

$$\log\left(\frac{p_i}{1-p_i}\right) = \text{logit}(p_i)$$

as $p_i \rightarrow 1$ $\text{logit}(p_i) \rightarrow +\infty$

as $p_i \rightarrow 0$ $\text{logit}(p_i) \rightarrow -\infty$

Extra office hour

①

($i=1, \dots, n$)

$(Y_i | p_i, \beta) \stackrel{I}{\sim} \text{Bernoulli}(p_i)$ \uparrow
3065

$$\log\left(\frac{p_i}{1-p_i}\right) = \beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \dots + \beta_k x_{ik}$$

old school rule:

try to avoid $k \leq 5$

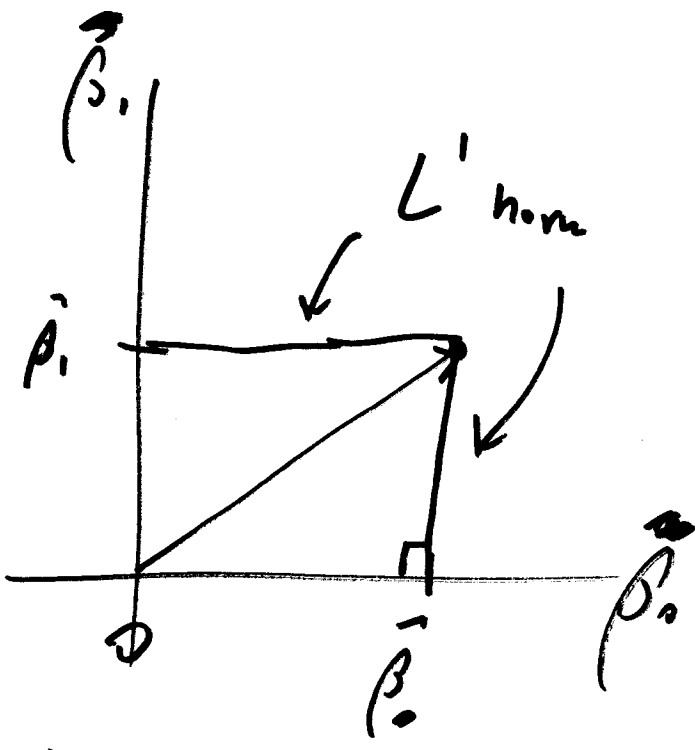
by that $\frac{n}{k} < 10$

$$1 + 57 + \binom{57}{2} = \underline{\underline{1654}}$$

GOF

regularization
↓

$$\log(\text{posterior}) = \log(\text{likelihood}) + \log(\text{prior})$$



$$\|\hat{\beta}\|_2 = \sqrt{\hat{\beta}_0^2 + \hat{\beta}_1^2}$$

(L² norm of $\hat{\beta}$)

$$\hat{\beta} = (\hat{\beta}_0, \hat{\beta}_1)$$

$$\|\hat{\beta}\|_1 = \frac{|\hat{\beta}_0| + |\hat{\beta}_1|}{(L_1, L_1)}$$

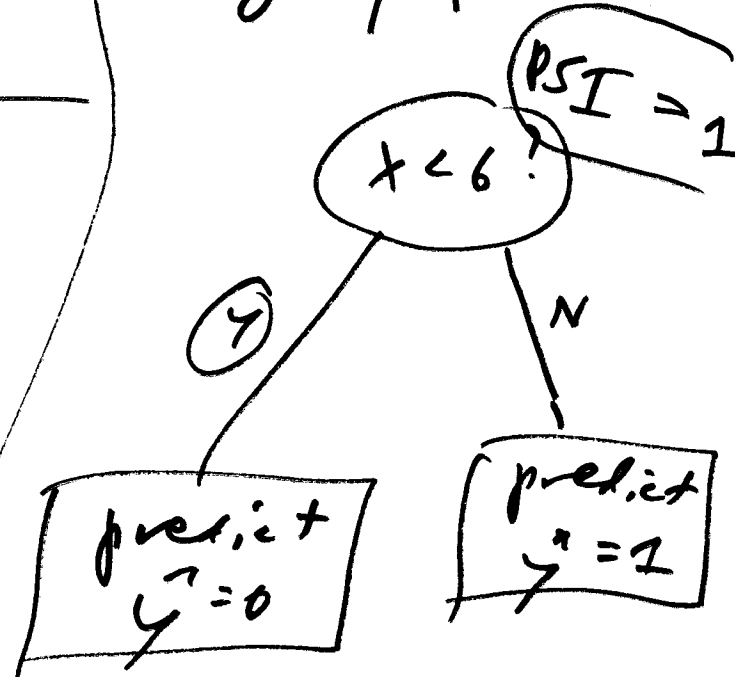
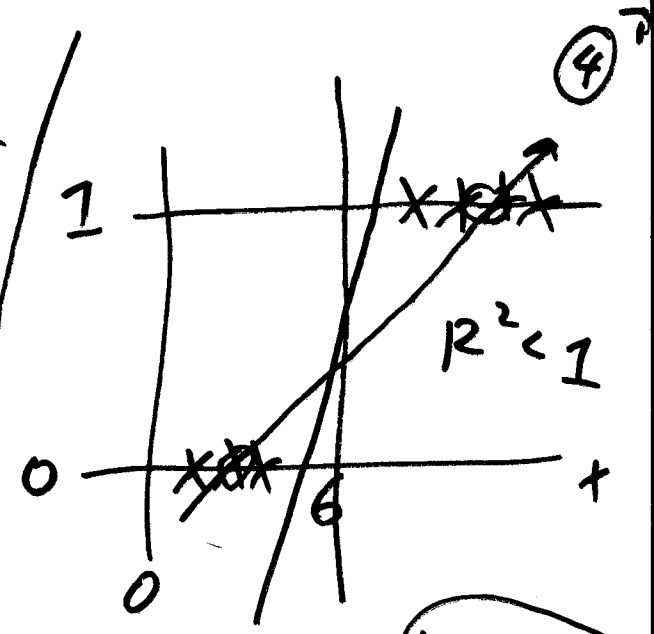
$\|\hat{\beta}\|_1 > \|\hat{\beta}\|_2$
 except when $\hat{\beta} = \underline{0}$

$(Y_i | p_i \in \mathcal{B}) \sim \text{Bernoulli}(p_i)$
 $(i = 1, \dots, n)$

$$\log\left(\frac{p_i}{1-p_i}\right) = \beta_0$$

$$\log\left(\frac{p_i}{1-p_i}\right) = \beta_0 + \beta_1 x_{i1} \begin{cases} 1 & \text{if } _ \\ 0 & \text{else} \end{cases}$$

	R^2	PSI
no predictive signal	0	0
as strength of predictive signal ↑	↑	↑
perfect predictive signal	1	1



$R^2 = \text{PSI}$ if
 ① OLS (linear) ^{THAT'S 2(A)}
 ② $y = \begin{cases} 1 \\ 0 \end{cases}$

$$y_i = I(x_i > 6)$$