

log1pmx(), bd0(), stirlerr() – Computing Poisson, Binomial, Gamma Probabilities in R

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Abstract

The auxiliary function `log1pmx()` (“log 1 plus minus x”), had been introduced when R’s `pgamma()` (incomplete Γ function) had been numerically improved by Morten Welinder’s contribution to R’s [PR#7307, in Jan. 2005](#)¹, it is mathematically defined as

$$\log1pmx(x) := \log(1+x) - x \approx -x^2/2 + x^3/3 - x^4/4 \pm \dots, \quad (1)$$

and for numerical evaluation, suffers from two levels of cancellations for small x , i.e., using `log1p(x)` for `log(1+x)` is not sufficient.

In 2000 already, Catherine Loader’s contributions for more accurate computation of binomial, Poisson and negative binomial probabilities, [Loader \(2000\)](#), had introduced auxiliary functions `bd0()` and `stirlerr()`, see below.

Much later, in R’s [PR#15628, in Jan. 2014](#)², Welinder noticed that in spite of Loader’s improvements, Poisson probabilities were not perfectly accurate (only ca. 13 accurate digits instead of $15.6 \approx \log_{10}(2^{52})$), relating the problem to somewhat imperfect computations in `bd0()`, which he proposed to address using `log1pmx()` on one hand, and additionally addressing cancellation by using *two* double precision numbers to store the result (his proposal of an `ebd0()` function).

Here, I address the problem of providing more accurate `bd0()` (and `stirlerr()` as well), applying Welinder’s proposal to use `log1pmx()`, but otherwise diverging from the proposal.

1 Introduction

According to R’s reference documentation, `help(dbinom)`, the binomial (point-mass) probabilities of the binomial distribution with `size = n` and `prob = p` has “density” (point probabilities)

$$p(x) := p(x; n, p) := \binom{n}{x} p^x (1-p)^{n-x}$$

for $x = 0, \dots, n$, and these are (in R function `dbinom()`) computed via Loader’s algorithm ([Loader \(2000\)](#)) which had improved accuracy considerably, also for R’s internal `dpois_raw()` function which is used further directly in `dpois()`, `dnbinom()`, `dgamma()`, the non-central `dbeta()` and `dchisq()` and even the *cumulative* $\Gamma()$ probabilities `pgamma()`

¹https://bugs.R-project.org/bugzilla/show_bug.cgi?id=7307#c6

²https://bugs.r-project.org/bugzilla/show_bug.cgi?id=15628

and hence indirectly e.g., for cumulative central and non-central chisquare probabilities (`pchisq()`).

Loader noticed that for large n , the usual way to compute $p(x; n, p)$ via its logarithm $\log(p(x; n, p)) = \log(n!) - \log(x!) - \log((n-x)!) + x \log(p) + (n-x) \log(1-p)$ was inaccurate, even when accurate $\log \Gamma(x) = \mathbf{lgamma}(x)$ values are available to get $\log(x!) = \log \Gamma(x+1)$, e.g., for $x = 10^6$, $n = 2 \times 10^6$, $p = 1/2$, about 7 digits accuracy were lost from cancellation (in subtraction of the log factorials).

Instead, she wrote

$$p(x; n, p) = p(x; n, \frac{x}{n}) \cdot e^{-D(x; n, p)}, \quad (2)$$

where the ‘‘Deviance’’ $D(\cdot)$ is defined as

$$\begin{aligned} D(x; n, p) &= \log p(x; n, \frac{x}{n}) - \log p(x; n, p) \\ &= x \log \left(\frac{x}{np} \right) + (n-x) \log \left(\frac{n-x}{n(1-p)} \right), \end{aligned} \quad (3)$$

and to avoid cancellation, $D(\cdot)$ has to be computed somewhat differently, namely – correcting notation wrt the original – using a *two*-argument version $D_0(\cdot)$:

$$\begin{aligned} D(x; n, p) &= np \tilde{D}_0 \left(\frac{x}{np} \right) + nq \tilde{D}_0 \left(\frac{n-x}{nq} \right) \\ &= D_0(x, np) + D_0(n-x, nq), \end{aligned} \quad (4)$$

where $q := 1 - p$ and

$$\tilde{D}_0(r) := r \log(r) + 1 - r \quad \text{and} \quad (5)$$

$$D_0(x, M) := M \cdot \tilde{D}_0(x/M) \quad (6)$$

$$= M \cdot \left(\frac{x}{M} \log \left(\frac{x}{M} \right) + 1 - \frac{x}{M} \right) = x \log \left(\frac{x}{M} \right) + M - x \quad (7)$$

Note that since $\lim_{x \downarrow 0} x \log x = 0$, setting

$$\tilde{D}_0(0) := 1 \quad \text{and} \quad (8)$$

$$D_0(0, M) := M \tilde{D}_0(0) = M \cdot 1 = M$$

defines $D_0(x, M)$ for all $x \geq 0$, $M > 0$.

The careful C function implementation of $D_0(x, M)$ is called `bd0(x, np)` in Loader’s C code and now R’s Mathlib at <https://svn.r-project.org/R/trunk/src/nmath/bd0.c>, mirrored, e.g., at [Winston Chen’s github mirror](#)³. In 2014, Morten Welinder suggested in [R’s PR#15628](#)⁴ that the current `bd0()` implementation is still inaccurate in some regions (mostly *not* in the one it has been carefully implemented to be accurate, i.e., when $x \approx M$) notably for computing Poisson probabilities, `dpois()` in R; see more below.

Evaluating of $p(x; n, p)$ in (2), in addition to $D(x; n, p)$ in (4) also needs $p(x; n, \frac{x}{n})$ where in turn, the Stirling De Moivre series is used:

$$\log n! = \frac{1}{2} \log(2\pi n) + n \log(n) - n + \delta(n), \quad \text{where the ‘‘Stirling error’’ } \delta(n) \text{ is} \quad (9)$$

$$\delta(n) := \log n! - \frac{1}{2} \log(2\pi n) - n \log(n) + n = \quad (10)$$

$$= \frac{1}{12n} - \frac{1}{360n^3} + \frac{1}{1260n^5} - \frac{1}{1680n^7} + \frac{1}{1188n^9} + O(n^{-11}). \quad (11)$$

³<https://github.com/wch/r-source/blob/trunk/src/nmath/bd0.c>

⁴https://bugs.r-project.org/bugzilla/show_bug.cgi?id=15628

Note that $\delta(n) \equiv \text{stirlerr}(n)$ in the C code provided by Loader and now in R's Mathlib at <https://svn.r-project.org/R/trunk/src/nmath/stirlerr.c>.

Note that for the binomial, x is an integer in $\{0, 1, \dots, n\}$ and $M = np \geq 0$, but the formulas (6), (7) for $D_0(x, M)$ apply and are needed, e.g., for `pgamma()` computations for general non-negative $(x, M > 0)$ where even $x = 0$ is well defined, see (8) above.

Further (see Loader), such a saddle point approach is needed for Poisson probabilities, as well, where

$$p_\lambda(x) = e^{-\lambda} \frac{\lambda^x}{x!} \tag{12}$$

$$\begin{aligned} \log p_\lambda(x) &= -\lambda + x \log \lambda - \underbrace{\log(x!)}_{\log(1/\sqrt{2\pi x}) - (x \log x - x + \delta(x))} \\ &= \log \frac{1}{\sqrt{2\pi x}} - x \log \frac{x}{\lambda} + x - \lambda - \delta(x), \end{aligned} \tag{13}$$

is re-expressed using $\delta(x)$ and from (7) $D_0(x, \lambda)$ as

$$p_\lambda(x) = \frac{1}{\sqrt{2\pi x}} e^{-\delta(x) - D_0(x, \lambda)} \tag{14}$$

Also, negative binomial probabilities, `dnbinom()`, TODO

Even for the t_ν density, `dt()`,

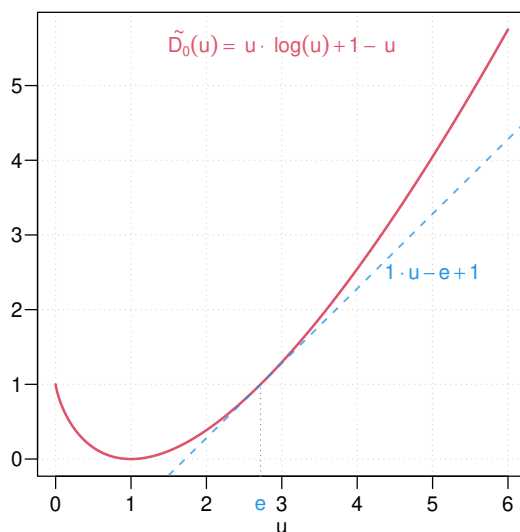
...but there have a direct approximations in package DPQ, currently functions `c_dt(nu)` and even more promisingly, `lb_chi(nu)`. TODO

2 Loader's "Binomial Deviance" $D_0(x, M) = \text{bd0}(x, M)$

Loader's "Binomial Deviance" function $D_0(x, M) = \text{bd0}(x, M)$ has been defined for $x, M > 0$ where the limit $x \rightarrow 0$ is allowed (even though not implemented in the original `bd0()`), here repeated from (6) :

$$\begin{aligned} D_0(x, M) &:= M \cdot \tilde{D}_0\left(\frac{x}{M}\right), \quad \text{where} \\ \tilde{D}_0(u) &:= u \log(u) + 1 - u = u(\log(u) - 1) + 1. \end{aligned}$$

Note the graph of $\tilde{D}_0(u)$,



has a double zero at $u = 1$, such that for large M and $x \approx M$, i.e., $\frac{x}{M} \approx 1$, the direct computation of $D_0(x, M) = M \cdot \tilde{D}_0(\frac{x}{M})$ is numerically problematic. Further,

$$D_0(x, M) = M \cdot \left(\frac{x}{M} \left(\log\left(\frac{x}{M}\right) - 1 \right) + 1 \right) = x \log\left(\frac{x}{M}\right) - x + M. \quad (15)$$

We can rewrite this, originally by e-mail from Martyn Plummer, then also indirectly from Morten Welinder's mentioning of `log1pmx()` in his PR notably for the important situation when $|x - M| \ll M$. Setting $t := (x - M)/M$, i.e., $|t| \ll 1$ for that situation, or equivalently, $\frac{x}{M} = 1 + t$. Using t ,

$$t := \frac{x - M}{M} \quad (16)$$

$$\begin{aligned} D_0(x, M) &= \underbrace{M \cdot (1 + t)}_x \log(1 + t) - \underbrace{t \cdot M}_{x - M} = M \cdot ((t + 1) \log(1 + t) - t) = \\ &= M \cdot p_1 l_1(t), \end{aligned} \quad (17)$$

where

$$p_1 l_1(t) := (t + 1) \log(1 + t) - t = \frac{t^2}{2} - \frac{t^3}{6} \pm \dots, \quad (18)$$

$$\begin{aligned} &= (\log(1 + t) - t) + t \cdot \log(1 + t) \\ &= \log1pmx(t) + t \cdot \log1p(t) \end{aligned} \quad (19)$$

where the Taylor series expansion is useful for small $|t|$,

$$p_1 l_1(t) = \frac{t^2}{2} - \frac{t^3}{6} + \frac{t^4}{12} \pm \dots = \sum_{n=2}^{\infty} \frac{(-t)^n}{n(n-1)} = \frac{t^2}{2} \sum_{n=2}^{\infty} \frac{(-t)^{n-2}}{n(n-1)/2} = \frac{t^2}{2} \sum_{n=0}^{\infty} \frac{(-t)^n}{\binom{n+2}{2}} = \quad (20)$$

$$= \frac{t^2}{2} \left(1 - t \left(\frac{1}{3} - t \left(\frac{1}{6} - t \left(\frac{1}{10} - t \left(\frac{1}{15} - \dots \right) \right) \right) \right) \right), \quad (21)$$

which we provide in DPQ via function `p1l1ser(t, k)` getting the first k terms, and the corresponding series approximation for

$$D_0(x, M) = \lim_{k \rightarrow \infty} p1l1ser\left(\frac{x - M}{M}, k, F = \frac{(x - M)^2}{M}\right), \quad (22)$$

where the approximation of course uses a finite k instead of the limit $k \rightarrow \infty$.

This Taylor series expansion is useful and nice, but may not even be needed typically, as both utility functions `log1pmx(t)` and `log1p(t)` are available implemented to be fully accurate for small t , $t \ll 1$, and (19), indeed, with $t = (x - M)/M$ the evaluation of

$$D_0(x, M) = M \cdot p_1 l_1(t) = M \cdot (\log1pmx(t) + t \cdot \log1p(t)), \quad (23)$$

seems quite accurate already on a wide range of (x, M) values.

Note that $x * \log1p(x)$ and `log1pmx()` have different signs, but also note that for small $|x|$, are well approximated by x^2 and $-x^2/2$, so their sum $p_1 l_1(x) = \log1pmx(x) + x \cdot \log1p(x)$ is approximately $x^2/2$ and numerically computing $x^2 - x^2/2$ should only lose 1 or 2 bits of precision.

References

Loader, C. (2000). Fast and accurate computation of binomial probabilities. Technical report, Lucent; Murray Hill, NJ USA.

```

> par(mfcol=1:2, mar = 0.1 + c(2.5, 3, 1, 2), mgp = c(1.5, 0.6, 0), las=1)
> p.p1l1(-7/8, 2, ylim = c(-1,2))
> arrows(.3,-.3, 0, 0); text(.3, -.3, "zoom in")
> p.p1l1(-1e-4, 1.5e-4, ylim=1e-8*c(-.6, 1), do.legend=FALSE)

```

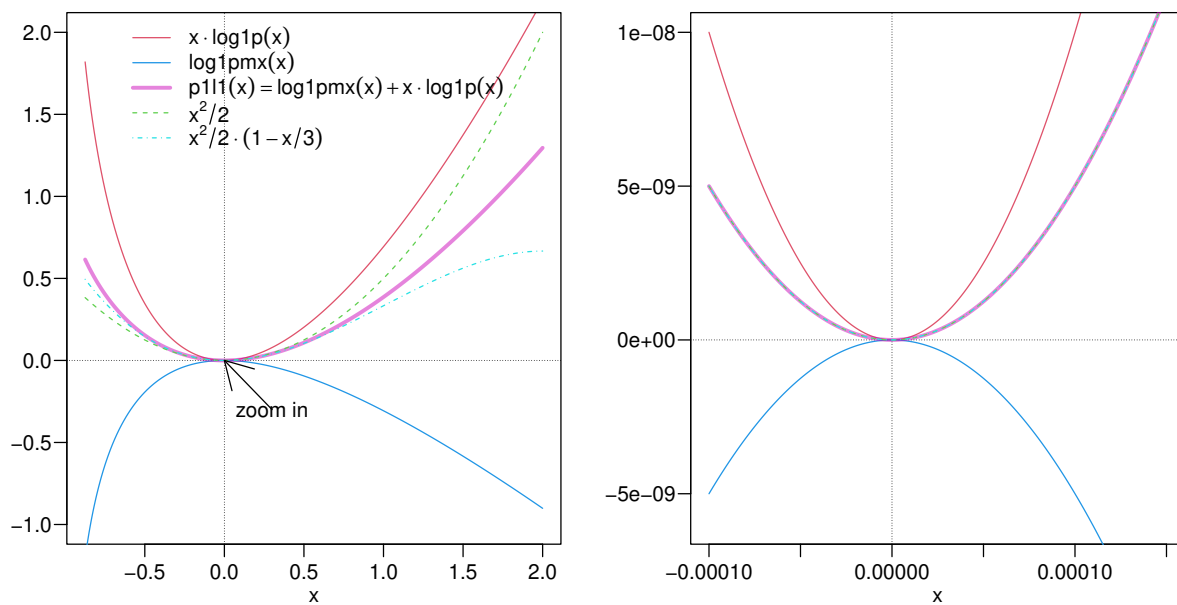


Figure 1: $p1l1()$ and its constituents; on the right, zoomed in 4 and 8 orders of magnitude, where the Taylor approximations $x^2/2$ and $x^2/2 - x^3/6$ are visually already perfect.