STAT 400 UIUC

Examples for 2.3

The k^{th} moment of X (the k^{th} moment of X about the origin), μ_k , is given by

$$\mu_k = \mathrm{E}(\mathrm{X}^k) = \sum_{\mathrm{all}\,x} x^k \cdot f(x)$$

The k^{th} central moment of X (the k^{th} moment of X about the mean), μ_k' , is given by

$$\mu_{k}' = E((X - \mu)^{k}) = \sum_{\text{all } x} (x - \mu)^{k} \cdot f(x)$$

The moment-generating function of X, $M_X(t)$, is given by

$$M_{X}(t) = E(e^{tX}) = \sum_{\text{all } x} e^{tx} \cdot f(x)$$

Theorem 1:
$$M'_X(0) = E(X)$$
 $M''_X(0) = E(X^2)$
 $M''_X(0) = E(X^k)$

Theorem 2: $M_{X_1}(t) = M_{X_2}(t)$ for some interval containing 0 $\Rightarrow f_{X_1}(x) = f_{X_2}(x)$

Theorem 3: Let Y = a X + b. Then $M_Y(t) = e^{bt} M_X(at)$

1. Suppose a random variable X has the following probability distribution:

f(x)	Find the moment-generating function of X, $M_X(t)$.
0.20	
0.40	$M_X(t) = E(e^{tX}) = \sum e^{tX} \cdot f(x)$
0.30	$\operatorname{all} x$
0.10	$= 0.20 e^{10 t} + 0.40 e^{11 t} + 0.30 e^{12 t} + 0.10 e^{13 t}.$
	$ \begin{array}{c} f(x) \\ 0.20 \\ 0.40 \\ 0.30 \\ 0.10 \end{array} $

2. Suppose the moment-generating function of a random variable X is

$$M_X(t) = 0.10 + 0.15 e^t + 0.20 e^{2t} + 0.25 e^{-3t} + 0.30 e^{5t}.$$

Find the expected value of X, E(X).

$$M'_{X}(t) = 0.15 e^{t} + 0.40 e^{2t} - 0.75 e^{-3t} + 1.50 e^{5t}.$$

$$E(X) = M'_{X}(0) = 0.15 + 0.40 - 0.75 + 1.50 = 1.30.$$

X	f(x)	$x \cdot f(x)$	
0	0.10	0	
1	0.15	0.15	
2	0.20	0.4	
-3	0.25	-0.75	$\mathbf{E}(\mathbf{X}) = \sum \mathbf{X} \cdot f(\mathbf{X}) =$
5	0.30	1.5	all x

OR

3. Suppose a discrete random variable X has the following probability distribution:

$$f(0) = P(X=0) = 2 - e^{1/2},$$
 $f(k) = P(X=k) = \frac{1}{2^k \cdot k!}, k = 1, 2, 3, ...$

a) Find the moment-generating function of X, $M_X(t)$.

$$M_{X}(t) = \sum_{\text{all } x} e^{tx} \cdot f(x) = 1 \cdot \left(2 - e^{1/2}\right) + \sum_{k=1}^{\infty} e^{tk} \cdot \frac{1}{2^{k} \cdot k!}$$
$$= \left(2 - e^{1/2}\right) + \sum_{k=1}^{\infty} \frac{\left(\frac{e^{t}}{2}\right)^{k}}{k!} = \left(2 - e^{1/2}\right) + \left(\frac{e^{e^{t}}}{2} - 1\right)$$
$$= 1 - e^{1/2} + e^{e^{t}}/2.$$

b) Find the expected value of X, E(X), and the variance of X, Var(X).

$$M_{X}'(t) = e^{e^{t}/2} \cdot e^{t}/2, \qquad E(X) = M_{X}'(0) = e^{1/2}/2.$$

$$M_{X}''(t) = e^{e^{t}/2} \cdot \left(e^{t}/2\right)^{2} + e^{e^{t}/2} \cdot e^{t}/2, \qquad E(X^{2}) = M_{X}''(0) = \frac{3}{4} \cdot e^{1/2}.$$

$$Var(X) = E(X^{2}) - [E(X)]^{2} = \frac{3}{4} \cdot e^{1/2} - \frac{1}{4} \cdot e.$$

4. Let X be a Binomial (n, p) random variable. Find the moment-generating function of X.

$$M_{\mathbf{X}}(t) = \sum_{k=0}^{n} e^{tk} \cdot {\binom{n}{k}} \cdot p^{k} \cdot (1-p)^{n-k}$$
$$= \sum_{k=0}^{n} {\binom{n}{k}} \cdot \left(p \cdot e^{t}\right)^{k} \cdot (1-p)^{n-k} = \left[(1-p) + p \cdot e^{t}\right]^{n}.$$

- 5. Let X be a geometric random variable with probability of "success" *p*.
- a) Find the moment-generating function of X.

$$\begin{split} \mathbf{M}_{\mathbf{X}}(t) &= \sum_{k=1}^{\infty} e^{tk} \cdot (1-p)^{k-1} \cdot p = p \cdot e^{t} \cdot \sum_{k=1}^{\infty} e^{t(k-1)} \cdot (1-p)^{k-1} \\ &= p \cdot e^{t} \cdot \sum_{n=0}^{\infty} \left[(1-p) \cdot e^{t} \right]^{n} = \frac{p \cdot e^{t}}{1-(1-p) \cdot e^{t}}, \quad t < -\ln(1-p). \end{split}$$

b) Use the moment-generating function of X to find E(X).

$$M'_{X}(t) = \frac{p \cdot e^{t} \cdot (1 - (1 - p) \cdot e^{t}) - p \cdot e^{t} \cdot (-(1 - p) \cdot e^{t})}{(1 - (1 - p) \cdot e^{t})^{2}}$$
$$= \frac{p \cdot e^{t}}{(1 - (1 - p) \cdot e^{t})^{2}}, \qquad t < -\ln(1 - p).$$
$$E(X) = M'_{X}(0) = \frac{p}{(p)^{2}} = \frac{1}{p}.$$

6. a) Find the moment-generating function of a Poisson random variable.

$$M_{X}(t) = \sum_{k=0}^{\infty} e^{tk} \cdot \frac{\lambda^{k} \cdot e^{-\lambda}}{k!} = e^{-\lambda} \cdot \sum_{k=0}^{\infty} \frac{\left(\lambda \cdot e^{t}\right)^{k}}{k!}$$
$$= e^{-\lambda} \cdot e^{\lambda \cdot e^{t}} = e^{\lambda \cdot (e^{t} - 1)}.$$

$$(\ln M_X(t))'|_{t=0} = E(X) = \mu_X$$

 $(\ln M_X(t))''|_{t=0} = E(X^2) - [E(X)]^2 = \sigma_X^2$

b) Find E(X) and Var(X), where X is a Poisson random variable.

$$\ln M_{X}(t) = \lambda (e^{t} - 1).$$

$$(\ln M_{X}(t))' = \lambda e^{t}.$$

$$(\ln M_{X}(t))' |_{t=0} = E(X) = \lambda.$$

$$(\ln M_{X}(t))'' = \lambda e^{t}.$$

$$(\ln M_{X}(t))'' |_{t=0} = Var(X) = \lambda.$$