

GAUSSIAN (NORMAL) DISTRIBUTION

$$X \sim \mathcal{N}(\mu, \sigma^2)$$

$$f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-(x-\mu)^2/(2\sigma^2)} \quad E[X] = \mu$$

$$\text{Var}(X) = \sigma^2$$

the standard normal is denoted by $\mathcal{N}(0, 1)$

↳ any Gaussian RV can be expressed as

$$X = \sigma Z + \mu, \quad Z \sim \mathcal{N}(0, 1)$$

a linear combination of independent Gaussians:

$$X \sim \mathcal{N}(\mu_X, \sigma_X^2), \quad Y \sim \mathcal{N}(\mu_Y, \sigma_Y^2), \quad Z = aX + bY$$

$$Z \sim \mathcal{N}(a\mu_X + b\mu_Y, a^2\sigma_X^2 + b^2\sigma_Y^2)$$

CENTRAL LIMIT THEOREM

for i.i.d. RVs x_1, x_2, \dots, x_n with $E(x_i) = \mu$, $\text{var}(x_i) = \sigma^2$,

define $s_n = \sum_{i=1}^n x_i$. then,

$$\frac{s_n - n\mu}{\sigma\sqrt{n}} \rightarrow \mathcal{N}(0,1) \quad \text{as } n \rightarrow \infty$$

equivalently, $\frac{s_n}{n} \rightarrow \mathcal{N}(\mu, \frac{\sigma^2}{n})$ as $n \rightarrow \infty$