Lsn10

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Admin/Questions

The kth moment (sometimes called the raw moment and denoted as μ'_k) of a random variable is defined as $E[Y^k]$ and can always be found through either:
or in the case of discrete random variables
However, another method of finding the moments can be found by calculating $E[e^{ty}]$. Why this expectation? Recall that the Taylor series expansion about 0 for e^x is:
So, assuming we have finite moments we can write:
The neat thing about this is if we consder the first derivative we end up with:
Now if we evaluate this function we have μ'_1 .
Continuing on in this manner, the second derivative yields:

Which again, evaluating at t = 0 gives us the second moment.

Suppose that the waiting time for the first customer to enter a retail shop after 9:00 A.M. is a random variable $Y \sim \text{Exp}(\theta)$. Let's prove the result in the back of our book, that the MGF is $\frac{1}{1-\theta t}$ and use it to find E[Y] and $E[Y^2]$

While this is all well and good, a more common use for MGFs is what's pointed out on pg 141 of our text. Specifically, if an MGF exists, it is *unique*. This gives us an alternative way to characterize a distribution outside of a pdf.

To really take advantage of this, we need a few additional facts proven about MGFs. Specifically, if Y is a random variable with MGF m(t) and U is given by U = aY + b, the MGF of U is $e^{tb}m(at)$. To see this, let's start at:

$$m_u(t) =$$
Next we substitute in $U = ay + b$

$$=$$
Now pull out and regroup to arrive at
$$=$$

So, using this result, let $Y \sim N(\mu, \sigma)$ and, using the result in the back of the book that the mgf of a Normal random variable is $\exp\left(\mu t + \frac{t^2\sigma^2}{2}\right)$, find the mgf and hence the distribution of X = -3Y + 4

Let's practice a bit.

Keeping in mind that the formula for the binomial expansion is $(x+y)^n = \sum_{i=0}^n \binom{n}{i} x^{n-i} y^i$, argue that the mgf for a binomial random variable is $(pe^t + (1-p))^n$

