] 1.87256043775286

## Standard Error of Regression Estimates:

**#STANDARD ERROR OF REGRESSION ESTIMATES:**

**sb=sqrt(diag(MSE\*solve(t(X)%\*%X)))**

**sb**

**> sb**

**sb = sqrt(diag(MSE\*(X<sup>T</sup>X)<sup>-1</sup>))**

```
(Intercept)          X          T2
T3
2.523004120596323 0.102584878232654 1.228965128623691
1.205623298657670
```

## Saturated Model:

**#SATURATED MODEL:**

**A=factor(1:length(K\$Y))**

**FMS=glm(Y~A,data=K)**

**summary(FMS)**

**anova(FMS)**

**anova(FMS,FM3)**

**> summary(FMS)**

```
Call:
glm(formula = Y ~ A, data = K)
```

```
Deviance Residuals:
 [1] 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	3.800000000000e+01	NA	NA	NA
A2	1.000000000000e+00	NA	NA	NA
A3	-2.000000000000e+00	NA	NA	NA
A4	7.000000000000e+00	NA	NA	NA
A5	-5.000000000000e+00	NA	NA	NA
A6	5.000000000000e+00	NA	NA	NA
A7	5.03060675734e-14	NA	NA	NA
A8	5.25195927430e-14	NA	NA	NA
A9	-1.100000000000e+01	NA	NA	NA
A10	-4.000000000000e+00	NA	NA	NA
A11	-1.400000000000e+01	NA	NA	NA
A12	-6.000000000000e+00	NA	NA	NA
A13	-7.000000000000e+00	NA	NA	NA
A14	-1.700000000000e+01	NA	NA	NA
A15	-1.000000000000e+01	NA	NA	NA

(Dispersion parameter for gaussian family taken to be NaN)

**zero SSE**

```
Null deviance: 6.464000000000e+02 on 14 degrees of freedom
Residual deviance: 9.504196186103e-27 on 0 degrees of freedom
AIC: -864.8235586097
```

**anova() fails in here**

Number of Fisher Scoring iterations: 1

**> anova(FMS)**

```
Error in if (dispersion == 1) Inf else object$df.residual :
missing value where TRUE/FALSE needed
```

**> anova(FMS,FM3)**

Analysis of Deviance Table

```
Model 1: Y ~ A
Model 2: Y ~ X + T
```

	Resid. Df	Resid. Dev	Df	Deviance
1	0	0.000000000000		
2	11	38.57130852341	-11	-38.57130852341

The saturated model contains a parameter for each data point and as a result has zero error variance.

**Null Model:**

```
#NULL MODEL:
NM=glm(Y~1,data=K)
summary(NM)
anova(NM)
anova(NM,FM3)
```

**> summary(NM)**

```
Call:
glm(formula = Y ~ 1, data = K)

Deviance Residuals:
    Min       1Q   Median       3Q      Max
-12.8      -4.3       0.2       4.2      11.2

Coefficients:
            Estimate      Std. Error t value Pr(>|t|)
(Intercept) 33.8000000000000  1.75445012414 19.2653 1.7849e-11 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for gaussian family taken to be 46.171428571428)

Null deviance: 646.4 on 14 degrees of freedom
Residual deviance: 646.4 on 14 degrees of freedom
AIC: 103.0186805877
```

**SSE for the Null model = Total SS**

```
Number of Fisher Scoring iterations: 2
```

**> anova(NM)**

```
Analysis of Deviance Table

Model: gaussian, link: identity

Response: Y

Terms added sequentially (first to last)
```

	Df	Deviance	Resid.	Df	Resid.	Dev
NULL				14		646.4

**> anova(NM,FM3)**

```
Analysis of Deviance Table

Model 1: Y ~ 1
Model 2: Y ~ X + T

  Resid. Df    Resid. Dev Df    Deviance
1       14 646.400000000
2       11 38.5713085234 3 607.8286914766
```









:6)