$$X, Y = f(X) \approx Y = (f(X)-Y)^{2} = \sum_{x,y} P(X=n,Y=y) (f(n+y)^{2})$$

$$EPE = E_{xy} [(f(X)-Y)^{2}] = \sum_{x,y} P(X=n,Y=y) (f(n+y)^{2})$$

$$I(x,y,f)$$

$$E_{xy}[l(x,y,f)] = E_{x} E_{y|x}[l(x,y,f)|x]$$

 $\widehat{P}(X=x,Y=y) = P(X=x) \cdot P(Y=y|X=x)$

$$EPE(f) = \sum_{n} \sum_{y} P(x=n) \cdot P(Y=y|X=n) \cdot (f(n)-y)^{2}$$

$$= \sum_{n} P(x=n) \cdot \sum_{y} P(Y=y|X=n) \cdot (f(n)-y)^{2}$$

$$= \sum_{n} P(x=n) \cdot P(Y=y|X=n) \cdot (f(n)-y)^{2}$$

Solve
for each
$$g$$
 the $x = x = x$ $= x$ $=$

the regression fundion $f(x) = E_{Y|X}[Y|X]$ $f^{lent}(x) = E[Y|X=n]$ 715 yi of those Observations for which Ni= 2 (1) K-nn (2) XY TX, T X, T NXP JUNX

Curse of dimensionality:
$$\frac{1}{4} = h \cdot 1$$

$$\frac{1}{4} = V \cdot 1$$

$$\frac{1}{1} \times_{1}$$

$$\frac{1}{1} \times_{1}$$

$$\frac{1}{1} \times_{1}$$

$$E[Y|X=n] = \beta^{T}n$$

$$X \sim N(0,[0])$$

$$Y|K \sim N(\beta^{T}X,1)$$

$$Y = \beta^{T}X + E$$

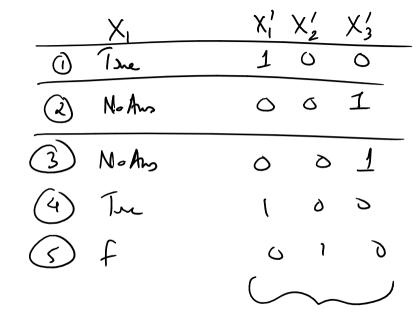
$$\uparrow N(0,$$

$$f(x) = \beta^{T} x \qquad \text{mon } EPE(\beta)$$

$$\beta = \left(E_{x}[x \times T]\right)^{T} \left(E_{xy}[x Y]\right)$$

$$\frac{\partial}{\partial \beta} = 0$$

$$Z = \begin{cases} X_1 + \beta_2 X_2 + \cdots & \beta_p \cdot X_p + \beta_0 1 \\ X_1 - X_p \end{cases}$$
New
$$Z = \begin{cases} X_1 - X_p \\ X_2 - X_p \\ X_1 - X_p \end{cases}$$
He particularly the state of the state



EPE for Classification GE{1 ... K3 $EPE(\hat{G}(x))$ $= E_{x} \sum_{k=1}^{\infty} L(G=k, \hat{G}(x)) \cdot P(G=k|X)$ $= E_{xG}[L(G, \hat{G}(x))]$

$$L(G=1,G(x)) \cdot P(G=1|X=x) + L(G=2,G(x)) \cdot P(G=2|X=x)$$

$$L(G=3,G(x)) \cdot P(G=3|X=x)$$

$$L(G=3,G(x)) \cdot P(G=3|X=x)$$