# STATS 200A: Homework #3

Professor Yingnian Wu Assignment: 1, 2

### Eric Chuu

### UID: 604406828

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## Problem 1

For  $X \sim \text{Binom}(n, p)$ , let  $\mu = np$  and  $\sigma^2 = np(1-p)$ . Let  $Z = (X - \mu)/\sigma$ . Prove that  $P(Z \in [a, b]) \rightarrow \int_a^b f(z)dz$ , where f(z) is the density of the standard normal distribution.

#### Solution

For general  $p \in (0,1)$ , let q = 1 - p. Let  $k = \mu + z\sigma = np + z\sqrt{npq} = np + d$ . Neglecting O(1/n) terms, we have

$$P(X = np + d) = \frac{n!}{(np+d)!(nq-d)!} p^{np+d} q^{nq-d}$$
(1)

$$\frac{\sqrt{2}n^n e^{-n} p^{np+d} q^{nq-d}}{(2)}$$

$$= \frac{\sqrt{2\pi(np+d)(np+d)^{np+d}e^{-np-d}\sqrt{2\pi(nq-d)(nq-d)^{nq-d}e^{-nq+d}}}}{\sqrt{2\pi(np+d)(np+d)^{np+d}e^{-np-d}\sqrt{2\pi(nq-d)(nq-d)^{nq-d}e^{-nq+d}}}}$$
(2)

$$=\frac{\sqrt{n}}{\sqrt{2\pi(np+d)(nq-d)}}\left(\frac{np}{np+d}\right)^{np+a}\left(\frac{nq}{nq-d}\right)^{nq-a}$$
(3)

Taking the log of both sides and using the Taylor expansion of  $\log(1+\delta) = \delta - \delta^2/2 + O(\delta^3)$ , we have

$$\log\left[\left(\frac{np}{np+d}\right)^{np+d}\left(\frac{nq}{nq-d}\right)^{nq-d}\right] \tag{4}$$

$$= -(np+d)\log\left(1+\frac{d}{np}\right) - (nq-d)\log\left(1-\frac{d}{nq}\right)$$
(5)

$$= (-np+d)\left(\frac{d}{np} - \frac{d^2}{2n^2p^2} + O\left(\frac{1}{n^3}\right)\right) - (nq-d)\left(-\frac{d}{nq} - \frac{d^2}{2n^2q^2} + O\left(\frac{1}{n^3}\right)\right)$$
(6)

$$= -d + \frac{d^2}{2np} - \frac{d^2}{2np} + O\left(\frac{1}{n^3}\right) + d + \frac{d^2}{2nq} - \frac{d^2}{nq} + O\left(\frac{1}{n^3}\right)$$
(7)

$$= -\frac{d^2}{2np} - \frac{d^2}{2nq} + O\left(\frac{1}{n^3}\right) \tag{8}$$

$$= -z^2 \left(\frac{npq}{2np} + \frac{npq}{nq}\right) \tag{9}$$

$$= -z^2 \left(\frac{q}{2} + \frac{p}{2}\right) \tag{10}$$

$$=\frac{-z^2}{2} \tag{11}$$

Note also that in equation (3), since k = np + d and n - k = n - np - d = nq - d, we can then express the first term of the product as

$$\frac{\sqrt{n}}{\sqrt{2\pi(np+d)(nq-d)}} = \sqrt{\frac{1}{2\pi\frac{1}{n}k(n-k)}} = \frac{1}{\sqrt{2\pi\frac{k}{n}n(1-\frac{k}{n})}}$$
(12)

Since k = np + d, then as  $n \to \infty$ , we have  $\frac{k}{n} \to p$ , so equation (12) gives us

$$\frac{1}{\sqrt{2\pi\frac{k}{n}n(1-\frac{k}{n})}} \to \frac{1}{\sqrt{2\pi pnq}}$$
(13)

Combining our conclusions from equations (3), (11), and (13), we have the following

$$P(X = np + d) = \frac{1}{\sqrt{2\pi npq}} e^{-z^2/2}$$
(14)

Taking  $\Delta z = \sqrt{npq}$ , equation (14) becomes

$$P(X = np + d) = \frac{1}{\sqrt{2\pi}} e^{-z^2/2} \Delta z$$
(15)

$$= f(z) \cdot \Delta z \tag{16}$$

Since we are calculating the probability of  $Z \in [a, b]$ , we consider the following quantities

$$Z = a = \frac{X - \mu}{\sigma}$$
  

$$\Rightarrow X = a\sigma + \mu$$
  

$$Z = b = \frac{X - \mu}{\sigma}$$
  

$$\Rightarrow X = b\sigma + \mu$$

Then we can calculate the probability that the point falls within this region

$$P(Z \in [a, b]) = P(X \in [a\sigma + \mu, b\sigma + \mu])$$

$$(17)$$

$$=\sum_{k=\mu+a\sigma}^{\mu+o\sigma} P(X=k)$$
(18)

$$=\sum_{z\in[a,b]}P(X=\mu+z\sigma)$$
(19)

$$=\sum_{z\in[a,b]}f(z)\cdot\Delta z\tag{20}$$

As  $n \to \infty$ ,  $\Delta z = \frac{1}{\sqrt{npq}} \to 0$ , and equation (20) evaluates to

$$\sum_{z \in [a,b]} f(z) \cdot \Delta z \to \int_a^b f(z) dz$$

which is exactly what we wanted to show.

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# Problem 2

Calculate the expectation and variance of the Poisson random variable based on the probability mass function.

#### Solution

Let X be a Poisson random variable with parameter . Then the PMF of X is given by:

$$P_X(k) = e^{-\lambda} \frac{\lambda^k}{k!}, \qquad k = 0, 1, \dots$$
 (21)

We can then use the PMF to calculate the expectation of X,

$$\begin{split} \mathbf{E}[X] &= \sum_{k=0}^{\infty} k e^{-\lambda} \frac{\lambda^k}{k!} \\ &= \sum_{k=1}^{\infty} k e^{-\lambda} \frac{\lambda^k}{k!} \\ &= \lambda e^{-\lambda} \sum_{k=1}^{\infty} \frac{\lambda^{k-1}}{(k-1)!} \\ &= \lambda e^{-\lambda} \sum_{i=0}^{\infty} \frac{\lambda^i}{i!} \\ &= \lambda e^{-\lambda} (1+\lambda+\frac{\lambda}{2!}+\cdots) \\ &= \lambda e^{-\lambda} e^{\lambda} \\ &= \lambda \end{split}$$

We calculate  $E[X^2]$  as well as an intermediate step for calculating the variance of X.

$$\begin{split} \mathbf{E}[X^2] &= \sum_{k=0}^{\infty} k^2 e^{-\lambda} \frac{\lambda^k}{k!} \\ &= \lambda e^{-\lambda} \sum_{k=1}^{\infty} k \frac{\lambda^{k-1}}{(k-1)!} \\ &= \lambda e^{-\lambda} \sum_{k=1}^{\infty} (k-1+1) \cdot \frac{\lambda^{k-1}}{(k-1)!} \\ &= \lambda e^{-\lambda} \left[ \sum_{k=1}^{\infty} (k-1) \frac{\lambda^{k-1}}{(k-1)!} + \sum_{k=1}^{\infty} \frac{\lambda^{k-1}}{(k-1)!} \right] \\ &= \lambda e^{-\lambda} \left[ \lambda \sum_{k=1}^{\infty} \frac{\lambda^{k-2}}{(k-2)!} + \sum_{j=0}^{\infty} \frac{\lambda^j}{j!} \right] \\ &= \lambda e^{-\lambda} \left[ \lambda \sum_{i=0}^{\infty} \frac{\lambda^i}{i!} + e^{\lambda} \right] \\ &= \lambda e^{-\lambda} \left( \lambda e^{\lambda} + e^{\lambda} \right) \\ &= \lambda^2 + \lambda \end{split}$$

Then the variance of X is given by

$$\operatorname{Var}(X) = \operatorname{E}[X^2] - \operatorname{E}[X]^2 = \lambda^2 + \lambda - \lambda^2 = \lambda$$