Lecture 9: Monte Carlo estimates of various statistics

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Estimating the probability density function via histograms



- Let X be a random variable and Y = g(X).
- We want to approximate the PDF p(y) of Y.



- Take \underline{M} small bins $[b_0, b_1], \ldots, [b_{M-1}, b_M]$ in the \underline{y} space.
- We will approximate p(y) with a constant inside each bin:

$$\hat{p}_{M}(y) = \sum_{j=1}^{M} \hat{c}_{j} 1_{[b_{j-1},b_{j}]}(y),$$

$$precewise approximation$$

• Where the c_j 's are coefficients to be determined.



• We will approximate p(y) with a constant inside each bin:

$$\hat{p}_{M}(y) = \sum_{j=1}^{M} c_{j} 1_{[b_{j-1},b_{j}]}(y),$$

• The constants are just:

$$c_{j} = p(b_{j-1} \leq Y \leq b_{j}) = F(b_{j}) - F(b_{j-1}) \qquad \text{know how to}$$

$$= F_{N}(b_{j}) - F_{N}(b_{j-1}) \qquad \text{find } F_{N}(y)$$



 So, we can approximate the constants with the empirical CDF:

$$\bar{c}_{j,N} := \bar{F}_N(b_j) - \bar{F}_N(b_{j-1}) \to c_j \text{ a.s.}$$

The intuitive meaning of this is:

$$\overline{c_{j,N}} = \frac{\text{number of samples that fall in bin } [b_{j-1}, b_j]}{N}$$



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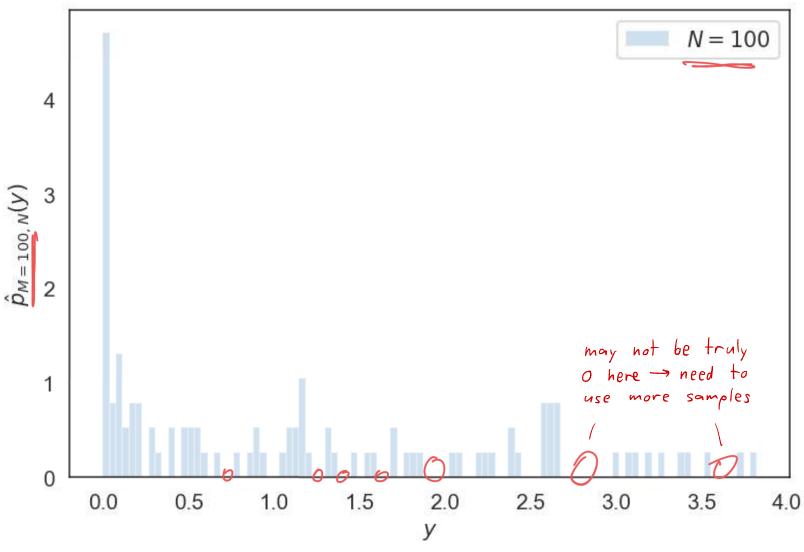
$$\bar{c}_{j,N} = \frac{\text{number of samples that fall in bin } [b_{j-1},b_j]}{N}$$

• Putting everything together our approximation becomes:

$$\hat{p}_{M,N}(y) = \sum_{j=1}^{M} \bar{c}_{j,N} \mathbf{1}_{[b_{j-1},b_j]}(y)$$
samples

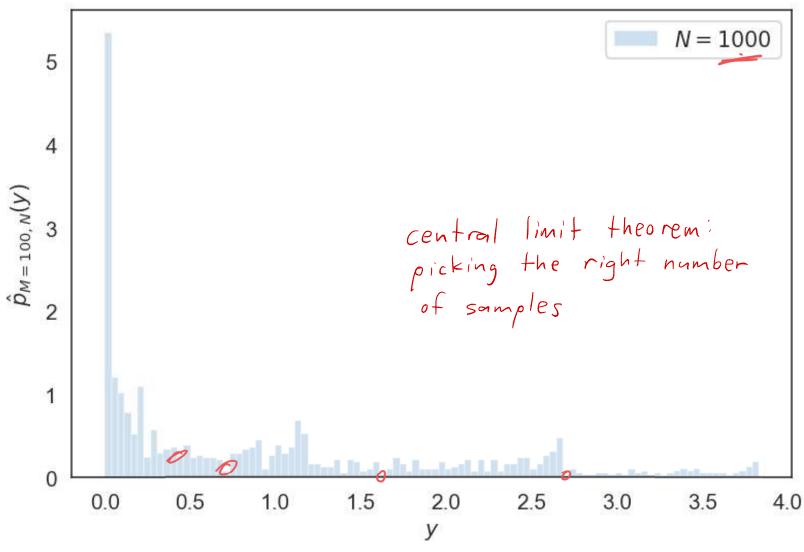


Example: 1D PDF





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