

# MATH2710 — Proof Presentation

Mike Medved

April 12th, 2023

## Introduction

In this report, we will prove that the sum of two independent normal random variables is also a normal random variable, such that:

$$\begin{aligned} X &\sim \mathcal{N}(\mu_X, \sigma_X^2) & Y &\sim \mathcal{N}(\mu_Y, \sigma_Y^2) \\ X + Y &= Z \sim \mathcal{N}(\mu_X + \mu_Y, \sigma_X^2 + \sigma_Y^2) \end{aligned}$$

## Preliminaries

### The Moment Generating Function

Let  $X$  be a random variable with distribution function  $F_X(x)$ . The moment generating function of  $X$  is defined as follows:

$$\phi_X(t) = \mathbb{E} [e^{tX}]$$

**Property:** When added, the moment generating functions of two independent random variables  $X$  and  $Y$  is equal to the product of their individual moment generating functions:

$$\phi_{X+Y}(t) = \phi_X(t) \cdot \phi_Y(t)$$

Since  $X$  and  $Y$  are independent with respect to one-another, we have the right to multiply their moment generating functions.

$$\begin{aligned} &\mathbb{E} [e^{tX}] \cdot \mathbb{E} [e^{tY}] \\ &= \mathbb{E} \left( e^{t(X+Y)} \right) \end{aligned}$$

## Proof

**Theorem.** Let  $X$  and  $Y$  be independent normal random variables with mean  $\mu$  and variance  $\sigma^2$ . Prove that the random variable  $X + Y = Z$  is a normal random variable  $Z \sim \mathcal{N}(\mu_X + \mu_Y, \sigma_X^2 + \sigma_Y^2)$ .

*Proof.* Let  $X, Y$  be two independent normal random variables such that:

$$X \sim \mathcal{N}(\mu_X, \sigma_X^2) \quad Y \sim \mathcal{N}(\mu_Y, \sigma_Y^2)$$

We aim to prove that the random variable  $Z = X + Y$  is also a normal random variable with mean  $\mu_X + \mu_Y$  and variance  $\sigma_X^2 + \sigma_Y^2$ .

In this case, we know that  $X, Y$  are both independent of one another, and are normal random variables. Therefore, we can use the sum of their moment generating functions to prove that  $Z$  is also a normal random variable.

Let  $t \in \mathbb{R}$ , then the moment generating function of  $X \sim \mathcal{N}(\mu_X, \sigma_X^2)$  is the following:

$$\phi_X(t) = \mathbb{E}[e^{tX}] = \exp\left(t\mu_X - \frac{\sigma_X^2 t^2}{2}\right)$$

Similarly, the moment generating function of  $Y \sim \mathcal{N}(\mu_Y, \sigma_Y^2)$  is the following:

$$\phi_Y(t) = \mathbb{E}[e^{tY}] = \exp\left(t\mu_Y - \frac{\sigma_Y^2 t^2}{2}\right)$$

Therefore, the moment generating function of  $X + Y$  will be:

$$\begin{aligned} \phi_{X+Y}(t) &= \phi_X(t) \cdot \phi_Y(t) \\ &= \exp\left(t\mu_X - \frac{\sigma_X^2 t^2}{2}\right) \cdot \exp\left(t\mu_Y - \frac{\sigma_Y^2 t^2}{2}\right) \\ &= \exp\left(t(\mu_X + \mu_Y) - \frac{(\sigma_X^2 + \sigma_Y^2) \cdot t^2}{2}\right) \end{aligned}$$

This is exactly the moment generating of a normal random variable  $\mathcal{N}(\mu_X + \mu_Y, \sigma_X^2 + \sigma_Y^2)$ , which is what we aimed to prove.

Hence,  $X + Y$  is a normal random variable  $Z \sim \mathcal{N}(\mu_X + \mu_Y, \sigma_X^2 + \sigma_Y^2)$ . □