

Problem 3

Poisson processes with unobs. heterog.

- a) Intensity:  $\lambda(t)dt = P(dN(t) = 1 | \mathcal{F}_{t-})$   
 i.e., the probability of having an event in a small interval  $[t, t+dt)$ , given the history up to and not including  $t$ .

For  $N_i(t)$ , events can only occur when  $t \leq \tau_i$ , but otherwise at rate  $\theta_i$  at each time.

i.e.

$$\lambda_i(t) dt = I(t \leq \tau_i) \theta_i dt$$

$$\text{so } \lambda_i(t) = I(t \leq \tau_i) \theta_i$$

For general formula:

$\lambda_i(t; \theta)$  equals  $I(t \leq \tau_i) \theta_i$  at each of the  $N_i(\tau_i)$  jumps of the process (i.e., when  $\Delta N_i(t) = 1$ ).

This gives the first factor,  $\theta_i^{N_i(\tau_i)}$

The second factor of ABG (5.4) has a factor  $e^{-\int \lambda_i(t) dt}$  from each process. From  $N_i(t)$  it is

$$e^{-\int_0^{\tau_i} \lambda_i(t) dt} = e^{-\int_0^{\tau_i} I(t \leq \tau_i) \theta_i dt} = \underline{e^{-\theta_i \tau_i}}$$

This completes the result



b) Given  $z_i$ , the likelihood contribution is  
 - by (a) -

$$(z_i; \theta) = N_i(\tau_i) e^{-z_i \theta \tau_i}$$

$$= \theta^{N_i(\tau_i)} z_i^{N_i(\tau_i)} e^{-z_i \theta \tau_i}$$

Taking expectation we get the unconditional contribution:

$$\theta^{N_i(\tau_i)} E \left[ z_i^{N_i(\tau_i)} e^{-z_i \theta \tau_i} \right]$$

$$= \theta^{N_i(\tau_i)} \cdot \frac{\Gamma(N_i(\tau_i) + \frac{1}{\delta})}{\Gamma(\frac{1}{\delta})} \cdot \frac{\left(\frac{1}{\delta}\right)^{\frac{1}{\delta}}}{\left(\theta \tau_i + \frac{1}{\delta}\right)^{N_i(\tau_i) + \frac{1}{\delta}}}$$

Use Appendix (last formula) with

$$a = N_i(\tau_i)$$

$$b = \theta \tau_i$$

$$k = \gamma = \frac{1}{\delta}$$

c) The likelihood for all the data is a product of the above for  $i=1, \dots, n$ , i.e.

$$\theta^N \prod_{i=1}^n \frac{\Gamma(N_i(\tau_i) + \frac{1}{\delta})}{\Gamma(\frac{1}{\delta})} \cdot \frac{1}{\delta^{\frac{n}{\delta}} \prod_{i=1}^n \left(\theta \tau_i + \frac{1}{\delta}\right)^{N_i(\tau_i) + \frac{1}{\delta}}}$$

Taking the log, we get the announced  $l(\theta, \delta)$

$$u_1(\theta, \delta) = \frac{\partial}{\partial \theta} l(\theta, \delta) = \frac{N}{\theta} - \sum_{i=1}^n \left(N_i(\tau_i) + \frac{1}{\delta}\right) \frac{\tau_i}{\theta \tau_i + \frac{1}{\delta}}$$



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If all  $\tau_i = \tau$ :

$$u(\theta, \delta) = \frac{N}{\theta} - \ln \left[ \sum_{i=1}^n \left( N_i(\tau) + \frac{1}{\delta} \right) \cdot \frac{\tau}{\theta\tau + \frac{1}{\delta}} \right]$$
$$= \frac{N}{\theta} - \left( N + \frac{n}{\delta} \right) \cdot \frac{\tau}{\theta\tau + \frac{1}{\delta}}$$

Set equal 0:

$$\frac{N}{\theta} = \frac{\tau \left( N + \frac{n}{\delta} \right)}{\theta\tau + \frac{1}{\delta}}$$

~~$$N\theta\tau + \frac{N}{\delta} = \theta\tau N + \theta\tau \cdot \frac{n}{\delta}$$~~

$$N = \theta\tau n$$

$$\underline{\underline{\hat{\theta} = \frac{N}{\tau n}}} = \frac{\text{\# events}}{\text{total time}} = \frac{\text{occurrences}}{\text{exposure}}$$

By inserting  $\hat{\theta} = \frac{N}{\tau n}$  in the log likelihood, we get a profile likelihood for  $\delta$ , which can be maximized (where the maximum may be at the value 0 for  $\delta$ ).



d) Usual procedure is

$$f(z_i | \text{data}) \propto f(\text{data} | z_i) f(z_i)$$

Here

$$f(\text{data} | z_i) \stackrel{(*)}{=} \theta(z_i; \theta) N_i(\tau_i) e^{-z_i \theta \tau_i}$$

from a)  
with  $\theta_i = z_i \theta$

$$f(z_i) = \frac{\left(\frac{1}{\delta}\right)^{\frac{1}{\delta}}}{\Gamma\left(\frac{1}{\delta}\right)} z_i^{\frac{1}{\delta}-1} e^{-\frac{1}{\delta} z_i}$$

Appendix

Multiply and collect terms with  $z_i$

$$\propto z_i^{N_i(\tau_i) + \frac{1}{\delta} - 1} e^{-z_i(\theta z_i + \frac{1}{\delta})}$$

$$\stackrel{(*)}{\sim} \text{gamma}\left(N_i(\tau_i) + \frac{1}{\delta}, \theta z_i + \frac{1}{\delta}\right)$$

by appendix

which proves (\*)

The recovered  $z_i$  are then computed as the expected value of this, where ML estimates for  $\theta, \delta$  are substituted.

$$\hat{z}_i = \frac{N_i(\tau_i) + \frac{1}{\delta}}{\theta \hat{z}_i + \frac{1}{\delta}}$$

