### 1 Regresja liniowa

$$Y_i = \beta_0 + \beta_1 X_i + \xi_i, \, \beta_0 \, \beta_1 \, \xi_i \sim N \left( 0, \sigma^2 \right) \, E(Y_i | X_i) = \beta_0 + \beta_1 X_i, \, Var(Y_i | X_i) = \sigma^2$$

# 2 Estymacja

$$\hat{Y} = b_0 + b_1 X_i, e_i = Y_i - \hat{Y} = Y_i - (b_0 + b_1 X_i) \ b_1 = \frac{\sum (X_i - \bar{X})(Y_i - \bar{Y})}{\sum (X_i - \bar{X})^2}, b_0 = \bar{Y} - b_1 \bar{X}$$

$$s^2 = \frac{\sum (Y_i - \hat{Y}_i)^2}{n - 2} = \frac{\sum e_i^2}{n - 2} = \frac{SSE}{dfE} = MSE$$

### 3 Przedziały ufności

#### 3.1 Dla $\beta_1$

$$b_{1} \sim N(\beta_{1}, \sigma^{2}(b_{1})), \ \sigma^{2}(b_{1}) = \frac{\sigma^{2}}{\sum (X_{i} - \bar{X})^{2}}, \ t = \frac{(b_{1} - \beta_{1})}{s(b_{1})} \sim t(n - 2), \ s^{2}(b_{1}) = \frac{s^{2}}{\sum (X_{i} - \bar{X})^{2}},$$

$$b_{1} \pm t_{c}s(b_{1}), \ t_{c} = (1 - \alpha/2, n - 2)$$

$$P(|z| \geq |t|), \ z \sim t(n - 2)$$

### 3.2 Dla $\beta_0$

$$b_{1} \sim N(\beta_{0}, \sigma^{2}(b_{0})),$$

$$\sigma^{2}(b_{0}) = \sigma^{2} \left[ \frac{1}{n} + \frac{\bar{X}^{2}}{\sum (X_{i} - \bar{X})^{2}} \right],$$

$$t = \frac{b_{0} + \beta_{0}}{s(b_{0})} \sim t(n - 2)$$

$$s^{2}(b_{0}) = s^{2} \left[ \frac{1}{n} + \frac{\bar{X}^{2}}{\sum (X_{i} - \bar{X})^{2}} \right],$$

$$b_{0} \pm t_{c}s(b_{0}), t_{c} = (1 - \alpha/2, n - 2),$$

$$t = \frac{b_{0} + \beta_{00}}{s(b_{0})}, P(|z| \geqslant |t|),$$

$$z \sim t(n - 2)$$

#### 4 Moc

$$t = \frac{b_1}{s(b_1)} \sim t(n-2, \delta), \ \delta = \frac{\beta_1}{\sigma(b_1)}, \\ b_1 \sim N(\beta_1, \sigma_2(b_1)), \\ \sigma^2(b_1) = \frac{\sigma^2}{\sum (X_i - \bar{X})^2},$$

# 5 Estymacja średniej

$$\begin{split} E(Y_h) &= \mu_h = \beta_0 + \beta_1 X_h, \\ \hat{\mu}_h &= b_0 + b_1 X_h, \\ \hat{\mu}_h &\sim N(\mu_h, \sigma^2), \\ \sigma^2(\hat{\mu}_h) &= \sigma^2 \left[ \frac{1}{n} + \frac{(X_h - \bar{X})^2}{(X_i - \bar{X})^2} \right], \\ s^2(\hat{\mu}_h) &= s^2 \left[ \frac{1}{n} + \frac{(X_h - \bar{X})^2}{\sum (X_i - \bar{X})^2} \right], \\ \frac{\hat{\mu}_h - E(Y_h)}{s(\hat{\mu}_h)} &\sim t(n-2) \\ \hat{\mu}_h &\pm t_c s(\hat{\mu}_h) \end{split}$$

## 6 Prognoza

$$Y_{h} = \beta_{0} + \beta_{1}X_{h} + \xi_{h},$$

$$s^{2}(pred) = s^{2} \left[ 1 + \frac{1}{n} + \frac{(X_{h} - \bar{X})^{2}}{\sum (X_{i} - \bar{X})^{2}} \right],$$

$$\frac{Y_{h} - \hat{\mu}_{h}}{s(pred)} \sim t(n - 2),$$

$$\hat{\mu}_{h} \pm t_{c}s(pred)$$

#### 7 ANOVA

$$\begin{split} SST &= \sum (Y_i - \bar{Y})^2, \, dfT = n-1, \, MST = \frac{SST}{dfT} \\ SSM &= \sum (\hat{Y}_i - \bar{Y})^2, \, dfM = 1, \, MSM = \frac{SSM}{dfM} \\ SSE &= \sum (Y_i - \hat{Y}_i)^2, \, dfE = n-2, \, MSE = \frac{SSE}{dfE} \\ F &= \frac{MSM}{MSE} \sim F(dfM, dfE) = F(1, n-2) \end{split}$$

# 8 Porównywanie modeli

$$F = \frac{\frac{SSE(R) - SSE(F)}{dfE(R) - dfE(F)}}{MSE(F)}$$

$$SSE(R) = \sum_{i} (Y_i - b_0)^2 = \sum_{i} (Y_i - \bar{Y})^2 = SST \ SSE(F) = SSE \ df E(R) = n - 1 \ df E(F) = n - 2$$

$$r^2 = SSM/SST = 1 - SSE/SST$$