

# Moment Generating Function (MGF): definition

We are interested in finding the Moment Generating Function (MGF) of a random variable  $X \sim N(\mu, \sigma^2)$ , which is denoted  $M_X(t)$  and is an alternative specification of its PDF, and defined as

$$M_X(t) = E[e^{tX}] = \int_{-\infty}^{+\infty} e^{tX} f(x) dx$$

for  $X$  being a continuous random variable.

Then  $n^{th}$  moment will be given by  $E[X^n] = \frac{d^n M_X(t)}{dt^n} \Big|_{t=0}$ , that is the  $n^{th}$  derivative of the MGF evaluated at  $t = 0$ .

# MGF of a standard Normal r.v.

To derive the MGF of an arbitrary Normal random variable, we first begin with the MGF of the random variable  $Z = \frac{X-\mu}{\sigma} \sim N(0,1)$ . We have

$$\begin{aligned} M_Z(t) &= E[e^{tZ}] = \int_{-\infty}^{+\infty} e^{tZ} f_{\mu,\sigma}(z) dz = \int_{-\infty}^{+\infty} e^{tZ} \frac{1}{\sqrt{2\pi}} e^{-\frac{Z^2}{2}} dz \\ &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} e^{-\frac{(Z^2-2tZ)}{2}} dz \\ &= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} e^{-\frac{(Z-t)^2}{2} + \frac{t^2}{2}} dz \\ &= e^{t^2/2} \underbrace{\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} e^{-\frac{(Z-t)^2}{2}} dz}_{=\sqrt{2\pi}} \\ &= e^{t^2/2} \end{aligned}$$

# MGF of a Normal r.v.

Then, to obtain the MGF of the random variable  $X = \mu + \sigma Z \sim N(\mu, \sigma^2)$ , we proceed as follows:

$$\begin{aligned}M_X(t) &= E[e^{tX}] \\&= E[e^{t(\mu + \sigma Z)}] \\&= E[e^{\mu t} e^{t\sigma Z}] \\&= e^{\mu t} E[e^{t\sigma Z}] \\&= e^{\mu t} M_Z(t\sigma) \\&= e^{\mu t} e^{(t\sigma)^2/2} \\&= e^{\frac{t^2\sigma^2}{2} + \mu t}\end{aligned}$$

# First and second moments of a Normal random variable

For instance the first and second moment are respectively given by

$$\begin{aligned}E[X] &= \frac{dM_X(t)}{dt} \Big|_{t=0} = M'_X(0) \\&= (t\sigma^2 + \mu)e^{\frac{t^2\sigma^2}{2} + \mu t} \\&= \mu\end{aligned}$$

$$\begin{aligned}E[X^2] &= \frac{d^2M_X(t)}{dt^2} \Big|_{t=0} = M''_X(0) \\&= (t\sigma^2 + \mu)^2 e^{\frac{t^2\sigma^2}{2} + \mu t} + \sigma^2 e^{\frac{t^2\sigma^2}{2} + \mu t} \\&= \mu^2 + \sigma^2\end{aligned}$$

So in our case, the variance of  $X \sim N(\mu, \sigma^2)$  is given by  $E[X^2] - (E[X])^2 = \mu^2 + \sigma^2 - (\mu)^2 = \sigma^2$ .

# MGF in R

```
mgf.normal <- function(t, sigma, mu){  
  mgf <- expression(exp(((t^2 * sigma^2)/2) + mu*t))  
  dmgt.dt <- D(mgf, "t")  
  ddmgt.dtt <- D(dmgt.dt, "t")  
  dddmgt.dttt <- D(ddmgt.dtt, "t")  
  first.moment <- eval(dmgt.dt)  
  second.moment <- eval(ddmgt.dtt)  
  third.moment <- eval(dddmgt.dttt)  
  output <- NULL  
  output$first.moment <- first.moment  
  output$second.moment <- second.moment  
  output$third.moment <- third.moment  
  return(output)  
}
```

```
mgf.normal(t = 0, sigma = 1, mu = 0)  
# $first.moment  
# [1] 0  
# $second.moment  
# [1] 1  
# $third.moment  
# [1] 0
```

```
mgf.normal(t = 0, sigma = 2, mu = 4)  
# $first.moment  
# [1] 4  
# $second.moment  
# [1] 20  
# $third.moment  
# [1] 112
```