Categorical independent variables and interactions with R[*]

```
> kars = read.table("http://www.utstat.toronto.edu/~brunner/302f13/code_n_data/lecture/mcars4.data")
> kars[1:4,]
  Cntry lper100k weight length
1    US     19.8   2178   5.92
2 Japan      9.9   1026   4.32
3    US     10.8   1188   4.27
4    US     12.5   1444   5.11
>
> attach(kars) # Variables are now available by name
> n = length(length); n
[1] 100
> # Make indicator dummy variables for Cntry
> # U.S. will be the reference category
> c1 = numeric(n); c1[Cntry=='Europ'] = 1
> table(c1,Cntry)
   Cntry
c1  Europ Japan US
  0     0    13 73
  1    14     0  0
> c2 = numeric(n); c2[Cntry=='Japan'] = 1
> table(c2,Cntry)
   Cntry
c2  Europ Japan US
  0    14     0 73
  1     0    13  0
>
> # Take a look at mean fuel consumption per country
> aggregate(lper100k,by=list(Cntry),FUN=mean)
  Group.1        x
1   Europ 10.17857
2   Japan 10.68462
3      US 12.96438
> # Must specify a LIST of grouping factors
```

On average, the U.S. cars seem to be using more fuel. Back it up with a hypothesis test.

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Origin	c1	c2	$E(Y\|\mathbf{X}=\mathbf{x}) = \beta_0 + \beta_1 C_1 + \beta_2 C_2$
Europe	1	0	$\beta_0 + \beta_1$
Japan	0	1	$\beta_0 + \beta_2$
U.S.	0	0	β_0

```
> # One-factor ANOVA to compare means
> justcountry = lm(lper100k ~ c1+c2)
> summary(justcountry)

Call:
lm(formula = lper100k ~ c1 + c2)

Residuals:
    Min      1Q  Median      3Q     Max 
-5.0644 -2.1644 -0.4644  2.5154  6.8356 

Coefficients:
            Estimate Std. Error t value Pr(>|t|)    
(Intercept)  12.9644     0.3651  35.511  < 2e-16 ***
c1           -2.7858     0.9101  -3.061  0.00285 ** 
c2           -2.2798     0.9390  -2.428  0.01703 *  
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1 

Residual standard error: 3.119 on 97 degrees of freedom
Multiple R-squared: 0.1203,	Adjusted R-squared: 0.1022 
F-statistic: 6.634 on 2 and 97 DF,  p-value: 0.001993 


> 
> # Which means are different?
> Have t-tests. What about Europe vs. Japan?
```

```
> # Repeating ...
> summary(justcountry)$coefficients
             Estimate Std. Error   t value     Pr(>|t|)
(Intercept) 12.964384  0.3650854 35.510547 2.167687e-57
c1          -2.785812  0.9101021 -3.060989 2.853779e-03
c2          -2.279768  0.9390140 -2.427832 1.703327e-02
>
```

$$T = \frac{\mathbf{a}'\widehat{\boldsymbol{\beta}} - \mathbf{a}'\boldsymbol{\beta}}{s\sqrt{\mathbf{a}'(\mathbf{X}'\mathbf{X})^{-1}\mathbf{a}}}$$

```
> # First replicate test of H0: beta1=0
> betahat = justcountry$coefficients; betahat
(Intercept)          c1          c2
  12.964384   -2.785812   -2.279768
> a1 = rbind(0,1,0); a1
     [,1]
[1,]    0
[2,]    1
[3,]    0
> V = vcov(justcountry) # MSE * (X'X)-inverse
> T1 = t(a1) %*% betahat / sqrt(t(a1) %*% V %*% a1)
> T1 = as.numeric(T1)
> T1; 2*(1-pt(abs(T1),97)) # 2-tailed p-value
[1] -3.060989
[1] 0.002853779
>

> # Now test H0: beta1 = beta2
> a = rbind(0,1,-1)
> T = as.numeric( t(a)%*%betahat/sqrt(t(a)%*%V%*%a) )
> pval = 2*(1-pt(abs(T),97))
> T; pval
[1] -0.4211978
[1] 0.6745425
```

Conclusion: American cars are getting fewer kilometers per litre on average than Japanese and European cars.

```
> # Get nicer-looking ANOVA summary table
> is.factor(Cntry)
[1] TRUE
> jc2 = aov(lper100k~Cntry); summary(jc2)  # aov is a wrapper for lm
            Df Sum Sq Mean Sq F value  Pr(>F)
Cntry        2  129.1   64.55   6.634 0.00199 **
Residuals   97  943.8    9.73
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1
>

>
> # The factor Cntry has dummy vars built in.
> # What are they?
> contrasts(Cntry) # Note alphabetical order
      Japan US
Europ     0  0
Japan     1  0
US        0  1
>
> summary(lm(lper100k~Cntry))


Call:
lm(formula = lper100k ~ Cntry)

Residuals:
    Min      1Q  Median      3Q     Max
-5.0644 -2.1644 -0.4644  2.5154  6.8356

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  10.1786     0.8337  12.209  < 2e-16 ***
CntryJapan    0.5060     1.2014   0.421  0.67454
CntryUS       2.7858     0.9101   3.061  0.00285 **
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

Residual standard error: 3.119 on 97 degrees of freedom
Multiple R-squared: 0.1203,   Adjusted R-squared: 0.1022
F-statistic: 6.634 on 2 and 97 DF,  p-value: 0.001993
```

```
> # You can select the dummy variable coding scheme.
> contr.treatment(3,base=2) # Category 2 is the reference category
  1 3
1 1 0
2 0 0
3 0 1

> # U.S. as reference category again
> Country = Cntry
> contrasts(Country) = contr.treatment(3,base=3)
> summary(lm(lper100k~Country))

Call:
lm(formula = lper100k ~ Country)

Residuals:
    Min      1Q  Median      3Q     Max 
-5.0644 -2.1644 -0.4644  2.5154  6.8356 

Coefficients:
            Estimate Std. Error t value Pr(>|t|)    
(Intercept)  12.9644     0.3651  35.511  < 2e-16 ***
Country1     -2.7858     0.9101  -3.061  0.00285 ** 
Country2     -2.2798     0.9390  -2.428  0.01703 *  
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1 

Residual standard error: 3.119 on 97 degrees of freedom
Multiple R-squared: 0.1203,	Adjusted R-squared: 0.1022 
F-statistic: 6.634 on 2 and 97 DF,  p-value: 0.001993
```

Include covariates

Origin	c1	c2	$E(Y\|\mathbf{X}=\mathbf{x}) = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 C_1 + \beta_4 C_2$
Europe	1	0	$(\beta_0 + \beta_3) + \beta_1 X_1 + \beta_2 X_2$
Japan	0	1	$(\beta_0 + \beta_4) + \beta_1 X_1 + \beta_2 X_2$
U.S.	0	0	$\beta_0 \quad + \beta_1 X_1 + \beta_2 X_2$

```
> # Include covariates
> fullmodel = lm(lper100k ~ weight+length+Country)
> summary(fullmodel) # Look carefully at the signs!

Call:
lm(formula = lper100k ~ weight + length + Country)

Residuals:
    Min      1Q  Median      3Q     Max 
-4.5063 -0.8813  0.0147  1.3043  2.9432 

Coefficients:
             Estimate Std. Error t value Pr(>|t|)    
(Intercept) -7.276937   3.006354  -2.421 0.017399 *  
weight       0.005457   0.001472   3.707 0.000352 ***
length       2.345968   0.980329   2.393 0.018676 *  
Country1     1.487722   0.575633   2.584 0.011274 *  
Country2     1.994239   0.584995   3.409 0.000958 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1 

Residual standard error: 1.703 on 95 degrees of freedom
Multiple R-squared: 0.7431,   Adjusted R-squared: 0.7323 
F-statistic: 68.71 on 4 and 95 DF,  p-value: < 2.2e-16 
> # Test car size controlling for country
> anova(justcountry,fullmodel) # Full vs reduced
Analysis of Variance Table

Model 1: lper100k ~ c1 + c2
Model 2: lper100k ~ weight + length + Country
  Res.Df    RSS Df Sum of Sq      F    Pr(>F)    
1     97 943.81                                  
2     95 275.61  2     668.2 115.16 < 2.2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1 
> # I advise using anova ONLY to compare full and reduced models
```

```
>
> # Might as well test country controlling for size too.
> justsize = lm(lper100k ~ weight+length); summary(justsize)

Call:
lm(formula = lper100k ~ weight + length)

Residuals:
    Min      1Q  Median      3Q     Max 
-4.3857 -1.0684 -0.0556  1.3077  4.0429 

Coefficients:
             Estimate Std. Error t value Pr(>|t|)   
(Intercept) -3.617472   2.958472  -1.223  0.22439   
weight       0.004949   0.001546   3.202  0.00185 **
length       1.835625   1.017349   1.804  0.07428 . 
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1 

Residual standard error: 1.804 on 97 degrees of freedom
Multiple R-squared: 0.7058,	Adjusted R-squared: 0.6997 
F-statistic: 116.4 on 2 and 97 DF,  p-value: < 2.2e-16 

>
> anova(justsize,fullmodel)
Analysis of Variance Table

Model 1: lper100k ~ weight + length
Model 2: lper100k ~ weight + length + Country
  Res.Df    RSS Df Sum of Sq      F   Pr(>F)   
1     97 315.64                                
2     95 275.61  2    40.035 6.8999 0.001592 **
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1 
>
```

```
> # Drop length and try including interactions
> eqslope    = lm(lper100k ~ weight+c1+c2)
> summary(eqslope)

Call:
lm(formula = lper100k ~ weight + c1 + c2)

Residuals:
    Min      1Q  Median      3Q     Max 
-5.0550 -0.4890  0.0138  1.2755  2.8316 

Coefficients:
              Estimate Std. Error t value Pr(>|t|)    
(Intercept) -0.4241768  0.9376017  -0.452  0.65200    
weight       0.0086939  0.0005942  14.631  < 2e-16 ***
c1           1.2127472  0.5777671   2.099  0.03844 *  
c2           1.8932896  0.5976631   3.168  0.00206 ** 
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1 

Residual standard error: 1.745 on 96 degrees of freedom
Multiple R-squared: 0.7276,	Adjusted R-squared: 0.7191 
F-statistic: 85.49 on 3 and 96 DF,  p-value: < 2.2e-16
```

Origin	C1	C2	$E(Y\|\mathbf{X}=\mathbf{x}) = \beta_0 + \beta_1 X_1 + \beta_3 C_1 + \beta_4 C_2 + \beta_5 X_1 C_1 + \beta_6 X_1 C_2$
Europe	1	0	$(\beta_0 + \beta_3) + (\beta_1+\beta_5)X_1$
Japan	0	1	$(\beta_0 + \beta_4) + (\beta_1+\beta_6)X_1$
U.S.	0	0	$\beta_0 + \beta_1 X_1$

```
> wc1 = weight*c1; wc2 = weight*c2
> uneqslope = lm(lper100k ~ weight+c1+c2+wc1+wc2)
> summary(uneqslope)
```

```
Call:
lm(formula = lper100k ~ weight + c1 + c2 + wc1 + wc2)

Residuals:
    Min      1Q  Median      3Q     Max
-4.8461 -0.5647 -0.1310  1.3273  2.6569

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)  0.4005480  0.9545858   0.420   0.6757
weight       0.0081583  0.0006065  13.452   <2e-16 ***
c1          -3.8072812  2.3485193  -1.621   0.1083
c2          -8.7126778  5.0437692  -1.727   0.0874 .
wc1          0.0044198  0.0020348   2.172   0.0324 *
wc2          0.0097631  0.0046908   2.081   0.0401 *
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1

Residual standard error: 1.687 on 94 degrees of freedom
Multiple R-squared: 0.7507,   Adjusted R-squared: 0.7375
F-statistic: 56.63 on 5 and 94 DF,  p-value: < 2.2e-16

> anova(eqslope,uneqslope)
Analysis of Variance Table

Model 1: lper100k ~ weight + c1 + c2
Model 2: lper100k ~ weight + c1 + c2 + wc1 + wc2
  Res.Df    RSS Df Sum of Sq      F Pr(>F)
1     96 292.22
2     94 267.43  2    24.793 4.3573 0.0155 *
---
Signif. codes:  0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1
```

The heavier the car, the greater the average fuel consumption. Rates of increase are greater for Japanese and European cars than for American cars.

```
> # Plot the regression lines
> yhat = uneqslope$fitted.values
> plot(weight,yhat,pch=' ',xlab='Weight in kg',
+ ylab='Fuel Consumption in lper100k')
> title('Weight and Estimated Fuel Consumption')
> lines(weight[Cntry=='US'],yhat[Cntry=='US'],lty=1)
> lines(weight[Cntry=='Europ'],yhat[Cntry=='Europ'],lty=2)
> lines(weight[Cntry=='Japan'],yhat[Cntry=='Japan'],lty=3)
> x1 = c(1800,2000); y1 = c(8,8); lines(x1,y1,lty=1); text(1700,8,'US   ')
> x2 = c(1800,2000); y2 = c(9,9); lines(x2,y2,lty=2); text(1700,9,'Japan')
> x3 = c(1800,2000); y3 = c(10,10); lines(x3,y3,lty=3); text(1700,10,'Europ')
```