

We consider modeling for a "crater-counting" data set consisting of

- N_{2+} locations, sizes and records of exactly which 2 or more of J analysts counted each of the corresponding craters, and
- for $j = 1, 2, \dots, J$, a number S_1^j of locations and sizes of possible craters counted only by analyst j , each of which may be either a real crater (missed by all other analysts) or a "phantom" non-crater seen only by analyst j (we'll call the number of actual craters in this group N_1^j and write $F_j = S_1^j - N_1^j$ for the number of phantoms seen by analyst j).

We'll do this in a way that is consistent with independent HPP models for locations of craters and phantoms across an area A , in particular with locations of real craters having intensity ρ and phantoms from analyst j having intensity ρ_j for $j = 1, 2, \dots, J$. We'll further allow that there it is possible for analysts to miss craters in their counting and that there are N_0 craters missed by all in the counting (and that ultimately it is $N = N_0 + N_{2+} + \sum_{j=1}^J N_1^j$ that is Poisson with mean ρA).

So, suppose that analyst j misses a real crater of size s in his/her counting with probability

$$p(s | \gamma_j)$$

(In more complex modeling, this probability could be a function of other things, like local conditions around the crater center under consideration perhaps specified in terms of a number of other centers within some distance, and perhaps an overall density/number of real craters.) Exactly what form to use for this is open to discussion. It could be something as simple as

$$p(s | \gamma_1, \gamma_2) = I[s < \gamma_1] + \gamma_2 I[s \geq \gamma_1]$$

a step function taking only the value 1 at and near 0 and some positive value γ_2 to the right of a cut point γ_1 . Another possible form that looks like it might be tractable is

$$p(s | \gamma_1, \gamma_2, \gamma_3) = I[s < \gamma_1] + I[s \geq \gamma_1] \left(\gamma_2 + (1 - \gamma_2) \exp\left(-\frac{s - \gamma_1}{\gamma_3}\right) \right)$$

This function is 1 at and below the threshold $s = \gamma_1$, it has limit γ_2 as $s \rightarrow \infty$, to which it decreases in exponential fashion with a rate parameter γ_3 .

Next, suppose that real craters have sizes that are iid with marginal density

$$f(s | \boldsymbol{\theta})$$

Further, suppose that phantom craters for analyst j have recorded/perceived sizes that are iid with marginal density

$$h(s | \boldsymbol{\eta}_j)$$

Plots in the Robbins paper seem to make values of a cdf (or 1 minus the cdf) for crater size look roughly linear over some interval on log-log scales. If a continuous cdf has

$$\ln F(s | \theta_0, \theta_1) = (\theta_0 + \theta_1 \ln s) I[\theta_0 + \theta_1 \ln s < 0]$$

then

$$F(s | \theta_0, \theta_1) = I[\theta_0 + \theta_1 \ln s < 0] \exp(\theta_0 + \theta_1 \ln s) + I[\theta_0 + \theta_1 \ln s \geq 0]$$

and

$$\begin{aligned} f(s | \theta_0, \theta_1) &= I[\theta_0 + \theta_1 \ln s < 0] \left(\frac{\theta_1}{s} \right) \exp(\theta_0 + \theta_1 \ln s) \\ &= \theta_1 \exp(\theta_0) s^{\theta_1 - 1} I \left[0 < s < \exp \left(-\frac{\theta_0}{\theta_1} \right) \right] \end{aligned}$$

and perhaps this form for $\theta_1 > 0$ and $\theta_0 < 0$ would work as a marginal pdf of real crater diameters.

A plausible parametric form for the density for phantom diameters is up for discussion. Something as simple as the $U(\eta_1, \eta_2)$ density might be used, as might the kind of form suggested above for real diameters.

In any event, the basic parameters of the modeling are

$$\boldsymbol{\theta}, \gamma_1, \gamma_2, \dots, \gamma_J, \boldsymbol{\eta}_1, \boldsymbol{\eta}_2, \dots, \boldsymbol{\eta}_J, \rho, \rho_1, \rho_2, \dots, \text{ and } \rho_J$$

Various simplified versions of what follows can be had by assuming that some parameter(s) is (are) fixed across j , or are known, etc.

We proceed to consider some basic quantities associated with the mechanisms for generating sizes and observing craters of a given size. In what follows, let

$$\mathbf{I} = (I_1, I_2, \dots, I_J)$$

be a vector of 0's and 1's. Define

$$q^{\mathbf{I}}(s | \gamma_1, \dots, \gamma_J) = \prod_{j=1}^J (1 - p(s | \gamma_j))^{I_j} (p(s | \gamma_j))^{1-I_j}$$

= the probability that a crater of size s is seen by exactly those analysts j with $I_j = 1$

so that in particular, with $\mathbf{0}$ a vector of 0's,

$$q^{\mathbf{0}}(s | \gamma_1, \dots, \gamma_J) = \prod_{j=1}^J p(s | \gamma_j)$$

= the probability that a crater of size s is not seen by any analyst

We'll let

$$q_1^j(s | \gamma_1, \dots, \gamma_J) = (1 - p(s | \gamma_j)) \prod_{j' \neq j} p(s | \gamma_{j'})$$

= the probability that a crater of size s is seen only by analyst j

and abbreviate as

$$q_{2+}(s | \gamma_1, \dots, \gamma_J) = 1 - q^{\mathbf{0}}(s | \gamma_1, \dots, \gamma_J) - \sum_{j=1}^J q_1^j(s | \gamma_1, \dots, \gamma_J)$$

= the probability that a crater of size s is seen by at least 2 analysts

These quantities are obviously all functions of size and depend upon the parameters $\gamma_1, \dots, \gamma_J$ that describe the analyst-specific crater-detection properties.

Versions of all the quantities q averaged across sizes of craters can be defined and depend upon both the parameters $\gamma_1, \dots, \gamma_J$ and upon the parameter $\boldsymbol{\theta}$ that describes the size distribution.

That is, we define

$$q^{\mathbf{I}}(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta}) = \int q^{\mathbf{I}}(s | \gamma_1, \dots, \gamma_J) f(s | \boldsymbol{\theta}) ds$$

= the probability of a crater detection pattern \mathbf{I} across the analysts

and

$$q^{\mathbf{0}}(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta}) = \int q^{\mathbf{0}}(s | \gamma_1, \dots, \gamma_J) f(s | \boldsymbol{\theta}) ds$$

= the probability a crater is seen by no analyst

and

$$q_1^j(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta}) = \int q_1^j(s | \gamma_1, \dots, \gamma_J) f(s | \boldsymbol{\theta}) ds$$

= the probability a crater is seen only by analyst j

and

$$q_{2+}(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta}) = 1 - q^0(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta}) - \sum_{j=1}^J q_1^j(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta})$$

= the probability that a crater is seen by at least 2 analysts

These integrals are useful in defining some important conditional probabilities and conditional densities. In particular, for an \mathbf{I} with at least 2 entries of 1,

$$\frac{q^{\mathbf{I}}(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta})}{q_{2+}(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta})} = \text{the conditional probability of crater-detection pattern } \mathbf{I} \text{ across analysts given that it is seen by at least 2 analysts}$$

Also, the conditional pdf of crater size given detection pattern \mathbf{I} across analysts is

$$\frac{q^{\mathbf{I}}(s | \gamma_1, \dots, \gamma_J) f(s | \boldsymbol{\theta})}{q^{\mathbf{I}}(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta})}$$

and in particular, the conditional pdf of crater size given detection by only analyst j is

$$\frac{q_1^j(s | \gamma_1, \dots, \gamma_J) f(s | \boldsymbol{\theta})}{q_1^j(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta})}$$

Conditional on the parameters $\boldsymbol{\theta}, \gamma_1, \gamma_2, \dots, \gamma_J, \boldsymbol{\eta}_1, \boldsymbol{\eta}_2, \dots, \boldsymbol{\eta}_J, \rho, \rho_1, \rho_2, \dots,$ and ρ_j , we suppose that there are independent random variables

$$N_0 \sim \text{Poisson}(q^0(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta}) \rho A)$$

and

$$N_{2+} \sim \text{Poisson}(q_{2+}(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta}) \rho A)$$

and for $j = 1, 2, \dots, J$

$$S_1^j \sim \text{Poisson}(q_1^j(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta}) \rho A + \rho_j A)$$

Then, for each $j = 1, 2, \dots, J$ suppose that there are latent variables $T_1^j, T_2^j, \dots, T_{S_1^j}^j$ that are (conditional on all before) independent Bernoulli variables with success probabilities

$$\frac{\rho q_1^j(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta})}{\rho q_1^j(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta}) + \rho_j}$$

(indicators of a possible crater being real and not a phantom). Then

$$N_1^j = \sum_{l=1}^{S_1^j} T_l^j \quad \text{and} \quad F^j = S_1^j - N_1^j$$

are independent Poisson variables, N_1^j (the number of real craters seen only by analyst j) with mean $q_1^j(\gamma_1, \dots, \gamma_J, \theta) \rho A$ and F^j (the number of phantoms reported by analyst j) with mean $\rho_j A$. Notice then that

$$N_0 + N_{2+} + \sum_{j=1}^J N_1^j$$

(as a sum of independent Poisson variables) is Poisson with mean

$$q^0(\gamma_1, \dots, \gamma_J, \theta) \rho A + q_{2+}(\gamma_1, \dots, \gamma_J, \theta) \rho A + \sum_{j=1}^J q_1^j(\gamma_1, \dots, \gamma_J, \theta) \rho A = \rho A$$

So we have here a probability structure for the counts that is consistent with the Poisson process assumptions, the crater size generating assumptions, and the analyst crater-detection assumptions. It remains to build the part of the model that will attach patterns of observer

detection **I** to the N_{2+} craters seen by 2 or more analysts and sizes to each of the $N_{2+} + \sum_{j=1}^J S_1^j$ observed potential craters and to the N_0 unobserved craters. This we can do using appropriate conditional probabilities and pdfs.

First, we suppose that conditional on all else, sizes of N_0 unobserved craters are iid according to the pdf

$$\frac{q^0(s | \gamma_1, \dots, \gamma_J) f(s | \theta)}{q^0(\gamma_1, \dots, \gamma_J, \theta)}$$

Then (still conditioned on all parameters and counts) independent of these sizes (independently for all $j = 1, 2, \dots, J$ and $l = 1, 2, \dots, S_1^j$) conditioned on $T_l^j = 1$ suppose that an l th potential crater seen only by analyst j has size from the density

$$\frac{q_1^j(s | \gamma_1, \dots, \gamma_J) f(s | \theta)}{q_1^j(\gamma_1, \dots, \gamma_J, \theta)}$$

while conditioned on $T_l^j = 0$ it has size from the density $h(s | \eta_j)$.

Finally, consider modeling of vectors of crater-detection patterns and sizes for the N_{2+} craters seen by 2 or more analysts. Conditional on the parameters and all else, these pairs can be taken as iid according to the recipe that first \mathbf{I} with at least 2 entries of 1 has pmf given by

$$m(\mathbf{I} | \gamma_1, \dots, \gamma_J, \boldsymbol{\theta}) = \frac{q^{\mathbf{I}}(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta})}{q_{2+}(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta})}$$

and subsequently a size is selected from the density

$$\frac{q^{\mathbf{I}}(s | \gamma_1, \dots, \gamma_J) f(s | \boldsymbol{\theta})}{q^{\mathbf{I}}(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta})}$$

What must be sampled in order to do a Bayes analysis here? Of course, one must begin with a prior of some kind for the model parameters, say

$$g(\boldsymbol{\theta}, \gamma_1, \gamma_2, \dots, \gamma_J, \boldsymbol{\eta}_1, \boldsymbol{\eta}_2, \dots, \boldsymbol{\eta}_J, \rho, \rho_1, \rho_2, \dots, \rho_J)$$

Then, the data and latent variables have "densities" conditional on the parameters that are as follows.

First, there are N_{2+} cases with at least 2 observers, and this has probability

$$\frac{\exp(-q_{2+}(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta}) \rho A) (q_{2+}(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta}) \rho A)^{N_{2+}}}{N_{2+}!}$$

Each such case l then has likelihood

$$\frac{q^{\mathbf{I}}(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta})}{q_{2+}(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta})} \cdot \frac{q^{\mathbf{I}}(s_l | \gamma_1, \dots, \gamma_J) f(s_l | \boldsymbol{\theta})}{q^{\mathbf{I}}(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta})}$$

So ultimately, the contribution to the likelihood from those cases is proportional to

$$\exp(-q_{2+}(\gamma_1, \dots, \gamma_J, \boldsymbol{\theta}) \rho A) \prod_l q^{\mathbf{I}}(s_l | \gamma_1, \dots, \gamma_J) f(s_l | \boldsymbol{\theta})$$

There is also a latent variable N_0 that has pmf

$$\frac{\exp\left(-q^0(\gamma_1, \dots, \gamma_J, \theta) \rho A\right) \left(q^0(\gamma_1, \dots, \gamma_J, \theta) \rho A\right)^{N_0}}{N_0!}$$

and this is the contribution to the "likelihood" from this variable.

Finally, there are S_1^j counts made by observer j alone. This count variable has pmf

$$\frac{\exp\left(-\left(q_1^j(\gamma_1, \dots, \gamma_J, \theta) \rho A + \rho_j A\right)\right) \left(q_1^j(\gamma_1, \dots, \gamma_J, \theta) \rho A + \rho_j A\right)^{S_1^j}}{S_1^j}$$

Each of the S_1^j cases (say, l) then has $T_l^j = 1$ with probability

$$\frac{q_1^j(\gamma_1, \dots, \gamma_J, \theta) \rho}{q_1^j(\gamma_1, \dots, \gamma_J, \theta) \rho + \rho_j}$$

and, of course, $T_l^j = 0$ with probability

$$\frac{\rho_j}{q_1^j(\gamma_1, \dots, \gamma_J, \theta) \rho + \rho_j}$$

Conditioned on $T_l^j = 1$ and all else, s_l has density

$$\frac{q_1^j(s | \gamma_1, \dots, \gamma_J) f(s | \theta)}{q_1^j(\gamma_1, \dots, \gamma_J, \theta)}$$

and conditioned on $T_l^j = 0$ it has density $h(s | \mathbf{n}_j)$. So the contribution to the likelihood for these cases is

$$\exp\left(-\left(q_1^j(\gamma_1, \dots, \gamma_J, \theta) \rho A + \rho_j A\right)\right) \times \prod_l \left(\rho q_1^j(s_l | \gamma_1, \dots, \gamma_J) f(s_l | \theta)\right)^{T_l^j} \times \prod_l \left(\rho_j h(s_l | \mathbf{n}_j)\right)^{1-T_l^j}$$

There is a product of this type for every j included in the likelihood.

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Notice that in an MCMC algorithm, a Gibbs step update for a T_l^j will be made using the conditional probability that $T_l^j = 1$, i.e.

$$\frac{q_l^j(s_l | \gamma_1, \dots, \gamma_J, \boldsymbol{\theta}) \rho f(s_l | \boldsymbol{\theta})}{q_l^j(s_l | \gamma_1, \dots, \gamma_J, \boldsymbol{\theta}) \rho f(s_l | \boldsymbol{\theta}) + \rho_j h(s_l | \boldsymbol{\eta}_j)}$$

and, of course, that $T_l^j = 0$, i.e.

$$\frac{\rho_j h(s_l | \boldsymbol{\eta}_j)}{q_l^j(s_l | \gamma_1, \dots, \gamma_J, \boldsymbol{\theta}) \rho f(s_l | \boldsymbol{\theta}) + \rho_j h(s_l | \boldsymbol{\eta}_j)}$$