## (2) Sum of two independent distributions

(a) Let  $X \sim \mathcal{P}(\lambda_1)$  and  $Y \sim \mathcal{P}(\lambda_2)$  be two independent Poisson random variables.

$$X + Y \sim \mathcal{P}(\lambda_1 + \lambda_2).$$

(b) Let U and V be two independent random variables with exponential distribution  $\exp(\lambda)$ . Show that

$$U + V \sim Gamma(2, \lambda)$$
 and  $\min\{U, V\} \sim \exp(2\lambda)$ .

*Hint*: It is useful to use moment generating functions. Recall, the pdf of a random variable  $X \sim Gamma(\alpha, \beta)$  is

$$f(x) = \begin{cases} \frac{x^{\alpha - 1}e^{-\frac{x}{\beta}}}{\Gamma(\alpha)\beta^{\alpha}} & x > 0\\ 0, & x \le 0 \end{cases},$$

and its mgf is of the form  $\left(\frac{1}{1-\beta t}\right)^{\alpha}$  for  $t<\frac{1}{\beta}$ . Particularly, the pdf of a random variable  $X\sim \exp(\lambda)=Gamma(1,\frac{1}{\lambda})$  is of the form

$$f(x) = \begin{cases} \lambda e^{-\lambda x}, & x \ge 0 \\ 0, & x \le 0 \end{cases}.$$

a) 
$$M_{\chi}(t) = \mathbb{E}(e^{t\chi}) = e^{-\lambda_{1}} \sum_{n=0}^{\infty} e^{tn} \frac{\lambda_{n}^{k}}{n!} = e^{-\lambda_{1}} \sum_{n=0}^{\infty} \frac{(\lambda_{1}e^{t})^{k}}{k!} = e^{-\lambda_{1}} \frac{\lambda_{2}e^{t}}{k!} = e^{-\lambda_{1}} e^{\lambda_{1}e^{t}} = e^{\lambda_{1}(e^{t}-1)}$$

$$M_{\chi}(t) M_{\gamma}(t) = e^{(\lambda_{1}+\lambda_{1})(e^{t}-1)} = M_{\chi+\gamma}(t)$$

b) 
$$M_{u}(t) = M_{v}(t) = \frac{1}{1-\lambda t}$$
,  $M_{u+v}(t) = M_{u}(t)M_{v}(t) = \frac{1}{(1-\lambda t)^{2}}$   
Hence,  $V+V\sim$  Gamma  $(2, \lambda)$ ,  $uxev$  exp $(\lambda) = Gamma(1, \lambda).!.!$   
 $uxM pdf f(x) = \frac{\lambda^{\alpha}}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x}$ ,  $O(x cooling)$