

Goal: to show that $E\left[\frac{1}{2}(y - \hat{y})^2\right]$ is minimized by

$$f^*(x) = E[Y | X=x] \text{ for any given } x.$$

(1) $E\left[\frac{1}{2}(y - \hat{y})^2\right]$ is minimized when $E[(y - \hat{y})^2]$ is min.

$$(2) E[(y - \hat{y})^2] = E[(\hat{y} - E[Y|X=x] + E[Y|X=x] - y)^2]$$

(3) let $a = \hat{y} - E[Y|X=x]$ and $b = E[Y|X=x] - y$.

$$(4) \text{ then } (2) = E[(a+b)^2] = E[a^2] + E[2ab] + E[b^2]$$

Starting with $E[b^2]$:

$$E[b^2] = E[(E[Y|X=x] - y)^2]. \text{ This is invariable; not dependent on } f^*(x)$$

$$(5) E[2ab] = 2E[ab] = 2E[\hat{y}E[Y|X=x] - \hat{y}y - E[Y|X=x]^2 + yE[Y|X=x]]$$

$$E[\hat{y}] = f^*(x) = E[Y|X=x]$$

$$\Rightarrow (5) = 2[E[Y|X=x]^2 - E[Y|X=x]y - E[Y|X=x]^2 + yE[Y|X=x]] = 0$$

Therefore $E\left[\frac{1}{2}(y - \hat{y})^2\right]$ is min when $E[a^2]$ is min

$$E[a^2] = E[(\hat{y} - E[Y|X=x])^2]$$

$$\arg\min_{\hat{y}} ((\hat{y} - E[Y|X=x])^2) = E[Y|X=x]$$