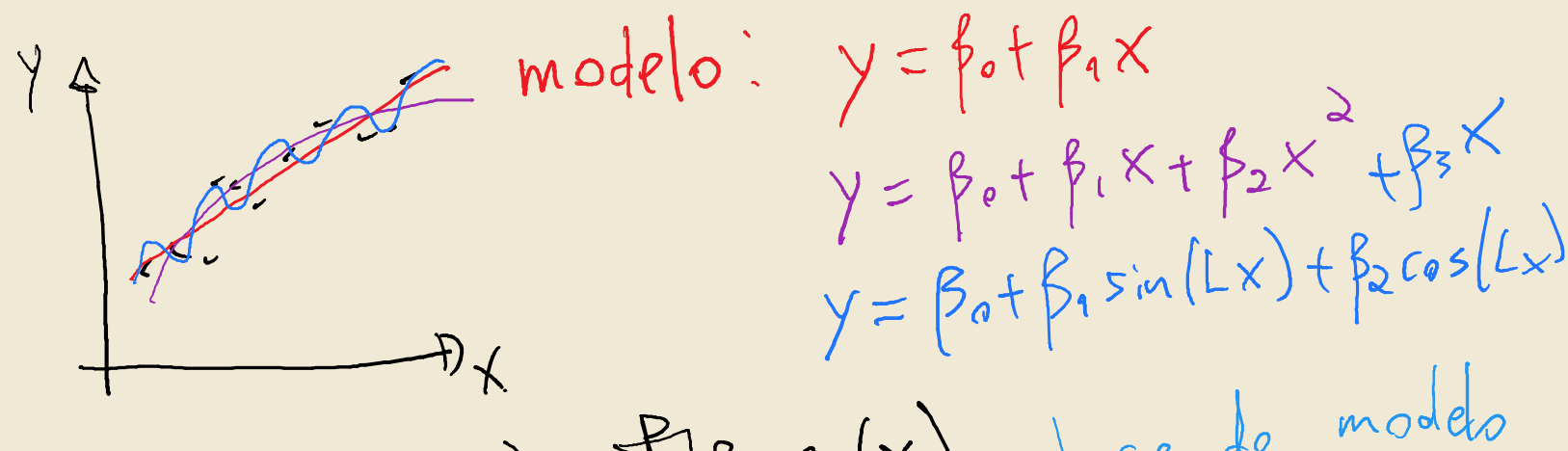


05/06 - Quadrados Mínimos



Modelo: $h(x; \beta) = \sum_{j=1}^p \beta_j \psi_j(x)$ base do modelo

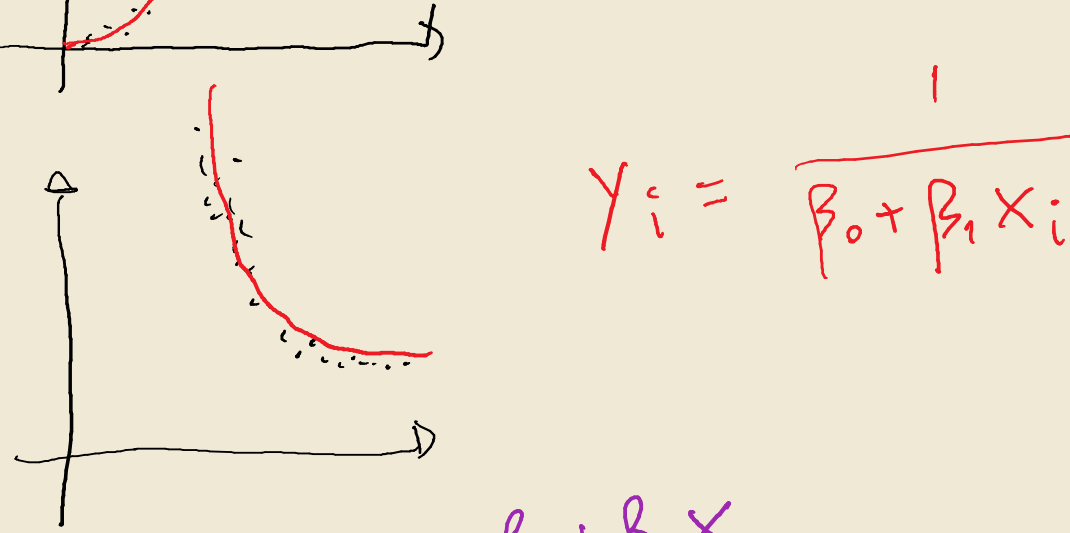
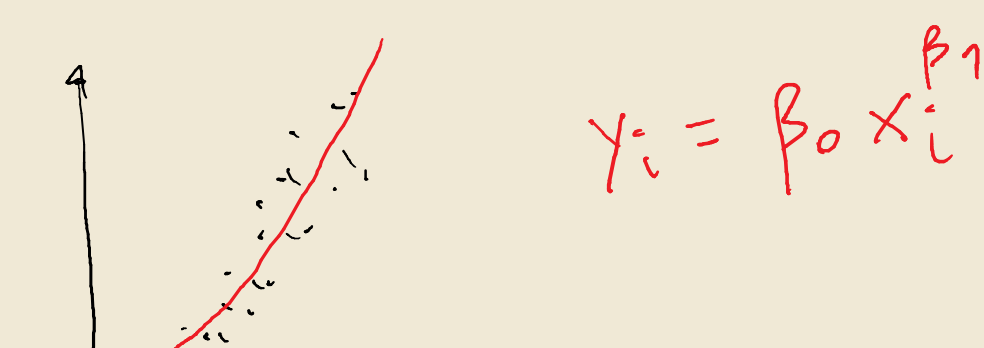
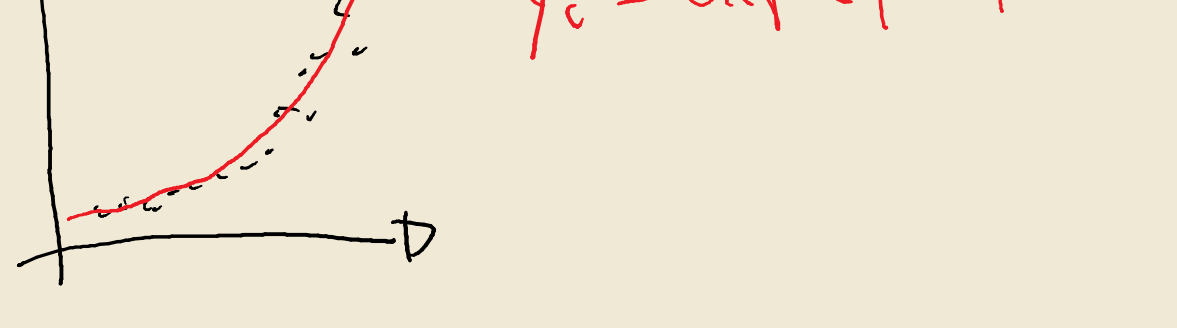
$$X = \begin{bmatrix} \psi_1(x_1) & \psi_2(x_1) & \dots & \psi_p(x_1) \\ \vdots & \vdots & & \vdots \\ \psi_1(x_n) & \psi_2(x_n) & \dots & \psi_p(x_n) \end{bmatrix} \in \mathbb{R}^{n \times p}$$

→ Jacobiana de $F(\beta) = y - X\beta$

Problema: $\min_{\beta} \frac{1}{2} \|y - X\beta\|^2$ resíduo $y - X\beta$

equiv. $X^T X \beta = X^T y$ equações normais

Não-linear linearizável



Ex.: $h(x; \beta) = e^{\beta_0 + \beta_1 x}$

$\ln h(x; \beta) = \beta_0 + \beta_1 x$

$\psi(x; \beta)$

$y_i \approx h(x_i; \beta) \Rightarrow \underbrace{\ln y_i}_{z_i} \approx \underbrace{\ln h(x_i; \beta)}_{\psi(x_i; \beta)} = \beta_0 + \beta_1 x_i$

Ajuste $\{(x_i, z_i), i=1, \dots, n\}$

Ex.:

x	1	2	3	4
y	2.1	4.4	7.3	15.9

Ajuste do modelo $y = e^{\beta_0 + \beta_1 x}$ ← n sei

Veja que $\ln y = \beta_0 + \beta_1 x$ daí sei

x	1	2	3	4
$\ln y$	0.7419	1.4816	1.9879	2.7663

Resolvendo criando X e y'

$X = \begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 3 \\ 1 & 4 \end{bmatrix}$

$\beta_0 + \beta_1 x$

$y' = \begin{bmatrix} 0.7419 \\ 1.4816 \\ 1.9879 \\ 2.7663 \end{bmatrix}$

$\ln y$

$X^T X = \begin{bmatrix} 4 & 10 \\ 10 & 30 \end{bmatrix} \quad X^T y' = \begin{bmatrix} 6.9777 \\ 20.734 \end{bmatrix}$

$(X^T X) \beta = X^T y \Rightarrow \beta = (X^T X)^{-1} X^T y$

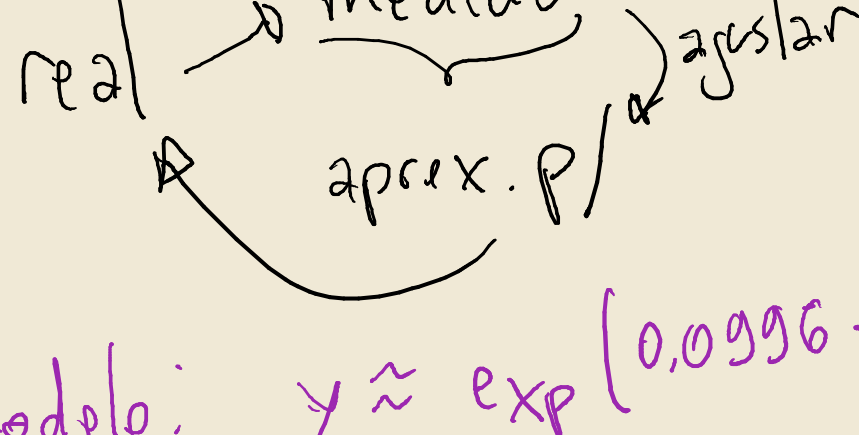
$\hat{\beta} = \frac{1}{20} \begin{bmatrix} 30 & -10 \\ -10 & 4 \end{bmatrix} \begin{bmatrix} 6.9777 \\ 20.734 \end{bmatrix} = \begin{bmatrix} 0.0996 \\ 0.6580 \end{bmatrix}$

$\hat{y} = \exp(X \hat{\beta})$ (y predito/estimado)

$\hat{y} = \exp \left(\begin{bmatrix} 1 & 1 \\ 1 & 2 \\ 1 & 3 \\ 1 & 4 \end{bmatrix} \begin{bmatrix} 0.0996 \\ 0.6580 \end{bmatrix} \right) = \begin{bmatrix} 2.1332 \\ 4.1190 \\ 7.9554 \\ 15.3573 \end{bmatrix}$

$\begin{matrix} 2.1 \\ 4.4 \\ 7.3 \\ 15.9 \end{matrix}$

resíduo: $y - \hat{y} = y - X \hat{\beta}$ (erro)



O modelo: $y \approx \exp(0.0996 + 0.6580x)$

Avaliando o modelo

SQR = Soma dos Quadrados dos Resíduos

$= \sum_{i=1}^n (y_i - \underbrace{h(x_i; \beta)}_{\hat{y}_i})^2$

Métrica R^2

$\bar{y} = \frac{1}{n} \sum_{i=1}^n y_i$

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

Soma Total

$R^2 = 1 - \frac{SQR}{SAT} = 1 - \frac{\sum_{i=1}^n (y_i - \hat{y}_i)^2}{\sum_{i=1}^n (y_i - \bar{y})^2}$

$R^2 \in [0, 1]$

Ex.: $y_i \approx \beta_0 x_i^{\beta_1}$

$\ln y_i \approx \underbrace{\ln \beta_0}_{\beta_0} + \underbrace{\beta_1}_{\beta_1} \ln x_i \rightarrow \text{linear em } x$

Ex.: $y_i \approx \frac{1}{\beta_0 + \beta_1 x_i}$

$\frac{1}{y_i} \approx \beta_0 + \beta_1 x_i$