**Linear Regression Formulas** Linear Regression Equation  $y = \alpha + \beta x + \epsilon$ Matrix form:  $\mathbf{Y} = \boldsymbol{\beta} \mathbf{X} + \boldsymbol{\epsilon}$ 

with  $E(Y) = \beta X$ 

# Acronyms and names

· TSS: Total Sum of Squares

· RSS: Residual Sum of Squares RSE: Residual Standard Error

MSS: Mean Sum of Squares

S<sub>x</sub>x: Corrected sum of squares of x

S<sub>v</sub> y: Corrected sum of squares of y

S<sub>x</sub> y: Corrected sum of products of xy

n: number of observations

• p: number of parameters (does not include intercept) Least square estimates (matrix form) Sum of squares function

$$S(\boldsymbol{\beta}) = \sum_{i=1}^{n} y_i + X_i \boldsymbol{\beta}$$

Estimates, matrix from

$$\begin{split} \widehat{\boldsymbol{\beta}} &= (\boldsymbol{X^TX})^{-1}\boldsymbol{X^TY} \\ RSS &= S(\widehat{\boldsymbol{\beta}}) = \boldsymbol{Y^TY} - \boldsymbol{Y^TX}\widehat{\boldsymbol{\beta}} \\ \sigma^2 &= \frac{RSS}{n-p}RSE \end{split}$$

$$= \sigma = \sqrt{\frac{RSS}{n-p}}$$

where  $\sigma^2$  is the variance, n-p is the degrees of freedom.

Estimates, non-matrix form

$$S_{xx} = \sum_{i=1}^{n} (x_i - \overline{x})^2$$

$$S_{yy} = \sum_{i=1}^{n} (y_i - \overline{y})^2$$

$$S_{xy} = \sum_{i=1}^{n} (x_i - \overline{x})(y_i - \overline{y})$$

$$\widehat{\beta} = \frac{S_{xy}}{S_{xx}}$$

$$\widehat{\alpha} = \overline{y} - \frac{S_{xy}}{S_{xx}} \overline{x}$$

Correlation

$$\rho(X,Y) = \frac{Cov(X,Y)}{\sqrt{Var(X)Var(Y)}}$$
 
$$r = \frac{S_{xy}}{\sqrt{S_{xx}S_{yy}}}$$

RSS,TSS, MSS

$$\begin{split} \widehat{\beta}_0 &= \overline{\mathbf{y}} \\ then \\ S(\widehat{\beta}_0) &= TSS \\ &= S_{yy} \\ &= \sum_{i=1}^n (y_i - \overline{\mathbf{y}})^2 \end{split}$$

$$RSS = \sum_{i=1}^{n} (y_i - \widehat{y})^2$$

TSS = MSS + RSS

$$\mathbb{R}^2$$
 (standard and adjusted)

$$R^2 = 1 - \frac{RSS}{TSS}$$

In the case of a  $\emph{simple}$  linear regression (one explanatory variable)  $R^2 = r^2$ 

$$R^{2}(adj) = 1 - \frac{\frac{RSS}{n-p-1}}{\frac{TSS}{n-1}}$$

$$R^{2}(adj) = 1 - (1 - R^{2}) \frac{n-1}{n-p-1}$$

#### Assumptions of a linear model

- A: the deterministic part of the model captures all the non-random structure in the data.
   B: the scale of the variability of the errors is constant at all values of the explanatory variables.
- C: errors are independent.
- . D: errors are normally distributed.
- E: the values of the explanatory variables are recorded without error.

#### Residuals

$$\widehat{\epsilon_i} = y_i - \widehat{(y_i)}$$

standardised residuals

$$r_i = \frac{\widehat{\epsilon}_i}{\sqrt{Var(\widehat{\epsilon}_i)}}$$

Inference for Regression Coefficients estimated standard error (e.s.e./s.e.)

$$e.s.e.(\widehat{\boldsymbol{\beta}}) = e.s.e(\widehat{\boldsymbol{\beta}}) = \sqrt{\frac{RSS}{n-p}} \boldsymbol{b}^T (\boldsymbol{X}^T \boldsymbol{X})^{-1} \boldsymbol{b}$$

pivotal function

$$\frac{\boldsymbol{b}^T \widehat{\boldsymbol{\beta}} - \boldsymbol{b}^T \boldsymbol{\beta}}{e.s.e(\widehat{\boldsymbol{\beta}})} \sim t(n-p; \frac{1+c}{2})$$

with c: confidence level ( $c = 0.95 \implies (c+1)/2 = 0.975$ ) confidence interval (for slope parametres)

$$\begin{split} \boldsymbol{b}^T \widehat{\boldsymbol{\beta}} \pm t(n-p; \frac{1+c}{2}) e.s.e.(\widehat{\boldsymbol{\beta}}) \\ \boldsymbol{b}^T \widehat{\boldsymbol{\beta}} \pm t(n-p; \frac{1+c}{2}) \sqrt{\frac{RSS}{n-p}} (\boldsymbol{b}^T (\boldsymbol{X}^T \boldsymbol{X})^{-1} \boldsymbol{b}) \end{split}$$

prediction interval (for predicted variable)

$$\boldsymbol{b}^T \widehat{\boldsymbol{\beta}} \pm t(n-p; \frac{1+c}{2}) \sqrt{\frac{RSS}{n-p} (\frac{1}{1} + \boldsymbol{b}^T (\boldsymbol{X}^T \boldsymbol{X})^{-1} \boldsymbol{b})}$$

$$\boldsymbol{Y} = \begin{pmatrix} y_{11} \\ \vdots \\ y_{1n_1} \\ \vdots \\ y_{21} \\ y_{2n_2} \end{pmatrix}, \boldsymbol{X} = \begin{pmatrix} 1 & (x_{11} - \overline{x_1}) & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 1 & (x_{1n_1} - \overline{x_1}) & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 1 & (x_{1n_1} - \overline{x_1}) & 0 & 0 \\ 0 & 0 - \overline{x_1}) & 1 & (x_{21} - \overline{x_2}) \\ 0 & 0 - \overline{x_1}) & 1 & (x_{2n_2} - \overline{x_2}) \end{pmatrix}, \boldsymbol{\beta} = \begin{pmatrix} \alpha_1 \\ \beta_1 \\ \alpha_2 \\ \beta_2 \end{pmatrix}$$

Least squares estimate

$$\widehat{\boldsymbol{\beta}} = \begin{pmatrix} \frac{S_{x_1 y_1}^{\overline{y}_1}}{S_{x_1 x_1}} \\ \frac{S_{x_2 y_2}}{S_{x_2 y_2}} \\ \frac{S_{x_2 y_2}}{S_{x_2 x_2}} \end{pmatrix}$$

$$RSS = RSS1 + RSS2$$

95% confidence interval for  $(\beta_1 - \beta_2)$ 

$$\widehat{\beta}_{1} - \widehat{\beta}_{2} \pm t(n_{1} + n_{2} - 4; 0.975) \sqrt{\frac{RSS_{1} + RSS_{2}}{n_{1} + n_{2} - 4}(\frac{1}{S_{x_{1}x_{2}}} + \frac{1}{S_{x_{1}x_{2}}})}$$

### Parallel lines

$$\boldsymbol{Y} = \begin{pmatrix} y_{11} \\ \vdots \\ y_{1n_{1}} \\ \vdots \\ y_{21} \\ \vdots \\ y_{2n_{2}} \end{pmatrix}, \boldsymbol{X} = \begin{pmatrix} 1 & 0 & (x_{11} - \overline{x}_{1}) \\ \vdots & \vdots & \vdots \\ 1 & 0 & (x_{1n_{1}} - \overline{x}_{1}) \\ 0 & 1 & (x_{21} - \overline{x}_{2}) \\ \vdots & \vdots & \vdots \\ 0 & 1 & (x_{2n_{2}} - \overline{x}_{2}) \end{pmatrix}, \boldsymbol{\beta} = \begin{pmatrix} \alpha_{1} \\ \alpha_{2} \\ \beta \end{pmatrix}$$

$$\widehat{\boldsymbol{\beta}} = \begin{pmatrix} \frac{\overline{y}_1}{y_2} \\ \frac{S_{x_1 y_1} + S_{x_2 y_2}}{S_{x_1 x_1} + S_{x_2 x_2}} \end{pmatrix}$$

95% confidence interval for  $\alpha_1 - \alpha_2 + \beta(\overline{x}_2 - \overline{x}_1)$ 

$$\begin{split} \pmb{b} = & \begin{pmatrix} 1 \\ -1 \\ \overline{x_2} - \overline{x_1} \end{pmatrix} \\ p = 3 \\ \pmb{b}^T \hat{\pmb{\beta}} \pm t(n_1 + n_2 - p; 0.975) \sqrt{\frac{RSS}{n_1 + n_2 - p}} \pmb{b}^T (\pmb{X}^T \pmb{X})^{-1} \pmb{b} \\ \pmb{b}^T \hat{\pmb{\beta}} \pm t(n_1 + n_2 - p; 0.975) \sqrt{\frac{RSS}{n_1 + n_2 - p}} (\frac{1}{n_1} + \frac{1}{n_2} + \frac{(\overline{x_2} - \overline{x_1})^2}{S_{x_1 x_1} + S_{x_2 x_2}}) \end{split}$$

$$(\pmb{X}^T \pmb{X})^{-1} = \begin{pmatrix} \frac{1}{n_1} & 0 & 0 \\ 0 & \frac{1}{n_2} & 0 \\ 0 & 0 & \frac{1}{S_{x_1 x_1} + S_{x_2 x_2}} \end{pmatrix}$$

## ANOVA table

Component	df	SS	MS	F value
Model	p-1	MSS	$\frac{MSS}{p-1}$	$\frac{\frac{MSS}{p-1}}{\frac{RSS}{n-p}}$
Residual Total	n-p $n-1$	RSS TSS	$\frac{RSS}{n-p}$	. p

- df: degrees of freedomSS: Sum of SquaresMS: Mean Squares

F statistic ( $H_0$ : all parameters = 0) (large F values  $\implies$  rejection of  $H_0$ )