Model of Super Spreading Events

Hannah Craddock

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Model

Let $x_t = y_t + z_t$

where

 $y_t = \text{Number of infecteds not by a super-spreading event (NSSE)}$

 $z_t = x_t - y_t =$ Number of infecteds infected by a super-spreading event (SSE)

Probability

$$P(x_t) = \sum_{y_t=0}^{x_t} P(y_t) \cdot P(z_t)$$

i. y_t - Non super-spreading events

 $y_t \sim Poisson(\alpha \cdot \lambda_t),$

$$p(y_t) = exp(-\alpha \cdot \lambda_t) \cdot \frac{1}{y_t!} \cdot (\alpha \cdot \lambda_t)^{y_t},$$

with

$$\lambda_t = \sum_{i=1}^{t-1} y_i \ (Gamma((t-i); k, \theta) - Gamma((t-i-1); k, \theta)$$

ii. z_t - super-spreading events

For each SSE event we get $Poisson(\gamma)$ infections

 $z_t|n_t \sim Poisson(\gamma \cdot n_t)$ = Number of infecteds by all Super-Spreading Events at time t

 $n_t \sim Poisson(\beta \cdot \lambda_t) = \text{Number of super-spreading events}$

$$p(z_t) = \sum_{n_t=0}^{\infty} p(n_t = n_t) \cdot p(z_t | n_t = n_t)$$

 $\therefore p(z_t)$ is a Poisson-Poisson Compound

Determining the Mean and Variance of z_t

$$\mathbb{E}(z_t) = \mathbb{E}(\mathbb{E}(z_t|n_t))$$

$$= \mathbb{E}(\gamma \cdot n_t)$$

$$= \gamma \mathbb{E}(n_t)$$

$$= \gamma \cdot \beta \cdot \lambda_t$$

 $\operatorname{Var}(\mathbf{z}_t) = \operatorname{Var}(\mathbb{E}(z_t|n_t)) + \mathbb{E}(\operatorname{Var}(z_t|n_t))$

As $z_t|n_t$ and n_t are poisson rvs the mean and variance are equivalent giving;

$$= \operatorname{Var}(\gamma \cdot n_t) + \mathbb{E}(\gamma \cdot n_t)$$
$$= \gamma^2 \cdot \beta \cdot \lambda_t + \gamma \cdot \beta \cdot \lambda_t$$

Poisson-Poisson Compound z_t in the form of a Negative Binomial RV

The mean and variance of the Poisson-Poisson Compound z_t is equated to that of a Negative Binomial distribution of size n with probability of success p and density given by;

$$\frac{\Gamma(x+n)}{\Gamma(n)\cdot x!}\cdot p^n(1-p)^x$$

for
$$x = 0, 1, 2, ..., n > 0$$
 and $0 .$

This represents the number of failures which occur in a sequence of Bernoulli trials before a target number of successes is reached. The mean and variance are respectfully;

$$\mu = \frac{n(1-p)}{p}, \qquad Var = \frac{n(1-p)}{p^2}$$

Equating the poisson-poisson compound to that of the negative binomial distribution gives;

$$\mathbf{I.} \ \mu = \frac{n(1-p)}{p} = \gamma \cdot \beta \cdot \lambda_t$$

II
$$Var = \frac{n(1-p)}{p^2} = \gamma^2 \cdot \beta \cdot \lambda_t + \gamma \cdot \beta \cdot \lambda_t$$

From I we can say that the size

$$n = \frac{p \cdot \gamma \cdot \beta \cdot \lambda_t}{1 - n} \qquad \text{III.}$$

Substituting III into II gives;

$$\frac{(\frac{p \cdot \gamma \cdot \beta \cdot \lambda_t}{1 - p}) \cdot (1 - p)}{p^2} = \gamma^2 \cdot \beta \cdot \lambda_t + \gamma \cdot \beta \cdot \lambda_t$$

Cancelling like terms gives;

$$\frac{(\cancel{p}\cdot\cancel{\gamma}\cdot\cancel{\beta}\cdot\cancel{\lambda_t})\cdot(1-p)}{p^{\cancel{t}}} = \gamma^{\cancel{t}}\cdot\cancel{\beta}\cdot\cancel{\lambda_t} + \gamma\cdot\cancel{\beta}\cdot\cancel{\lambda_t}$$

$$\frac{1}{n} = \gamma + 1$$

$$1 = p \cdot \gamma + p$$

$$1=p\cdot \gamma + p$$

$$\therefore p = \frac{1}{\gamma + 1} \quad \textbf{IV.}$$

Substituting IV into III gives;

$$n = \frac{\frac{1}{\gamma + 1} \cdot \gamma \cdot \beta \cdot \lambda_t}{1 - \frac{1}{\gamma + 1}}$$

$$n = \frac{\frac{\gamma \cdot \beta \cdot \lambda_t}{\gamma + 1}}{\frac{\gamma + 1 - 1}{\gamma + 1}}$$

$$n = \frac{\frac{\cancel{\gamma} \cdot \beta \cdot \lambda_t}{\cancel{\gamma} + 1}}{\frac{\cancel{\gamma}}{\cancel{\gamma} + 1}}$$

$$\therefore n = \beta \cdot \lambda_t$$

 z_t can thus be written as a Negative Binomial of size n with probability of success p, i.e

$$z_t \sim \text{NB}(\beta \cdot \lambda_t, \frac{1}{\gamma + 1})$$

$$p(z_t) = \frac{\Gamma(x+\beta \cdot \lambda_t)}{\Gamma(\beta \cdot \lambda_t) \cdot z_t!} \cdot \frac{1}{\gamma+1}^{\beta \cdot \lambda_t} \cdot (1 - \frac{1}{\gamma+1})^{z_t} \qquad \text{for } z_t = 0, 1, 2, ..., n > 0 \text{ and } 0$$

Likelihood

$$\begin{split} P(x_t) &= \sum_{y_t=0}^{x_t} P(y_t) \cdot P(z_t) \\ L(\alpha, \beta, \gamma, \lambda | \ \mathbf{x_t}) &= \sum_{y_t=0}^{x_t} \prod_{t=1}^{N days} P(y_t) \cdot P(z_t) \\ &= \sum_{y_t=0}^{x_t} \prod_{t=1}^{N days} exp(-\alpha \cdot \lambda_t) \cdot \frac{1}{y_t!} \cdot (\alpha \cdot \lambda_t)^{y_t} \times \frac{\Gamma(z_t + \beta \cdot \lambda_t)}{\Gamma(\beta \cdot \lambda_t) \cdot z_t!} \cdot \frac{1}{\gamma + 1} \cdot (1 - \frac{1}{\gamma + 1})^{z_t} \\ &= \sum_{y_t=0}^{x_t} \prod_{t=1}^{N days} exp(-\alpha \cdot \lambda_t) \cdot \frac{1}{y_t!} \cdot (\alpha \cdot \lambda_t)^{y_t} \times \frac{\Gamma(z_t + \beta \cdot \lambda_t)}{\Gamma(\beta \cdot \lambda_t) \cdot z_t!} \cdot \frac{1}{\gamma + 1} \cdot (\frac{\gamma}{\gamma + 1})^{z_t} \end{split}$$

Log Likelihood

$$l(\alpha, \beta, \gamma, \lambda | \mathbf{x_t}) = \sum_{y_t=0}^{x_t} \prod_{t=1}^{Ndays} ln\left(P(y_t) \cdot P(z_t)\right)$$

$$= \sum_{y_t=0}^{x_t} \prod_{t=1}^{Ndays} ln\left(exp(-\alpha \cdot \lambda_t) \cdot \frac{1}{y_t!} \cdot (\alpha \cdot \lambda_t)^{y_t} \times \frac{\Gamma(z_t + \beta \cdot \lambda_t)}{\Gamma(\beta \cdot \lambda_t) \cdot z_t!} \cdot \frac{1}{\gamma + 1}^{\beta \cdot \lambda_t} \cdot (\frac{\gamma}{\gamma + 1})^{z_t}\right)$$

Using the fact that the logarithm of a product is the sum of the logarithms gives;

$$\begin{split} l(\alpha,\beta,\gamma,\lambda|\ \mathbf{x_t}) &= \sum_{y_t=0}^{x_t} \left(-\sum_{t=1}^{Ndays} \alpha \cdot \lambda_t \right. \\ &+ \sum_{t=1}^{Ndays} ln(\frac{1}{y_t!}) \right. \\ &+ \sum_{t=1}^{Ndays} ln(y_t \cdot \alpha \cdot \lambda_t) \right. \\ &+ \sum_{y_t=0}^{Ndays} \beta \cdot \lambda_t \cdot ln(\frac{1}{\gamma+1}) \right. \\ &+ \sum_{t=1}^{Ndays} z_t \cdot ln(\frac{\gamma}{\gamma+1}) \bigg) \end{split}$$

Simplifying this gives;

$$\begin{split} l(\alpha,\beta,\gamma,\lambda_t|\ \mathbf{x_t}) &= \sum_{y_t=0}^{x_t} \left(-\alpha \sum_{t=1}^{Ndays} \lambda_t \ + \sum_{t=1}^{Ndays} ln(\frac{1}{y_t!}) \ + \sum_{t=1}^{Ndays} ln(y_t \cdot \alpha \cdot \lambda_t) \ + \sum_{t=1}^{Ndays} ln(\frac{\Gamma(z_t + \beta \cdot \lambda_t)}{\Gamma(\beta \cdot \lambda_t) \cdot z_t!}) \right) \\ &+ \ \beta \cdot ln(\frac{1}{\gamma+1}) \sum_{t=1}^{Ndays} \lambda_t \ + \ ln(\frac{\gamma}{\gamma+1}) \sum_{t=1}^{Ndays} z_t \right) \end{split}$$