

## Central Limit Theorem

Suppose  $X_1, X_2, \dots$  is an infinite sequence of independent random variables (not necessarily Normal) that each have mean  $\mu$  and variance  $\sigma^2$ . Fix "a" then the probability that the sum of the first  $n$  of the random variables, suitably scaled and shifted, is less than  $a$  will go to the same probability that a standard Normal random variable is less than  $a$  (as  $n \rightarrow \infty$ ).

$$\lim P\left(\frac{X_1 + \dots + X_n - n\mu}{\sqrt{n\sigma^2}} \leq a\right) = \int_{-\infty}^a \frac{1}{\sqrt{2\pi}} e^{-z^2/2} dz = F_z(a)$$

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The Central Limit Theorem is a limiting result. In practice, we often apply it to finite but large collections of independent random variables, to model their sum by a standard Normal random variable.

Two caveats: The CLT works better and better, the larger  $n$  is, and there is no cutoff for which it works completely above some particular  $n$ .

We must be sure to get the scaling and shifting correct, if we are going to apply the CLT properly.