

The probability distribution of the sum/average of three random variables is the linear convolution of the individual probability distribution of these three random variables.

Regardless of the shape of the individual distributions, please perform simply a "chain" of linear convolutions to illustrate the Gaussianity claim made by the central limit theorem.

Random variables : $(Y = X_1 + X_2 + \dots + X_n)$

using Y_1 , Y_2 , and Y_3 as our RV's

$$Y_1 + Y_2 + Y_3 = \mathcal{Z} f_{Y_1}(y) * \mathcal{Z} f_{Y_2}(y) * \mathcal{Z} f_{Y_3}(y)$$

$$= \mathcal{Z} (f_{Y_1}(y) * f_{Y_2}(y) * f_{Y_3}(y))$$

$$= \mathcal{Z} \left[\underbrace{\left(\int_{-\infty}^{\infty} f_{Y_1}(y) f_{Y_2}(x-y) dy \right)}_{f_A(x)} * f_{Y_3}(y) \right]$$

~~$$= \mathcal{Z} \left[\int_{-\infty}^{\infty} f_A(x) f_{Y_3}(x-y) dy \right]$$~~

$$f_2(z) = \int_{-\infty}^{\infty} f_X(z-y) f_Y(y) dy$$

$$= 3 \exp(-z)$$

~~$$= \mathcal{Z} [x \exp(-x) * f_{Y_3}(y)]$$~~

$$= \mathcal{Z} \left[\int_{-\infty}^{\infty} x \exp(-x) f_{Y_3}(x-y) dy \right]$$

$$= \mathcal{Z} [x^2 \exp(-x)]$$

$$= \mathcal{Z} [x^2 e^{-x}]$$

Central Limit Thm

$$\mu_{\bar{x}} = \mu$$

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

$$Z = \frac{\bar{x} - \mu_x}{\sigma_{\bar{x}}}$$

$$= \frac{\bar{x} - \mu}{\frac{\sigma}{\sqrt{n}}}$$