CENG 222 Statistical Methods for Computer Engineering

Week 3

Chapter 3 Families of discrete distributions

Bernoulli distribution

- A random variable with two possible values, 0 and 1, is called a *Bernoulli variable*
- The distribution of such a r.v. is called the *Bernoulli distribution*
- Any random experiment with a binary outcome is called a *Bernoulli trial*
- Generic outcome names: *successes* and *failures*

Not equally likely outcomes

- In general, f(1) = f(0) = 0.5 does NOT hold when the binary outcomes are not equally likely
- If f(1) = p, what is E(X) and Var(X)?

What about "non 0-1", binary outcomes?

- Example:
 - What if the two possible outcomes are 5 and 9 with f(5) = 0.3 and f(9) = 0.7?
 - What is the expected value?

What about "non 0-1", binary outcomes?

- Example:
 - What if the two possible outcomes are 5 and 9 with f(5) = 0.3 and f(9) = 0.7?
 - What is the expected value?
 - It is just a shifted and rescaled standard Bernoulli trial.
 - X = 4B + 5
 - $E(X) = E(4B + 5) = 4E(B) + 5 = 4 \cdot 0.7 + 5 = 7.8$

Binomial distribution

• Number of successes in a sequence of independent Bernoulli trials

-n: number of trials

-p: probability of success

•
$$f_x(x) = P(X = x) = \binom{n}{x} p^x q^{n-x}$$

- Expected value and variance:
 - A binomial variable X is a sum of n independent Bernoulli trials.

$$-E(X) = np, Var(X) = npq$$

Using distribution tables

- Table A2, *cdf* of Binomial distribution
- *pdf* can be obtained by difference of two consecutive entries
- Example 3.16
- Example 3.17

Using *binocdf*(*x*,*n*,*p*) function of MATLAB

Geometric distribution

- The number of Bernoulli trials needed to get the first success
- The support is the set of integers $[1..\infty]$
- $f_x(x) = P(X = x) = pq^{x-1}$
- The support is unbounded - Check that $\sum_{x} f_{x}(x) = \sum_{x=1}^{\infty} p(1-p)^{x-1} = 1$
- Expected value and variance:

 $-E(X) = 1/p, Var(X) = (1-p)/p^2$

Geometric distribution

- Example 3.20 St. Petersburg Paradox
- Gambling with a guaranteed strategy to win a desired amount
 - Even when *p* is less then 0.5!
 - Start with the desired amount
 - Double betting amount every time you loose
 - Stop when you win the first time
 - E.g if p=0.2 the expected number of bets to win is 5!

Geometric distribution

- So what's the paradox?
- What is the amount of money, *Y*, needed to follow the strategy?
 - $-Y = D2^{X-1}$ where *D* is the desired amount and *X* is the number of bets needed to win.
 - $-E(Y) = \text{infinity when } p \le 0.5 \text{ (the paradox)}$

Negative Binomial distribution

- In a sequence of independent Bernoulli trials, the number of trials needed to obtain *k* successes
 - It can be considered as the *inverse* of the Binomial, where, we now fix the number of successes and count the number of trials *n* to reach that number of successes
- It is a generalization of the Geometric distribution

Negative Binomial distribution

•
$$f_x(x) = P(X = x) = {\binom{x-1}{k-1}} p^k q^{x-k}$$

- Expected value and variance:
 - A negative binomial variable X is a sum of k independent Geometric variables.
 - $-E(X) = k/p, Var(X) = k(1-p)/p^2$
- Example 3.21
 - -k = 12, p = 0.95, P(X > 15) = ?
 - $-P(X>15) = 1 F_X(15)$
 - Can be solved by using the Binomial distribution with n = 15, p = 0.95, $P(Y < 12) = F_Y(11)$.

Poisson distribution

- The number of rare events occurring within a fixed period of time
- It has a single parameter
 - $-\lambda$: frequency, average number of events

$$-f_x(x) = e^{-\lambda} \frac{\lambda^x}{x!}$$
$$-E(X) = \lambda, Var(X) = \lambda$$

• Example 3.22

Poisson approximation of Binomial distribution

- Poisson distribution can be used to approximate Binomial distribution when *n* is large and *p* is small
 - E.g., $n \ge 30$ and $p \le 0.05$
 - $-np = \lambda$
- Example 3.25 The Birthday Problem