Lecture 27

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Recap:

$$S = X_1 + X_2 + \dots + X_N$$
$$M_S(t) = M_N(\ln M_X(t))$$

If $S \sim \text{Compound Poisson}(\lambda)$, then $M_S(t) = e^{\lambda(M_X(t)-1)}$.

Recall Composition Theorem, $N_j \sim \mathcal{P}(\lambda_j)$, where $j = 1, \dots, m$, all independent,

$$\implies \sum_{j=1}^{m} N_j \sim \mathcal{P}\left(\sum_{j=1}^{m} \lambda_j\right)$$

Defn: General Composition Theorem:

Suppose $S_j \sim \text{Compound Poisson}(\lambda_j)$ (all independent) with claim size cdf F_j . Then,

$$S = \sum_{i=1}^{m} S_i$$

is Compound Poisson($\sum_{j=1}^{m} \lambda_j$) with claim size cdf

$$\sum_{j=1}^{m} \left(\frac{\lambda_j}{\sum_{i=1}^{m} \lambda_i} \right) F_j = \frac{\sum_{j=1}^{m} \lambda_j F_j}{\sum_{j=1}^{m} \lambda_j}$$

$$M_{S}(t) = \prod_{j=1}^{m} M_{S_{j}}(t)$$

$$= \prod_{j=1}^{m} e^{\lambda_{j}(M_{j}(t)-1)}$$

$$= e^{\sum_{j=1}^{m} (\lambda_{j}(M_{j}(t)-1))}$$

$$= e^{(\sum_{j=1}^{m} \lambda_{j}) \left(\frac{\sum_{j=1}^{m} \lambda_{j} M_{j}(t)}{\sum_{j=1}^{m} \lambda_{j}} - 1\right)}$$

(write M_j as the mgf corresponding to F_j)

 $\implies S \sim C \mathcal{P}\left(\sum_{j=1}^{m} \lambda_j\right)$ with X's cdf.

$$F = \frac{\sum_{j=1}^{m} \lambda_j M_j(t)}{\sum_{j=1}^{m} \lambda_j}$$

Aside: Suppose X has cdf. Then,

$$\begin{split} M_X(t) &= \mathbb{E}(e^{Xt}) \\ &= \int_{-\infty}^{\infty} e^{xt} \, dF(x) \\ &= \int_{-\infty}^{\infty} e^{xt} \frac{\sum_{j=1}^{m} \lambda_j \, dF_j(x)}{\sum_{j=1}^{m} \lambda_j} \\ &= \frac{\sum_{j=1}^{m} \lambda_j \int_{-\infty}^{\infty} e^{xt} \, dF_j(x)}{\sum_{j=1}^{m} \lambda_j} \\ &= \frac{\sum_{j=1}^{m} \lambda_j M_j(t)}{\sum_{j=1}^{m} \lambda_j} \end{split}$$

E.g.: Suppose $S_A \sim C\mathcal{P}(2)$ r.v. with claim size distribution

$$\mathbb{P}(X_A = 1) = 0.6 = 1 - \mathbb{P}(X_A = 2),$$

and suppose $S_B \sim C\mathcal{P}(1)$ r.v. with claim size distribution

$$\mathbb{P}(X_B = 1) = 0.7 = 1 - P(X_B = 3).$$

(a) Compute the mgf of $S = S_A + S_B$.

$$M_{S_A}(t) = e^{\lambda_A(M_A(t)-1)}$$

$$= e^{2(0.6e^t + 0.4e^{2t} - 1)}$$

$$= e^{1.2e^t + 0.8e^{2t} - 2}$$

$$M_{S_B}(t) = e^{0.7e^t + 0.3e^{3t} - 1}$$

$$M_S(t) = e^{1.9e^t + 0.8e^{2t} + 0.3e^{3t} - 3}$$
$$= e^{3(\frac{19}{30}e^t + \frac{4}{15}e^{2t} + \frac{1}{10}e^{3t} - 1)}$$

(b) State S's distribution.

 $S \sim C \mathcal{P}(3)$ with claim size distribution $\mathbb{P}(X=1) = \frac{19}{30}$, $\mathbb{P}(X=2) = \frac{4}{15}$, $\mathbb{P}(X=3) = \frac{1}{10}$.

$$\mathbb{P}(X = 1) = \frac{2}{3} \cdot \mathbb{P}(X_A = 1) + \frac{1}{3} \cdot \mathbb{P}(X_B = 1)$$
$$= \frac{2}{3}(0.6) + \frac{1}{3}(0.7) = \frac{19}{30}$$

$$\mathbb{P}(X=2) = \frac{2}{3} \cdot \mathbb{P}(X_A = 2) + \frac{1}{3} \cdot \mathbb{P}(X_B = 2)$$
$$= \frac{2}{3}(0.4) + \frac{1}{3} \cdot 0 = \frac{4}{15}$$

$$\begin{split} M_{S_A}(t) &= e^{\lambda_A(M_A(t)-1)} \\ &= e^{2(0.6e^t + 0.4e^{2t} - 1)} \\ &= e^{1.2e^t + 0.8e^{2t} - 2} \end{split}$$

$$M_{S_B}(t) = e^{0.7e^t + 0.3e^{3t} - 1}$$

$$M_S(t) = e^{1.9e^t + 0.8e^{2t} + 0.3e^{3t} - 3}$$
$$= e^{3(\frac{19}{30}e^t + \frac{4}{15}e^{2t} + \frac{1}{10}e^{3t} - 1)}$$

(c) Write $S=1\cdot N_1+2\cdot N_2+3\cdot N_3$ and compute $\mathbb{P}(S=x),\,x=0,1,2,3$

$$N_1$$
 = the number of size 1

$$\sim \mathcal{P}\left(3\cdot\frac{19}{30}\right) = \mathcal{P}\left(\frac{19}{10}\right)$$

 N_2 = the number of size 2

$$\sim \mathcal{P}\left(3 \cdot \frac{4}{15}\right) = \mathcal{P}\left(\frac{4}{5}\right) = \mathcal{P}(0.8)$$

 N_3 = the number of size 3

$$\sim \mathcal{P}\left(3 \cdot \frac{1}{10}\right) = \mathcal{P}(0.3)$$

$$S_A \sim C \mathcal{P}(1), \qquad \mathbb{P}(X_A = 1) = 1$$

$$S_B \sim C \mathcal{P}(3), \qquad \mathbb{P}(X_B = 1) = \frac{1}{2} = 3\mathbb{P}(X_B = 3)$$

$$\mathbb{P}(S=3) = \mathbb{P}(S_A=0)\mathbb{P}(S_B=3) + \dots$$

$$\mathbb{P}(S_B = 3) = \mathbb{P}(N_B = 1, X_B = 3) + \mathbb{P}(N_B = 3)(\mathbb{P}(X_B = 1))^3$$