

Supplementary Material

Tatsuru Kobayashi*

Graduate School of
Information Science and Technology,
University of Tokyo
7-3-1 Hongo, Bunkyo-ku
Tokyo 113-8656
Japan

Kumiko Tanaka-Ishii†

Research Center for
Advanced Science and Technology,
University of Tokyo
4-6-1 Komaba, Meguro-ku
Tokyo 153-8904
Japan

A Mathematical Proof of Taylor Exponent

Here we show that the Taylor exponent of an independent and identically distributed (i.i.d.) process is 0.5. A proof in a more general form is shown in (Eisler, Bartos, and Kertész, 2007). This is a known mathematical fact, as found previously in (Yule, 1968).

Proposition 1. *The Taylor exponent of a sequence generated by an i.i.d. process is 0.5.*

Proof. Consider i.i.d. random variables $X_1, \dots, X_i, \dots, X_N$, where i denotes the location within a text. For a specific word $w_k \in W$, with W being the set of words, let p_k denote the probability of occurrence of word w_k , i.e., $\mathbb{P}(X_i = w_k) = p_k$ (for all i). Naturally, the expectation \mathbb{E} and variance \mathbb{V} of the count of w_k for X_i are the following:

$$\mathbb{E}[X_i] = p_k, \quad (1)$$

$$\mathbb{V}[X_i] = p_k(1 - p_k), \quad (2)$$

which only depend on the constant p_k . With window size Δt , $\mu_k = \Delta t \mathbb{E}[X_i]$. Note that $\sigma_k^2 = \Delta t \mathbb{V}[X_i]$, because

$$\begin{aligned} \sigma_k^2 &= \mathbb{V} \left[\sum_{i=1}^{\Delta t} X_i \right] \\ &= \mathbb{E} \left[\left(\sum_{i=1}^{\Delta t} (X_i - p_k) \right)^2 \right] \\ &= \mathbb{E} \left[\sum_{i=1}^{\Delta t} (X_i - p_k)^2 \right. \\ &\quad \left. + 2 \sum_{i \neq j} (X_i - p_k)(X_j - p_k) \right] \end{aligned}$$

$$\begin{aligned} &= \mathbb{E} \left[\sum_{i=1}^{\Delta t} (X_i - p_k)^2 \right] \\ &= \sum_{i=1}^{\Delta t} \mathbb{V}[X_i] \\ &= \Delta t \mathbb{V}[X_i]. \end{aligned}$$

Furthermore, note that $\mathbb{E}[(X_i - p_k)(X_j - p_k)] = 0$ for every i, j with $i \neq j$, because X_i and X_j are independent of each other and (1) holds. Therefore, Taylor exponent α of an i.i.d. process is 0.5, because

$$\sigma_k^2 = \frac{\mathbb{V}[X_i]}{\mathbb{E}[X_i]} \mu_k.$$

□

References

- Eisler, Zoltán, Imre Bartos, and János Kertész. 2007. Fluctuation scaling in complex systems: Taylor's law and beyond. *Advances in Physics*, pages 89–142.
- Yule, G. Udny. 1968. *The Statistical Study of Literary Vocabulary*. Archon Books.

*kobayashi@cl.rcast.u-tokyo.ac.jp
†kumiko@cl.rcast.u-tokyo.ac.jp