

Corollaries of BV Decomposition

Cor. 1 Assume additive noise ;

if $\exists w^*, g(x) = f(x; w^*) = \langle w^*, x \rangle$
 then $(g(x) - E_\epsilon[f_{LS}|x_0])^2 = 0$

$$\begin{aligned}
 & g^2 - 2g E_\epsilon[f_{LS}|x_0] + E_\epsilon[f_{LS}|x_0]^2 \\
 &= g^2 - 2g E_\epsilon[x_0(X^T X)^{-1} X^T (g + \epsilon)] + E_\epsilon[x_0(X^T X)^{-1} X^T (g + \epsilon)]^2 \\
 &= g^2 - 2g E_\epsilon[x_0(X^T X)^{-1} X^T g] + E_\epsilon[x_0(X^T X)^{-1} X^T g]^2 \\
 &= g^2 - 2g E_\epsilon[\underbrace{x_0(X^T X)^{-1} X^T X^T w^*}_I] + E_\epsilon[\underbrace{x_0(X^T X)^{-1} X^T X^T w^*}_I]^2 \\
 &= g^2 - 2g E_\epsilon[g] + E_\epsilon[g]^2 \\
 &= 0 \quad \text{OR: show } g(x) - E_\epsilon[f_{LS}|x_0] = 0 \text{ directly}
 \end{aligned}$$

Cor 2. $\text{Var}[f_{LS}|x_0] = \langle h, h \rangle \sigma^2$
 $h = x_0(X^T X)^{-1} X^T$

$$\begin{aligned}
 \text{Var}[f_{LS}|x_0] &= E_\epsilon[f_{LS}^2] - E_\epsilon[f_{LS}]^2 \\
 &= E_\epsilon[\langle h, y \rangle^2] - E_\epsilon[\langle h, y \rangle]^2 \\
 &= E_\epsilon[\langle h, g + \epsilon \rangle^2] - E_\epsilon[\langle h, y \rangle]^2 \\
 &= E_\epsilon[\langle h, g \rangle^2 + \langle h, \epsilon \rangle^2 + 2\langle h, g \rangle \langle h, \epsilon \rangle] - E_\epsilon[\langle h, g + \epsilon \rangle]^2 \\
 &= \langle h, g \rangle^2 + E_\epsilon[\langle h, \epsilon \rangle^2] + 0 - E_\epsilon[\langle h, g + \epsilon \rangle]^2 \\
 &= \langle h, g \rangle^2 - E_\epsilon[\langle h, g \rangle]^2 + E_\epsilon[\langle h, \epsilon \rangle^2] + 0 \\
 &= 0 + E_\epsilon[\langle h, \epsilon \rangle^2] + 0 \\
 &= 0 + E_\epsilon[\text{tr}[h^T \epsilon \epsilon^T h]] \\
 &= E_\epsilon[\text{tr}[h h^T \epsilon \epsilon^T]] = \text{tr}(\tilde{a} \tilde{b}) = \text{tr}(\tilde{b} \tilde{a}) \\
 &= \text{tr}[E_\epsilon[h h^T \cdot \sigma^2 I]]
 \end{aligned}$$

$$\begin{aligned} &= \text{tr}[hh^T E_\epsilon [E_\epsilon^T E]] \\ &= \text{tr}[hh^T \cdot \theta^2 I] \quad \leftarrow \text{quadratic rule} \\ &\qquad\qquad\qquad \searrow I \in \mathbb{R}^{n \times n} \\ &= \theta^2 \text{tr}[h^T h] = \text{tr}(ab^T) = \text{tr}[b^T a] \\ &= \theta^2 \cdot \langle h, h \rangle \end{aligned}$$

$$\begin{aligned}\text{Or, } E_e[\langle h, e \rangle^2] &= E_e[h^T e e^T h] \\ &= h^T E_e[e e^T] h \\ &= h^T \cdot h \cdot \sigma^2 = \langle h, h \rangle \cdot \sigma^2\end{aligned}$$