$\begin{array}{c} {\rm Package\ sads} \\ {\rm Modeling\ Species\ Abundance\ Distributions} \\ {\rm LOGBOOK} \end{array}$

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Warning

Though this document will hopefully become a comprehensive vignette, currently it is a draft logbook used to clarify some aspects of the theoretical background of this package to developers, and to guide the next steps.

Poisson Sample

Given that there is n individuals of a given species in a community and that a fraction a of these individuals were randomly sampled, the probability of getting y individuals in the sample is given by the Poisson distribution

$$f(y) = \frac{(a n)^y e^{-a n}}{y!}$$
 (1)

Where the the single parameter of the Poisson λ correspond to the expected value E[y] and thus is substituted by the product a n. Now suppose that the abundance n in the community is also a random variable that can be described by some probability distribution g(n). Then the unconditional probability of getting y individuals in the sample is given by the integral

$$h(y) = \int^{\Omega} f(y)g(n) dn$$
 (2)

Which is the sad model we are looking for. Although there are analytic solutions for many cases, our goal is to write a R function that solves numerically equation 2 in a generic way, that is, a single function whith an argument for g(n). This function would then allow to compound any probability density function with the Poisson distribution. Taking this generalization further, we can substitute the Poisson distribution by another sampling model, e.g. the negative binomial to model aggregated sampling.

Our first step is to construct a function for a particular case for which the analytical solution is known, which allow us to check the results of the numerical integration.

Compounding with the exponential

Taking for g(n) the exponential

$$q(n) = \lambda e^{-n\lambda} \tag{3}$$

The equation 2 solves to:

$$h(y) = \frac{\lambda a^y}{(a+\lambda)^{y+1}} \tag{4}$$

Using logarithms in the intermediary calculations to avoid overflow, we can define this function in $\,{\sf R}\,{\sf as}$

The equivalent function for numerical integration in Ris¹:

```
> dsad <- Vectorize(FUN=
+ function(y,a,lambda){
+ poi <- function(y,n){</pre>
```

+ f1 <- function(n){

+ dexp(n,rate=lambda) * poi(y,n)
+

+ }

integrate(f1,0,Inf, abs.tol = .Machine\$double.eps^0.25/1000000)\$value

+ }, + "y")

The functions seems to give the same results

```
> dsad(0:10,a=0.1,lambda=0.01)
```

- [1] 0.09090909 0.08264463 0.07513148 0.06830135 0.06209213 0.05644739
- [7] 0.05131581 0.04665074 0.04240976 0.03855433 0.03504939
- > dpoix(0:10,frac=0.1,rate=0.01)
 - [1] 0.09090909 0.08264463 0.07513148 0.06830135 0.06209213 0.05644739
 - [7] 0.05131581 0.04665074 0.04240976 0.03855433 0.03504939

 $^{^1\}mathrm{This}$ function must be passed to <code>Vectorize</code> to make <code>integrate</code> work with more than one value of <code>y</code>

But as the value of y increases the numeric solution goes to zero. This effect is circumvented by the argument abs.tol, but it is not yet solved for small values of frac and rate, e.g.

- > dsad(c(200,20000),0.1,0.00001)
- [1] 6.319212e-64 0.000000e+00
- > dpoix(c(200,20000),0.1,0.00001)
- [1] 9.801016e-05 1.353353e-05

The figure 1 shows that this effect occurs even for moderate values of λ . The next step is to investigate this, and to correct the numerical integration.

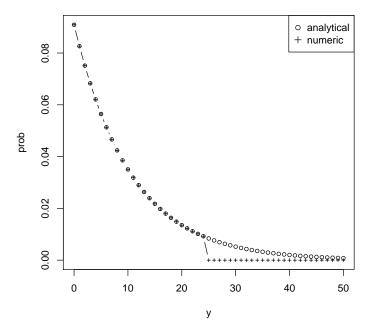


Figura 1: Probability value returned by the function dpoix, which calculates the Poisson-exponential analytically and the function dsad, which calculate the compound distribution by numerical integration.

Numerical integration with the function integrate

The first step on investigating this integration is verify the details of the integrate function. From the R-help of the function, we find that it imple-

ments the adaptative quadrature method. Additionally to the expression being integrated and the limits of integration, integrate also allow to specify the absolute(abs.tol) and relative (rel.tol) accuracy of the function used to approximate the integrand.²

As was said above, the numeric solution goes to zero as the value of y increases. And this effect is circumvented specifying a value for abs.tol. Although this approach conduct to a numerical result equivalent to the one otained analytically, it is easily found that for ranges(of the parameter y and the values of frac and rate) out of the considerated, then the explicit definition of abs.tol does not help at all. Alternativaly, a similar effect is observated when modifying the upper limit of integration and it is interesting to note that the effects are observated even for wide ranges of the parameters. May be the change on abs.tol induct a modification in the numerial estimation of Inf³.

A first guess on replace Inf

We now propose another way to circumvent the problem, it consist in manipulate the upper limit of integration Inf. Such procedure is not recommended according to R-help but for our purposes it seems to produce good results.

With a few trials we found that substitutig Inf by particular values, this can make the numerical solution matches the anlytical. It is observated that increasing the upper limit, when frac and rate are getting lower, conduct to better aproximation of the numerical integration. Considering frac = 10^{-1} and rate = 10^{-5} , with $200 \le y \ge 20000$, a reasonably good choice is replace Inf by k*y/(a+lambda) with k=1.9920941. Figure 3 shows the analytical and numerical curves visually matching. Figure 2 is the result when integral is calculated with abs.tol.

```
y1 < -dpoix(x,frac,rate)
```

In the above case the value of k was manually tunned, hopefully giving exact results for:

```
> dsad(c(200,20000),0.1,0.00001)
```

[1] 9.801016e-05 1.353353e-05

> dpoix(c(200,20000),0.1,0.00001)

[1] 9.801016e-05 1.353353e-05

For ranges out of the cited above, then the value of k, or even the entire solution, may be defined more consistently. This result can be improved, for wide ranges of frac, rate and y, if we specify the upper limit of integration in an another alternative manner. This time the idea is to split the total range of any parameter, y for example, in two or more parts and then specify the numerical integration for each part, as is shown below.

 $^{{\}it 2http://www.amtp.cam.ac.uk/lab/people/sd/lectures/nummeth 98/integration.htm}$

³http://en.scientificcommons.org/43526693

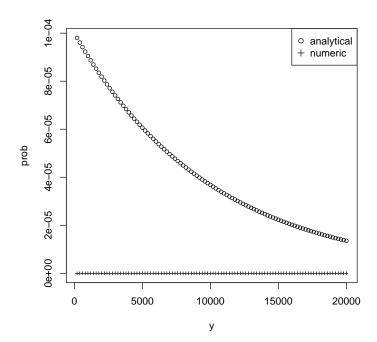


Figura 2: Probability value returned by the function dpoix and the function dsad. Considering frac = 10^{-1} and rate = 10^{-5} , with $200 \le y \ge 20000$

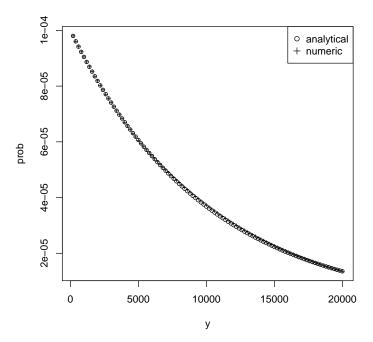


Figura 3: Probability value returned by the function dpoix and the function dsad. Considering frac = 10^{-1} and rate = 10^{-5} , with $200 \le y \ge 20000$

The above expression is an attempt to correct the result from the numerical integration provided by the standart integrate R function. The limits for each y interval were selected to exemplify.

In fact, this is not a final solution yet, but results are improved as shows Figure 4. Unfortunately both curves do not match for all ranges. Figure 8 show an example when it does not work, for rate = 10^{-6} and frac = 10^{-1} . Another methods/functions for numerical integration must be tested.

A general criteria to set the upper integration limit

An abundance value that is larger than the abundances of all or most of the species in the community is a natural upper limit for the integration. This value can be set with the quantile functions of the probability distibutions. For

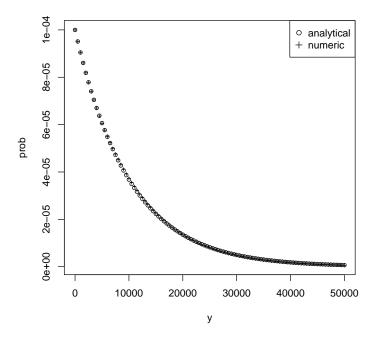


Figura 4: Probability value returned by the function dpoix and the function dsad. Considering frac = 10^{-1} and rate = 10^{-5} , with $0 \le y \ge 50000$

example, for a community sad that follow a gamma distribution with parameters rate=0.000075 and shape=0.75 4 , 99.99% of the species will have an abundance below:

> qgamma(0.9999,rate=0.000075, shape=0.75)

[1] 112629.0

This provides an abundance high enough to be used as upper integration limit, and that can be used with any community sad distribution, since all density functions has the [q] quantile function. Initial tests suggest that this may provide a general solution (see below in the section on generalization of the function), but systematic evaluation with a wide range of values is still to be done.

 $^{^4\}mathrm{These}$ are the values used by Green & Plotkin fig. 1

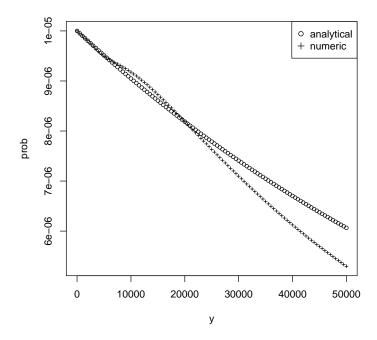


Figura 5: Probability value returned by the function dpoix and the function dsad. Considering frac = 10^{-1} and rate = 10^{-6} , with $0 \le y \ge 50000$

Numerical Integration with function GLIntegrate

Package distr provide an object oriented implementation of distributions. Additionaly, package distrEx provide extensions of package distr. That extensions include two functions for integration: GLIntegrate and distrExIntegrate.

The R-help description of the GLIntegrate function says it use Gauss-Legendre quadrature over a finite interval, in contrast with our problem that has an infinite upper limit.

The distrExIntegrate function does not have the same limitation relative to the interval of integration. But, as the description of the function explain, it does a numerical integration via integrate, in case integrate fails a Gauss-Legendre quadrature is performed.

Figures 6 and 7 shows the result from using GLIntegrate and distrExIntegrate respectively. As expected, from the description of the function, distrExIntegrate exibits a shape like the one got from integrate.

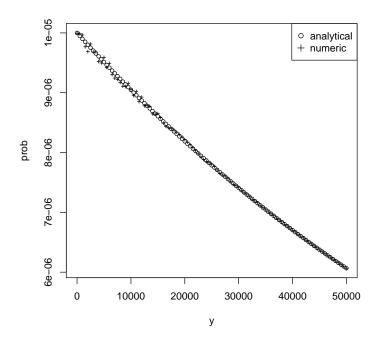


Figura 6: Probability value returned by the function dpoix and the function dsad. Considering frac = 10^{-1} and rate = 10^{-6} , with $0 \le y \ge 50000$. For the upper limit of integration it was used 10^6 . For the

Numerical Integration with function adaptIntegrate

The function adaptIntegrate come with the Cubature package. Cubature is a package for mutidimensional integration over hypercubes.

Installation can be performed with:

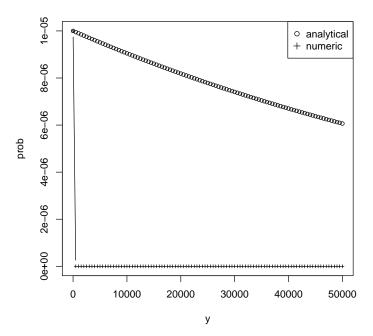


Figura 7: Probability value returned by the function dpoix and the function dsad. Considering frac = 10^{-1} and rate = 10^{-6} , with $0 \le y \ge 50000$. For the upper limit of integration it was used Inf.

- + adaptIntegrate(f1,0,10^6)\$integral
- + },
- + "y")

Some aspects of the numerical integration

One important aspect to verify when dealing with Probability Density Functions - PDFs, is that the sum of probabilities over the entire length is 1. This is tested for both dpoix and dsad with the R function sum.

Compounding with the gamma

Taking for g(n) the gamma

$$g(n) = \frac{\lambda^{\beta} n^{\beta - 1} e^{-\lambda n}}{\Gamma(\beta)}$$
 (5)

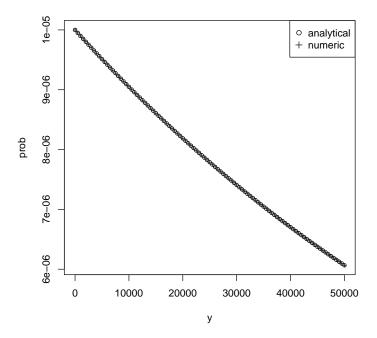


Figura 8: Cubature. Probability value returned by the function dpoix and the function dsad. Considering frac = 10^{-1} and rate = 10^{-6} , with $0 \le y \ge 50000$

The equation 2 solves to:

$$h(y) = \frac{a^y \lambda^{\beta} \Gamma(y+\beta)}{y! \Gamma(\beta) (a+\lambda)^{y+\beta}}$$
 (6)

We can define this function in $\,{\sf R}\,{\sf as}\,$

```
> dpoig <- function(x, frac, rate, shape) {
+    b <- (frac^x)*(rate^shape)*gamma(x+shape)
+    c <- factorial(x)*gamma(shape)*(frac+rate)^(x+shape)
+    b/c
+ }</pre>
```

The equivalent function for numerical integration in Ris:

```
> dsad_gamma <- Vectorize(FUN=
+ function(y,a,lambda,shape=0.75,fn=2){
+ poi <- function(y,n){
+ w <- y*log(a*n)-lfactorial(y)-a*n</pre>
```

```
exp(w)
    if(fn==1){
      f1 <- function(n){</pre>
        dexp(n,rate=lambda) * poi(y,n)
    }else{
      f1 <- function(n){</pre>
        dgamma(n, shape, rate = lambda, scale = 1/rate) * poi(y,n)
    }
    if (y<10){
            integrate(f1,0,100/max(a,lambda))$value
    }else if (y<50){
            k <- 6
            integrate(f1,0,k*y/(a+lambda))$value
    }else{
            k <- 1.9920941
            integrate(f1,0,k*y/(a+lambda))$value
    }
+ },
+ "y")
  The functions seems to give the same results
> frac <- 10^-1
> rate <- 10^-1
> x <- c(1, 10, 50, 100)
> shape <- 0.75
> dsad_gamma(x,frac,rate,shape,fn=2)
[1] 2.229763e-01 2.640021e-04 1.617665e-16 1.209309e-31
> dpoig(x,frac,rate,shape)
[1] 2.229763e-01 2.640021e-04 1.617665e-16 1.209309e-31
```

General Function

The purpose is to generalize the function to allow the combination of Poisson or Binomial Negative sampling with any distribution in the community. The current version of this generalized function includes the upper limit set by the quantile criteria, and some tweaks to avoid some of the biases related above:

```
> dsad <- function(y,frac,sad,samp="Poisson",log=FALSE,upper=0.9999,...){
+ qcom <- paste("q",deparse(substitute(sad)),sep="")
+ dcom <- paste("d",deparse(substitute(sad)),sep="")</pre>
```

```
+ dots <- c(as.name("n"),list(...))
+ uplim <- do.call(qcom,c(upper,list(...)))
+ f1 <- function(z){
+ f2 <- function(n){
+ t1 <- do.call(dcom,dots)
+ t2 <- dpois(z,frac*n)
+ ifelse(t1==0|t2==0,0,t1*t2*1e12)
+ }
+ integrate(f2,0,uplim,rel.tol=sqrt(.Machine$double.eps),subdivisions=500)$value
+ }
+ res <- sapply(y,f1)/(1e12*upper)
+ if(log)log(res) else res
+ }</pre>
```

Some observations:

- First working version substituted the old version of dsad
- A first version of the help page is already done, see for details
- The function no uses dpois for the Poisson distribution
- rel.tol and subdivisions increased following https://stat.ethz.ch/pipermail/r-help/attachments/20100921/56db7df0/attachment.pl
- log argument added; necessary to model fitting by maximum likelihood
- arguments to be passed to the community distribution choosen by sad vary, and are all under They should be named.
- Old functions removed from the directory

To Do List

- 1. Functions with analytical solution for Poisson-truncated hyperbolic
- 2. To use dnbinom to check dpoig: a compound Poisson-gamma is a negative binomial for integers, so they must give the same results.
- 3. To set the parameters of dpoig to match the parameters of dnbinom
- 4. To standardize arguments and their names across the density functions.
- 5. To check the thoroughly compound solutions of the generic function dsads against their analytical counterparts (dpoix, dpoig ...)
- 6. After all checks with Poisson sampling were done and OK, proceed to set the functions for Binomial Sampling (Table 2 of Green & Plotkin).