

(a)

$$(i) T = \sum_{i=1}^n a_i X_i$$

$$E(T) = E\left(\sum_{i=1}^n a_i X_i\right) = \sum_{i=1}^n a_i E(X_i)$$

$$E(X_i) = \mu \Rightarrow ET = \sum_{i=1}^n a_i \cdot \mu$$

$$\text{to satisfy } ET = \mu \Rightarrow \sum_{i=1}^n a_i = 1$$

$$(ii) \text{var}(T) = \sum_{i=1}^n a_i^2 \text{var}(X_i) = \sum_{i=1}^n a_i^2 \sigma^2$$

$$\text{create } F = \sigma^2 \sum_{i=1}^n a_i^2 + \lambda \left(\sum_{i=1}^n a_i - 1 \right)$$

$$\frac{\partial F}{\partial a_i} = 2\sigma^2 a_i + \lambda = 0 \Rightarrow a_i = -\frac{\lambda}{2\sigma^2}$$

$$\Rightarrow n a_i = 1 \quad \therefore a_i = \frac{1}{n}$$

This makes the estimator T behave as a mean which consistent with my guess.

$$(iii) \text{var}(T) = \sum_{i=1}^n a_i^2 \text{var}(X_i) = \sum_{i=1}^n \left(\frac{1+b}{n}\right)^2 \sigma^2$$
$$= \frac{\sigma^2}{n} (1+b)^2$$

$$\frac{\partial \text{var}(T)}{\partial b} = 0 \Rightarrow b = -1$$

$$\Rightarrow T = 0$$

has less MSE but high bias
using this is unreasonable

$$(iv) \text{MSE}(T) = 0$$

$$\text{MSE}(x) = \frac{\mu^2}{n}$$

Can't use this choice in practice

(b)

$$\begin{aligned} \text{(i)} \quad [\text{bias}(\hat{m}(x_0))]^2 &= (f(x_0) - y_{x_0})^2 \\ &= \left(f(x_0) - \frac{1}{K} \sum_{i=1}^K y_i \right)^2 \\ &= \left(f(x_0) - \frac{1}{K} \sum_{i=1}^K (f(z_i) + \varepsilon_i) \right)^2 \\ &= \left(f(x_0) - \frac{1}{K} \sum_{i=1}^K f(z_i) - \frac{1}{K} \sum_{i=1}^K \varepsilon_i \right)^2 \quad \Leftarrow \text{mean}(\varepsilon) = 0 \\ &= \left(f(x_0) - \frac{1}{K} \sum_{i=1}^K f(z_i) \right)^2 \end{aligned}$$

$$\begin{aligned} \text{(ii)} \quad \text{var}(\hat{m}(x_0)) &= E((\hat{m}(x_0) - f(x_0))^2) \\ &= E\left(\left(\frac{1}{K} \sum_{i=1}^K y_i - f(x_0)\right)^2\right) \\ &= E\left(\left(\frac{1}{K} \sum_{i=1}^K f(z_i) - f(x_0) + \frac{1}{K} \sum_{i=1}^K \varepsilon_i\right)^2\right) \quad \Leftarrow \text{mean}(\varepsilon) = 0 \\ &= E\left(\left(\frac{1}{K} \sum_{i=1}^K f(z_i) - f(x_0)\right)^2\right) \end{aligned}$$

$$\begin{aligned} \text{(iii)} \quad \text{MSE} &= \text{bias}^2(\hat{m}(x_0)) + \text{var}(\hat{m}(x_0)) + \varepsilon^2 \\ &= \left(f(x_0) - \frac{1}{K} \sum_{i=1}^K f(z_i)\right)^2 + E\left(\left(\frac{1}{K} \sum_{i=1}^K f(z_i) - f(x_0)\right)^2\right) + \varepsilon^2 \end{aligned}$$

when $K=1$ it is low bias and high variance

when $K=\infty$ it is high bias and low variance

The larger the K , the ~~lower~~ ^{higher} the bias
There must be an K that makes the ^{bias and} variance the lowest
and when greater or less than this optimal K , it cause the bias
and variance to become larger.