

Moment Generation Function (MGF) for the Normal distribution

The MGF is defined as $E e^{tX}$ which for the Normal $N(\mu, \sigma)$ becomes

$$M(t) = \int \frac{1}{\sigma\sqrt{2\pi}} e^{tx} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

or

$$M(t) = \int \frac{1}{\sigma\sqrt{2\pi}} \exp\left(tx + \frac{-(x-\mu)^2}{2\sigma^2}\right) dx$$

Now, recall that we don't know how to integrate difficult functions like this so we need to use a trick, namely to try and see if can rewrite the argument of the exponential function as “ $-(x - \text{stuff})^2 / (\text{more stuff})$ ” which has the same form as the argument of the normal distribution which we know what integrates to. (Note that for the purpose of integrating it is only the x -variable that “matters.”) This is known as “completing the square.”

So let us concentrate on the term $tx + \frac{-(x-\mu)^2}{2\sigma^2}$ and try to collect all terms involving x :

$$\begin{aligned} tx + \frac{-(x-\mu)^2}{2\sigma^2} &= \frac{1}{2\sigma^2}(-x^2 - 2\mu x + \mu^2) + 2tx\sigma^2 = \frac{-1}{2\sigma^2}(x^2 - 2\mu x + \mu^2 - 2tx\sigma^2) = \\ &= \frac{-1}{2\sigma^2}(x^2 - 2x(\mu + t\sigma^2) + \mu^2) = \frac{-1}{2\sigma^2}[(x - (\mu + t\sigma^2))^2 - (\mu + t\sigma^2)^2 + \mu^2] = \\ &= \frac{-1}{2\sigma^2}[(x - (\mu + t\sigma^2))^2 - 2\mu t\sigma^2 - t^2\sigma^4] = \frac{-(x - (\mu + t\sigma^2))^2}{2\sigma^2} + \frac{2\mu t\sigma^2 + t^2\sigma^4}{2\sigma^2} = \\ &= \frac{-(x - (\mu + t\sigma^2))^2}{2\sigma^2} + \mu t + \frac{1}{2}t^2\sigma^2 = \end{aligned}$$

So, finally, we get

$$\begin{aligned} M(t) &= \frac{1}{\sigma\sqrt{2\pi}} \int e^{\frac{-(x-(\mu+t\sigma^2))^2}{2\sigma^2} + \mu t + \frac{1}{2}t^2\sigma^2} dx = \frac{1}{\sigma\sqrt{2\pi}} \int e^{\mu t + \frac{1}{2}t^2\sigma^2} e^{-\frac{(x-(\mu+t\sigma^2))^2}{2\sigma^2}} dx = \\ &= e^{\mu t + \frac{1}{2}t^2\sigma^2} \frac{1}{\sigma\sqrt{2\pi}} \int e^{-\frac{(x-(\mu+t\sigma^2))^2}{2\sigma^2}} dx \end{aligned}$$

which, since the last term is a normal density, gives

$$M(t) = e^{\mu t + \frac{1}{2}t^2\sigma^2}$$

MGF for Poisson

$$E e^{tX} = \sum_{x=0}^{\infty} e^{tx} e^{-\lambda} \frac{\lambda^x}{x!} = \sum_{x=0}^{\infty} (e^t)^x e^{-\lambda} \frac{\lambda^x}{x!} = e^{-\lambda} \sum_{x=0}^{\infty} \frac{(e^t \lambda)^x}{x!} = e^{-\lambda} e^{e^t \lambda} = e^{\lambda(e^t - 1)}$$

MGF for Binomial

$$\begin{aligned} E e^{tX} &= \sum_{x=0}^n e^{tx} \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x} = \sum_{x=0}^n \frac{n!}{x!(n-x)!} (e^t p)^x (1-p)^{n-x} \\ &= [e^t p + (1-p)]^n = [1 + (e^t - 1)p]^n \end{aligned}$$

MGF for Exponential

$$E e^{tX} = \frac{1}{\theta} \int_0^{\infty} e^{tx} e^{-x/\theta} = \frac{1}{\theta} \int_0^{\infty} e^{-x(1/\theta - t)} = \frac{1}{\theta} \int_0^{\infty} e^{-x(\frac{1-\theta t}{\theta})} = \frac{1}{\theta} \frac{\theta}{1-\theta t} = \frac{1}{1-\theta t}$$