

## Central Limit Theorem

Let  $X_1, \dots, X_n$  be a sequence of mutually independent and identically distributed random variables with means  $\eta$  and variances  $\sigma^2$ . Let

$$X = \sum_{j=1}^n X_j$$

Define a normalized random variable

$$Y = \frac{X - n\eta}{\sigma n^{1/2}}$$

Then the distribution function of  $Y$  converges to a zero mean, unit variance Gaussian distribution as  $n \rightarrow \infty$ .

### Proof

$$\Phi_Y(\omega) = E\{e^{i\omega Y}\} = E\{e^{i\Omega(X - n\eta)}\} = E\left\{e^{i\Omega \sum_{j=1}^n (X_j - \eta)}\right\} = E\left\{\prod_{j=1}^n e^{i\Omega(X_j - \eta)}\right\},$$

where

$$\Omega = \frac{\omega}{\sigma n^{1/2}}.$$

Noting that  $X_1, \dots, X_n$  are independent random variables, it follows that

$$\begin{aligned} \Phi_Y(\omega) &= \prod_{j=1}^n E\{e^{i\Omega(X_j - \eta)}\} = \prod_{j=1}^n \left( e^{-i\Omega\eta} E\{e^{i\Omega X_j}\} \right) \\ &= \prod_{j=1}^n \left( e^{-i\Omega\eta} \Phi_{X_j}(\Omega) \right) = \left( e^{-i\Omega\eta} \Phi_{X_j}(\Omega) \right)^n. \end{aligned}$$

Note that the characteristic functions are identical. that is.

$$\Phi_{X_j}(\Omega) = \Phi_{X_1}(\Omega) = \dots = \Phi(\Omega).$$

As  $n \rightarrow \infty$ ,  $\Omega \rightarrow 0$ , it follows that

$$\Phi(\Omega) = \sum_k m_k \frac{(i\Omega)^k}{k!}, \quad m_k = E\{X^k\},$$

and

$$\begin{aligned}\Phi_Y(\omega) &= \left[ \left( 1 - i\Omega\eta - \frac{\Omega^2\eta^2}{2} \dots \right) \left( 1 + i\Omega\eta - \frac{\sigma^2 + \eta^2}{2} \Omega^2 \dots \right) \right]^n \\ &= \left[ 1 + \Omega^2\eta^2 - \frac{\Omega^2\eta^2}{2} - \frac{\sigma^2 + \eta^2}{2} \Omega^2 \dots \right]^n \\ &= \left( 1 - \frac{\sigma^2\eta^2}{2} \dots \right)^n\end{aligned}$$

Noting

$$\lim_{N \rightarrow \infty} \left( 1 + \frac{c}{N} \right)^N \rightarrow e^c,$$

it then follows that

$$\Phi_Y(\omega) = \left( 1 - \frac{\omega^2}{2n} \right)^n \rightarrow e^{-\omega^2/2}$$

which is the characteristic function of a zero mean, unit variance normal random variable.