

# **The SDM is not the territory: species distribution models must account for finite abundances**

[M.D. Catchen](#)<sup>1,2</sup>

<sup>1</sup> McGill University   <sup>2</sup> Québec Centre for Biodiversity Sciences

**Correspondance to:**

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**Abstract:** A representation of a thing is not the same as that thing.

1 Nature does not prepare distributions, only states.

2 ET Jaynes

3 I would warn you that I do not attribute to nature either beauty or deformity, order  
4 or confusion. Only in relation to our imagination can things be called beautiful or  
5 ugly, well-ordered or confused

6 a common misquote of *Baruch Spinoza*, assembled from translated parts of his  
7 *Ethics (Part I)*

8 Species do not *actually* have distributions. This may seem a radical claim, given the rise of  
9 species distribution modeling as both a field of study and imperative for ecosystem management  
10 over the last several decades. But consider that species are composed of discrete objects—  
11 individual organisms that occupy points in space and which move through time. The location  
12 of every individual organism of a particular species at a particular time is an observable value,  
13 which we could feasibly write down. In most cases the number of individuals of a species  
14 becomes large enough that this is no longer practical.

15 A distribution is not some inherent property of a given species, but a conceptual framework  
16 that we invoke because we know that sampling of species locations is incomplete, and in most  
17 contexts these location of the individuals observed in this sample will change as species move  
18 after they are observed. The goal of a species distribution model (SDM) is instead to take a  
19 set of coordinates of observed occurrence of a species  $\mathbf{O} = \{\vec{o}_1, \vec{o}_2, \dots\}$  and to best describe  
20 a distribution  $D$  such that the true coordinates of the individuals of that species, denoted  $\mathbf{X} =$   
21  $\{\vec{x}_1, \vec{x}_2, \dots\}$  are likely to have been drawn from this distribution  $D$ . Note that typically  $|\mathbf{O}| \ll$   
22  $|\mathbf{X}|$ , as is the reason we don't try to measure the location every individual in the first place  
23 (that being said, for charismatic megafauna that are nearly extinct, this *is* what we do, precisely  
24 because it is feasible). Yet this should not be mistaken for the distribution  $D$  being an inherent  
25 but latent “property” of species.

26 Many approaches have been taken to design SDMs, but almost universally the output of an SDM  
27 is a raster, where the value of each location/cell  $i$ , denoted  $p_i$ , forms a distribution as  $\sum_i p(i) = 1$ .  
28 The value of a cell is often referred to in plain language as “occurrence probability.” But what

is meant by this?—is it the probability conditional on observing an individual that it will be observed at that location? Or is it the probability that an observer would find an individual of this species at location if they “look hard enough?”

This semantic confusion is a by-product of using a distribution as a tool to model something that is discrete — the finite number of individuals of a species that exist across space. Regardless of the paradigm used to design the model predicting occurrence probability, the framing of *occurrence probability* as existing per unit space is fundamentally a *frequentist* view of probability, as this does not consider that a finite number of samples from this spatial distribution are unlikely to produce, and instead imposes the idea of a “long-run” probability of occurrences.

. A more appropriate way to view this would be the probability you observe an individual at a location  $\vec{x}$  as conditional on there being  $N$  total individuals of a given species across the entire spatial domain,  $p = P(\vec{x}|N)$ — we illustrate this using a “sandbox” SDM in the next section.

## An illustration

What is the value for which  $p(x)$  is non-zero, but *effectively* 0?

The goal in this section is to determine how the abundance of a species is  $N$  effects the meaning of the occurrence probability

species all occur in cells of the raster with a probability-value  $A_{xy}$  that is greater than some threshold. Dare I say it, but this section may contain multiple integrals.

Consider an SDM where the probability of occurrence of a species is given for each location  $x$  is given by  $P(x)$ . Assume the rank-frequency distribution values of  $P(x)$  follow an exponential distribution, with pdf  $f(x) = \lambda e^{-\lambda x}$ . What is the probability that for  $N$  observations of this species, that all of them occur in cells above some threshold value  $\epsilon$ ?

[Figure 1 about here.]

We start by determining what the probability of a single observation happening *below*  $\epsilon$ . Assume  $O \sim \text{Exp}(\lambda)$ . Then

$$P(O < \epsilon) = \int_{x^*}^{\infty} f(x)dx$$

54 From this we see  $\epsilon = \lambda e^{-\lambda x^*}$  which implies

$$\Rightarrow x^* = \frac{1}{\lambda} \ln \left( \frac{\lambda}{\epsilon} \right)$$

55 substituting into first line and integrating, because the exponential distribution is nice this cosmic  
56 gumbo now reduces to

$$P(O < \epsilon) = \frac{\epsilon}{\lambda}$$

57 Next, we take this result and plug it back into our original question, which is the probability that  
58 none of  $N$  observations occur below  $\epsilon$  which we can express as

$$59 \text{ Beroulli} \left( N, \left( 1 - \frac{\epsilon}{\lambda} \right)^N \right)$$

60 which looks like

61 [Figure 2 about here.]

62 As the mass of probability becomes more “evenly spread” across the entire spatial domain, the  
63 probability of all individuals being observed in locations with  $p > \epsilon$  goes down, as there are  
64 more cells with  $p \leq \epsilon$ .

## 65 **Test if continuous approx of space holds for various raster sizes**

66 In this section we risk falling into the mind-projection fallacy again, as in reality, an SDM is  
67 described by a finite  $n \times m$  raster where the values of the raster at an index  $(x, y)$  and does not  
68 “truly” follow an exponential distribution as assumed above.

## **An example: use real data and make an SDM, and report different maps based on simulating occurrence**

### **Conc**

Jaynes on the mind-projection fallacy:

In studying probability theory, it was vaguely troubling to see reference to “Gaussian random variables,” or “stochastic processes,” or “stationary time-series,” or “disorder,” as if the property of being Gaussian, random, stochastic, stationary, or disorderly is a real property, like the property of possessing mass or length, existing in Nature... As soon as the error had a definite name and description, it was much easier to recognize. Once one has grasped the idea, one sees the Mind Projection Fallacy everywhere; what we have been taught as deep wisdom, is stripped of its pretensions and seen to be instead a foolish non sequitur. The error occurs in two complementary forms, which we might indicate thus:

A): My own imagination -> Real property of Nature

B): My own ignorance -> Nature is indeterminate

“Our own ignorance implies nature is indeterminate.” This is why we build SDMs. Clearly the locations of the individuals of a species at any point in time is a measurable property of the world for which there cannot be more than one realized value. But we cannot sample this entire thing, so we take a subset of it and aim to estimate the this latent “species distribution” in order to predict where one might observe a species.

This pattern is common in the history of science. To develop on an example raised by Jaynes—quantum mechanics has an object that much like a species distribution model: the wave function  $\psi$  describing the probability of observing a particle across space. A misinterpretation of the wave function, according to Jaynes, is that often one assumes that the distribution of where observers see a particle is an inherent property of that particle, rather than being a construct of human imagination created to make predictions based on the information we have observed

95 about that particle. The most (in)famous example of this is likely Schrodinger's cat: often pre-  
96 sented as the lens that the cat is somehow *both* alive and dead at the same time—a quintessential  
97 of the mind-projection fallacy as described above. The state of the external world cannot be as-  
98 sumed indeterminate for the sole reason that we lack the information to fully describe it. This  
99 is equivalent to saying if one is in New York, then for oneself London becomes a multiverse of  
100 the possible worlds which are only realized upon one's return.

101 Is “probability” is a fixed property of nature rather than an abstraction used describe what we  
102 can say about a system given a set of information? me, personally, i don't know.

## 103 **References**

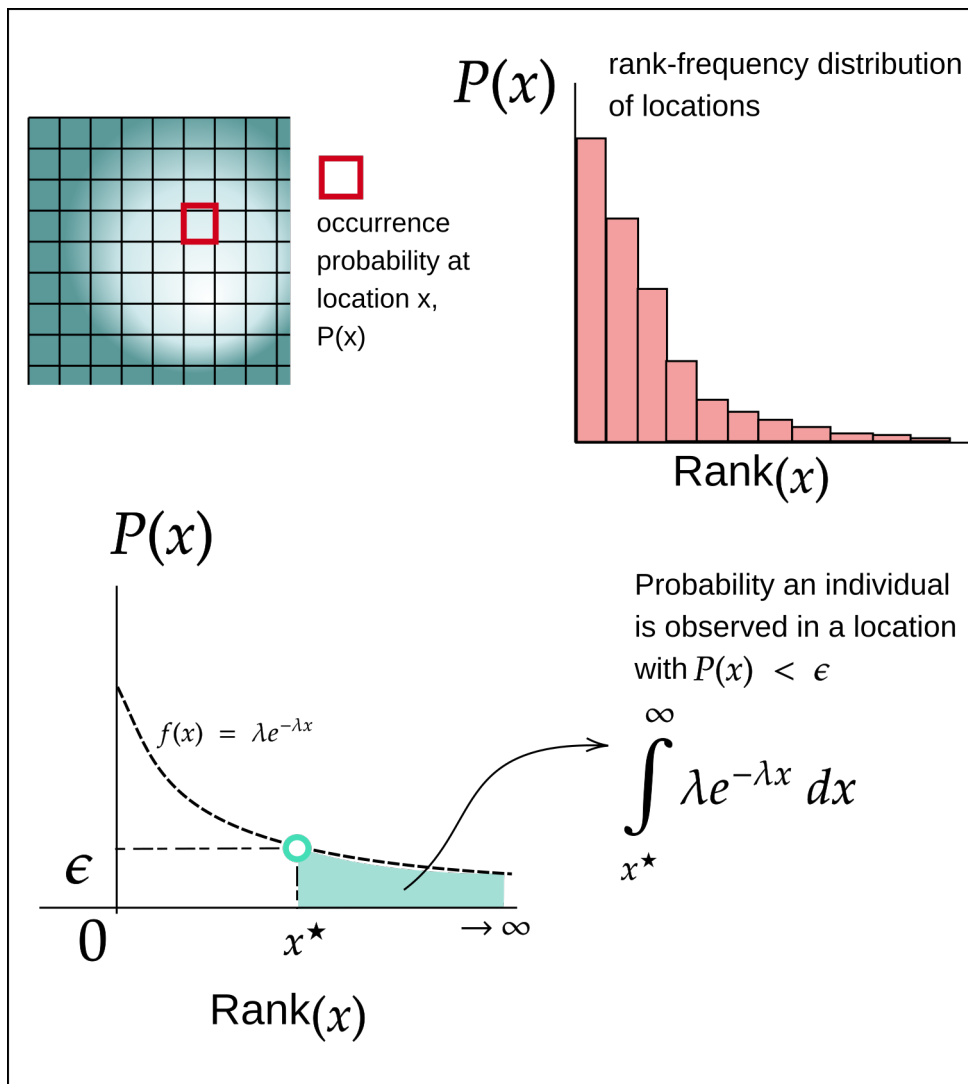


Figure 1: todo



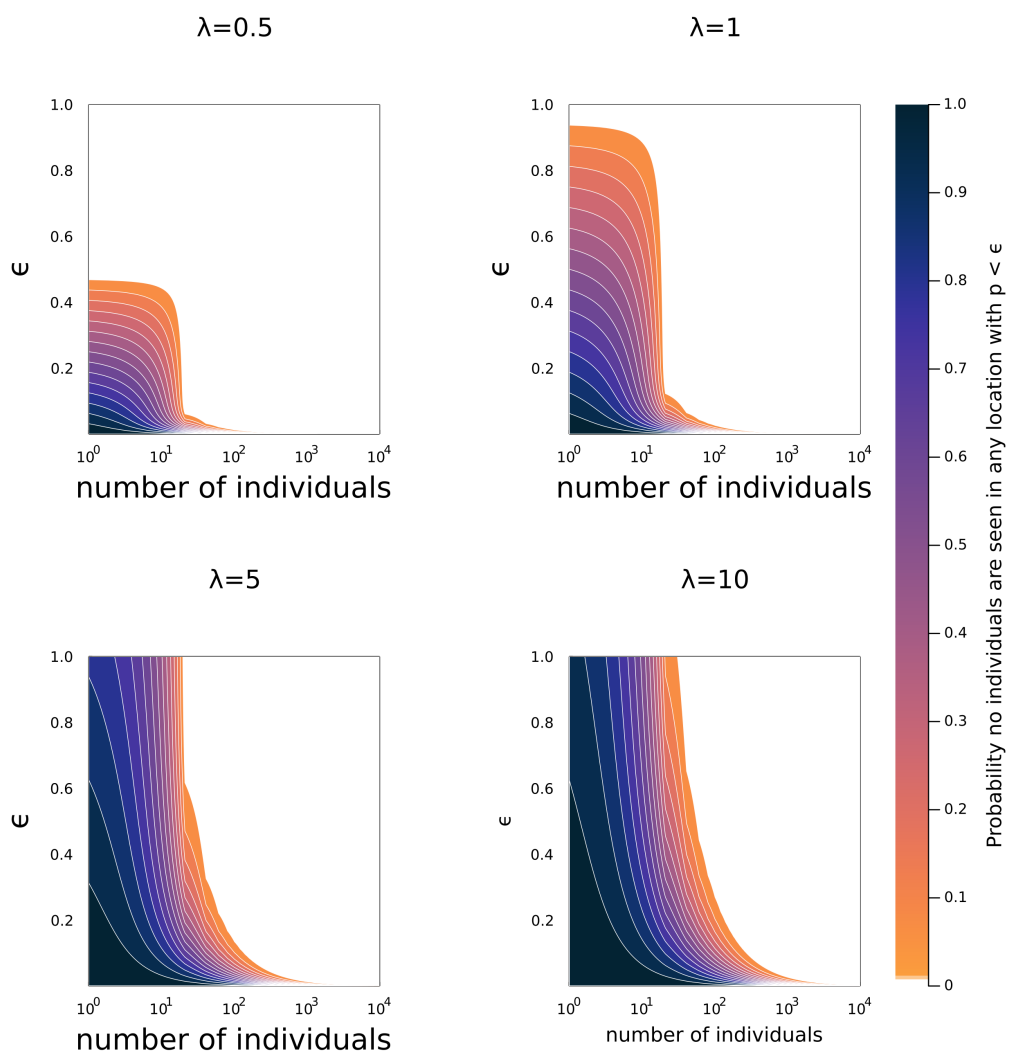


Figure 2: todo