

Statistical modelling

#1.b Central limit theorem

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Null distribution

When we perform an hypothesis test, we need to know the behaviour of the statistic under the null hypothesis in order to draw a conclusion (reject/fail to reject \mathcal{H}_0)

The test statistic is often

✚ a Wald statistic (mean or maximum likelihood estimator)

Under regularity conditions and for n sufficiently large, the null distribution is approximately normal. Why?

Central Limit Theorem (informal)

Let Y_1, \dots, Y_n be a random sample from a distribution with

- + expectation μ ,
- + (finite) variance σ^2 .

If n is large, the mean \bar{Y}_n approximately follows a normal distribution with mean μ and variance σ^2/n .

$$\bar{Y}_n \dot{\sim} \text{No}(\mu, \sigma^2/n)$$

Central Limit Theorem (formal)

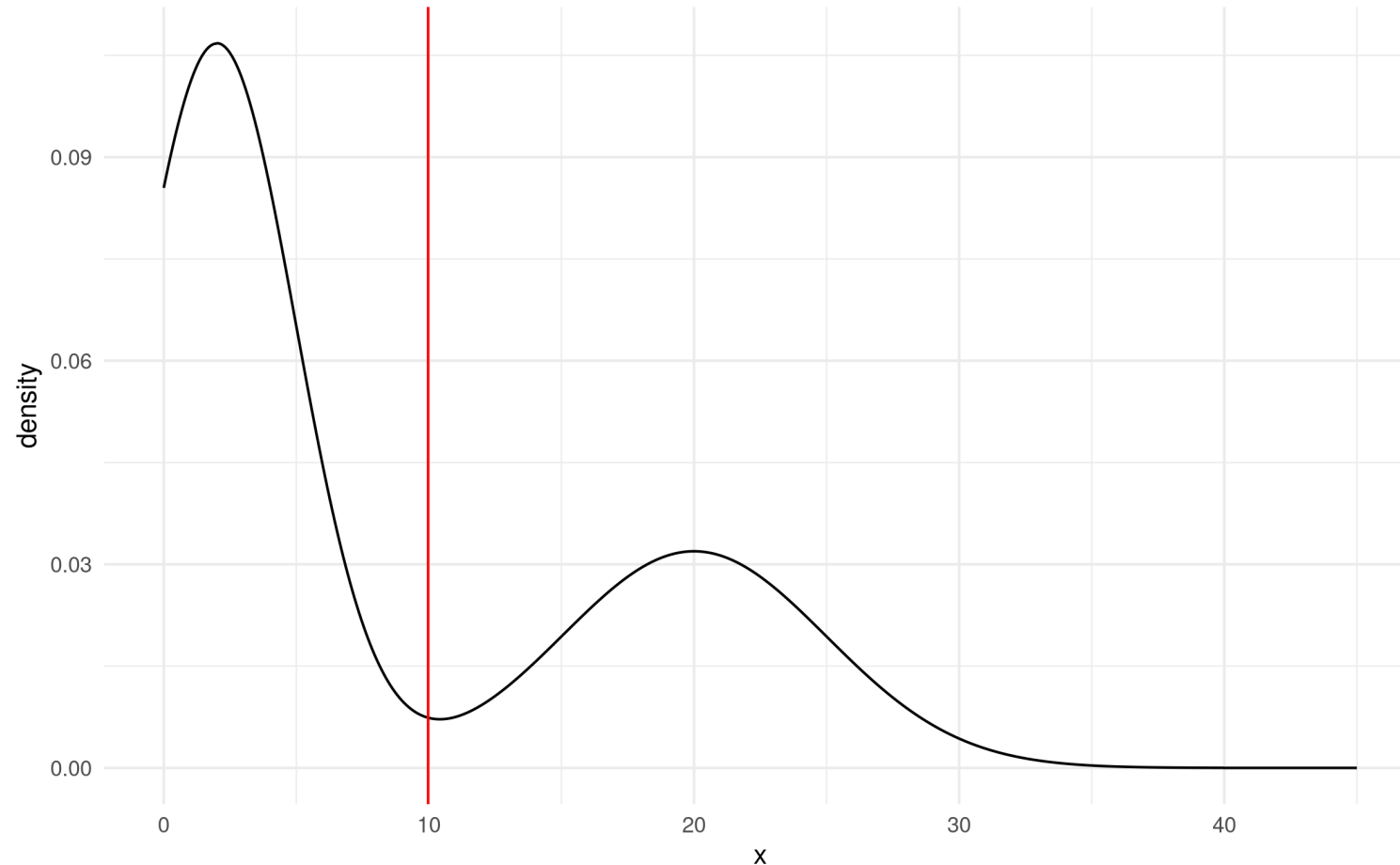
Let Y_1, \dots, Y_n be independent and identically distributed random variables with distribution F and finite variance and let $\bar{Y}_n = n^{-1} \sum_{i=1}^n Y_i$ denote the sample mean.

For any $y \in \mathbb{R}$, the mean converges in distribution to a normal distribution, i.e.,

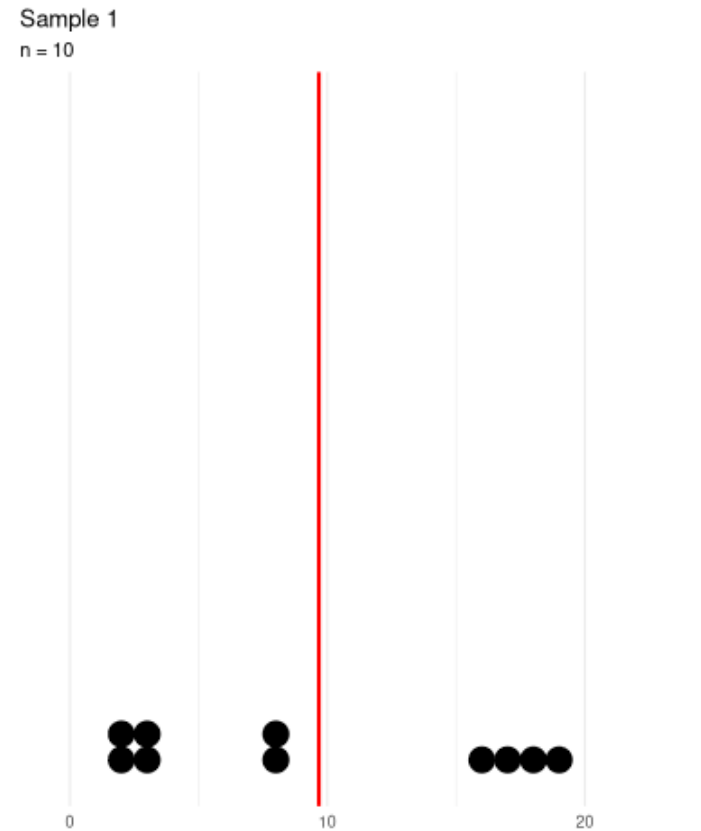
$$\lim_{n \rightarrow \infty} \mathbf{P} \left(\sqrt{n} \frac{\bar{Y}_n - \mu}{\sigma} \leq y \right) = \Phi(y)$$

where $\Phi(y)$ is the distribution function of $\mathbf{No}(0, 1)$.

Let's represent graphically the central limit theorem by drawing samples repeatedly from the following distribution (left truncated, multimodal, etc.)

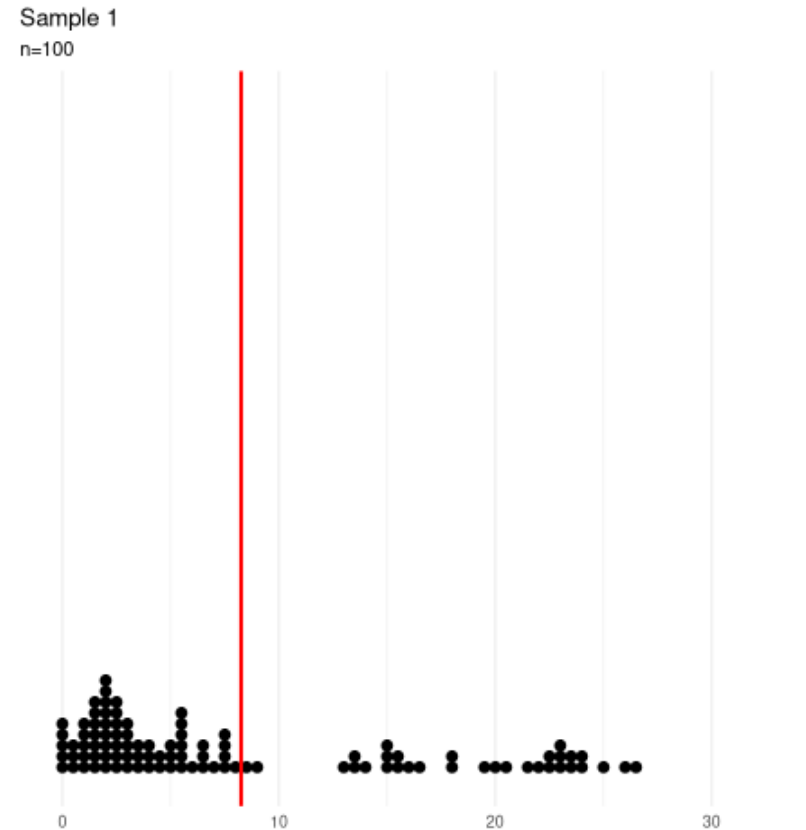


Let's draw **20** random samples of size $n = 10$.



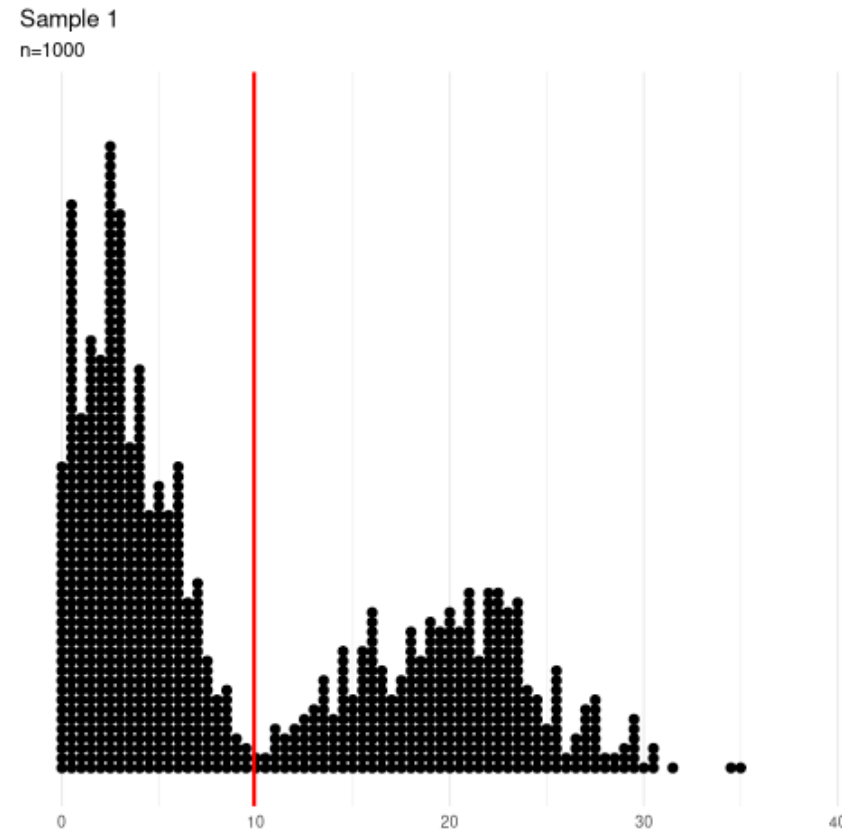
Dot plot of random sample of size $n = 10$ and sample mean (vertical red line)

If we increase the sample size to $n = 100$, the variability of the sample mean decreases.



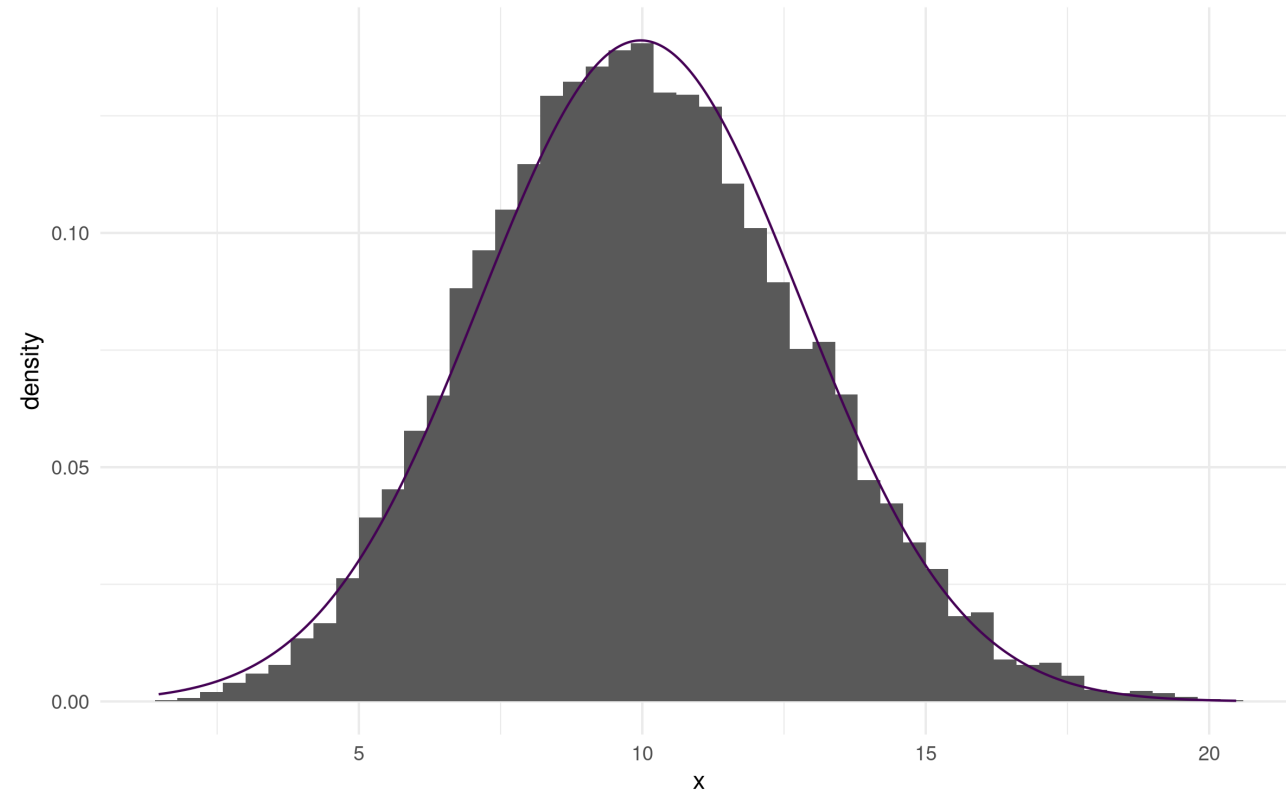
Dot plot of random sample of size $n = 100$ and sample mean (vertical red line)

Same thing, this time with $n = 1000$ observations per sample.



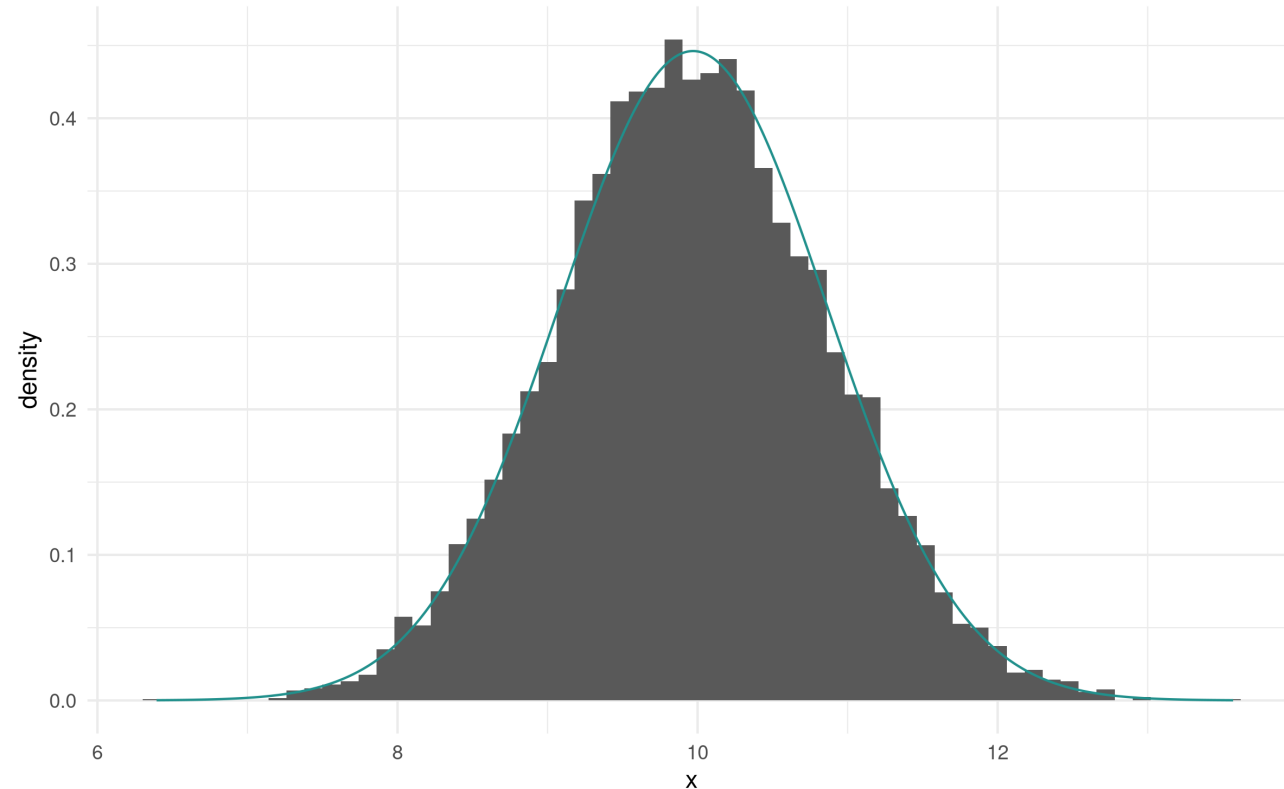
Dot plot of random sample of size $n = 1000$ and sample mean (vertical red line)

If we draw an histogram of the means (vertical red lines), what do we obtain?



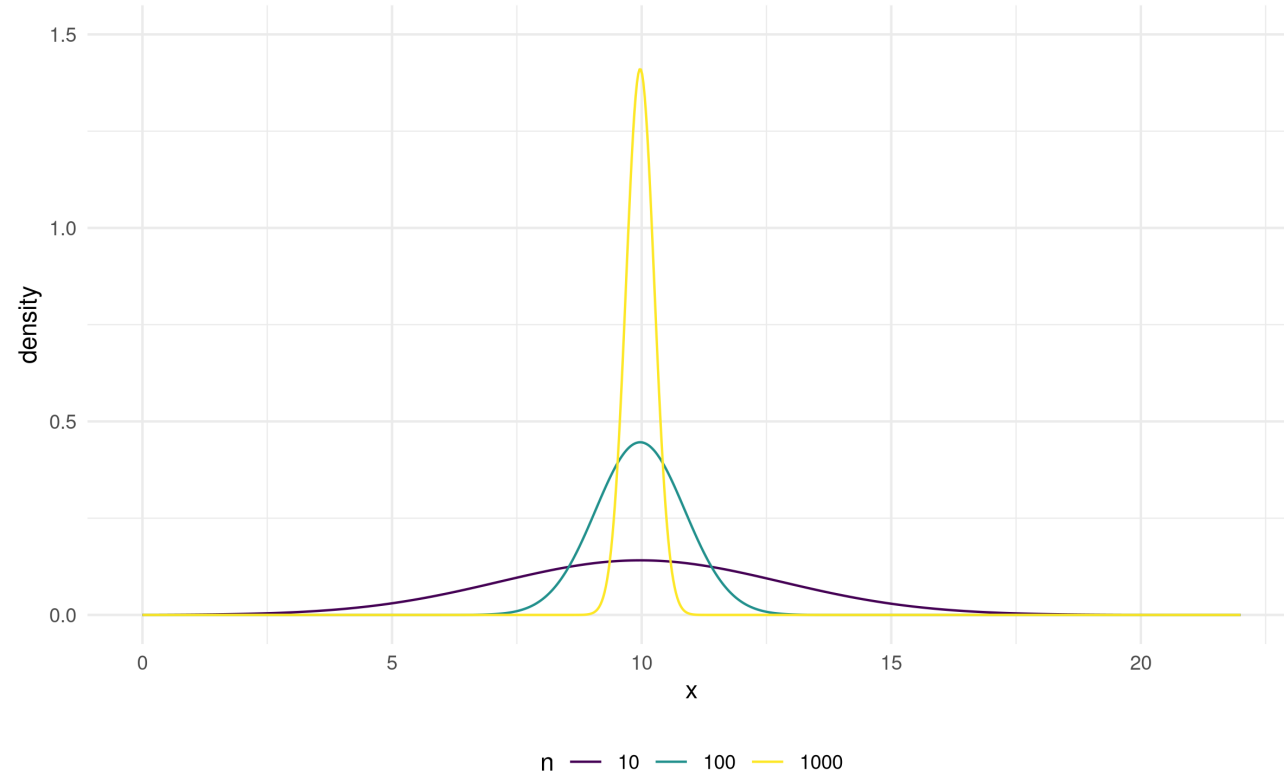
Histogram of the empirical distribution of sample means of $n = 10$ observations and CLT normal approximation.

The quality of the CLT approximation improves when the sample size n increases



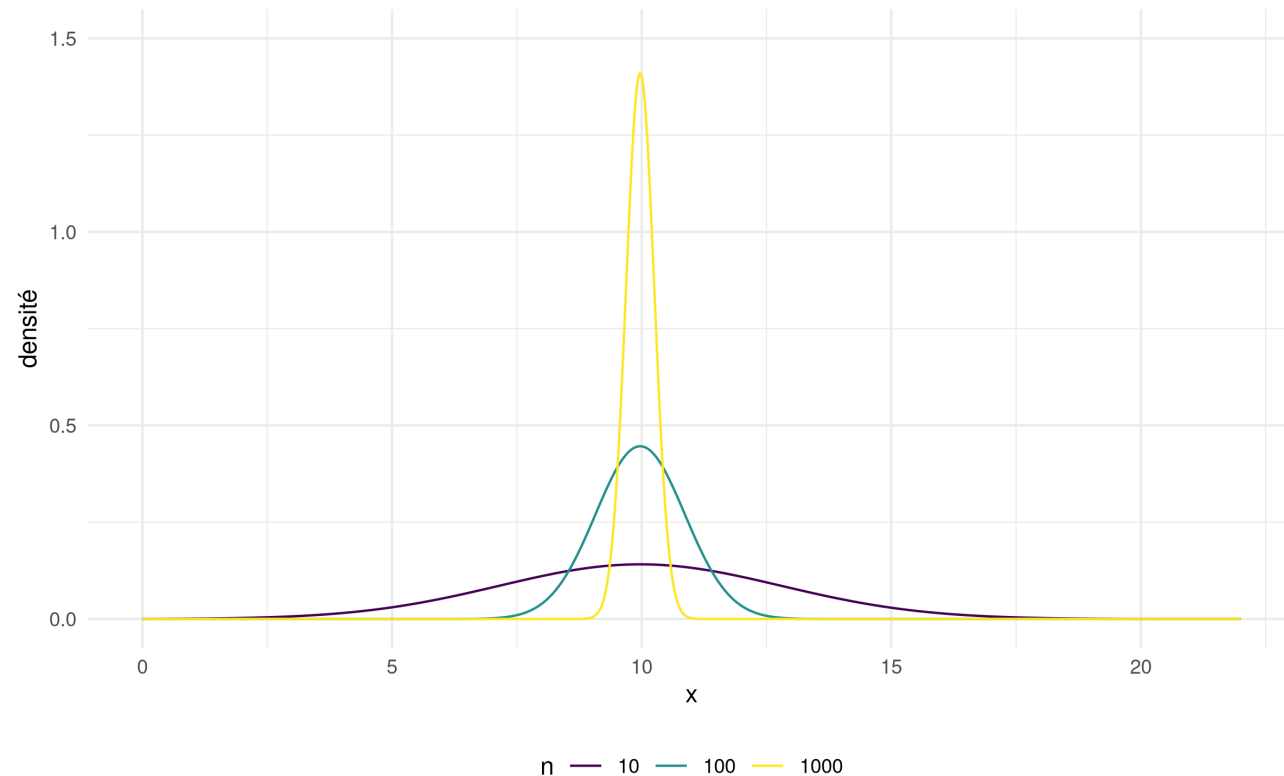
Histogram of the empirical distribution of sample means of $n = 100$ observations and CLT normal approximation.

Convergence is faster near the mean than in the tails of the distribution.



Histogram of the empirical distribution of sample means of $n = 1000$ observations and CLT normal approximation. observations.

The variance of the sample mean \bar{Y}_n when $\text{Va}(Y_i) = \sigma^2 (i = 1, \dots, n)$ is roughly σ^2/n .



Normal approximation of the mean for different sample sizes.