

FYS-STK3155/4155, lecture
September 22, 2025

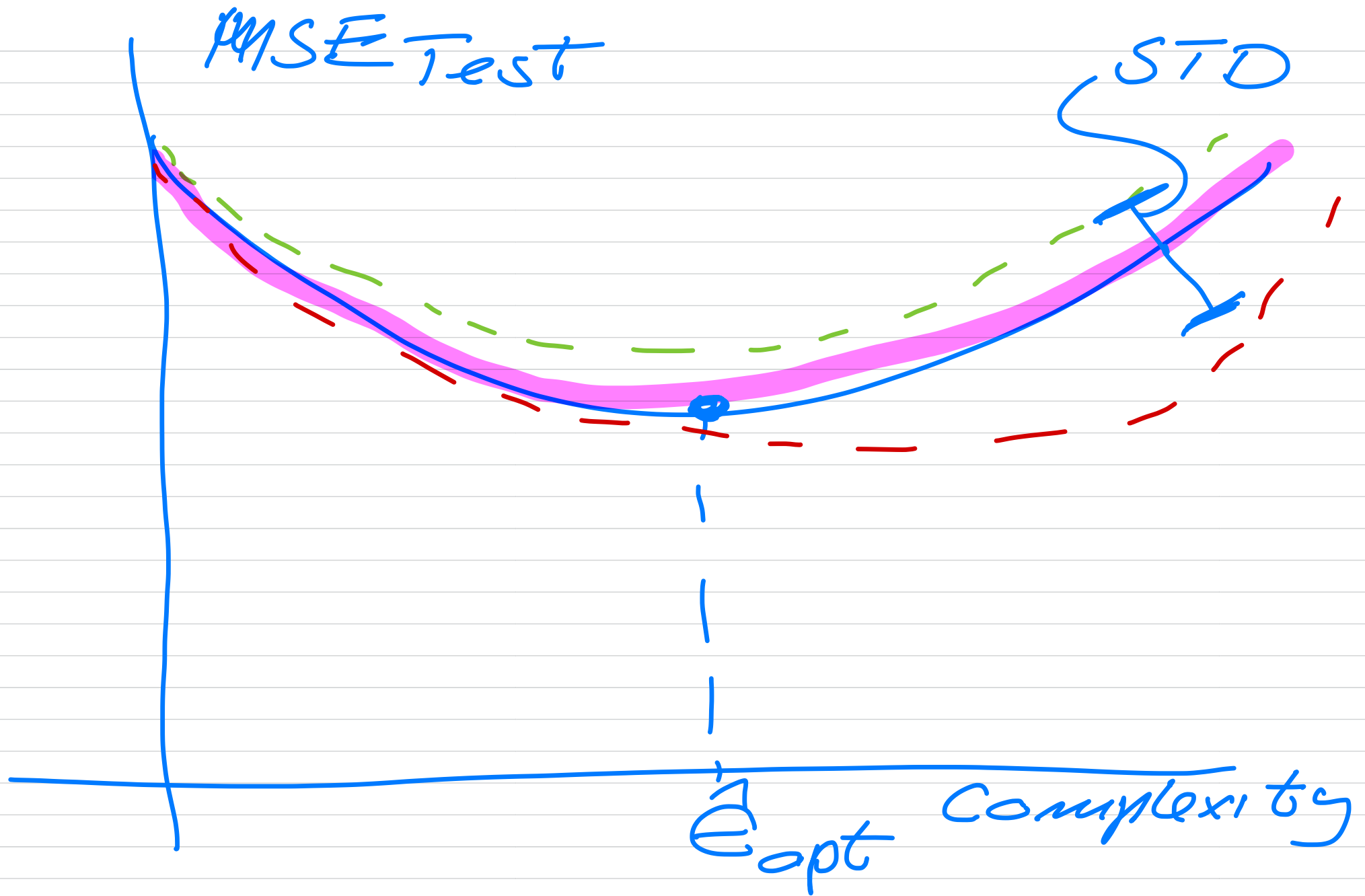
FYS-STK3155/4155 September 22

Resampling methods-

- Bootstrap
- cross-validation

Classification problems

- Logistic regression

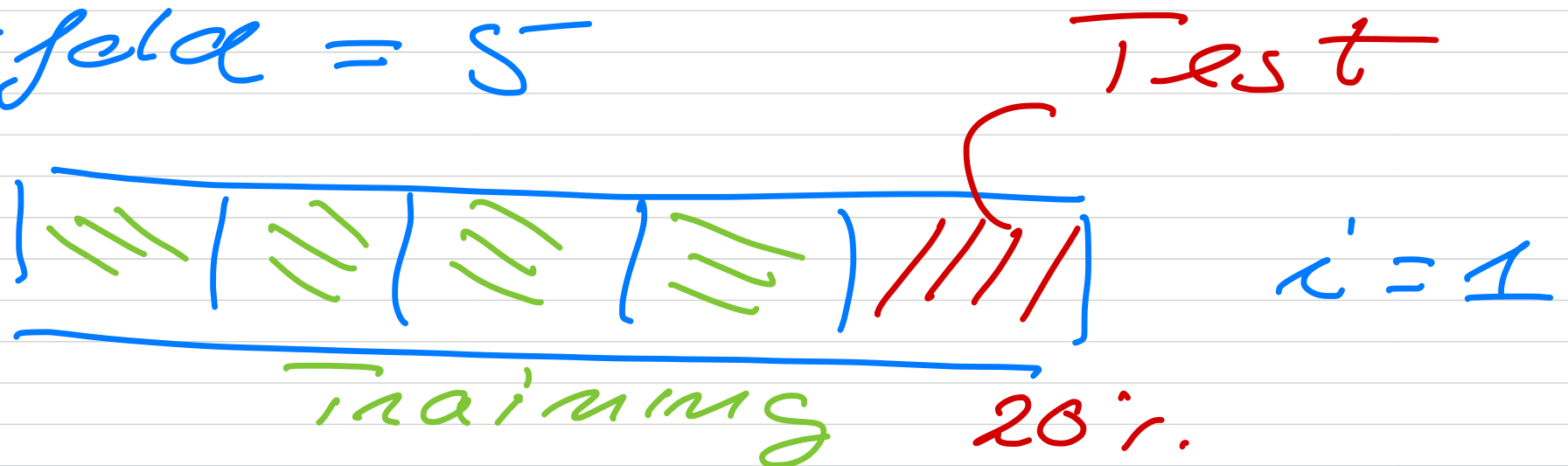


Cross-validation for resampling

- $kfold$ = number of folds

$kfold \sim 5-20$

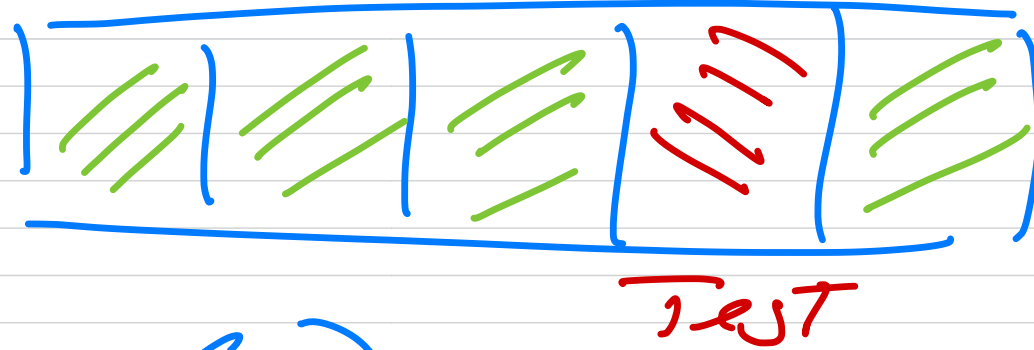
$kfold = 5$



$MSE_{Test}(1)$

LOOCV (only one in test set)

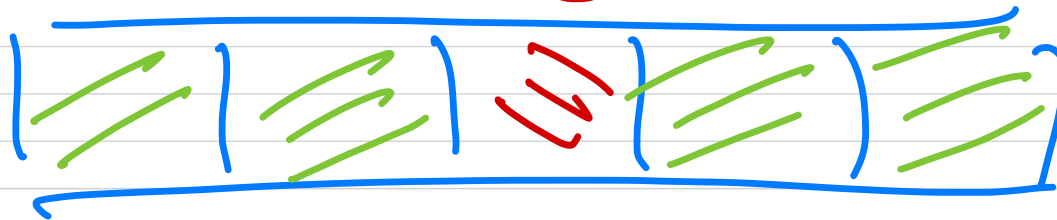
$$\lambda = 2$$



$$MSE_{Test}(2)$$

Test

$$\lambda = 3$$



$$MSE_{Test}(3)$$

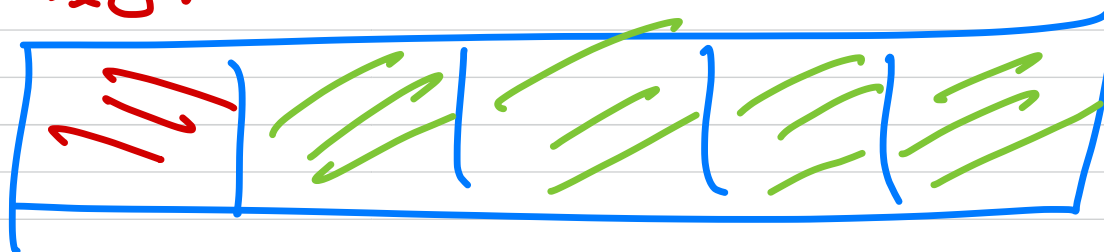
,

,

$$MSE_{Test}(4)$$

Test

$$\lambda = 5$$



$$MSE_{Test}(5)$$

$$\overline{MSE}_{Test} = \frac{1}{N} \sum_{k=1}^N MSE_{Test}(k)$$

X

Logistic regression

linear regression continuous

$$y_i = f(x_i) + \varepsilon_i$$

$$\varepsilon_i \sim N(0, \sigma^2)$$

$$x_i \in (-\infty, \infty)$$

$$([x_a, x_b])$$

$$y_i \in (-\infty, \infty)$$

$$([y_a, y_b])$$

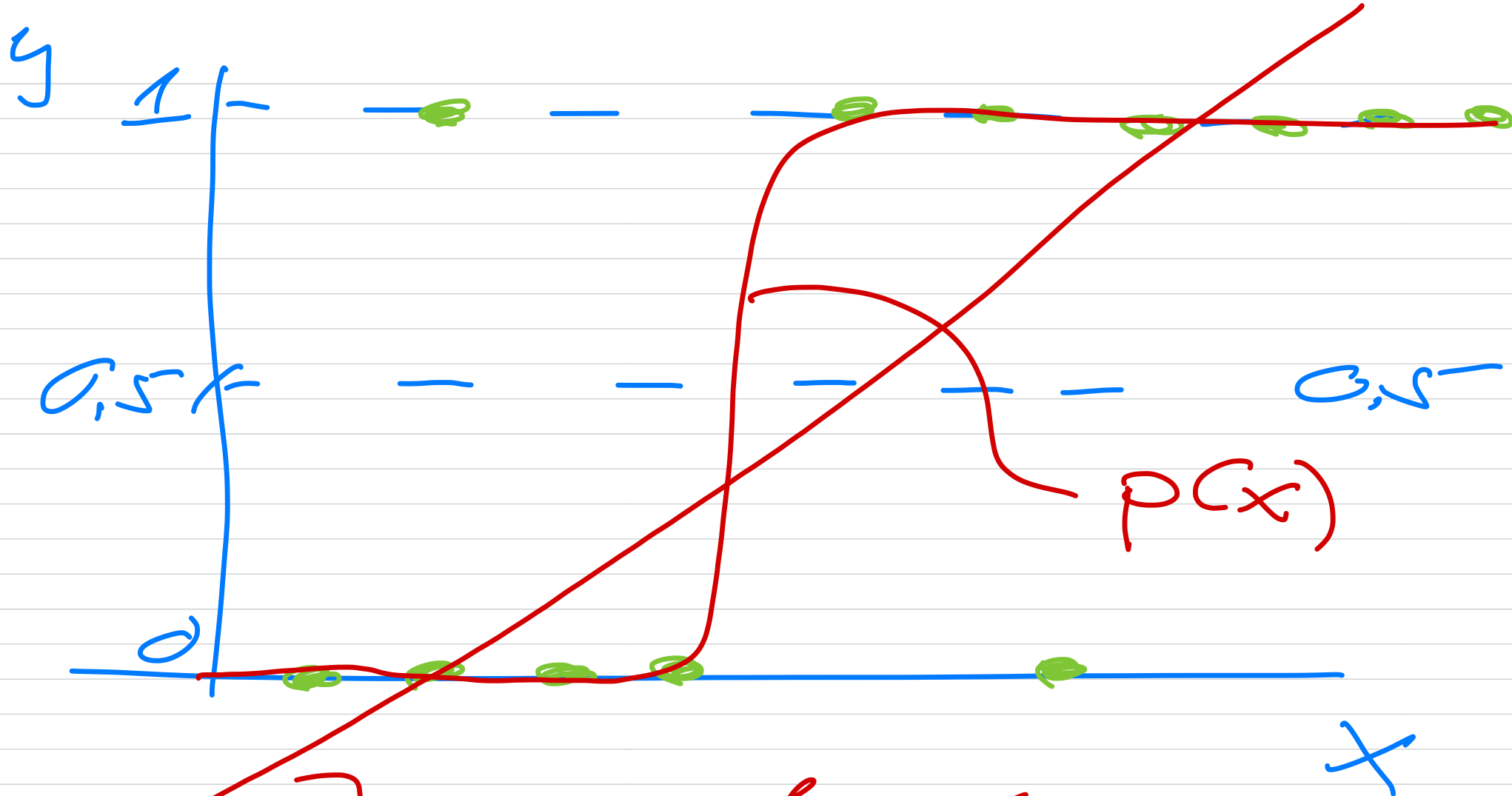
$$y_i \approx \underbrace{\sum x_{ij} \theta_j}_{\tilde{y}_i \approx f(x_i)} + \varepsilon_i$$

First order polynomial

$$y_i \approx \theta_0 + \theta_1 x_i + \varepsilon_i$$

What if y_i are discrete?

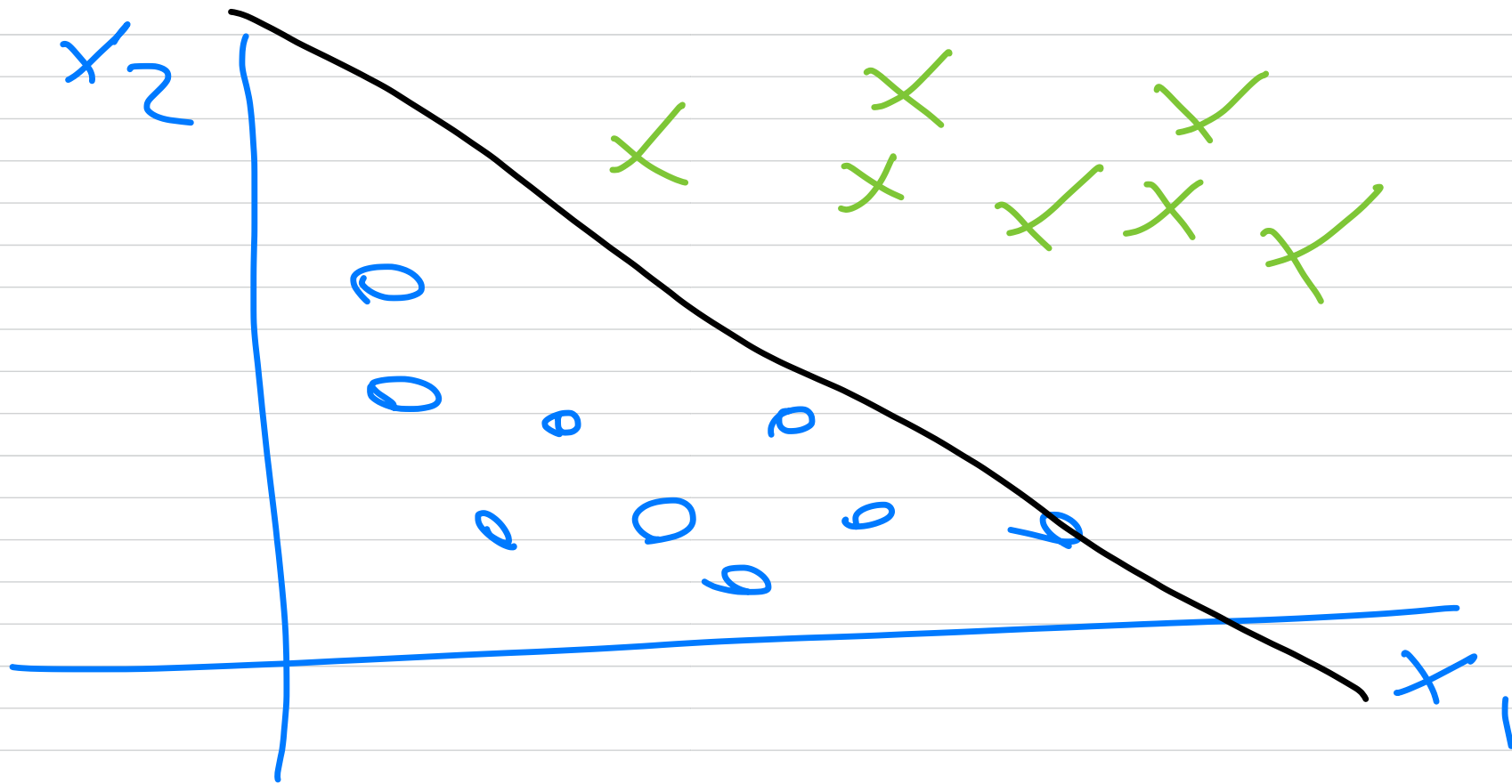
Example $y_i = \{0, 1\}$
Binary case



~ goes from

$y = -\infty$ to $y = +\infty$

if $p(x_i) \leq 0.5$, then $y_i = 0$
 else $p(x_i) > 0.5$ then $y_i = 1$



$$y_i = f(x_i) + \varepsilon_i'$$

replace

$$y_i = p(x_i) + \varepsilon_i'$$

$$0 \leq p(x_i) \leq 1$$

can be interpreted as a probability

$$\int_{x \in D} p(x) dx = 1$$

$$\left(\sum_{i \in D} p(x_i) = 1 \right)$$

$$x \in [0, 1] ; p(1) = 1$$

$$p(x_i) \leq p(x_j) \text{ if } x_i \leq x_j$$

Example: Sigmoid function

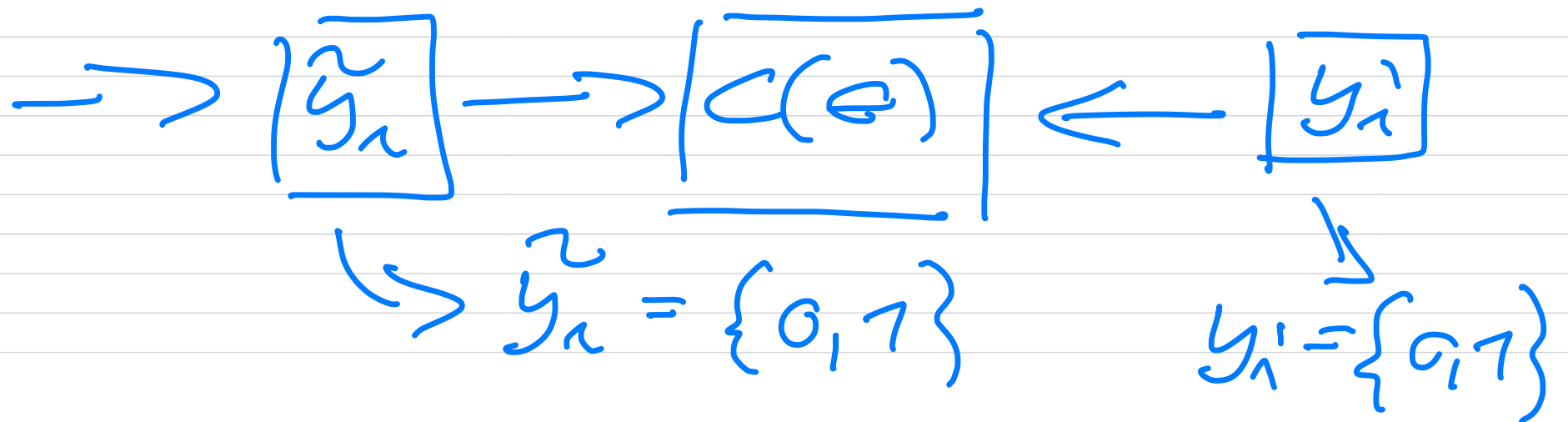
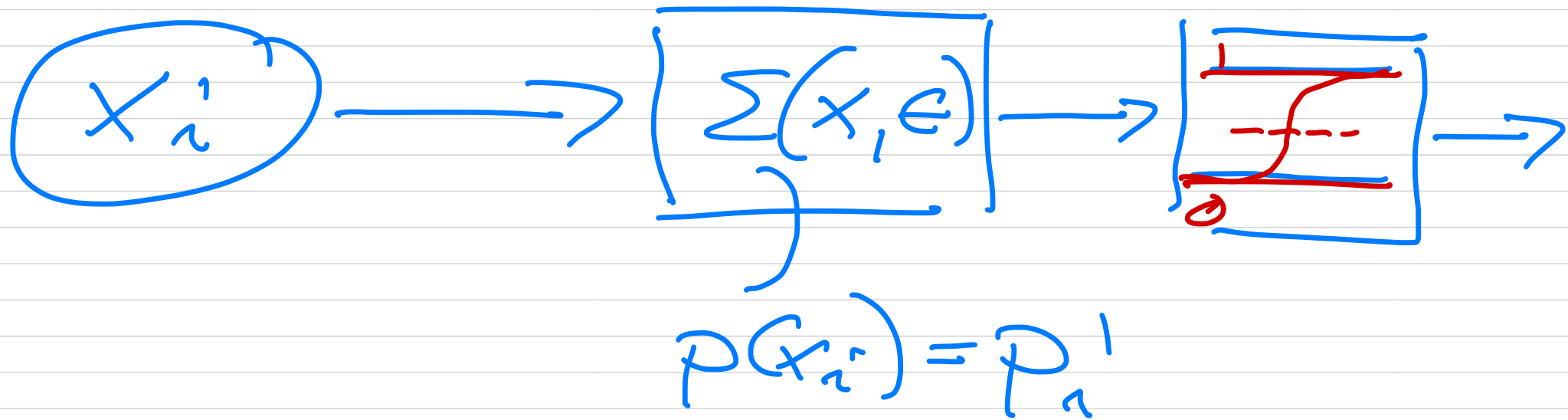
$$p(x) = \frac{e^{\theta x}}{1 + e^{\theta x}}$$

$$y = 1 ; \quad p(x) = \frac{e^{\theta x}}{1 + e^{\theta x}}$$

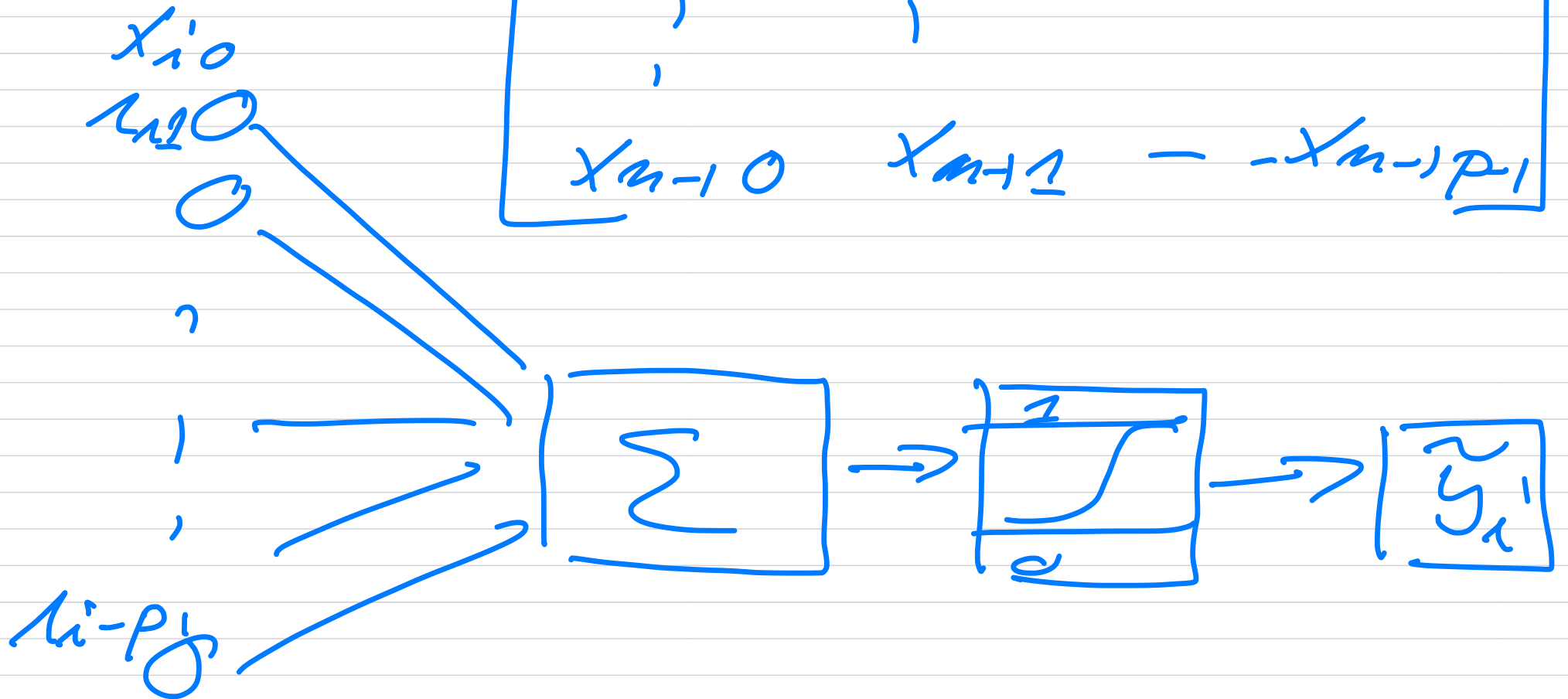
$$y = 0 ; \quad 1 - p(x) = \frac{1}{1 + e^{\theta x}}$$

$$\sum_i p(x_i) = p + (1 - p) \\ = 1$$

Graphical representation



$$X = \begin{bmatrix} x_{00} & x_{01} & \dots & x_{0p-1} \\ x_{10} & 1 & & \\ x_{20} & 1 & & \\ \vdots & \vdots & & \\ x_{n-10} & x_{n-11} & \dots & x_{n-1p-1} \end{bmatrix}$$



Model

$$p(x_i') = \frac{e^{\beta_0 + \beta_1 x_i'}}{1 + e^{\beta_0 + \beta_1 x_i'}}$$

$$D = \left\{ (x_0, y_0), (x_1, y_1), \dots, (x_{n-1}, y_{n-1}) \right\}$$

$$y_i' = p(x_i') + \varepsilon_i' = p_i' + \varepsilon_i'$$

$$P(D|e) = C(e)$$

$$P(D|e) = \prod_{i=0}^{n-1} P_i^{y_i} (1-P_i)^{1-y_i}$$

$$y_i = 1 \quad ; \quad P_i$$

$$y_i = 0 \quad ; \quad 1-P_i$$

$$\frac{\partial P(D|e)}{\partial \theta_j} = 0$$

$$C(\theta) =$$

$$-\log P(D|\theta)$$

$$\hat{\theta} = \arg \min_{\theta \in \mathbb{R}^p} (-\log P(D|\theta))$$

$$\prod_{i=0}^{n-1} p_i^{y_i} (1-p_i)^{1-y_i}$$

$$C(e) = - \sum_{i=0}^{n-1} \left[g_i' \log p_i' + (1-g_i') \log (1-p_i') \right]$$

$$\log \left[\frac{e^{\theta_0 + \theta_1 x_i'}}{1 + e^{\theta_0 + \theta_1 x_i'}} \right]$$

$$= (\theta_0 + \theta_1 x_i')$$

$$- \log (1 + e^{\theta_0 + \theta_1 x_i'})$$

$$C(\theta) = - \sum_{i=0}^{n-1} \left[g_i'(\theta_0 + \theta_1 x_i') - \log(1 + e^{\theta_0 + \theta_1 x_i'}) \right]$$

$$\frac{\partial C}{\partial \theta_0} = 0 \quad \wedge \quad \frac{\partial C}{\partial \theta_1} = 0$$

$$\frac{\partial C}{\partial \theta_0} = 0 = - \sum_{i=0}^{n-1} (g_i' - p_i')$$

$$\frac{\partial C}{\partial \theta_1} = 0 = - \sum_{i=0}^{n-1} x_i' (y_i' - p_i)$$

$$\frac{\partial C}{\partial \theta} = - X^T (\vec{y} - \vec{p})$$

depends
on θ

$$p(x_i') = \frac{e^{\theta_0 + \theta_1 x_i'}}{1 + e^{\theta_0 + \theta_1 x_i'}}$$

$$\frac{\partial^2 C}{\partial \theta^2} = X^T W X$$

$$w_{ii} = p_i (1 - p_i)$$

$$w_{ij} \ (i \neq j) = 0$$

$$E^{(n+1)} = E^{(n)} - \eta \left(\frac{\partial C}{\partial \theta} \right)_{\theta = \theta^{(n)}}$$