Problem

For example, for a parameter θ , you already have got the unbiased estimation $\hat{\theta}$ which means $E(\hat{\theta}) = \theta$. But what if you want to get a point estimation of θ^k ? In general, $\hat{\theta}^k$ is no an unbiased estimation of θ^k . i.e., $E(\hat{\theta}^k) \neq \theta^k$ in general. How to do this?

Method

A natural way to deal with this is to first see the difference between $E(\hat{\theta}^k)$ and θ^k . A (almost only) way is to use taylor expansion.

$$E(\hat{\theta}^{k}) = E[(\hat{\theta} - \theta + \theta)^{k}] \approx E[\theta^{k} + k\theta^{k-1}(\hat{\theta} - \theta) + \frac{1}{2}k(k-1)\theta^{k-2}(\hat{\theta} - \theta)^{2}]$$

$$= E[\theta^{k}] + k\theta^{k-1}E(\hat{\theta} - \theta) + \frac{1}{2}k(k-1)\theta^{k-2}E[(\hat{\theta} - \theta)^{2}]$$

$$= \theta^{k} + 0 + \frac{1}{2}k(k-1)\theta^{k-2}E[(\hat{\theta} - \theta)^{2}]$$

$$= \theta^{k} + \frac{1}{2}k(k-1)\theta^{k-2}E[\hat{\theta}^{2} - 2\hat{\theta}\theta + \theta^{2}] = \theta^{k} + \frac{1}{2}k(k-1)\theta^{k-2}var(\hat{\theta})$$

This means that, (approximately) , the point estimator $\hat{\theta}^k - \frac{1}{2}k(k-1) \, \theta^{k-2} \, var(\hat{\theta})$ can approximately provide a unbiased estimation of θ^k , that is (just rearrange the above results) $\theta^k = E[\hat{\theta}^k] - \frac{1}{2}k(k-1) \, \theta^{k-2} \, var(\hat{\theta})$. But, what is θ^{k-2} ? naturally, we need to do this recursively.

Procedure

We already know that

$$\theta = E(\hat{\theta}) \text{ (1)}$$

$$\theta^{2} = E(\hat{\theta})^{2} = E(\hat{\theta}^{2}) - var(\hat{\theta}) \text{ (2)}$$

$$\theta^{3} = E(\hat{\theta}^{3}) - \frac{1}{2}3(3-1) \theta var(\hat{\theta}) \text{ (3)}$$

plug in the expression of θ in (1)

$$\theta^4 = E(\hat{\theta}^4) - \frac{1}{2}4(4-1) \theta^2 var(\hat{\theta})$$
 (4)

Plut in the expression of $\,\theta^2\,$ in (2)....

Doing this recursively, we can finally get a corrected estimation of θ^k