Supplemental note for Week 2 Part 3

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Binomial distribution \rightarrow Normal distribution 1

Start from Eq.(C6) shown below.

$$P(n) = \frac{M!}{n!(M-n)!}p^n(1-p)^{M-n}$$
 (1)

$$\ln P(n) = \ln M! - \ln n! - \ln(M-n)! + n \ln p + (M-n) \ln(1-p) \tag{2}$$

$$= M(\ln M - 1) - n(\ln n - 1) - (M - n)(\ln(M - n) - 1) \tag{3}$$

$$+n\ln p + (M-n)\ln(1-p) \tag{4}$$

Here we used Stirling's approximation valid for large n.

$$ln n! \simeq n(ln n - 1)$$
(5)

Take the 1st derivative of the above equation in terms of n.

$$\frac{d \ln P}{dn} = -(\ln n + 1) + (\ln(M - n) + 1) + \ln p - \ln(1 - p)$$

$$= -\ln n + \ln(M - n) + \ln p - \ln(1 - p)$$
(6)

$$= -\ln n + \ln(M - n) + \ln p - \ln(1 - p) \tag{7}$$

$$= \ln\left[\frac{M-n}{n}\right] - \ln\left[\frac{1-p}{p}\right] \tag{8}$$

When n and M are both large, the peak in P(n) and also in $\ln P(n)$ is very sharp around the mean value $n = \langle N \rangle$, and thus

$$\frac{d\ln P}{dn}\Big|_{n=\langle N\rangle} = \ln \left[\frac{M-\langle N\rangle}{\langle N\rangle}\right] - \ln \left[\frac{1-p}{p}\right] = 0 \tag{9}$$

$$\therefore \frac{M - \langle N \rangle}{\langle N \rangle} = \frac{1 - p}{p} \tag{10}$$

$$\langle N \rangle = Mp.$$
 (11)

Consider a Taylor expansion of $\ln P(n)$ around the mean $\langle N \rangle$ by defining $n = \langle N \rangle + \delta n$.

$$\ln P(n) = \ln(\langle N \rangle) + \left. \frac{d \ln P}{dn} \right|_{n = \langle N \rangle} \delta n + \frac{1}{2} \left. \frac{d^2 \ln P}{dn^2} \right|_{n = \langle N \rangle} \delta n^2 + \cdots \tag{12}$$

From Eq.(7),

$$\frac{d^2 \ln P}{dn^2} = -\frac{1}{n} - \frac{1}{M-n} \tag{13}$$

$$\frac{d^2 \ln P}{dn^2} = -\frac{1}{n} - \frac{1}{M-n}$$

$$\frac{d^2 \ln P}{dn^2} \Big|_{n=\langle N \rangle} = -\frac{1}{Mp} - \frac{1}{M-Mp} = -\frac{1}{Mp(1-p)} = -\sigma^{-2}.$$
(13)

Terminate Eq.(12) at the 2nd order in terms of δn , and use Eqs.(9) and (14).

$$P(n) = const. \times \exp\left(-\frac{\delta n^2}{2\sigma^2}\right) \tag{15}$$

Determine const. so that that $\int_{-\infty}^{\infty} P(n)dn = 1$, we finally obtain the following normal distribution function with $\langle N \rangle = Mp$ and $\sigma^2 = Mp(1-p)$.

$$P(n) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(n-\langle N\rangle)^2}{2\sigma^2}\right)$$
 (16)

${\bf 2} \quad {\bf Binomial \ distribution} \rightarrow {\bf Poisson \ distribution}$

We now consider the limit of $M \to \infty$ while $\langle N \rangle = Mp = a$ remains constant. Notice that the following approximations hold.

$$M - n \simeq M \tag{17}$$

$$\frac{M!}{(M-n)!} = M(M-1)\cdots(M-n+1) \simeq M^n$$
 (18)

Again start from Eq.(C6), we can derive Poisson distribution as shown below with $\langle N \rangle = Mp = a$ and $\sigma^2 = Mp(1-p) \simeq a$.

$$P(n) = \frac{M!}{n!(M-n)!}p^n(1-p)^{M-n}$$
(19)

$$\simeq \frac{1}{n!} M^n \left(\frac{a}{M}\right)^n \left(1 - \frac{a}{M}\right)^M$$
 (20)

$$\simeq \frac{a^n e^{-a}}{n!} \tag{21}$$

Here we used

$$\lim_{M \to \infty} \left(1 - \frac{a}{M} \right)^M = e^{-a}. \tag{22}$$