

1. Let Z be poisson RV with mean μ , $\mu \geq 1$, then

$$(a) \Pr(Z = \mu + h) = \frac{e^{-\mu} \mu^{(\mu+h)}}{(\mu+h)!} \geq \Pr(Z = \mu - h - 1) = \frac{e^{-\mu} \mu^{(\mu-h-1)}}{(\mu-h-1)!}$$

$$\Pr(X=r) = \frac{e^{-\mu} \mu^r}{r!} \quad \text{for } 0 \leq h \leq \mu - 1$$

$$\Pr(Z = \mu + h) = \frac{e^{-\mu} \mu^{(\mu+h)}}{(\mu+h)!} \geq \Pr(Z = \mu - h - 1) = \frac{e^{-\mu} \mu^{(\mu-h-1)}}{(\mu-h-1)!}$$

$$\mu^{2h+1} \geq \frac{(\mu+h)!}{(\mu-h-1)!} \quad \mu^{(2h+1)} \geq \prod_{k=\mu-h}^{\mu+h} k$$

$$\mu^{2h} \geq (\mu-h)(\mu-h+1) \dots (\mu-1)(\mu+1) \dots (\mu+h)$$

$$\mu^{2h} \geq (\mu^2 - h^2)(\mu^2 - (h-1)^2) \dots (\mu^2 - 1)$$

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4 4 4

(b)

$$\Pr(Z \geq \mu) = \sum_{h \geq 0} \Pr(Z = \mu + h) \geq \sum_{h \geq 0} \Pr(Z = \mu - h - 1)$$

$$= 1 - \sum_{h \geq 0} \Pr(Z = \mu + h)$$

$$\sum_{h \geq 0} \Pr(Z = \mu + h) \geq 1$$

$$\sum_{h \geq 0} \Pr(Z = \mu + h) = \Pr(Z \geq \mu) \geq \frac{1}{2}$$

2,

$$X \sim \text{Po}(\mu)$$

Assume $n \rightarrow \infty$

$$X = Y + Z$$

$$\Rightarrow Y \sim \text{Bin}(n, p), \quad np = \mu_Y$$

$$Z \sim \text{Bin}(n, 1-p) \quad n(1-p) = \mu_Z$$

$$X \sim \text{Po}(\mu)$$

$$Pr(Y=k) = Pr(Y=k | X=n) Pr(X=n)$$

$$Pr(Y=k \cap Z=n-k) = Pr(Y=k | Z=n-k) Pr(Z=n-k)$$

$$= Pr(Y=k) Pr(Z=n-k)$$

$$Pr(Y=k) = \sum_{n=k}^{\infty} Pr(Y=k | X=n) Pr(X=n)$$

$$= \sum_{n=k}^{\infty} \binom{n}{k} p^k (1-p)^{n-k} \frac{e^{-\mu} \mu^n}{n!}$$

$$= \sum_{n=k}^{\infty} \frac{n!}{n!(n-k)!k!} p^k (1-p)^{n-k} \frac{e^{-\mu} \mu^n}{n!}$$

$$= \frac{\mu^k e^{-\mu} p^k}{k!} \sum_{n=k}^{\infty} \frac{[(1-p)\mu]^{n-k}}{(n-k)!} = \frac{(\mu p)^k e^{-\mu}}{k!} \sum_{i=0}^{\infty} \frac{[(1-p)\mu]^i}{i!}$$

$$= \frac{(\mu p)^k e^{-\mu}}{k!} \cdot e^{(1-p)\mu} = \frac{(\mu p)^k e^{-p\mu}}{k!}$$

let $k+h=n$

$$Pr(Z=n-k) = Pr(Z=h) = \sum_{n=h}^{\infty} \binom{n}{h} (1-p)^h p^{n-h} \frac{e^{-\mu} \mu^n}{n!}$$

$$= \frac{[(1-p)\mu]^h e^{-(1-p)\mu}}{h!} = \frac{(\mu(1-p))^h e^{-\mu}}{h!} \cdot e^{p\mu}$$

 $p\mu = \mu(1-p)$

$$Pr(Z=h) Pr(Y=k) = \frac{[(\mu(1-p))^h e^{-(1-p)\mu}] \cdot (\mu p)^k e^{-p\mu}}{h! \cdot k!} = \frac{e^{-\mu} \mu^h (1-p)^h p^k}{h! \cdot k!}$$

$$= Pr(Z=h | X=h+k) Pr(X=h+k) = \frac{n!}{(n-h)!h!} (1-p)^h p^{n-h} \frac{e^{-\mu} \mu^n}{n!} = \frac{e^{-\mu} \mu^h (1-p)^h p^k}{h! \cdot k!}$$

$$\text{let } h+k=n$$

$$\begin{aligned}
 (a) \quad E[X_1 X_2 \dots X_k] &= \sum_{a_1=0,1} \sum_{a_2=0,1} \dots \sum_{a_k=0,1} X_1 X_2 \dots X_k \cdot \Pr(X_1=a_1) \Pr(X_2=a_2) \dots \Pr(X_k=a_k) \\
 &= 1 \cdot \Pr(X_1=1) \cap \Pr(X_2=1) \cap \dots \cap \Pr(X_k=1) = \left(1 - \frac{k}{n}\right)^n
 \end{aligned}$$

$$\begin{aligned}
 E[Y_1 Y_2 \dots Y_k] &= \sum_{a_1=0,1} \sum_{a_2=0,1} \dots \sum_{a_k=0,1} Y_1 Y_2 \dots Y_k \cdot \Pr(Y_1=a_1) \cap \Pr(Y_2=a_2) \dots \Pr(Y_k=a_k) \\
 &= 1 \cdot \Pr(Y_1=1) \cap \Pr(Y_2=1) \cap \dots \cap \Pr(Y_k=1) \cap \Pr(Y_k=a_k) \\
 &= \left(1 - \frac{1}{n}\right)^k = \left(1 - \frac{1}{n}\right)^{nk}
 \end{aligned}$$

$$\Rightarrow \left(1 - \frac{k}{n}\right) \leq \left(1 - \frac{1}{n}\right)^k \quad \frac{d}{dk} \left(1 - \frac{k}{n}\right) = -\frac{1}{n}$$

$$\begin{aligned}
 (b) \quad X_1^{j_1} X_2^{j_2} \dots X_k^{j_k} &= X_1 X_2 \dots X_k \\
 \text{since } X_i^{j_i} &= 1 \text{ or } 0, \quad 1^k = 1, 0^k = 0 \\
 \frac{d}{dk} \left(1 - \frac{1}{n}\right)^k &= \frac{d}{dk} e^{k \ln(1 - \frac{1}{n})} = \ln(1 - \frac{1}{n})
 \end{aligned}$$

$$\begin{aligned}
 (c) \quad E[e^{tX}] &= \sum_{i=0}^{\infty} \frac{t^i}{i!} E[X^i] \quad E[e^{tY}] = \sum_{i=0}^{\infty} \frac{t^i}{i!} E[Y^i] \\
 \text{Taylor: } &\sum_{i=0}^{\infty} \frac{f^{(i)}(a)}{i!} (x-a)^i
 \end{aligned}$$

$$\text{Goal: } E[X^i] \leq E[Y^i]$$

$$\begin{aligned}
 E\left[\left(\sum_{j=1}^n X_j^{j_i}\right)^i\right] &= E\left[\sum_{j=1}^n \left(\prod_{j=1}^i X_j^{j_i}\right)\right] = E\left[\sum_{j=1}^n \left(\prod_{j=1}^i X_j\right)\right] \\
 &= \sum \left(E\left[\prod_{j=1}^i X_j\right]\right) \leq \sum \left(E\left[\prod_{j=1}^i Y_j\right]\right)
 \end{aligned}$$

(D)

$$\begin{aligned}
 \Pr(X \geq (1+\delta)E[X]) &= \Pr(e^{tX} \geq e^{t(1+\delta)E[X]}) \leq \frac{E[e^{tX}]}{e^{t(1+\delta)E[X]}} \\
 &= \frac{\prod_{i=1}^n E[e^{tX_i}]}{e^{t(1+\delta)E[X]}} = \frac{\prod_{i=1}^n (1 + (e^t - 1)p)}{e^{t(1+\delta)E[X]}} = \frac{e^{(e^t - 1)np}}{e^{t(1+\delta)E[X]}} = \frac{e^{(e^t - 1)np}}{e^{t(1+\delta)np}} = \frac{e^{(e^t - 1 - (1+\delta))np}}{1} \\
 \min\left(\frac{e^{(e^t - 1)np}}{e^{t(1+\delta)np}}\right) &= \min((e^t - 1)np - t(1+\delta)np) \quad \frac{d}{dt} (e^t - 1)np - t(1+\delta)np \\
 t = \ln \frac{1+\delta}{np} \quad t + \ln np &= \ln(1+\delta) \ln e^{np} = \ln(1+\delta) = e^t np - (1+\delta)np \\
 &\leftarrow e^t np - (1+\delta)np = 0
 \end{aligned}$$

4.

$$\begin{aligned}
 (a) \quad & E[f(Y_1^{(m)}, \dots, Y_n^{(m)})] \geq E[f(X_1^{(m)}, \dots, X_n^{(m)})] \Pr\left(\sum_{i=1}^n Y_i^{(m)} \geq m\right) \\
 & E[f(Y_1^{(m)}, \dots, Y_n^{(m)})] \\
 &= \sum_{k=0}^{\infty} E[f(Y_1^{(m)}, \dots, Y_n^{(m)}) | \sum_{i=1}^n Y_i^{(m)} = k] \cdot \Pr\left(\sum_{i=1}^n Y_i^{(m)} = k\right) \\
 &\geq E[f(Y_1^{(m)}, \dots, Y_n^{(m)}) | \sum_{i=1}^n Y_i^{(m)} \geq m] \cdot \Pr\left(\sum_{i=1}^n Y_i^{(m)} \geq m\right) \\
 &= \sum f(Y_1^{(m)}, \dots, Y_n^{(m)}) \cdot \Pr((Y_1^{(m)}, \dots, Y_n^{(m)}) = (k_1, \dots, k_n) | \sum_{i=1}^n k_i \geq m) \cdot \Pr\left(\sum_{i=1}^n Y_i^{(m)} \geq m\right) \\
 &= \sum f(X_1^{(m)}, \dots, X_n^{(m)}) \cdot \Pr((X_1^{(m)}, \dots, X_n^{(m)}) = (k_1, \dots, k_n) | \sum_{i=1}^n k_i \geq m) \cdot \Pr\left(\sum_{i=1}^n Y_i^{(m)} \geq m\right) \\
 &= E[f(X_1^{(m)}, \dots, X_n^{(m)})] \cdot \Pr\left(\sum_{i=1}^n Y_i^{(m)} \geq m\right)
 \end{aligned}$$

$$\begin{aligned}
 \Rightarrow & \Pr((Y_1^{(m)}, \dots, Y_n^{(m)}) = (k_1, \dots, k_n) | \sum_{i=1}^n k_i = k) \\
 &= \frac{\Pr(Y_1^{(m)} = k_1) \wedge \Pr(Y_2^{(m)} = k_2) \dots \wedge \Pr(Y_n^{(m)} = k_n)}{\Pr(\sum_{i=1}^n k_i = k)}
 \end{aligned}$$

$$\begin{aligned}
 &= \frac{\prod_{i=1}^n \frac{e^{-m} \cdot m^{k_i}}{k_i!}}{e^{-m} \cdot m^k} = \frac{\frac{e^{-m}}{m^k} \cdot \frac{m^{k_1}}{k_1!} \dots \frac{m^{k_n}}{k_n!}}{e^{-m} \cdot m^k} = \frac{k!}{k_1! k_2! \dots k_n! \cdot n^k}
 \end{aligned}$$

$$\begin{aligned}
 \Rightarrow & \Pr((X_1^{(m)}, \dots, X_n^{(m)}) = (k_1, \dots, k_n) | \sum_{i=1}^n k_i = k) \\
 &= \frac{\Pr(X_1^{(m)} = k_1) \wedge \dots \wedge \Pr(X_n^{(m)} = k_n)}{\Pr(\sum_{i=1}^n k_i = k)} = \frac{k!}{k_1! k_2! \dots k_n! \cdot n^k}
 \end{aligned}$$

$$\therefore \Pr((Y_1^{(m)}, \dots, Y_n^{(m)}) = (k_1, \dots, k_n) | \sum_{i=1}^n k_i = k) = \Pr((X_1^{(m)}, \dots, X_n^{(m)}) = (k_1, \dots, k_n) | \sum_{i=1}^n k_i = k)$$

(b)

According to 1.

$$Pr(Z \geq m) \geq \frac{1}{2}, \quad Z \sim P_0(m), \quad m \geq 1$$

$$\text{Let } \sum_{i=1}^n Y_i^{(m)} = Y^{(m)}, \quad \text{since } Y_i^{(m)} \sim P_0\left(\frac{m}{n}\right)$$

$$E[Y^{(m)}] = n \cdot \frac{m}{n} = m \Rightarrow Y^{(m)} \sim P_0(m)$$

$$\therefore Pr(Y^{(m)} \geq m) \geq \frac{1}{2}$$

$$\begin{aligned} E[f(Y_1^{(m)}, \dots, Y_n^{(m)})] &\geq E[f(X_1^{(m)}, \dots, X_n^{(m)})] \cdot Pr\left(\sum_{i=1}^n Y_i^{(m)} \geq m\right) \\ &= E[f(X_1^{(m)}, \dots, X_n^{(m)})] \cdot Pr(Y^{(m)} \geq m) \\ &\geq \frac{1}{2} E[f(X_1^{(m)}, \dots, X_n^{(m)})] \end{aligned}$$

$$\therefore E[f(X_1^{(m)}, \dots, X_n^{(m)})] \leq 2 E[f(Y_1^{(m)}, \dots, Y_n^{(m)})] \quad \psi$$