

# Exer. 1.1 :

$$1. \quad \eta(z, x) = E(l(y, z) | X=x)$$

$$R(f) = E(\eta(f(x), X))$$

$$2. \quad \eta(a+bx, x) = E[l(a+\beta x + \varepsilon), a+bx] \quad \boxed{X=x}$$

↑  
This only  $\varepsilon$ , which is  $\perp X$

Quadratic loss  
L-loss ;

If

$$l(y, z) = (y-z)^2$$

$$= E \left( \left( \underbrace{(\alpha - a) + (\beta - b)x}_{\otimes} + \varepsilon \right)^2 \right)$$

$$= E \left( \otimes^2 + 2 \otimes \varepsilon + \varepsilon^2 \right)$$

$$= \otimes^2 + 2 \otimes \underbrace{E(\varepsilon)}_{=0} + \underbrace{E(\varepsilon^2)}_{\sigma^2}$$

$$= \left[ (\alpha - a) + (\beta - b)x \right]^2 + \sigma^2$$

$$(\alpha - a)^2 + 2(\alpha - a)(\beta - b)x + (\beta - b)^2 x^2$$

• If  $f_{a,b}(x) = a + bx$ , then  $R(f_{a,b}) = E \left[ \left[ (\alpha - a) + (\beta - b)x \right]^2 + \sigma^2 \right]$

$$R(f_{a,b}) = \sigma^2 + (\alpha - a)^2 + 2(\alpha - a)(\beta - b)E(x) + (\beta - b)^2 E(x^2)$$

Exer. 1.2  $\mathbb{E}(\hat{R}_n(f)) = R(f)$

Empirical risk is an unbiased estimator of  $R(f)$ .

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Exer 1.3

1. Pick a measurable  $f$ . We have

$$R(f) \stackrel{\uparrow}{=} \mathbb{E} \left( \underbrace{\ell(f(x), x)}_{\forall} \right) \geq \mathbb{E} \left( \ell(f^*(x), x) \right) = R^*$$

Ex 1.1  
Q 1

$$\min_z \ell(z, x) = \ell(f^*(x), x)$$



$$2. \blacktriangleright R(f) = \mathbb{P}(Y \neq f(x))$$

$$\blacktriangleright r(z; x) = \mathbb{P}(Y \neq z | X=x) \quad z \in \{0, 1\}$$

$$= \begin{cases} \mathbb{P}(Y=0 | X=x) & \text{if } z=1 \\ \mathbb{P}(Y=1 | X=x) & \text{if } z=0 \end{cases}$$

Note:  $\mathbb{P}(Y=0 | X=x) + \mathbb{P}(Y=1 | X=x) = 1$

$$r(f^*(x), x) = \mathbb{P}(Y=0 | X=x) \wedge \mathbb{P}(Y=1 | X=x)$$

↑  
means min

So  $f^*(x) = \begin{cases} 1 & \text{if } \mathbb{P}(Y=0|X=x) < \frac{1}{2} \Leftrightarrow \mathbb{P}(Y=1|X=x) > \frac{1}{2} \\ 0 & \text{if } \mathbb{P}(Y=1|X=x) < \frac{1}{2} \Leftrightarrow \mathbb{P}(Y=0|X=x) > \frac{1}{2} \\ \begin{pmatrix} 0 \text{ or } 1 \\ \text{at random} \end{pmatrix} & \text{if } \mathbb{P}(Y=0|X=x) = \frac{1}{2} \end{cases}$

$$R(f^*) = \mathbb{P}(Y \neq f^*(X)) = \mathbb{E} \left( \mathbb{P}(Y=0|X) \wedge \mathbb{P}(Y=1|X) \right)$$

► Excess Risk:  $R(f) - R^*$

$$3. \blacktriangleright R(f) = \mathbb{E}[(Y - f(X))^2]$$

$$r(z, x) = \mathbb{E}[(Y - z)^2 | X=x]$$

$$\underbrace{Y^2 - 2zY + z^2}_{\mathbb{E}(Y^2 | X=x) - 2z \mathbb{E}(Y | X=x) + z^2}$$

$$= \underbrace{\mathbb{E}(Y^2 | X=x) - 2z \mathbb{E}(Y | X=x) + z^2}_{} \quad \checkmark \quad \cap$$

is a quadratic fun. of  $z$

that looks like 

It has a unique min at  $z = \mathbb{E}(Y | X=x) = f^*(x)$