

$$p = n = \text{data}$$

$$\text{delta}_y_i = y[i] - \hat{y}[i]$$

for (j in 1:(p+1)) {

$$W[j] = W[j] + \text{delta}_y_i [j]$$

02/19

AKA

$p = 1$

$y = \mathbb{R}$, p covariants

$$H = \{ \vec{w} \cdot \vec{x} : \vec{w} \in \mathbb{R}^{p+1} \} = \{ w_0 + w_1 x : w_0, w_1 \in \mathbb{R} \}$$

$A = \text{OLS}$

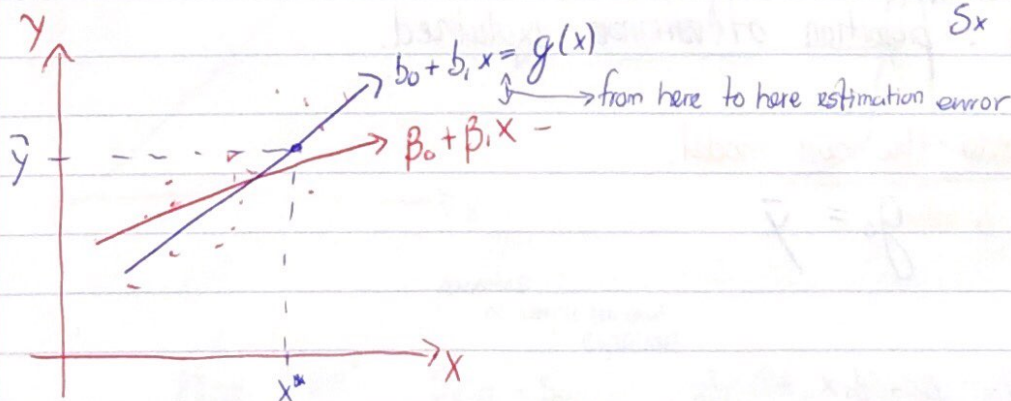
algorithm minimize SSE

$$g(x) = b_0 + b_1 x$$

$$b_0 = \bar{y} - r \frac{s_y}{s_x} \bar{x}$$

$$h^*(x) = \beta_0 + \beta_1 x$$

$$b_1 = r \frac{s_y}{s_x}$$



$$y_i = 3.5$$

$$\frac{1.6}{4} = 0.4$$

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Model perform metric
how well does g predict?

$$SSE = \sum_{i=1}^n e_i^2 = \sum (y_i - \hat{y})^2 \quad (\text{units } y\text{-square})$$

$$MSE = \frac{1}{n-2} SSE \quad (\quad " \quad " \quad ")$$

mean square error

$$RMSE = \sqrt{MSE} \quad (\text{Units } y)$$

approximate interpretation

g is RMSE of from y on average.

$g(x) \pm 2 \cdot RMSE$ is a 95% \rightarrow Empirical Rule
Confidence set for y . (rule of thumb)

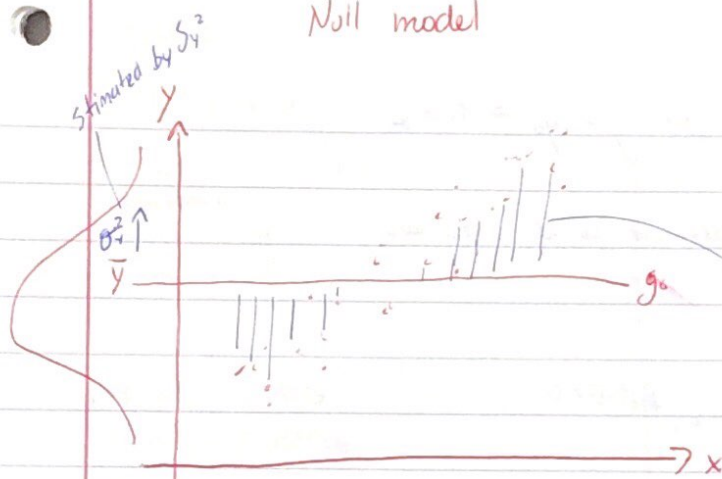
R^2 sample
 R : 1 proportion of ^{response} variance explained.

Consider the null model.

$$g_0 = \bar{y}$$

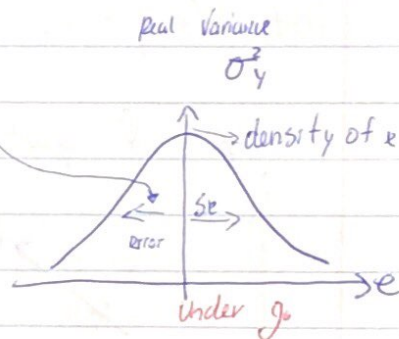
$$y = \beta_0 + \beta_1 x$$

Noit model



Sample variance

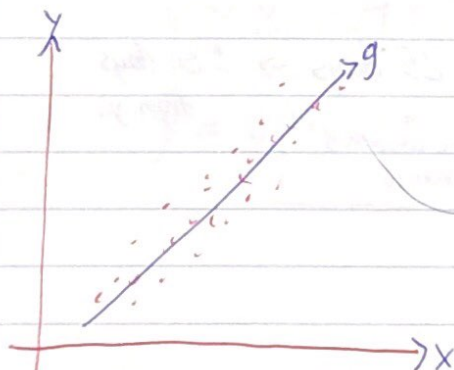
$$S_y^2 = \frac{1}{n-1} \sum (y_i - \bar{y})^2$$



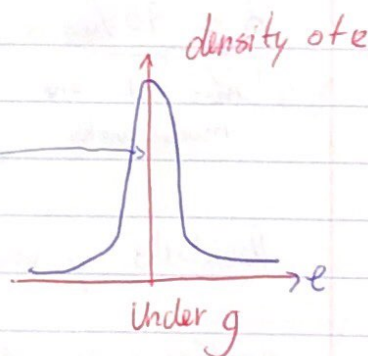
$$S_y^2 = \frac{1}{n-1} \sum (y_i - \bar{y})^2$$

SSE₀ Sum of squares of g₀

SST = sum squares total



Very small errors
small density



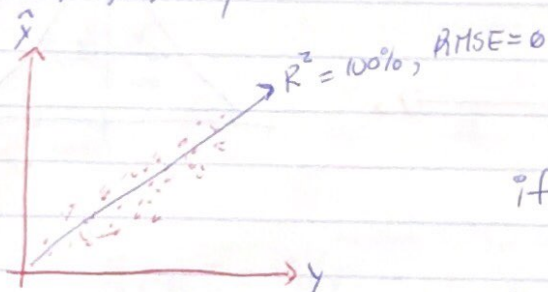
percentage of sample variance explained

$$R^2 = \frac{\Delta se^2}{S_{e,0}^2} = \frac{S_{e,0}^2 - S_e^2}{S_{e,0}^2} = \frac{\frac{1}{n-1} SSE_0 - \frac{1}{n-1} SSE}{\frac{1}{n-1} SSE_0} = \frac{SSE}{SST} = 1 - \frac{SSE}{SST}$$

$$\max (R^2) = 1 \quad \text{If } g = g_0 \Rightarrow R^2 = 0$$

$\inf \{R^2\} = -\infty \rightarrow$ because it can be in any plane

y & \hat{y} plot



if $R^2 \uparrow \Rightarrow$ then $RMSE \downarrow$

\rightarrow as you approach 100% ERROR go down.

y : # payments day late

$$R^2 = 98\%$$

this is how the model works

$$RMSE = 25 \text{ days} \rightarrow \pm 50 \text{ days from } y.$$

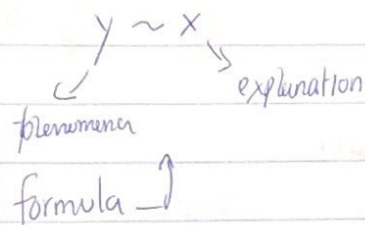
we care about this the variance

Residuals y and g

errors y and g^*

$RMSE \Rightarrow$ how far off you are from y

Robdio



in pack stats function \rightarrow lm
sigma \Rightarrow RMSE

$p=1$

$X_{\text{raw}} \in \{\text{red, green}\}$

coded $\begin{matrix} \downarrow & \downarrow \\ 0 & 1 \end{matrix}$

$\Rightarrow x \in \{0, 1\}$ binary

$\mathbb{I}_{\text{green}} = 1$

$y \sim x$ A: OLS

$$\hat{y} = g(x_{\text{raw}}) = \begin{cases} \bar{y}_r & \text{if } x_{\text{raw}} = 0 \\ \bar{y}_g & \text{if } x_{\text{raw}} = 1 \end{cases} = \bar{y}_r + (\bar{y}_g - \bar{y}_r)x$$

$\bar{y}_r + (\bar{y}_g - \bar{y}_r)x$

\bar{y}_r \bar{y}_g