

# 1

(a) Find the  $\mathbb{E}[\bar{X}(N)]$  analytically. Does it really estimate the mean of  $X$ ?

We have that

$$\bar{X}(N) = \frac{1}{N} \sum_{i=1}^N X_i, \quad X_i \sim \text{Uniform}(0, 1).$$

We first compute the expectation of  $X_i$ . We have

$$\begin{aligned} \mathbb{E}[X_i] &= \int_{-\infty}^{\infty} x \cdot p(X=x) \, dx \\ &= \int_0^1 x \cdot p(X=x) \, dx \\ &= \int_0^1 x \cdot \frac{1}{1-0} \, dx \\ &= \int_0^1 x \, dx \\ &= \left[ \frac{1}{2}x^2 \right]_0^1 \\ &= \frac{1}{2}(1^2 - 0^2) \\ &= \frac{1}{2} \end{aligned} \tag{1}$$

Therefore, we can compute the expectation of  $\bar{X}(N)$  as follows:

$$\begin{aligned} \mathbb{E}[\bar{X}(N)] &= \mathbb{E}\left[\frac{1}{n} \sum_{i=1}^n X_i\right] \\ &= \frac{1}{n} \sum_{i=1}^n \mathbb{E}[X_i] \\ &\stackrel{(1)}{=} \frac{1}{n} \sum_{i=1}^n \frac{1}{2} \\ &= \frac{1}{n} \cdot n \cdot \frac{1}{2} \\ &= \frac{1}{2} \end{aligned}$$

Therefore, the expected value of  $\mathbb{E}[\bar{X}(N)]$  estimates the true mean of  $X$ .

(b) Find the  $\text{Var}(\bar{X}(N))$  analytically. Hint: find the variance of a single  $X_i$  first.

Let us first compute the expectation of  $\mathbb{E}[X_i^2]$ :

$$\begin{aligned}
 \mathbb{E}[X_i^2] &= \int_0^1 x^2 \cdot p(X = x) \, dx \\
 &= \int_0^1 x^2 \cdot \frac{1}{1-0} \, dx \\
 &= \int_0^1 x^2 \, dx \\
 &= \left[ \frac{1}{3} \cdot x^3 \right]_0^1 \\
 &= \frac{1}{3}(1 - 0) \\
 &= \frac{1}{3}
 \end{aligned} \tag{2}$$

We can now compute the variance of a  $X_i$  as follows:

$$\begin{aligned}
 \text{Var}(X_i) &= \mathbb{E}[(X_i - \mathbb{E}[X_i])^2] \\
 &= \mathbb{E}\left[(X_i - \frac{1}{2})^2\right] \\
 &= \mathbb{E}\left[X_i^2 + \frac{1^2}{2^2} - 2 \cdot X_i \cdot \frac{1}{2}\right] \\
 &= \mathbb{E}\left[X_i^2 + \frac{1^2}{2^2} - X_i\right] \\
 &= \mathbb{E}[X_i^2] + \mathbb{E}\left[\frac{1}{4}\right] - \mathbb{E}[X_i] \\
 &= \mathbb{E}[X_i^2] + \frac{1}{4} - \frac{1}{2} \\
 &\stackrel{(2)}{=} \frac{1}{3} + \frac{1}{4} - \frac{1}{2} \\
 &= \frac{4}{12} + \frac{3}{12} - \frac{6}{12} \\
 &= \frac{1}{12}.
 \end{aligned}$$

We can now compute the variance of  $\bar{X}(N)$  as follows:

$$\begin{aligned}
 \text{Var}(\bar{X}(N)) &= \text{Var}\left(\frac{1}{N} \sum_{i=1}^N X_i\right) \\
 &= \frac{1}{N^2} \cdot \text{Var}\left(\sum_{i=1}^N X_i\right) \\
 &= \frac{1}{N^2} \cdot \sum_{i=1}^N \text{Var}(X_i) \\
 &= \frac{1}{N^2} \cdot \sum_{i=1}^N \frac{1}{12} \\
 &= \frac{1}{N^2} \cdot N \cdot \frac{1}{12} \\
 &= \frac{1}{N} \cdot \frac{1}{12}
 \end{aligned} \tag{3}$$

where Equation 3 holds as each drawing  $X_i$  is independent of each other drawing  $X_j$  ( $\Pr(X_i | X_j) = \Pr(X_i)$ ), and we can therefore use the property of the variance for i.i.d. variables  $\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y)$ .

Therefore, in the limit, we have

$$\lim_{n \rightarrow \infty} \text{Var}(\bar{X}(N)) = \lim_{n \rightarrow \infty} \frac{1}{N} \cdot \frac{1}{12} = 0$$

with a linear convergence rate.

**(c)** The probability density function of  $\hat{X}(N)$  is given as

$$f_{\hat{X}(N)}(x) = \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot x^{\frac{N-1}{2}} \cdot (1-x)^{\frac{N-1}{2}} .$$

Let us now compute the expectation of  $\hat{X}(N)$ :

$$\begin{aligned}
 \mathbb{E}[\hat{X}(N)] &= \int_0^1 x \cdot f_{\hat{X}(N)}(x) \, dx \\
 &= \int_0^1 x \cdot \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot x^{\frac{N-1}{2}} \cdot (1-x)^{\frac{N-1}{2}} \, dx \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \int_0^1 x \cdot x^{\frac{N-1}{2}} \cdot (1-x)^{\frac{N-1}{2}} \, dx \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \int_0^1 x^{\frac{N-1}{2}+1} \cdot (1-x)^{\frac{N-1}{2}} \, dx \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \int_0^1 x^{\frac{N+1}{2}} \cdot (1-x)^{\frac{N-1}{2}} \, dx \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \frac{\frac{N+1}{2}! \frac{N-1}{2}!}{(\frac{N+1}{2} + \frac{N-1}{2} + 1)!} \tag{4}
 \end{aligned}$$

$$\begin{aligned}
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \frac{\frac{N+1}{2}! \frac{N-1}{2}!}{(\frac{N-1}{2} + \frac{N-1}{2} + 1 + 1)!} \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \frac{\frac{N+1}{2}! \frac{N-1}{2}!}{(2 \cdot \frac{N-1}{2} + 2)!} \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \frac{\frac{N+1}{2}! \frac{N-1}{2}!}{(N-1+2)!} \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \frac{\frac{N+1}{2}! \frac{N-1}{2}!}{(N+1)!} \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{1}{2} \cdot \frac{(\frac{N+1}{2})! \frac{N-1}{2}!}{N!} \\
 &= \frac{N!}{(\frac{N-1}{2})! (N - \frac{N-1}{2})!} \cdot \frac{1}{2} \cdot \frac{(\frac{N+1}{2})! \frac{N-1}{2}!}{N!} \\
 &= \frac{1}{(\frac{N-1}{2})! (N - \frac{N-1}{2})!} \cdot \frac{1}{2} \cdot \frac{(\frac{N+1}{2})! \frac{N-1}{2}!}{1} \\
 &= \frac{1}{(N - \frac{N-1}{2})!} \cdot \frac{1}{2} \cdot \frac{(\frac{N+1}{2})!}{1} \\
 &= \frac{1}{(\frac{N+1}{2})!} \cdot \frac{1}{2} \cdot \frac{(\frac{N+1}{2})!}{1} \tag{5} \\
 &= \frac{1}{2}
 \end{aligned}$$

where we used the given known integral rule to obtain Equation 4, and Equation 5 results from

$$\begin{aligned}
 N - \frac{N-1}{2} &= \frac{2N}{2} - \frac{N-1}{2} \\
 &= \frac{N+N-(N-1)}{2} \\
 &= \frac{N+1}{2}.
 \end{aligned}$$

Therefore, the expectation of the of  $\hat{X}(N)$  is  $\frac{1}{2}$ , i.e., the true mean of  $X$ .

(d) Find the  $\text{Var}(\hat{X}(N))$  analytically. Which distribution has the better variance  $\bar{X}(N)$  or  $\hat{X}(N)$ ?

We compute the expectation of  $\hat{X}(N)^2$ :

$$\begin{aligned}
 \mathbb{E}[\hat{X}(N)^2] &= \int_0^1 x^2 \cdot f_{\hat{X}(N)}(x) \, dx \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \int_0^1 x^2 \cdot x^{\frac{N-1}{2}} \cdot (1-x)^{\frac{N-1}{2}} \, dx \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \int_0^1 x^{\frac{N-1}{2}+2} \cdot (1-x)^{\frac{N-1}{2}} \, dx \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \int_0^1 x^{\frac{N+3}{2}} \cdot (1-x)^{\frac{N-1}{2}} \, dx \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \frac{\frac{N+3}{2}! \frac{N-1}{2}!}{(\frac{N+3}{2} + \frac{N-1}{2} + 1)!} \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \frac{\frac{N+3}{2}! \frac{N-1}{2}!}{(\frac{N-1}{2} + \frac{N-1}{2} + 1 + 2)!} \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \frac{\frac{N+3}{2}! \frac{N-1}{2}!}{(2 \cdot \frac{N-1}{2} + 3)!} \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \frac{\frac{N+3}{2}! \frac{N-1}{2}!}{(N-1+3)!} \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \frac{\frac{N+3}{2}! \frac{N-1}{2}!}{(N+2)!} \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{N+1}{2} \cdot \frac{\frac{N+3}{2}! \frac{N-1}{2}!}{(N+2) \cdot (N+1)!} \\
 &= \binom{N}{\frac{N-1}{2}} \cdot \frac{1}{2} \cdot \frac{\frac{N+3}{2}! \frac{N-1}{2}!}{(N+2) \cdot N!} \\
 &= \frac{N!}{(\frac{N-1}{2})! (N - \frac{N-1}{2})!} \cdot \frac{1}{2} \cdot \frac{\frac{N+3}{2}! \frac{N-1}{2}!}{(N+2) \cdot N!} \\
 &= \frac{1}{(\frac{N-1}{2})! (N - \frac{N-1}{2})!} \cdot \frac{1}{2} \cdot \frac{\frac{N+3}{2}! \frac{N-1}{2}!}{(N+2)} \\
 &= \frac{1}{(N - \frac{N-1}{2})!} \cdot \frac{1}{2} \cdot \frac{\frac{N+3}{2}!}{(N+2)} \\
 &= \frac{1}{\frac{N+1}{2}!} \cdot \frac{1}{2} \cdot \frac{\frac{N+3}{2}!}{(N+2)} \\
 &= \frac{1}{\frac{N+1}{2}!} \cdot \frac{1}{2} \cdot \frac{\frac{N+3}{2} \cdot (\frac{N+1}{2})!}{(N+2)} \\
 &= \frac{1}{2} \cdot \frac{\frac{N+3}{2}}{(N+2)} \\
 &= \frac{1}{2} \cdot \frac{\frac{N+3}{2} \cdot 2}{(N+2) \cdot 2} \\
 &= \frac{1}{2} \cdot \frac{N+3}{(N+2) \cdot 2} \\
 &= \frac{1}{2} \cdot \frac{N+2+1}{(N+2) \cdot 2} \\
 &= \frac{1}{2} \cdot \left( \frac{N+2}{(N+2) \cdot 2} + \frac{1}{(N+2) \cdot 2} \right)
 \end{aligned}$$

$$\begin{aligned} &= \frac{1}{2} \cdot \left( \frac{1}{2} + \frac{1}{2N+4} \right) \\ &= \frac{1}{4} + \frac{1}{4N+8} \end{aligned}$$

We can finally compute the variance of  $\hat{X}(N)$  as follows:

$$\begin{aligned} \text{Var}(\hat{X}(N)) &= \mathbb{E}[\hat{X}(N)^2] - \mathbb{E}[\hat{X}(N)]^2 \\ &= \mathbb{E}[\hat{X}(N)^2] - \left( \frac{1}{2} \right)^2 \\ &= \mathbb{E}[\hat{X}(N)^2] - \frac{1}{4} \\ &= \frac{1}{4} + \frac{1}{4N+8} - \frac{1}{4} \\ &= \frac{1}{4N+8} \end{aligned}$$

which, in the limit, converges to zero:

$$\lim_{N \rightarrow \infty} \text{Var}(\hat{X}(N)) = \lim_{N \rightarrow \infty} \frac{1}{4N+8} = 0$$

As the convergence rate of  $\text{Var}(\hat{X}(N))$  is faster than  $\text{Var}(\bar{X}(N))$ , and that both have the same (correct) expectation, we can conclude that  $\hat{X}(N)$  has the better variance.