

2012 Theory I #1

$$1a) EV = E\left(E\left[I_{N \geq 0} \sum_{i=1}^N X_i \mid N\right]\right) = \sum_{k=1}^{\infty} k \lambda P(N=k)$$

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

$$\text{Var}[V] = E\left(\text{Var}\left[I_{N \geq 0} \sum_{i=1}^N X_i \mid N\right]\right) + \text{Var}\left(E\left[I_{N \geq 0} \sum_{i=1}^N X_i \mid N\right]\right)$$

$$= \sum_{k=1}^{\infty} k \lambda \frac{\mu^k e^{-\mu}}{k!} + \text{Var}[\mu \lambda] = \mu \lambda + \mu \lambda^2 = \mu \lambda(1 + \lambda)$$

$$1b.i) N_k \sim \text{pois}(k), \quad X_1^{(k)}, X_2^{(k)}, \dots \stackrel{iid}{\sim} \text{pois}(h/k)$$

$$U_k = I_{N_k \geq 0} \sum_{i=1}^N X_i^{(k)}$$

$$EV_k = E\left(E\left(I_{N_k \geq 0} \sum_{i=1}^N X_i^{(k)} \mid N_k\right)\right) = E\left[N_k \cdot \frac{h}{k}\right] = h$$

$$\text{Var}[U_k] = E\left(\text{Var}\left[I_{N_k \geq 0} \sum_{i=1}^N X_i^{(k)} \mid N\right]\right) + \text{Var}\left(E\left[I_{N_k \geq 0} \sum_{i=1}^N X_i^{(k)} \mid N\right]\right)$$

$$= E\left[N_k \frac{h}{k}\right] + \text{Var}\left[N_k \frac{h}{k}\right] = h + \frac{h^2}{k^2} k = h + \frac{h^2}{k} = \frac{h(k+h)}{k}$$

Thus,

$$\lim_{k \rightarrow \infty} EV_k = h, \quad \lim_{k \rightarrow \infty} \text{Var}[U_k] = h$$

$$\text{b.ii)} \quad P(X_i^{(k)} \neq D_i^{(k)}) = 1 - P(X_i^{(k)} = D_i^{(k)}) = 1 - [P(X_i^{(k)} = 0) + P(X_i^{(k)} = 1)]$$

$$= 1 - \frac{\lambda_k^0 e^{-\lambda_k}}{0!} - \frac{\lambda_k^1 e^{-\lambda_k}}{1!} = 1 - e^{-\lambda_k} - \lambda_k e^{-\lambda_k}$$

$$e^{-\lambda_k} = \sum_{p=0}^{\infty} \frac{(-\lambda_k)^p}{p!} = 1 - \lambda_k + \frac{\lambda_k^2}{2} - \frac{\lambda_k^3}{6} + \dots$$

$$= 1 - [1 - \lambda_k + o(\lambda_k)] - \lambda_k [1 - \lambda_k + o(\lambda_k)] = \lambda_k^2 + o(\lambda_k) + o(\lambda_k^2)$$

$$= \lambda_k^2 + o(\lambda_k) = \lambda_k^2 (1 + o(\lambda_k^{-1}))$$

↙ This term doesn't match

Proof by contr.
if it were possible,

1b.iii) We notice that $\text{LHS} > \text{RHS} \Rightarrow I(U \neq T) = 1$. Now $\{U \neq T\} = \{N > 0\} \cap \left\{ \bigcup_{i=1}^N X_i \neq D_i \right\}$. Let Z be the r.v. denoting the value of the RHS. But for any $\omega \in \{N > 0\} \cap \left\{ \bigcup_{i=1}^N X_i \neq D_i \right\}$, $Z(\omega) \geq 1$ so a contradiction is reached.

Next,

$$P(|U-T| > \epsilon) \leq P(U \neq T) \leq P(Z \geq 1) = 1 - P(Z=0)$$

(3)

* Use iterated expectation;
cleaner

now,

$$P(Z=0) = P(N=0) + E[P(Z=0 | N=n \geq 1)]$$

$$P(Z=0 | N=n \geq 1) = P(X_1=D_1, \dots, X_n=D_n | N=n \geq 1)$$

$$= [P(X_1=D_1)]^N = [1 - P(X_1 \neq D_1)]^N = [1 - \lambda_k^2 + o(\lambda_k)]^N$$

$$= P(N=0) + E[(1 - \lambda_k^2 + o(\lambda_k))^N | N \geq 1]$$

$$= e^{-\mu_k} + \sum_{x=1}^{\infty} [1 - \lambda_k^2 + o(\lambda_k)]^x \frac{\mu_k^x e^{-\mu_k}}{x!}$$

$$= \sum_{x=0}^{\infty} [1 - \lambda_k^2 + o(\lambda_k)]^x \frac{\mu_k^x e^{-\mu_k}}{x!} = e^{-\mu_k} \exp \{ [1 - \lambda_k^2 + o(\lambda_k)] \mu_k \}$$

$$= \exp \left\{ -\frac{\lambda^2}{k} + o(1) \right\} \rightarrow 1 \text{ as } k \rightarrow \infty$$

Thus

$$P(|U-T| > \varepsilon) \leq 1 - P(Z=0) \rightarrow 0 \text{ as } k \rightarrow \infty$$

$$I[T - \sum_{i=1}^k D_i > 0] = I[\bigcup_{i \in \{k, N\}} \{X_i \neq 0\}]$$

$$\Rightarrow P(T - \varepsilon D_i > 0) \leq E[N-k] P(\lambda_i > 0)$$

1b.iv) $E[T - \sum_{i=1}^k D_i] = 0$

$$P\left(\left|T - \sum_{i=1}^k D_i\right| > \varepsilon\right) \leq \frac{\text{Var}[T - \sum_{i=1}^k D_i]}{\varepsilon^2}$$

$$= \frac{1}{\varepsilon^2} \text{Var}\left[\sum_{i=1}^{\max(N, k)} D_i\right] \leq \frac{1}{\varepsilon^2} \text{Var}\left[\sum_{i=1}^{\max(N, k)} D_i\right]$$

$$= \frac{1}{\varepsilon^2} \left\{ E \left(\text{Var} \left[\sum_{i=1}^{N-k} D_i \mid N \right] \right) + \text{Var} \left(E \left[\sum_{i=1}^{N-k} D_i \mid N \right] \right) \right\}$$

$$= \frac{1}{\varepsilon^2} \left\{ E \left(|N-k| \text{Var}[D_1] \right) + \text{Var} \left(|N-k| ED_1 \right) \right\}$$

$$ED_1 = P(X_1 = 1) = \lambda_k e^{-\lambda_k}, \quad \text{Var}[D_1] = ED_1^2 - (ED_1)^2$$

$$= ED_1 - (ED_1)^2 = ED_1(1 - ED_1) = \lambda_k e^{-\lambda_k}(1 - \lambda_k e^{-\lambda_k})$$

$$= \frac{1}{\varepsilon^2} \left\{ \text{Var}[D_1] E[|N-k|] + (ED_1)^2 \text{Var}[|N-k|] \right\}$$

$$\leq \frac{1}{\varepsilon^2} \left\{ \text{Var}[D_1] E[|N|+|k|] + (ED_1)^2 \left(E[(N-k)^2] - (E[|N-k|])^2 \right) \right\}$$

$$\leq \frac{1}{\varepsilon^2} \left\{ \text{Var}[D_1] E[|N|+|k|] + (ED_1)^2 E[(N-k)^2] \right\}$$

$$= \frac{1}{\varepsilon^2} \left\{ 2\text{Var}[D_1] EN + (ED_1)^2 (EN^2 - (EN)^2) \right\}$$

$$= \frac{1}{\varepsilon^2} \left\{ 2\lambda_k e^{-\lambda_k} (1 - \lambda_k e^{-\lambda_k}) M_k + \lambda_k^2 e^{-2\lambda_k} M_k \right\} \quad \begin{matrix} \text{Treated } D_i \text{ as} \\ \text{Bernoulli} \end{matrix}$$

$$= \frac{1}{\varepsilon^2} \left\{ 2\lambda_k e^{-\lambda_k} (1 - \lambda_k e^{-\lambda_k}) + \frac{\lambda_k^2}{k} e^{-\lambda_k} \right\} \rightarrow 0$$

1b.v) $D_i \sim \text{bern}(\lambda_k e^{-\lambda_k})$ so that $\sum_{i=1}^k D_i \sim \text{bin}(k, \lambda_k e^{-\lambda_k})$

Then

$$M_{Y_k}(t) = [1 - \lambda_k e^{-\lambda_k} + \lambda_k e^{-\lambda_k} e^t]^k$$

$$= \left[1 + \frac{1}{k} h(-e^{-h/k} + e^{-h/k} e^t) \right]^k$$

$$\rightarrow \exp \left\{ 1 + \lim_{k \rightarrow \infty} h(-e^{-h/k} + e^{-h/k} e^t) \right\}$$

$$= \exp \{ h(e^t - 1) \}$$

1c.i) $EU = E(E[U|N]) = E[Nk] = h$

$$\text{Var}[U] = E(\text{Var}[U|N]) + \text{Var}(E[U|N])$$

$$= E[Nk] + \text{Var}[Nk] = h + k^2 \cdot \frac{h}{k} = h + hk = h(1+k)$$

Then take $h \rightarrow \infty$

1c.ii) $P(U=0) = E[P(U=0|N)] = E[(P(X_1=0))^N]$

$$= E[e^{-N\lambda_k}] = \sum_{x=0}^{\infty} e^{-x\lambda_k} \cdot \frac{\lambda_k^x e^{-\lambda_k}}{x!} = \sum_{x=0}^{\infty} \frac{(e^{-\lambda_k} \lambda_k)^x e^{-\lambda_k}}{x!}$$

$$= e^{-\lambda_k} \exp \{ e^{-\lambda_k} \lambda_k \} = e^{-h/k} \exp \{ e^{-k} \cdot \frac{h}{k} \} \rightarrow 1$$