

Example of “treatment contrasts” used by R in estimating ANOVA coefficients

The first example shows a simple numerical design matrix in R (no factors) for the groups “1”, “a”, “b”, “ab”.

```
> resp <- rnorm(4,0,1)
> data <- data.frame(resp=resp,
+                     A=c(0,1,0,1),
+                     B=c(0,0,1,1))
> data
      resp A B
1  0.3112628 0 0
2 -0.7630832 1 0
3 -1.7276357 0 1
4 -0.5787976 1 1
> lm1.out <- lm(resp~A*B,data=data)
```

Note that the columns of the design matrix have zeros where our effect contrasts had “-1”s. The results (fitted values and tests) are equivalent although this design matrix does not have orthogonal columns and the contrast coefficients do not sum to zero as can be seen from $X^T X$.

```
> X <- model.matrix(lm1.out)
> X
      (Intercept) A B A:B
1             1 0 0  0
2             1 1 0  0
3             1 0 1  0
4             1 1 1  1
> X%*%t(X)

      1 2 3 4
1 1 1 1 1 1
2 1 2 1 2
3 1 1 2 2
4 1 2 2 4
> coef <- solve(t(X)%*%X)%*%t(X)%*%data$resp
> coef

      [,1]
(Intercept)  0.3112628
A           -1.0743461
B           -2.0388985
A:B          2.2231841
> lm1.out$coef
(Intercept)      A      B      A:B
0.3112628 -1.0743461 -2.0388985  2.2231841
```

An equivalent example with factors rather than numeric design

```
> data <- data.frame(resp=resp,
+                     A=c("f", "m", "f", "m"),
+                     B=c("f", "f", "m", "m"))
> data
      resp A B
1  0.3112628 f f
2 -0.7630832 m f
3 -1.7276357 f m
4 -0.5787976 m m
```

R drops the first group. Since f comes before m in the alphabet the contrast for "f" is dropped.

```
> contrasts(data$A)
      m
f 0
m 1

> contrasts(data$B)
      m
f 0
m 1

> lm2.out <- lm(resp~A*B,data=data)
> X <- as.num(model.matrix(lm2.out))
> X
      [,1] [,2] [,3] [,4]
[1,]      1      0      0      0
[2,]      1      1      0      0
[3,]      1      0      1      0
[4,]      1      1      1      1
```

Note that the coefficients calculated using the model matrix and those from the R fit match and match the previous example.

```
> coef <- solve(t(X)%*%X)%*%t(X)%*%data$resp
> coef
      [,1]
[1,]  0.3112628
[2,] -1.0743461
[3,] -2.0388985
[4,]  2.2231841
> lm2.out$coef
(Intercept)          Am          Bm          Am:Bm
  0.3112628  -1.0743461  -2.0388985   2.2231841
> lm1.out$coef
(Intercept)          A          B          A:B
  0.3112628  -1.0743461  -2.0388985   2.2231841
```

Example of a 3x2 factorial. A has levels "low", "med" and "high" and B has levels "low" and "high".

```
> resp <- rnorm(6,0,1)
> data <- data.frame(resp=resp,
+                     A=c("low","med","high",
+                          "low","med","high"),
+                     B=c("low","low","low",
+                          "high","high","high"))
```

I'm specifying the factor levels here so we get them in order.

```
> data$A <- factor(data$A,levels=c("low","med","high"))
> data$B <- factor(data$B,levels=c("low","high"))
> data
      resp    A    B
1 -0.2601737 low low
2  0.1730960 med low
3 -0.7590816 high low
4  0.9682281 low high
5 -0.4736121 med high
6 -0.3177872 high high
>
> contrasts(data$A)
      med high
low     0     0
med     1     0
high    0     1
> contrasts(data$B)
      high
low     0
high    1
>
> lm.out <- lm(resp~A*B,data=data)
>
> X <- model.matrix(lm.out)
> X
(Intercept) Amed Ahigh Bhigh Amed:Bhigh Ahigh:Bhigh
1           1    0    0    0           0           0
2           1    1    0    0           0           0
3           1    0    1    0           0           0
4           1    0    0    1           0           0
5           1    1    0    1           1           0
6           1    0    1    1           0           1
attr("contrasts")
attr("contrasts")$A
[1] "contr.treatment"

attr("contrasts")$B
[1] "contr.treatment"
```

```

> X%*%t(X)

      1 2 3 4 5 6
1 1 1 1 1 1 1
2 1 2 1 1 2 1
3 1 1 2 1 1 2
4 1 1 1 2 2 2
5 1 2 1 2 4 2
6 1 1 2 2 2 4
> coef <- solve(t(X)%*%X)%*%t(X)%*%data$resp
> coef

      [,1]
(Intercept) -0.2601737
Amed         0.4332697
Ahigh        -0.4989079
Bhigh         1.2284018
Amed:Bhigh   -1.8751099
Ahigh:Bhigh  -0.7871074
> lm.out$coef
(Intercept)      Amed      Ahigh      Bhigh  Amed:Bhigh  Ahigh:Bhigh
-0.2601737    0.4332697   -0.4989079    1.2284018   -1.8751099   -0.7871074
> summary(lm.out)

```

Residuals:

ALL 6 residuals are 0: no residual degrees of freedom!

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	-0.2602	NA	NA	NA
Amed	0.4333	NA	NA	NA
Ahigh	-0.4989	NA	NA	NA
Bhigh	1.2284	NA	NA	NA
Amed:Bhigh	-1.8751	NA	NA	NA
Ahigh:Bhigh	-0.7871	NA	NA	NA

If A is a blocking variable and if we assume no iteration we can do testing.

```

> lm.out <- lm(resp~A+B,data=data)
> summary(lm.out)
Coefficients:
      Estimate Std. Error t value Pr(>|t|)
(Intercept)   0.1835     0.5436   0.338   0.768
Amed          -0.5043     0.6658  -0.757   0.528
Ahigh         -0.8925     0.6658  -1.340   0.312
Bhigh          0.3410     0.5436   0.627   0.595

```

```
> anova(lm.out)
```

Analysis of Variance Table

Response: resp

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
A	2	0.80098	0.40049	0.9035	0.5254
B	1	0.17442	0.17442	0.3935	0.5945
Residuals	2	0.88655	0.44328		

R can create many kinds of contrasts. Note that the polynomial and Helmert contrasts are orthogonal.

```
> contr.treatment(2)
  2
1 0
2 1

> contr.treatment(5)
  2 3 4 5
1 0 0 0 0
2 1 0 0 0
3 0 1 0 0
4 0 0 1 0
5 0 0 0 1

> con <- contr.poly(5)
> con
      .L      .Q      .C      ^4
[1,] -0.6324555  0.5345225 -3.162278e-01  0.1195229
[2,] -0.3162278 -0.2672612  6.324555e-01 -0.4780914
[3,]  0.0000000 -0.5345225 -4.095972e-16  0.7171372
[4,]  0.3162278 -0.2672612 -6.324555e-01 -0.4780914
[5,]  0.6324555  0.5345225  3.162278e-01  0.1195229
> round(t(tt)%*%tt,8)
      .L .Q .C ^4
.L  1  0  0  0
.Q  0  1  0  0
.C  0  0  1  0
^4  0  0  0  1
>
> contr.sum(5)
[,1] [,2] [,3] [,4]
1    1    0    0    0
2    0    1    0    0
3    0    0    1    0
4    0    0    0    1
5   -1   -1   -1   -1
>
> con <- contr.helmert(5)
[,1] [,2] [,3] [,4]
1   -1   -1   -1   -1
2    1   -1   -1   -1
3    0    2   -1   -1
4    0    0    3   -1
5    0    0    0    4
```

Here is the same example with polynomial contrasts for B.

```
> contrasts(data$A) <- contr.poly(3)
> contrasts(data$B) <- contr.treatment(2)
> data
      resp      A      B
1 -0.2601737 low low
2  0.1730960 med low
3 -0.7590816 high low
4  0.9682281 low high
5 -0.4736121 med high
6 -0.3177872 high high
> lm.out <- lm(resp~A+B,data=data)
> round(model.matrix(lm.out),4)
      (Intercept)      A.L      A.Q B2
1             1 -0.7071  0.4082  0
2             1  0.0000 -0.8165  0
3             1  0.7071  0.4082  0
4             1 -0.7071  0.4082  1
5             1  0.0000 -0.8165  1
6             1  0.7071  0.4082  1
attr(,"contrasts")
attr(,"contrasts")$A
      .L      .Q
low -7.071068e-01  0.4082483
med -7.850462e-17 -0.8164966
high 7.071068e-01  0.4082483

attr(,"contrasts")$B
      2
low 0
high 1

> summary(lm.out)

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)  -0.2821      0.3844  -0.734   0.539
A.L           -0.6311      0.4708  -1.340   0.312
A.Q            0.0474      0.4708   0.101   0.929
B2             0.3410      0.5436   0.627   0.595

Residual standard error: 0.6658 on 2 degrees of freedom
Multiple R-Squared:  0.5239,    Adjusted R-squared:  -0.1904
F-statistic: 0.7335 on 3 and 2 DF,  p-value: 0.6208

> anova(lm.out)
Analysis of Variance Table

Response: resp
      Df Sum Sq Mean Sq F value Pr(>F)
A       2  0.80098  0.40049   0.9035 0.5254
B       1  0.17442  0.17442   0.3935 0.5945
Residuals  2  0.88655  0.44328
```

Here is an example with more than one observation per “cell”.

```
> resp <- rnorm(12,0,1)
> data <- data.frame(resp=resp,
+                    A=c( "low","low"
+                        , "med","med",
+                        "high","high",
+                        "low","low"
+                        , "med","med",
+                        "high","high"),
+                    B=c(
+                        "low","low","low",
+                        "low","low","low",
+                        "high","high","high",
+                        "high","high","high"))
>
> data$A <- factor(data$A,levels=c("low","med","high"))
> data$B <- factor(data$B,levels=c("low","high"))
> data
      resp    A    B
1  3.33542288 low low
2 -0.02190004 low low
3 -0.34680683 med low
4 -2.11507430 med low
5  0.13964009 high low
6  0.79374788 high low
7  0.29565337 low high
8 -2.28921947 low high
9  1.50376653 med high
10 0.28410049 med high
11 0.57231880 high high
12 -0.76745620 high high
> lm.out <- lm(resp~A*B,data=data)
> round(model.matrix(lm.out),4)
      (Intercept) Amed Ahigh Bhigh Amed:Bhigh Ahigh:Bhigh
1             1     0     0     0             0             0
2             1     0     0     0             0             0
3             1     1     0     0             0             0
4             1     1     0     0             0             0
5             1     0     1     0             0             0
6             1     0     1     0             0             0
7             1     0     0     1             0             0
8             1     0     0     1             0             0
9             1     1     0     1             1             0
10            1     1     0     1             1             0
11            1     0     1     1             0             1
12            1     0     1     1             0             1
attr(,"contrasts")
attr(,"contrasts")$A
[1] "contr.treatment"

attr(,"contrasts")$B
[1] "contr.treatment"
```

```
> summary(lm.out)
```

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	1.657	1.016	1.630	0.1542
Amed	-2.888	1.437	-2.009	0.0913 .
Ahigh	-1.190	1.437	-0.828	0.4394
Bhigh	-2.654	1.437	-1.846	0.1144
Amed:Bhigh	4.778	2.033	2.351	0.0570 .
Ahigh:Bhigh	2.089	2.033	1.028	0.3437

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

Residual standard error: 1.437 on 6 degrees of freedom
 Multiple R-Squared: 0.5001, Adjusted R-squared: 0.08353
 F-statistic: 1.201 on 5 and 6 DF, p-value: 0.4087

```
> anova(lm.out)
```

Analysis of Variance Table

Response: resp

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
A	2	0.5257	0.2629	0.1272	0.8828
B	1	0.3982	0.3982	0.1927	0.6760
A:B	2	11.4766	5.7383	2.7777	0.1400
Residuals	6	12.3952	2.0659		

```
>
```