Y categorical (not numeric)

ground truth: for given x,

P(Xo) = P(Y=11 X=Xo)

Yandownness

Some Xo Can give different values of Y irreducible error

want
$$\widehat{T}(X_o)$$
 —) returns the category of Y

most likely at x.

measure error

misclassification rate : + 2, 1 {Y, + Y; } between 0 and 1

accuracy = 1- misclassification rate

Confusion matrix (k=2)

		true 1 0		a = # samples where b =		Y; = 1 Y; = 0
pred		a	6	c = d =	Ŷ; =0 Ŷ: -0	Yi=1 Yi=0
	0	C	d	n=a+b+c+d		1,70

misclassification rate = b+c accuracy = a+0

before, used CU to estimate MSPE now use CV to estimate misclassification vate

k=2Compute P(Y=1|X) P(Y=0|X)

predict Y=1 if P(Y=1/x)>0.5 & threshold

Why not consider different values of threshold? might wont to consider different threshold

L, can get smaller misclassification error with different threshold

linear regression gives Yany number ER want to get P(Ŷ=1) instead probability p 0 ≤ p ≤ 1

$$\begin{array}{ccc}
x \in R & \longrightarrow & \frac{e^{x}}{1 + e^{x}} \in (0, 1) \\
\hat{Y} = \hat{\beta}_{0} + \hat{\beta}_{1} \times & \longrightarrow & \frac{e^{x}p(\hat{\beta}_{0} + \hat{\beta}_{1} \times)}{1 + e^{x}p(\hat{\beta}_{0} + \hat{\beta}_{1} \times)} \in (0, 1) \\
P(\hat{Y} = 1 \mid X)
\end{array}$$

$$P(\hat{Y}=1|X) = \frac{\exp(\hat{\beta}_0 + \hat{\beta}_1 X)}{1 + \exp(\hat{\beta}_0 + \hat{\beta}_1 X)} \longrightarrow 1 \text{ as } X \to \infty$$

B, >0 than X increasing means P(Y=1/X) increases also $\hat{\beta}$. CO then x increasing means $P(\hat{Y}=1|X)$ decreases