

## 0.1 (Not) choosing a Renyi order

The entropy of a R.V.  $X$  tells us how difficult it is to predict, which is more or less exactly what we want to quantify (the more diverse a community is, the harder it should be to predict what a random sample from said community will contain). And it relates nicely to a tangible quantity by way of the uniform distribution over an outcome set of size  $n$ , i.e.

$$H(\mathbf{u}^n) = -\sum_{i=1}^n p_i \log p_i = -\log \frac{1}{n} \sum_{i=1}^n \frac{1}{n} = \log n \Leftrightarrow \exp H(\mathbf{u}^n) = n. \quad (1)$$

This is a nice interpretation -  $\exp H(\mathbf{x})$  is the number of *uniformly distributed* species that would be required to produce the same entropy as our specific community  $\mathbf{x}$ . Generalizing further, we get the  $q$ -parametrized equivalent species count

$$F_q(\mathbf{x}) := \exp \left[ \frac{1}{1-q} \log \sum_i x_i^q \right], \quad (2)$$

from which we can recover the trivial species count simply by inserting  $q = 0$ :

$$F_0(\mathbf{x}) = \exp \left[ \frac{1}{1} \log \sum_i x_i^0 \right] = \exp \log n = n. \quad (3)$$

Different values of  $q$  gives different metrics, reflecting various different properties of the distribution. If it were true that  $F_q(\mathbf{x}_1) > F_q(\mathbf{x}_2)$  for some  $q$  implies that  $F_q(\mathbf{x}_1) > F_q(\mathbf{x}_2)$  for all  $q$ , then the choice of  $q$  wouldn't matter much, since the internal ordering would be preserved. But we know this isn't the case. The choice of  $q$  matters. But since we can't sensibly argue for one choice being superior to another, there's only unbiased thing we can do - include *all possible* values of  $q$ ! Alas, the straight integral is divergent, while the asymptotic average

$$\lim_{q \rightarrow \infty} \frac{1}{q} \int F_q(\mathbf{x}) dq \quad (4)$$

is pointless for another reason - it vanishes for any  $\mathbf{x}$ . To see why, note that  $\lim_{q \rightarrow \infty} \frac{1}{q} F_q(\mathbf{x}) \approx \frac{1}{q} \exp \frac{-1}{q} \log x_{\max}^q = \frac{1}{q} \frac{1}{x_{\max}}$ .

However, by multiplying the integrand with (almost) *any* convergent function of  $q$ , we get something finite, but bounded. In fact, defining a diversity metric

$$\bar{F}(\mathbf{x}) := \int_0^\infty g(q) F_q(\mathbf{x}) dq \quad (5)$$

with *any* integrable function  $g$  that is strictly positive and monotonically decreasing for all  $q \geq 0$  will produce a finite and positive diversity metric well-defined for any distribution  $\mathbf{x}$ . An example would be the Gaussian  $e^{-q^2}$  or most decaying exponentials (or indeed any function that decays more rapidly than  $1/x$  asymptotically).

I don't know if it seems alien and messy at first glance, but it actually turns out to be quite elegant and should share a lot of nice qualities with the Laplace transform among other things.