



UNIVERSITÄT ZU LÜBECK
INSTITUT FÜR INFORMATIONSSYSTEME

On the Behaviour of Weighted Permutation Entropy on Fractional Brownian Motion in the Univariate and Multivariate Setting

Authors: Marisa Mohr, Florian Wilhelm, Ralf Möller

Marisa Mohr

FLAIRS, 18. May 2021

Abstract

- › Investigation of qualitative behaviour of the fractional Brownian motion
- › Permutation Entropy quantifies complexity
- › Weighted Permutation Entropy takes amplitudes within time series into account
- › Univariate context although many real-world challenges are multivariate

Contribution

- (I) Introduction of multivariate weighted permutation entropy (MWPE)
- (II) Investigation of the behaviour of weighted permutation entropy on both univariate and multivariate fractional Brownian motion

(Multivariate) Fractional Brownian Motion

Definition (Multivariate fractional Brownian motion [1])

An m -multivariate process $((X^i(t))_{i=1}^m)_{t \in \mathbb{R}}$ is called *multivariate fractional Brownian motion* (mfBm) with Hurst parameter $H = (H_1; \dots; H_m)$, $H_i \in (0; 1)$ for $i = 1; \dots; m$, and denoted as $B_H^m(t)$, if it is

- › **Gaussian distributed**,
- › **self-similar** with Hurst parameter H , i.e.,

$$(X_1(\cdot); \dots; X_m(\cdot))_{t \in \mathbb{R}} \stackrel{H}{=} ({}^{H_1}X_1(\cdot); \dots; {}^{H_m}X_m(\cdot))_{t \in \mathbb{R}}$$

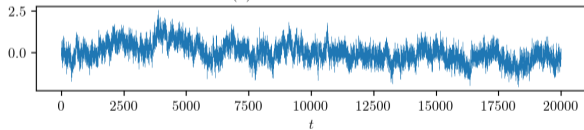
for any $\lambda > 0$, where $\stackrel{H}{=}$ denotes the equality of finite-dimensional distributions and it has

- › **stationary** increments, i.e., $B_H^m(t) - B_H^m(s) \stackrel{H}{=} B_H^m(t - s)$:

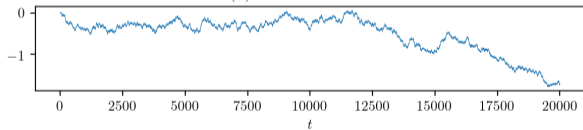
(Multivariate) Fractional Brownian Motion

- › $H < 1/2$: anti-persistence property and negatively correlated increments
- › $H = 1/2$: ordinary Brownian motion
- › $H > 1/2$: persistence property and positively correlated increments
- › $H \neq 1/2$: smoothness, less irregular and more trendy

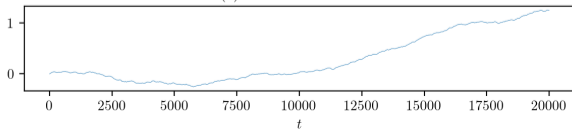
(a) fBm with $H = 0.1$



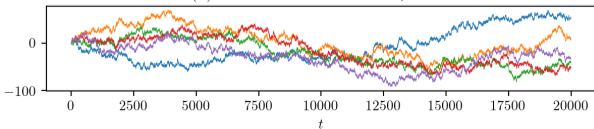
(b) fBm with $H = 0.5$



(c) fBm with $H = 0.8$



(d) mfBm with $m = 4$ and $H_i = 0.4$



Ordinal Pattern Symbolization

Ordinal patterns describe the total order between two or more neighbours, encoded by permutations.



Figure: All possible ordinal patterns of order $d = 3$.

Definition

A vector $(x_1; \dots; x_d) \in \mathbb{R}^d$ has ordinal pattern $(r_1; \dots; r_d) \in \mathbb{N}^d$ of order $d \in \mathbb{N}$ if $x_{r_1} \leq \dots \leq x_{r_d}$ and $r_{l-1} > r_l$ in the case $x_{r_{l-1}} = x_{r_l}$.

Ordinal Pattern in Time Series

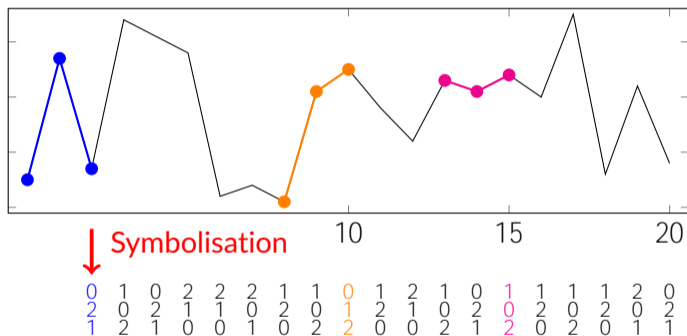


Figure: Ordinal pattern determination of order $d = 3$ and time delay $\tau = 1$ at any time point $t \in [d + 1; T]$.

Permutation Entropy

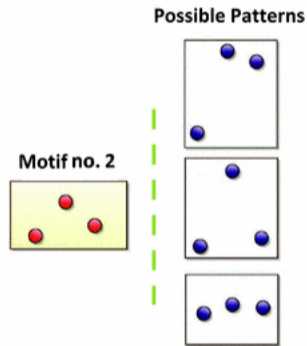
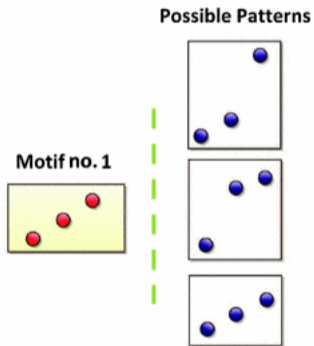
Definition (Permutation entropy [2])

The *permutation entropy (PE)* of order $d \geq 2$ and delay $\tau \geq 1$ of a univariate time series $x = (x_t)_{t=1}^T$, $T \geq N$ is defined by

$$PE_{d,\tau}(x) = -\sum_{j=1}^{d!} p_j^{(d,\tau)} \ln p_j^{(d,\tau)} \quad (1)$$

where $p_j^{(d,\tau)}$ is the frequency of ordinal pattern j in the time series.

Different Amplitudes [3]



Weighted Permutation Entropy:
Differentiates between different patterns of
a given motif in the amplitudes

Weighted Permutation Entropy

Definition (Weighted permutation entropy [3])

The *weighted permutation entropy* (WPE) of an univariate time series $x = (x_t)_{t=1}^T$, $T \geq N$, with order $d \geq N$ and delay $\tau \geq N$ is defined as

$$\text{WPE}_{d;\tau}(x) = \frac{1}{d!} \sum_j wp_j^{d;\tau} \ln wp_j^{d;\tau} \quad (2)$$

with

$$wp_j^{d;\tau} = \frac{\sum_{t=\tau}^T w_t \mathbb{P}[(x_{t-(d-1)\tau}, \dots, x_{t-\tau}, x_t) \text{ has pattern } j]}{\sum_{t=\tau}^T w_t \mathbb{P}[(x_{t-(d-1)\tau}, \dots, x_{t-\tau}, x_t)]}; \quad (3)$$

where $w_t = \frac{1}{d} \sum_{k=1}^d (x_{t-(k-1)\tau} - \bar{x}_t^{d;\tau})^2$ is the empirical variance of the sub-sequence and $\bar{x}_t^{d;\tau}$ denotes the arithmetic mean that is $\bar{x}_t^{d;\tau} = \frac{1}{d} \sum_{k=1}^d x_{t-(k-1)\tau}$ and $[x] = 1$ if x , 0 otherwise.

WPE on fBm

Theorem (1)

The WPE of order $d = 2$ for all delays $2 \leq N$ on fBm $B_H(t)$ is given by

$$\text{WPE}_{2;}(B_H(t)) = \ln(1=2): \quad (4)$$

Proof

WPE differs from PE in that the ordinal patterns are weighted depending on their position t according to Eq. (3). For a weight w_t of order $d = 2$, i.e., of two time steps x_{t-1} and x_t , we have

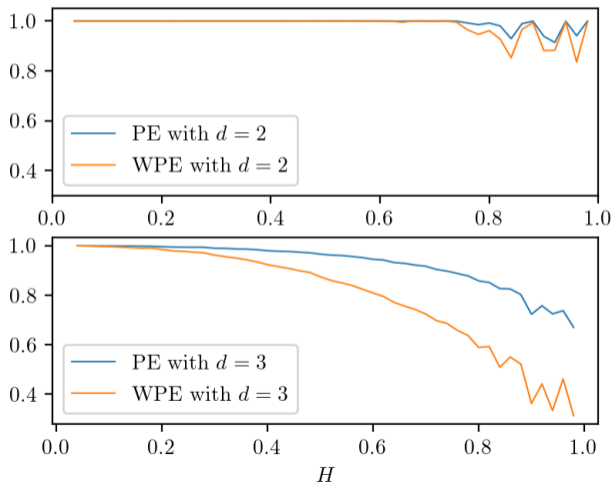
$$w_t = \frac{1}{2} \sum_{k=1}^2 (x_{t-(k-1)\tau} - \bar{x}_t^{2,\tau})^2 = \frac{1}{2} (x_t - x_{t-\tau})^2. \quad (5)$$

Since $x_t \sim B_H(t)$, we conclude from stationarity that

$$\frac{1}{2} (B_H(t) - B_H(t-\tau))^2 \sim \frac{1}{2} (B_H(\tau))^2 \quad (6)$$

with $\text{Var}(B_H(\tau)) = \sigma^2 \tau^{2H}$ and $\sigma^2 = \text{Var}(B_H(1)) = s^{2H}$. Consequently, the weights w_t are independently distributed from t , i.e., $w_t \sim \mathcal{N}(0, \frac{1}{2}(s\tau)^{2H})$. Considering the distribution of all possible realisations of fBm, we see from the use of the weights w_t from Eq. (3) in the calculation of WPE that it cancels out for a constant delay $\tau \in \mathbb{N}$.

WPE on fBm



In many fields of applications, multivariate measurements are performed.

Multivariate Weighted Permutation Entropy

- \triangleright For each variable $i = 1; \dots; m$ and for each ordinal pattern $j = 1; \dots; d!$ select all time steps in the time series $t \in [d + 1; T]$, for which the variable time pair $(i; t)$ has the ordinal pattern j .
- \triangleright Add up the weights w_t , i.e., $w_{ij} = \sum_{t=d+1}^T w_t$ for all selected ordinal pattern vectors j and for each variable $i = 1; \dots; m$. Note that the total count of weights w_t^i for each variable i is $n_i := T - (d - 1)$.
- \triangleright Divide the weighted sum w_{ij} by the total sum of all m weights to obtain the weighted frequencies for every pattern j .
- \triangleright Store the weighted frequencies in a *weighted pooling matrix* $P_w^{;d} \in \mathbb{R}^{m \times d!}$, which reflects the weighted distribution of the ordinal patterns in the multivariate time series across its m variables.

Multivariate Weighted Permutation Entropy

Definition (Multivariate weighted permutation entropy)

The *multivariate weighted permutation entropy* (MWPE) of a multivariate time series

$X = ((x_t^j)_{j=1}^m)_{t=1}^T$ is defined as PE of the marginal weighted frequencies

$P_{w_j}^{;d} = \frac{1}{m} \sum_{i=1}^m P_{w_{ij}}^{;d}$ for $j = 1, \dots, d!$ describing the distribution of the weighted ordinal pattern and can be calculated as

$$\text{MWPE}_{d;}(X) = \sum_j^{d!} P_{w_j}^{;d} \ln P_{w_j}^{;d} \quad (7)$$

MWPE in mfBm

Theorem (2)

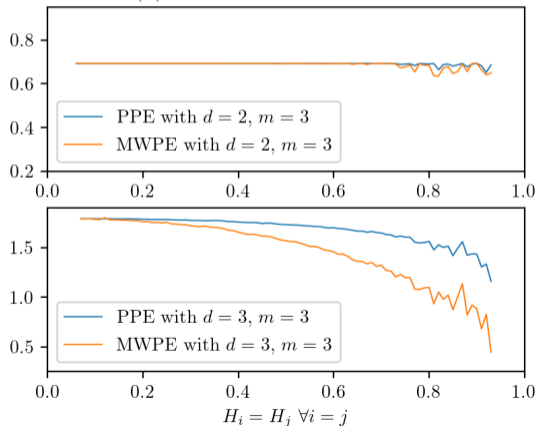
The MWPE of order $d = 2$ for all delays $2 \leq N$ on mfBm $B_H^m(t)$ is given by

$$\text{MWPE}(B_H^m(t)) = \ln(1=2) \quad (8)$$

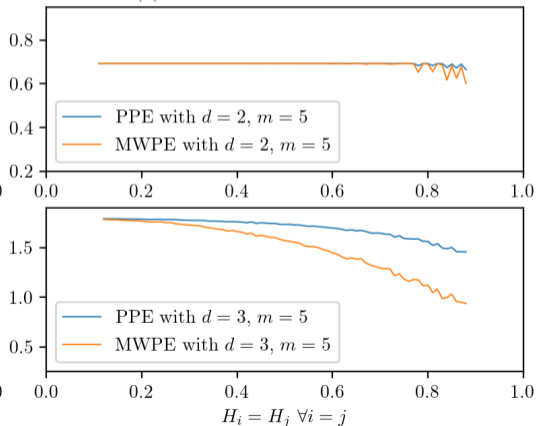
for all Hurst parameters H_i and variable-dimensions m .

MWPE on fBm

(b) PPE and MWPE on 3-fBm



(c) PPE and MWPE on 5-fBm



Conclusion

- › Definition of MWPE
- › In case of order $d = 2$, we have shown that WPE or MWPE on fBm or mfBm is constant, i.e., there is no effect of weighting
- › In case of order $d = 3$, we have shown that strictly ascending or strictly descending ordinal patterns $(0;1;2)$ or $(2;1;0)$ have greater weights and thus a greater impact on WPE and MWPE than the other four possible ordinal patterns

Future Work

- › Application for the estimation of the Hurst parameter or in classical data analysis and machine learning tasks, such as outlier detection or time series classification
- › Investigation of other orders $d > 3$ (*In Submission*)
- › Consideration of cross-correlations of variables (*In Submission*)

References

- [1] Pierre-Olivier Amblard, Jean-François Coeurjolly, Frédéric Lavancier, and Anne Philippe. Basic properties of the Multivariate Fractional Brownian Motion. *Séminaires et congrès*, 28:65–87, 2013.
- [2] Christoph Bandt and Bernd Pompe. Permutation Entropy: A Natural Complexity Measure for Time Series. *Physical Review Letters*, 88(17):174102, April 2002. Publisher: American Physical Society.
- [3] Bilal Fadlallah, BD Chen, Andreas Keil, and Jose Principe. Weighted-permutation entropy: A complexity measure for time series incorporating amplitude information. *Physical review. E, Statistical, nonlinear, and soft matter physics*, 87:022911, 02 2013.

