## Handout 10.1. Analysis of variance

## Two-way ANOVA

Consider the model which decomposes observations into a general level, a row effect, a column effect and a noise term.

$$x_{ij} = \mu + \alpha_i + \beta_j + \epsilon_{ij}, \epsilon_{ij} \sim N(0, \sigma^2)$$

in which  $\sum \alpha_i = 0$ , and  $\sum \beta_j = 0$ . Let  $x_{ij}$  denote the *j*th observation in the *i*th row,  $\bar{x}_i$  the mean of the ith row,  $\bar{x}_j$  the mean of the jth column, and  $\bar{x}_i$  the overall mean.

```
data(heart.rate)
attach(heart.rate)
heart.rate
```

```
> heart.rate <- data.frame(hr = c(96,110,89,95,128,100,72,79,100, 92,106,86,78,124,98,68,75,106,86,108,85,78,118,100,67,74, 104,92,114,83,83,118,94,71,74,102), subj=gl(9,1,36), time=gl(4,9,36,labels=c(0,30,60,120)))</pre>
```

The gl (generate levels) function is specially designed for generating patterned factors for balanced experimental designs. It has three arguments: the number of levels, the block length (how many times each level should repeat), and the total length of the result. The two patterns in the data frame are thus

```
> ql(9,1,36)
[1] 1 2 3 4 5 6 7 8 9 1 2 3 4 5 6 7 8 9 1 2 3 4 5 6 7 8 9 1 2 3
[31] 4 5 6 7 8 9
Levels: 1 2 3 4 5 6 7 8 9
> ql(4,9,36,labels=c(0,30,60,120))
[1] 0 0 0 0 0 0 0 0 30 30 30 30 30
[16] 30 30 30 60 60 60 60 60 60 60 60 60 120 120 120
[31] 120 120 120 120 120 120
Levels: 0 30 60 120
> attach(heart.rate)
> anova(lm(hr~subj+time))
Analysis of Variance Table
Response: hr
         Df Sum Sq Mean Sq F value Pr(>F)
          8 8966.6 1120.82 90.6391 4.863e-16
subj
                            4.0696 0.01802 *
          3 151.0
                     50.32
Residuals 24 296.8 12.37
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

## 2.1 graphics for repeated measurements

interaction.plot(time,subj,hr)

## 1. ANOVA table in the regression analysis

The variation between and within groups for a one-way analysis of variance generalizes to model variation

SSmodel= 
$$\sum_{i} (\hat{y}_i - \hat{y}_i)^2$$

and residual variation

$$SSres = \sum_{i} (y_i - \hat{y}_i)^2$$

which partition the total variation

SStotal= 
$$\sum_{i} (y_i - \hat{y})^2$$

This applies only when the model contains an intercept. The role of the group means in the one-way classification is taken over by the fitted values in the more general linear model.

The analysis of variance table corresponding to a regression analysis can be extracted with the function anova, just as for one- and two- way analyses of variance.

```
data(thuesen)
attach (thuesen)
lm.velo<-lm(short.velocity~blood.glucose)</pre>
summary(lm.velo)
Call:
lm(formula = short.velocity ~ blood.glucose)
Residuals:
    Min
              10 Median
                                30
                                        Max
-0.40141 -0.14760 -0.02202 0.03001
                                    0.43490
Coefficients:
             Estimate Std. Error t value Pr(>|t|)
(Intercept) 1.09781 0.11748 9.345 6.26e-09 ***
blood.glucose 0.02196 0.01045 2.101
                                           0.0479 *
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
Residual standard error: 0.2167 on 21 degrees of freedom
  (1 observation deleted due to missingness)
Multiple R-squared: 0.1737, Adjusted R-squared: 0.1343
F-statistic: 4.414 on 1 and 21 DF, p-value: 0.0479
```

Notice that the F test gives the same p-value as the t test for a zero slope. It is the same F test that gets printed at the end of the summary output Residual standard error is the square root of residual mean squares, namely 0.2167=sqrt(0.04696). R^2 is the proportion of the total sum of squares explained by the regression line, that is 0.1737=0.2073/(0.2073+0.9861)