E4.2: A distribution on $\mathbb{N}$ related to zeta function

Xinyu Mao

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Let $s > 1$, and define $\zeta(s) := \sum_{n \in \mathbb{N}} \frac{1}{n^s}$, as usual. Let X and Y be independent $\mathbb{N}$-valued random variables with

$$\mathcal{P}(X = n) = \mathcal{P}(Y = n) = \frac{n^{-s}}{\zeta(s)}.$$

Define events $E_n := \{X \text{ is divisible by } n\}$. One can easily see that

$$\mathcal{P}(E_n) = \sum_{i \in \mathbb{N}} \mathcal{P}(X = ni) = \sum_{i \in \mathbb{N}} \frac{(ni)^{-s}}{\zeta(s)} = \frac{n^{-s}}{\zeta(s)} \sum_{i \in \mathbb{N}} i^{-s} = n^{-s}. \tag{1}$$

Everything is set up now, and we shall explore some interesting properties of X and Y.

Proposition 1. $(E_p : p \in \mathbb{P})$ *are independent with $\mathbb{P}$ denoting all prime numbers.*

Proof. Let $S := \{p_1, p_2, \ldots, p_m\}$ be a finite subset of $\mathbb{P}$. By definition of E_n and Eq. (1), we immediately have

$$\mathcal{P}\left(\bigcap_{i=1}^{m} E_i\right) = \mathcal{P}\left(E_{p_1 p_2 \cdots p_m}\right) = (p_1 p_2 \cdots p_m)^{-s} = \prod_{i=1}^{m} \mathcal{P}(E_i),$$

which implies $(E_p : p \in \mathbb{P})$ are independent. $\qquad\square$

We can use this observation to cast light on **Euler's formula**

$$\frac{1}{\zeta(s)} = \prod_{p \in \mathbb{P}} (1 - \frac{1}{p^s}). \tag{2}$$

The LHS of Eq. (2) is obviously $\mathcal{P}(X = 1)$. The event that $X = 1$ can also be described as $E := \{p \nmid X, \forall p \in \mathbb{P}\} = \bigcap_{p \in \mathbb{P}} E_p^c$. Combining Eq. (1) and Proposition 1, we have $\mathcal{P}(E) = \prod_{p \in \mathbb{P}} \mathcal{P}(E_p^c) = \prod_{p \in \mathbb{P}}(1 - \frac{1}{p^s})$.

Clearly, $(E_{p^2}, p \in \mathbb{P})$ is independent as well(the argument is similar to Proposition 1). Hence,

$$
\begin{aligned}
\mathcal{P}(\text{no square other than 1 divides } X) &= \mathcal{P}\left(\bigcap_{p \in \mathbb{P}} E_{p^2}^c\right) \\
&= \prod_{p \in \mathbb{P}}(1 - \frac{1}{p^{2s}}) = \frac{1}{\zeta(2s)},
\end{aligned}
\tag{3}
$$

where the last equality follows from Eq. (2).

Define r.v. $H := \gcd(X, Y)$, and we shall show that

Proposition 2. $\mathcal{P}(H = n) = \frac{n^{-2s}}{\zeta(2s)}$.

Proof. We need the concept of conditional probability from elementary theory. Note that

$$
\mathcal{P}\left(X/n = k | E_n\right) = \frac{(nk)^{-s}}{\sum_{i \geq 1}(ni)^{-s}} = \frac{k^{-s}}{\zeta(s)}.
$$

Hence, $(X/n | E_n)$ has the same distribution as X. Let F_n be the event that n divides Y, then $(Y/n | E_n)$ also has the same distribution. Thus,

$$
\begin{aligned}
\mathcal{P}(H = n) &= \mathcal{P}\left(E_n \cap F_n \cap \gcd(\frac{X}{n}, \frac{Y}{n}) = 1\right) \\
&= \mathcal{P}(E_n)\mathcal{P}(F_n)\mathcal{P}\left(\gcd(\frac{X}{n}, \frac{Y}{n}) = 1 | E_n \cap F_n\right) \\
&= n^{-s} \cdot n^{-s} \cdot \mathcal{P}(\gcd(X, Y) = 1).
\end{aligned}
\tag{4}
$$

Since $(E_p, F_p : p \in \mathbb{P})$ is independent,

$$
\mathcal{P}(\gcd(X, Y) = 1) = \mathcal{P}(\bigcap_{p \in \mathbb{P}}(E_p \cap F_p)^c) = \prod_{p \in \mathbb{P}}(1 - \frac{1}{p^{2s}}) = \frac{1}{\zeta(2s)}.
$$

Plugging this into Eq. (4), we get the desired result. $\qquad \square$

However, we have not rigorously defined conditional probability yet. Can we avoid using conditional probability? The answer is yes, but we need a useful tool,

Möbius function, which is defined as

$$\mu : \mathbb{N} \to \{-1, 0, 1\}$$

$$n \mapsto \begin{cases} (-1)^m, \text{if } n = p_1 p_2 \cdots p_m, \text{ where } (p_i)_{i \in [m]} \text{ are distinct prime numbers;} \\ 0, \text{otherwise.} \end{cases}$$

Here we list two facts about μ without giving detailed proof [1].

Proposition 3. *(i)* $\sum_{d|n} \mu(d) = [n = 1]$.

(ii) $\sum_{n \in \mathbb{N}} \frac{\mu(n)}{n^s} = \prod_{p \in \mathbb{P}} \frac{1}{(1-p^s)} = \frac{1}{\zeta(s)}$.

Proof. (i) can be proved by counting the number of -1 and $+1$ among all $\mu(d)$. (ii) is simply expansion of the product. $\square$

Now we are well-equipped to give the coup de grace.

Another proof of Proposition 2. Let G_n be the event that X, Y are both divisible by n, which is equivalent to $n|H$. Clearly, $\mathcal{P}(G_n) = n^{-2s}$. By Proposition 3 (i),

$$\mathcal{P}(H = n) = \sum_{k=1}^{\infty} \mathcal{P}(H = kn) \sum_{d|k} \mu(d) = \sum_{d=1}^{\infty} \mu(d) \sum_{i=1}^{\infty} \mathcal{P}(H = idn). \qquad (*)$$

Note that $\sum_{i=1}^{\infty} \mathcal{P}(H = in) = \mathcal{P}(G_n)$ for all $n \in \mathbb{N}$. Hence,

$$(*) = \sum_{d=1}^{\infty} \mu(d) \mathcal{P}(G_{dn}) = \sum_{d=1}^{\infty} \mu(d)(dn)^{-2s}$$

$$= n^{-2s} \sum_{d=1}^{\infty} \mu(d) d^{-2s}$$

$$= \frac{n^{-2s}}{\zeta(2s)},$$

where the last equality follows from Proposition 3 (ii). $\square$

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[1] This kind of technique is called **Möbius inversion**. More information can be found at `https://math.berkeley.edu/~stankova/MathCircle/Multiplicative.pdf`

Remark. If we take $s = 1$(this makes no sense) in Eq. (3), we get the probability that X is square-free is $\frac{1}{\zeta(2)}$. In fact, the **natural density** of square-free numbers in $\mathbb{N}$ is exactly $\frac{1}{\zeta(2)}$! This is not just a coincidence. Formally, let A be a subset of $\mathbb{N}$, we define the natural density of A by the limit

$$\boldsymbol{d}(A) := \lim_{n \to \infty} \frac{|A \cap [n]|}{n},$$

and **analytic density** of A by the limit

$$\lim_{s \to 1^+} \frac{1}{\zeta(s)} \sum_{n \in A} n^{-s}.$$

Then we have

Theorem 4. *If natural density exists, then analytic density exists and they are equal.*

A proof can be found in [1] Chapter III.

References

[1] Gérald Tenenbaum. *Introduction to analytic and probabilistic number theory*, volume 163. American Mathematical Soc., 2015.