Lecture Topics: 5.6 Moment Generating Functions, 5.7 Limit Theorems

[Tags: MGFs]

1. We’ll practice using MGFs.
   a. Let \( X \sim \text{Geo}(p) \). Give a formula for \( M_X(t) \), and specify for which values of \( t \) the formula converges. Recall the geometric series formula: \( \sum_{k=0}^{\infty} r^k = \frac{1}{1-r} \) holds for \(|r| < 1\).
   b. Let \( Y \sim \text{NegBin}(r,p) \). Give a formula for \( M_Y(t) \).
   c. Set up formulas to compute \( E[Y] \) and \( \text{Var}(Y) \) using the MGF \( M_Y \) (but don’t compute anything).

**Solution:**

a. We have \( p_X(k) = (1-p)^{k-1}p \) for \( k = 1, 2, 3, \ldots \), so

\[
M_X(t) = E[e^{tX}] = \sum_{k=1}^{\infty} e^{tk} p_X(k) = \sum_{k=1}^{\infty} e^{tk} (1-p)^{k-1}p = \sum_{k=1}^{\infty} e^{tk} e^{t} e^{(1-p)^{k-1}}p
\]

\[
= pe^t \sum_{k=1}^{\infty} e^{t(k-1)}(1-p)^{k-1} = pe^t \sum_{k=1}^{\infty} ((1-p)e^t)^{k-1}
\]

\[
= \frac{pe^t}{1 - (1-p)e^t} \quad \text{[geometric series]}
\]

Provided \( |(1-p)e^t| < 1 \rightarrow e^t < \frac{1}{|1-p|} \rightarrow t < -\log(1-p) \) to converge.

b. Since \( Y = \sum_{i=1}^{r} X_i \) for \( X_i \sim \text{Geo}(p) \) iid, the MGF of the sum is the product of the MGFs:

\[
M_Y(t) = M_{\sum_{i=1}^{r} X_i}(t) = \prod_{i=1}^{r} M_{X_i}(t) = \prod_{i=1}^{r} \frac{pe^t}{1 - (1-p)e^t} = \left( \frac{pe^t}{1 - (1-p)e^t} \right)^r
\]

c. We have

\[
E[Y] = M_Y'(0)
\]
\[
E[Y^2] = M_Y''(0)
\]
\[
\text{Var}(Y) = E[Y^2] - E[Y]^2
\]
[Tags: CLT]

2. Use the CLT to approximate the following probabilities. Don’t forget to apply the continuity correction (only if necessary).

   a. Suppose we roll a fair 10-sided die until we get 100 sevens. What is the probability it takes at least 1050 rolls until this happens?

   b. Let $X$ be the sum of 10,000 real numbers, and $Y$ be the same sum, but with each number rounded to the nearest integer before summing. If the fractions rounded off are independent and each one is uniformly distributed over $(-0.5, +0.5)$, use the Central Limit Theorem to estimate the probability that $|X - Y| > 50$. Noticing that $|X - Y|$ could have been as great as 5,000, look at your answer and think about what it says. (As a small example with sums of 4 real numbers, suppose that $X = 3.2 + 1.92 + (-3.6) + 5.7$. Then $Y$ would be the sum of each of those terms when rounded to the nearest integer: $Y = 3 + 2 + (-4) + 6 = 7$. So, $|X - Y| = 1.3$. The fractions rounded off in this case are $(0.2, -0.08, 0.4, -0.3)$ and the assumption is that these fractions are independent and uniformly distributed in the real interval $(-0.5, +0.5)$.

Solution:

3. We know that $X \sim \text{NegBin} \left(r = 100, p = \frac{1}{10} \right)$, so $E[X] = \frac{r}{p} = 1000$ and $\text{Var}(X) = \frac{r(1-p)}{p^2} = 9000$. By the CLT, $X \approx N(\mu = 1000, \sigma^2 = 9000)$ since we have the sum of 100 iid Geometric rvs. We have to apply the continuity correction since we are approximating a discrete distribution (NegBin) with a continuous one (Normal).

$$P(X \geq 1050) = P(X \geq 1049.5) = 1 - P(X \leq 1049.5)$$

$$= 1 - P \left( \frac{X - 1000}{\sqrt{9000}} \leq \frac{1049.5 - 1000}{\sqrt{9000}} \right)$$

$$= 1 - P(Z \leq 0.52178) = 1 - \Phi(0.52) = 1 - 0.6985 = 0.3015$$

Imagine summing that Negative Binomial PMF...yuckkk

b. Let $R_1, ..., R_n \sim \text{Unif} (-0.5, +0.5)$ be the roundoff errors ($n = 10,000$). Then,

$$E[R_i] = \frac{a + b}{2} = \frac{-0.5 + 0.5}{2} = 0 \quad \text{Var}(R_i) = \frac{(0.5 - (-0.5))^2}{12} = \frac{1}{12}$$

Hence, $R = \sum_{i=1}^{n} R_i$ has $E[R] = 0$ and $\text{Var}(R) = \frac{10000}{12}$. By the CLT, $R \approx N\left(\mu = 0, \sigma^2 = \frac{10000}{12}\right)$. Note we don’t use a continuity correction here since we are approximating a continuous variable.

$$P(|R| > 50) = P(R > 50) + P(R < -50)$$

$$= 2P(R > 50) \approx 2P \left( \frac{R - 0}{\sqrt{\frac{10000}{12}}} \geq \frac{50 - 0}{\sqrt{\frac{10000}{12}}} \right)$$

$$= 2P(Z \geq 1.732) = 2(1 - \Phi(1.732)) = 0.0832$$
Megha has a health condition that requires unpredictable amounts of medication. Every day, there is a 20% chance that she feels perfectly fine and requires no medicine. Otherwise, she needs to take a dose of medication. The necessary dose is equally likely to be any value in the continuous range 1 to 5 ounces. How much medicine she needs on any given day is independent of all other days. Megha’s insurance will fully cover 90 ounces of medicine for each 30-day period. What is the probability that 90 ounces will be enough for the next 30 days? Make your life easier by using Central Limit Theorem.

**Solution:**

Let $X_1, ..., X_{30}$ be the amount of medicine she needs per day. Let $D_1, ..., D_{30}$ be the event she needs a dose on day $i$.

$$E[X_i] = E[X_i|D_i]P(D_i) + E[X_i|D_i^c]P(D_i^c) = \left(\frac{1+5}{2}\right) \cdot 0.8 + 0 \cdot 0.2 = 2.4$$

By LOTUS:

$$E[X_i^2] = E[X_i^2|D_i]P(D_i) + E[X_i^2|D_i^c]P(D_i^c) = \left(\int_1^5 x^2 \frac{1}{4} dx\right) 0.8 + 0 \cdot 0.2 \approx 8.267$$

Hence, $Var(X_i) = E[X_i^2] - E[X_i]^2 \approx 8.267 - 2.4^2 = 2.507$

The total dosage is $X = \sum_{i=1}^{30} X_i$, so $E[X] = 30 \cdot 2.4 = 72$ and $Var(X) = 30 \cdot 2.507 = 75.21$. By the CLT, since $X$ is the sum of iid variables, $X \approx N(\mu = 72, \sigma^2 = 75.21)$, and

$$P(X < 90) = P\left(\frac{X - 72}{\sqrt{75.21}} < \frac{90 - 72}{\sqrt{75.21}}\right) \approx P(Z \leq 2.0755) = \Phi(2.0755) \approx 0.98$$