

$$\bar{x} \in \mathbb{R}^d \quad y \in \mathbb{R}$$

$$L(y, f(\bar{x})) = (y - f(\bar{x}))^2$$

$$\underbrace{EPE[f]}_{\min} = \mathbb{E}_{p(\bar{x}, y)} [L(y, f(\bar{x}))] = \int (y - f(\bar{x}))^2 p(\bar{x}, y) d\bar{x} dy =$$

$$= \int \int (y - f(\bar{x}))^2 p(y|\bar{x}) p(\bar{x}) dy d\bar{x} =$$

$$= \int p(\bar{x}) \left(\int (y - f(\bar{x}))^2 p(y|\bar{x}) dy \right) d\bar{x}$$

$\downarrow f(\bar{x})$

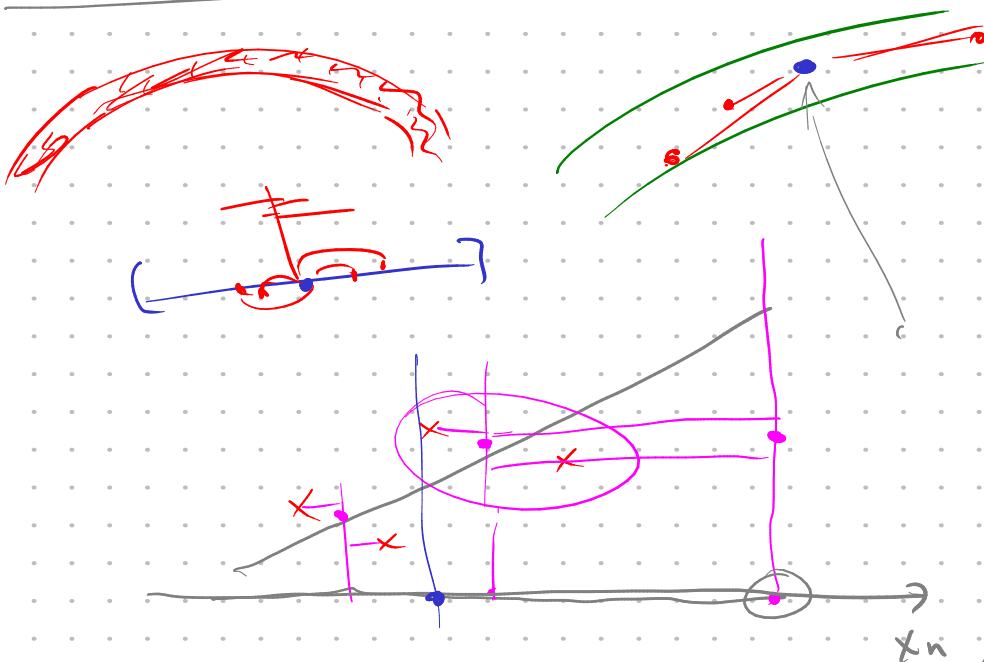
$$\int (y - a)^2 p(y) dy \xrightarrow{a} \min$$

$$a = \mathbb{E}_p[y]$$

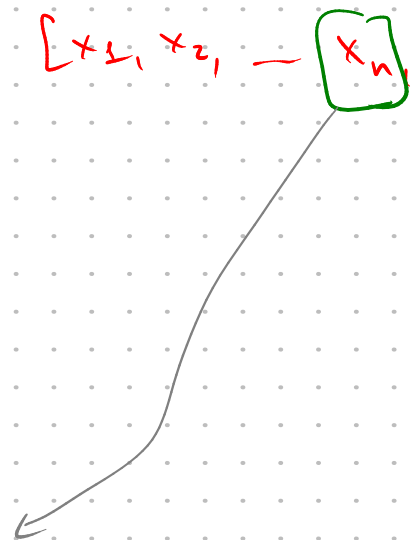
min

$$\hat{f}(\bar{x}) = \mathbb{E}_{p(y|\bar{x})}[y]$$

regression function

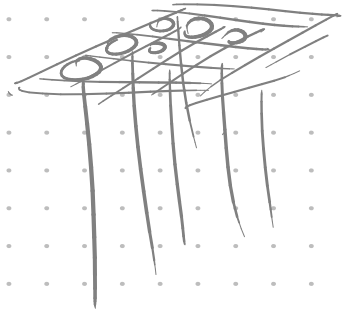
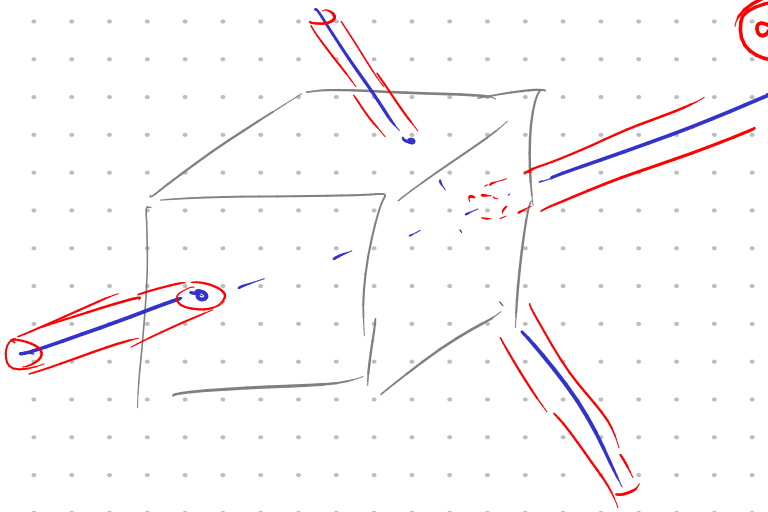
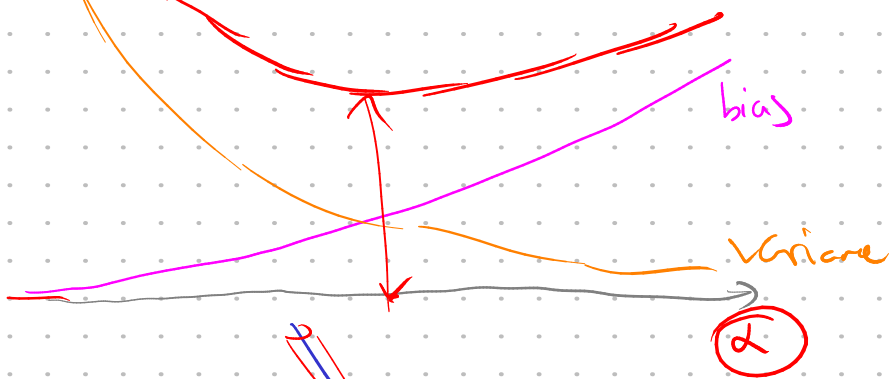


$$[x_1, x_2, \dots, x_n, \dots, x_N]$$

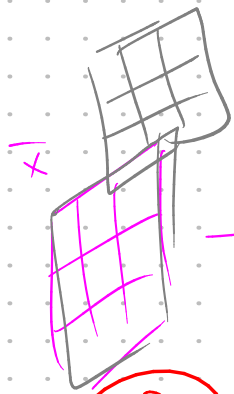
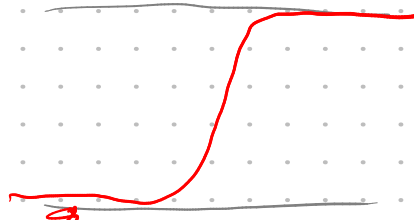


$$EPE[f] = \text{Bias}^2 + \text{Variance} + \text{Noise}$$

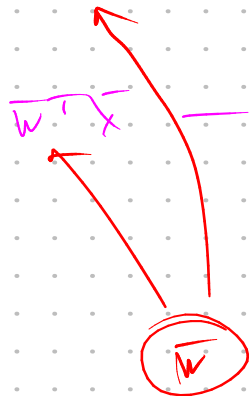
$$E[(f - \hat{f})^2] = E[(f - E\hat{f})^2] + E[(E\hat{f} - \hat{f})^2]$$



$$\bar{w}^T \bar{x}$$



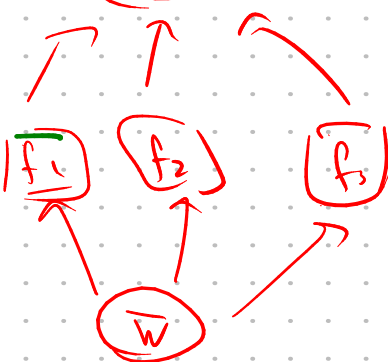
$$\bar{w}^T \bar{x}$$



$$h \rightarrow h(\bar{w}^T \bar{x})$$

$$\nabla_{\bar{w}} = h'(\bar{w}^T \bar{x}) \cdot \bar{x}$$

$$\nabla_{\bar{w}} = \sum_{\text{grid}} h'(\bar{w}^T \bar{x}) \bar{x}$$



$$\nabla_{\bar{w}} F = \left[\frac{\partial F}{\partial f_1} \right] \nabla_{\bar{w}} f_1 + \left[\frac{\partial F}{\partial f_2} \right] \nabla_{\bar{w}} f_2 + \left[\frac{\partial F}{\partial f_3} \right] \nabla_{\bar{w}} f_3$$

RoI Align

