

§ Improving MC

P1

§1. OMC

⊛ Given random estimator $\hat{\alpha}$ of α ^{unknown}

$$\text{Bias} = E[\hat{\alpha}] - \alpha$$

$$\begin{aligned} \text{MSE} &= E|\hat{\alpha} - \alpha|^2 \\ &= |\text{Bias}|^2 + \text{Var}(\hat{\alpha}) \end{aligned}$$

⊛ The smaller MSE, the better MC.

ex To find $\alpha = E[X]$, ^{Law(X)}
 $10^6 X^{-1} \sim \mu$

OMC

$$\hat{\alpha}_n = \frac{1}{n} \sum_{i=1}^n X_i, \quad X_i \sim \mu, \text{ iid}$$

① $\hat{\alpha}_n$ is unbiased

$$\begin{aligned} \text{② } \text{MSE} &= \text{Var}(\hat{\alpha}_n) = E|\hat{\alpha}_n|^2 - (E\hat{\alpha}_n)^2 \\ &= E|\hat{\alpha}_n|^2 - \alpha^2 \end{aligned}$$

"Smaller 2nd Moment, Better MC"
of $\hat{\alpha}_n$

$$\begin{aligned}
 (3) \quad E(\hat{\alpha}_n)^2 &= E\left(\frac{X_1 + X_2 + \dots + X_n}{n}\right)^2 \\
 &= \frac{1}{n^2} E\left(\sum_{i=1}^n X_i^2\right) = \frac{1}{n} E X^2 + \frac{n-1}{n} \alpha^2
 \end{aligned}$$

"Smaller $E X^2$, the better MC"

§ 2 Control variates

Assumption

(1) want $E[X] = \alpha$?

(2) $E[Y] = 0$, $E[XY]$ is known
 $E[Y^2] > 0$, is known

(3) $1P_X^{-1}, 1P_Y^{-1}$ sampling available

Think

$$\alpha = E[X]$$

$$= E[X - bY] \quad \text{for any constant } b.$$

$$\approx \frac{1}{n} \sum_{i=1}^n (X_i - bY_i) \triangleq \hat{\alpha}_n(b)$$

ex $\forall b \in \mathbb{R}$, $\hat{\alpha}_n(b)$ is unbiased.

we want choose smart b^* s.t.

$$\min_b \mathbb{E}[(X - bY)^2] = \mathbb{E}[(X - b^*Y)^2]$$

i.e.

$$b^* = \arg \min_b \mathbb{E}[(X - bY)^2]$$

Let $L^2(P)$ be the space, defined by

$$L^2(P) = \{X \mid \mathbb{E}[X^2] < \infty\}$$

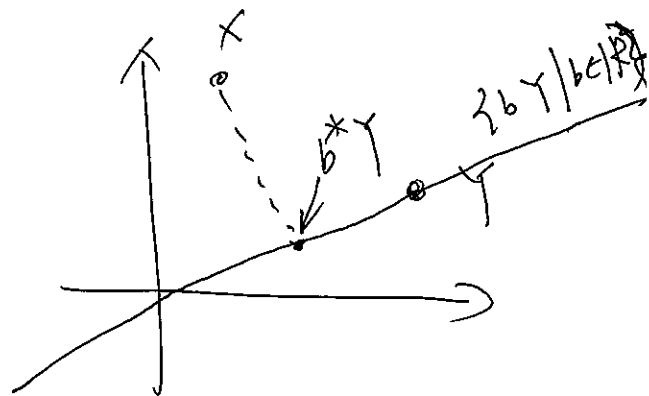
Then $L^2(P)$ is a Hilbert space with

$$\text{inner product } \langle X, Y \rangle \triangleq \mathbb{E}[XY]$$

$$\text{norm } \|X\|^2 \triangleq \langle X, X \rangle = \mathbb{E}[X^2]$$

We shall find b^* s.t.

$$(X - b^*Y) \perp \underbrace{\{bY \mid b \in \mathbb{R}^2\}}_{\text{line}}$$



$$(X - b^*Y) \perp Y$$

$$\mathbb{E}(X - b^*Y) \cdot Y = 0$$

$$\mathbb{E}(XY) - b^* \mathbb{E}Y^2 = 0$$

$$b^* = \frac{\mathbb{E}(XY)}{\mathbb{E}Y^2} = \frac{\langle X, Y \rangle}{\|Y\|^2}$$

Algo

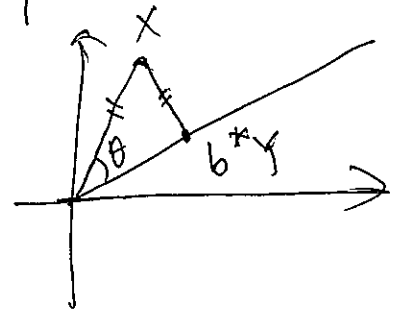
① Compute

$$b^* = \frac{E(XY)}{EY^2}$$

② Compute

$$\hat{X}_n(b^*) = \frac{1}{n} \sum_{i=1}^n (x_i - b^* y_i)$$

Q How much does $\hat{X}_n(b^*)$ improve from $\hat{X}_n(0)$?



A

$$\frac{\|X - b^*Y\|^2}{\|X\|^2} = \sin^2 \theta$$

$$= 1 - \cos^2 \theta$$

$$= 1 - \frac{\langle X, b^*Y \rangle^2}{\|X\|^2 \cdot \|b^*Y\|^2}$$

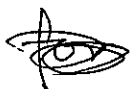
$$= 1 - \left(\frac{\langle X, Y \rangle}{\|X\| \cdot \|Y\|} \right)^2$$

$$= 1 - \rho_{X,Y}^2$$

→ correlation

Rk

"Control variate" improves "CMC"



Rk

The Bigger $|P_{xy}|$, The better improvement by "control variates" from "OMC"

* Extension for $E[Y] = a \neq 0$

we shall use $Y - a \rightarrow Y$.

b/c $E[Y - a] = 0$

So,

$$b^* = \frac{E[X(Y-a)]}{E[(Y-a)^2]} = \frac{E[X(Y-a)]}{E[(Y-a)^2]}$$

$$= \frac{E(XY) - E[X] \cdot E[Y]}{\text{Var}(Y)}$$

$$b^* = \frac{\text{Cov}(X, Y)}{\text{Var}(Y)}$$

Algo (If $\text{Cov}(X, Y)$, $\text{Var}(Y)$ is known)

Then

① Compute b^*

② Sampling (x_i, y_i) ,

③ Compute

$$\hat{\alpha}_n^* = \frac{1}{n} \sum_{i=1}^n (x_i - b^*(y_i - a))$$

Extension

If $\text{cov}(x, y)$, $\text{var}(y)$ is not known, then

use $\text{cov}(x, y) \approx \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})$

$$\text{var}(y) \approx \frac{1}{n-1} \sum_{i=1}^n (y_i - \bar{y})^2$$

where

$$\bar{x} = \sum x_i / n \quad \bar{y} = \sum y_i / n$$

Phj

Compute AAC with GAC
as its control variates.

(Para):

$$S_0 = 100 \quad k = 110 \quad r = 0.0475$$

$$\sigma = 0.2, \quad T = 1, \quad n = 5$$

§2.1 S (Importance Sampling)

17.

Goal

$$\alpha = E[h(X)] \quad \text{where } X \sim f \\ = \int h(x) f(x) dx.$$

ex Let

$$h(x) = 100 \cdot I_{(0, \frac{1}{100})}$$

$$f(x) = I_{(0, 1)}$$

$$\alpha = E[h(X)] = 1$$

① MC ($n=10$)

$$\hat{\alpha}_{10} = \frac{1}{10} \sum_{i=1}^{10} h(X_i), \quad X_i \sim f$$

Q. $\Pr(\hat{\alpha}_{10} = 0) = ?$

A. $\left(\frac{99}{100}\right)^{10} = \text{Big}$

② IS ($n=10$):

$$\text{Let } g(x) = b I_{(0, \frac{1}{100})} + \left(1 - \frac{b}{100}\right) I_{(\frac{1}{100}, 1)} \\ + \frac{100-b}{99} I_{(\frac{1}{100}, 1)}$$

ex verify. g is pdf.

$$\hat{\alpha}_{10} = \frac{1}{10} \sum_{i=1}^{10} h(X_i)$$