

# Permutation Complexity Measures for Time Series

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## Abstract

Permutation entropy has become a standard tool in time-series analysis that exploits the temporal properties of these data sets. Many current applications use an approach based on Shannon entropy, which implicitly assumes an underlying uniform distribution on patterns. In this paper, we consider several additional null models for time series data and determine the corresponding permutation distributions. This allows us to compare real-world data to more complex generative processes. Additionally, building on recent results of Martinez, we define a measure of complexity that allows us to characterize when a random walk is an appropriate model for a time series.

# Outline

- 1 Introduction
- 2 Background and Motivation
- 3 Null Models
- 4 Random Walk Metrics
- 5 Walks on  $S_n$
- 6 Conclusion

# Permutation Patterns

$$\begin{pmatrix} 3 & 6 & 7 & 9 & 4 & 8 & 1 & 2 & 5 \\ 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 \end{pmatrix}$$

# Permutation Patterns (123)

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$$\begin{pmatrix} 3 & 6 & 7 & 9 & 4 & 8 & 1 & 2 & 5 \\ 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 \end{pmatrix}$$

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# Permutation Patterns (312)

$$\begin{pmatrix} 3 & 6 & 7 & 9 & 4 & 8 & 1 & 2 & 5 \\ 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 \end{pmatrix}$$



# Iterated Maps

Given a function  $f : [0, 1] \rightarrow [0, 1]$  and a point  $x \in [0, 1]$ , consider the behavior of  $\{x, f(x), f(f(x)), f(f(f(x))), \dots\}$ .

# Iterated Maps

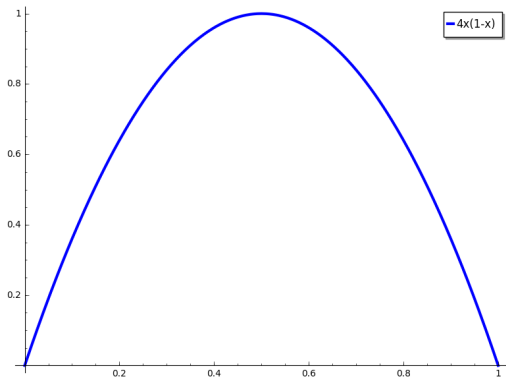
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## Example

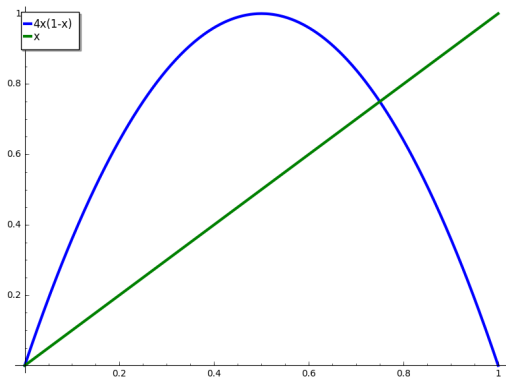
Let  $f(x) = 4x(1 - x)$  and  $x_0 = .2$ . Then, the list of values is:

$[0.20, 0.64, 0.92, 0.28, 0.82, 0.58, 0.97, 0.11, 0.40, \dots]$ .

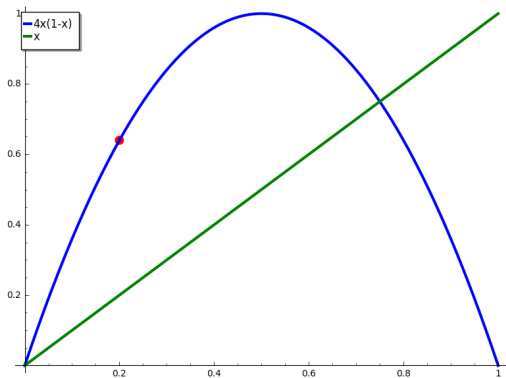
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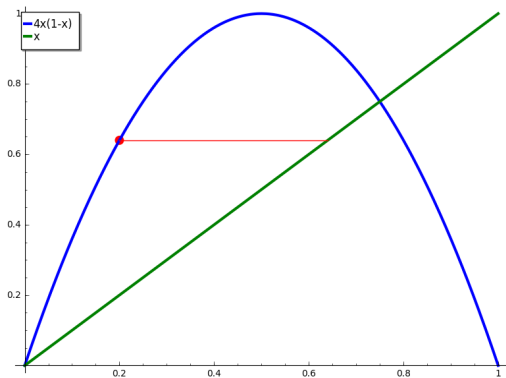
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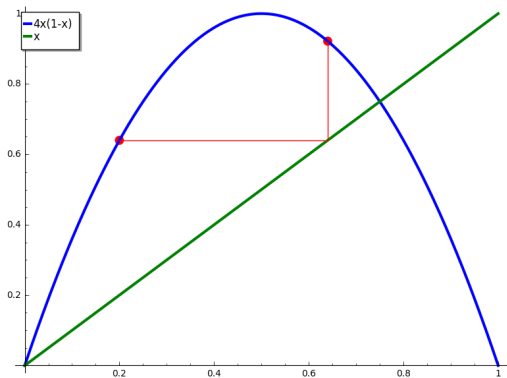
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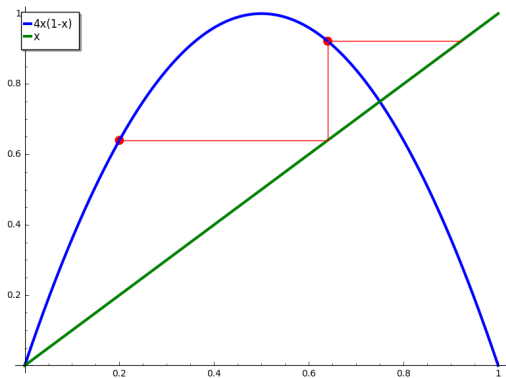
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# Iterated Example (12)

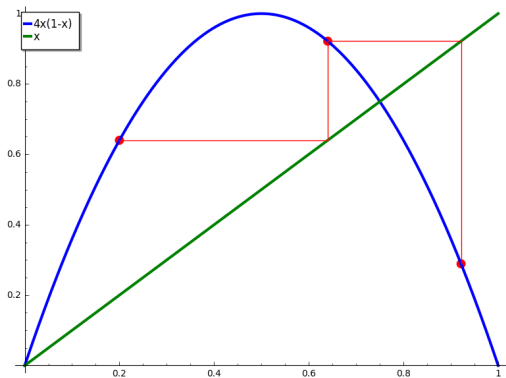


# Iterated Example (12)

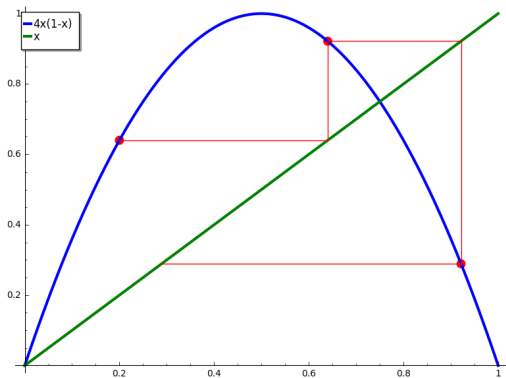




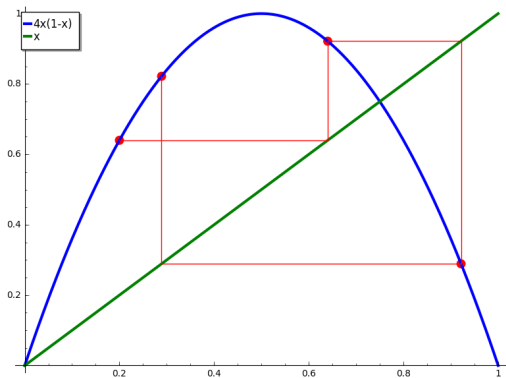
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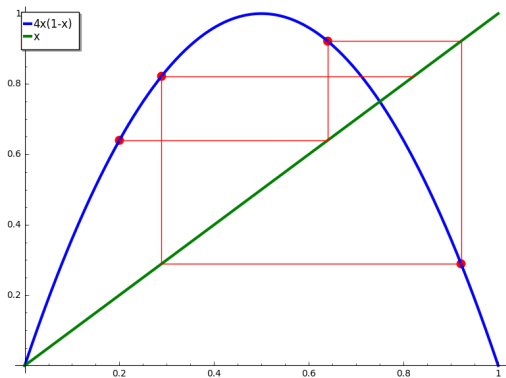
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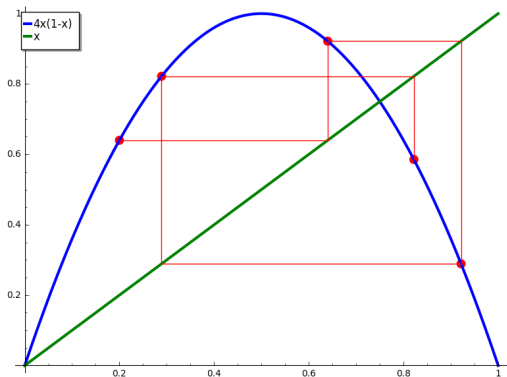
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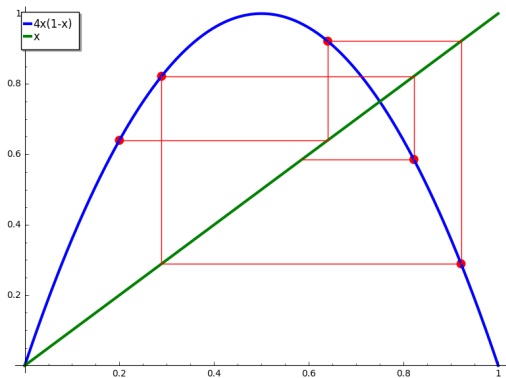
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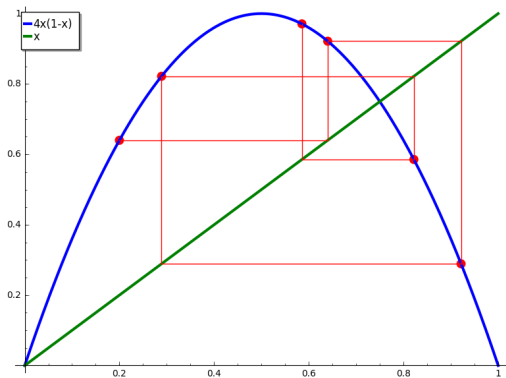
# Iterated Example (35142)



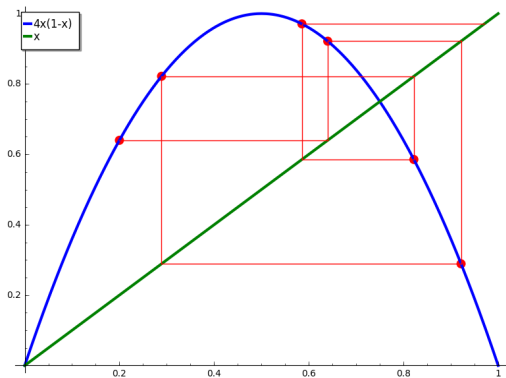
# Iterated Example (35142)



# Iterated Example (351426)

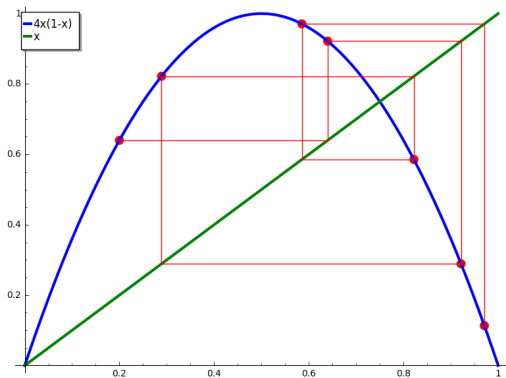


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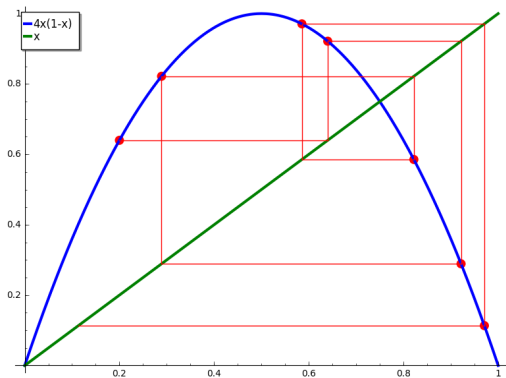




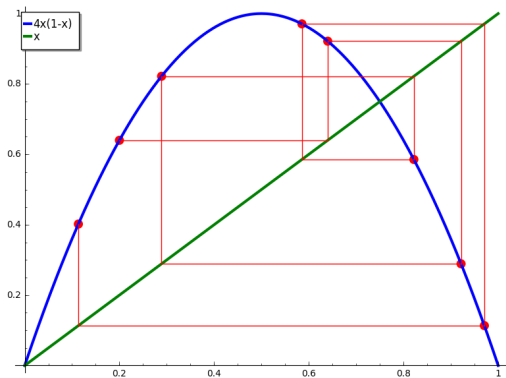
# Iterated Example (4625371)



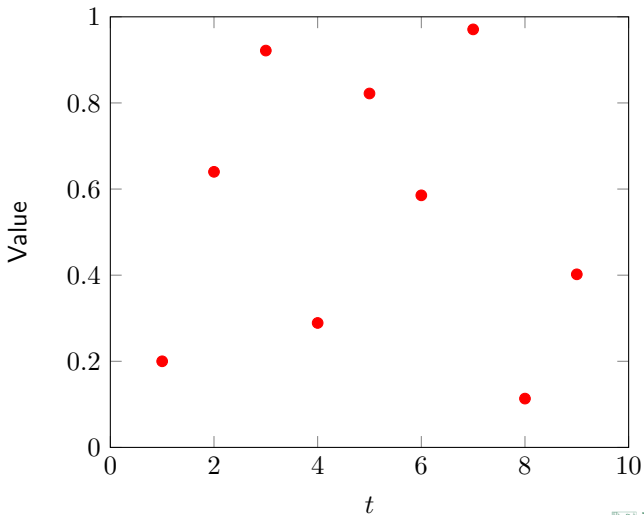
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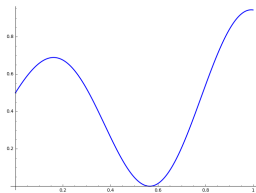
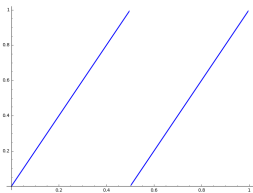
# Iterated Example (57264813)



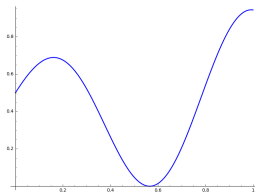
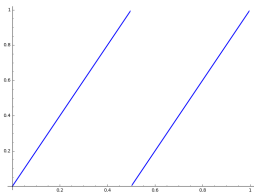
# Iterated Example (268375914)



# Forbidden Patterns



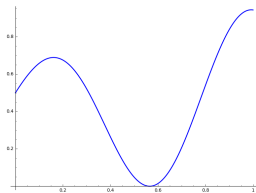
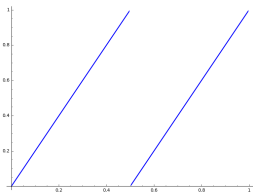
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## Definition (Topological Entropy)

$$TE = \lim_{n \rightarrow \infty} \frac{\log(|\text{Allow}(f)|)}{n - 1}$$

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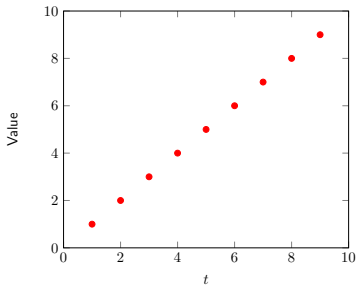
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- C. BANDT AND B. POMPE: *Permutation entropy: A natural complexity measure for time series*, Phys. Rev. Lett. 88, 174102 (2002).
- C. BANDT, G. KELLER, AND B. POMPE: *Entropy of interval map s via permutations*, Nonlinearity 15, 1595 (2002).

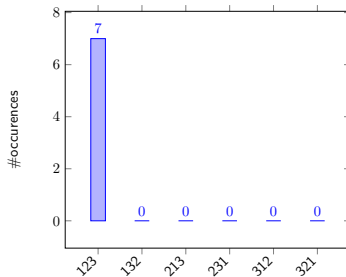
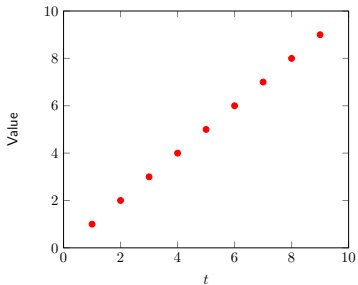


# Simple Time Series

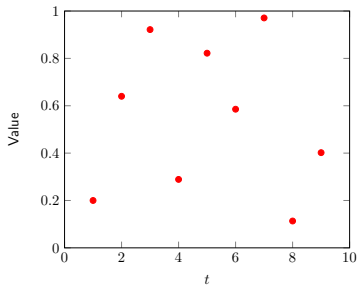




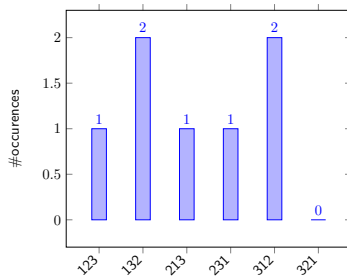
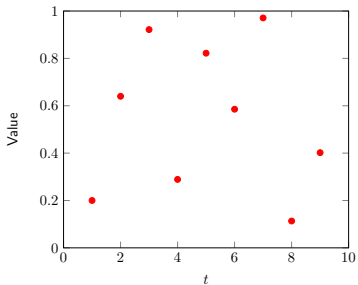
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# Time Series Complexity

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$$PE(\{X_i\}) = -\frac{1}{\log(N!)} \sum_{\pi \in S_n} p_\pi \log(p_\pi)$$

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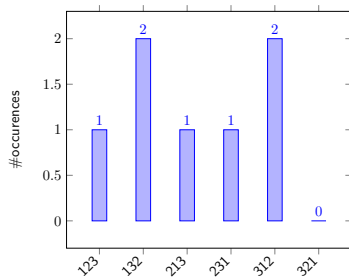
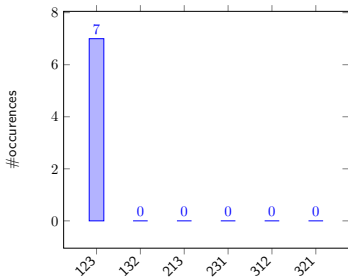
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- M. ZANIN, L. ZUNINO, O.A. ROSSO, AND D. PAPO : *Permutation Entropy and Its Main Biomedical and Econophysics Applications: A Review*, Entropy 2012, 14, 1553-1577.
- M. ZANIN: *Forbidden patterns in financial time series*, Chaos 18 (2008) 013119.
- C. BANDT: *Permutation Entropy and Order Patterns in Long Time Series*, Time Series Analysis and Forecasting, Springer, 2016.

# Histogram Comparison



# I.I.D. Variables

The uniformity assumption is already associated to a null model.

## Lemma

*Let  $\{X_i\}$  be a set of I.I.D. random variables, then  $(\pi, t)$  and  $(\tau, s)$  are independent random variables if and only if  $|t - s| > n$ .*



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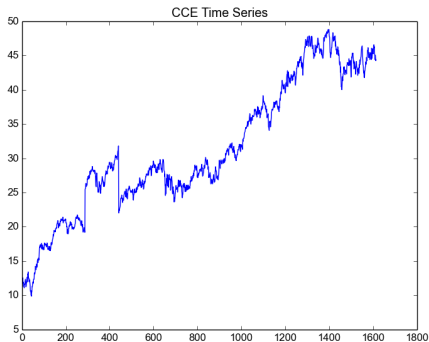
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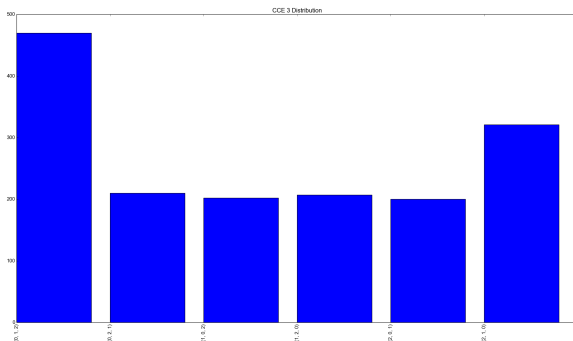
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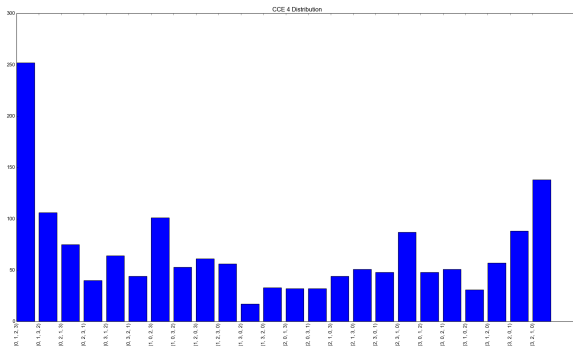
# Stock Data (Closing Prices)



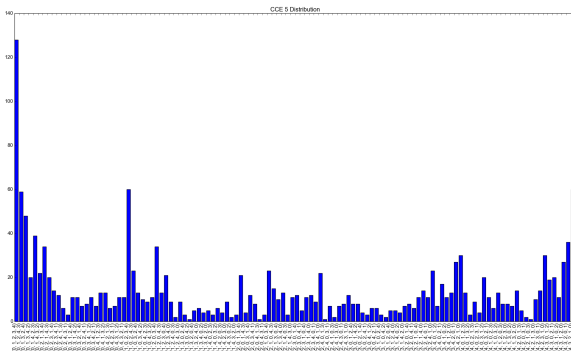
# Stock Data (n=3)



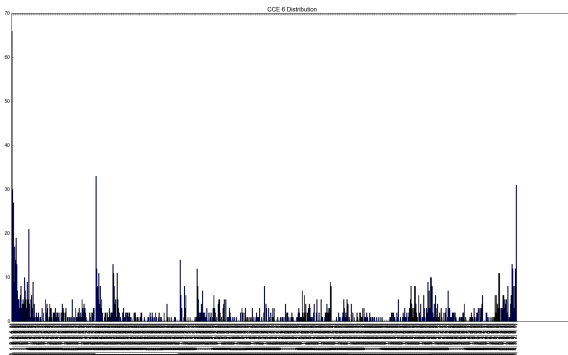
# Stock Data (n=4)



# Stock Data (n=5)



# Stock Data (n=6)



# Random Walk Models

## Definition (Random Walk)

Let  $\{X_i\}$  be a set of I.I.D. random variables and define  $\{Z_i\}$  by

$$Z_j = \sum_{i=0}^j X_j.$$

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## Theorem

*If  $\{Z_i\}$  are defined as above then either  $123 \dots n$  or  $n(n-1)(n-2) \dots 1$  occurs with the highest probability.*



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## Corollary

*If  $\{Z_i\}$  are defined as above and  $n \geq 3$  then the expected distribution of permutations is not uniform.*

# Uniform Walks

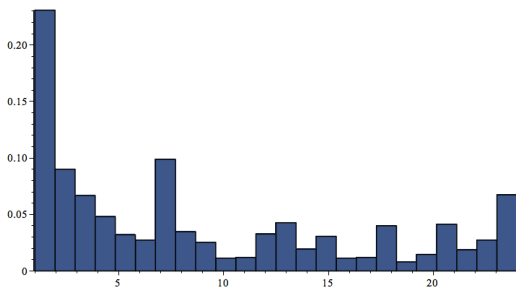
## Definition

Let the  $\{X_i\}$  be defined as uniform random variables over  $[b - 1, b]$  where  $0 < b < 1$  and define  $\{Z_i\}$  with  $Z_j = \sum_{i=0}^j X_i$ .

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# Expected Distributions

## Theorem

Let  $\{Z_i\}$  be a random walk as defined above, then the expected distributions of permutations for  $S_3$  and  $S_4$  are characterized by the following:

$\mathbb{P}(123) = b^2$	$\mathbb{P}(321) = (1 - b)^2$
$\mathbb{P}(132) + \mathbb{P}(231) = (1 - b)b$	$\mathbb{P}(213) + \mathbb{P}(312) = (1 - b)b$
$\mathbb{P}(1234) = b^3$	
$\mathbb{P}(4321) = (1 - b)^3$	
$\mathbb{P}(1243) + \mathbb{P}(1342) + \mathbb{P}(2341) = (1 - b)b^2$	
$\mathbb{P}(1432) + \mathbb{P}(2431) + \mathbb{P}(3421) = (1 - b)^2b$	
$\mathbb{P}(2134) + \mathbb{P}(3124) + \mathbb{P}(4123) = (1 - b)b^2$	
$\mathbb{P}(3214) + \mathbb{P}(4213) + \mathbb{P}(4312) = (1 - b)^2b$	
$\mathbb{P}(1324) + \mathbb{P}(1423) + \mathbb{P}(2314) + \mathbb{P}(2413) + \mathbb{P}(3412) = (1 - b)b^2$	
$\mathbb{P}(4231) + \mathbb{P}(3241) + \mathbb{P}(4132) + \mathbb{P}(3142) + \mathbb{P}(2143) = (1 - b)^2b$	

# Hyperplanes

## Example

In order for the pattern 1342 to appear in the time series we need the following inequalities to hold:

- $X_1 > 0$
- $X_2 > 0$
- $X_3 < 0$
- $X_3 > X_2$
- $X_3 < X_1 + X_2$

# Hyperplanes

## Example

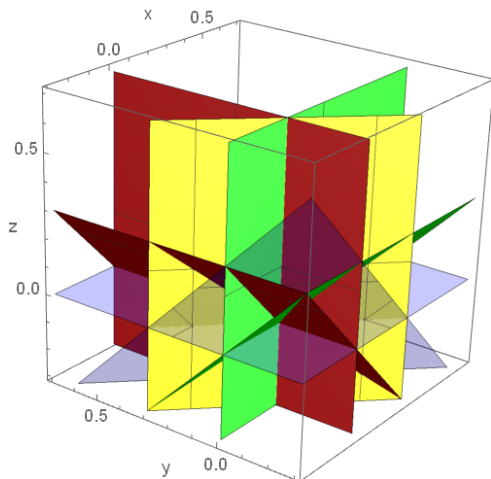
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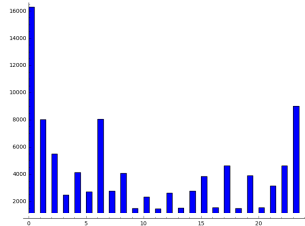
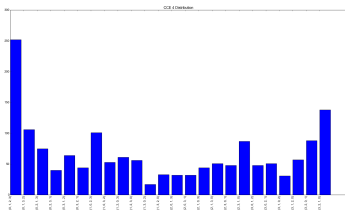
$\pi$	$\mathbb{P}(\pi)$	$\pi$	$\mathbb{P}(\pi)$
123	$b^2$	132	$(b-1) - \frac{3}{2}(b-1)^2$
213	$(b-1) - \frac{3}{2}(b-1)^2$	231	$\frac{1}{2}(b-1)^2$
312	$\frac{1}{2}(b-1)^2$	321	$(b-1)^2$



# Integration Regions



# Stock Comparison





## Equivalence Classes

### Theorem (Martinez 2015)

*Let  $\pi, \tau \in S_n$ . The  $\mathbb{P}(\pi) = \mathbb{P}(\tau)$  for every probability distribution on the  $\{X_i\}$  if and only if  $\pi$  and  $\tau$  are equivalent under repeated application of the reverse complement operation on bounded cylindrical blocks.*

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{1234}, {1243, 2134}, {1324}, {1342, 3124}, {1423, 2314}, {1432, 2143, 3214}, {2341, 3412, 4123}, {2413}, {2431, 4213}, {4231}, {3142}, {3241, 4132}, {3421, 4312}, {4321}
{12345}, {14325}, {21354}, {21453}, {25314}, {41352}, {45312}, {52341}, {54321}, {12543, 32145}, {13245, 12435}, {13425, 14235}, {15243, 32415}, {15342, 42315}, {15432, 43215}, {21345, 12354}, {21435, 13254}, {21543, 32154}, {23145, 12534}, {23415, 15234}, {24153, 31524}, {24315, 15324}, {24513, 35124}, {24531, 53124}, {25134, 23514}, {25341, 52314}, {25413, 35214}, {25431, 53214}, {31245, 12453}, {31425, 14253}, {31542, 42153}, {32514, 25143}, {32541, 52143}, {35142, 42513}, {35241, 52413}, {41235, 13452}, {41253, 31452}, {41325, 14352}, {41523, 34152}, {41532, 43152}, {42135, 13542}, {42351, 51342}, {45231, 53412}, {51324, 24351}, {51423, 34251}, {51432, 43251}, {53142, 42531}, {53241, 52431}, {54123, 34521}, {54132, 43521}, {54213, 35421}, {54231, 53421}, {54312, 45321}, {31254}, {43125, 14532}, {34125, 14523}, {13524, 24135}, {35412, 52134, 45213, 23541}, {51243, 32451}, {43512, 45132}, {23451, 45123, 34512, 51234}, {21534, 23154, 15423, 34215}

# Related Metrics

## Definition

To measure how closely the distribution matches the conditions of Martinez we compute

$$g_n(T) = \sum_{\Lambda_i \subset S_n} \sum_{\pi \in \Lambda_i} p_\pi |p_\pi - \mu_i|,$$

or “equivalently”,

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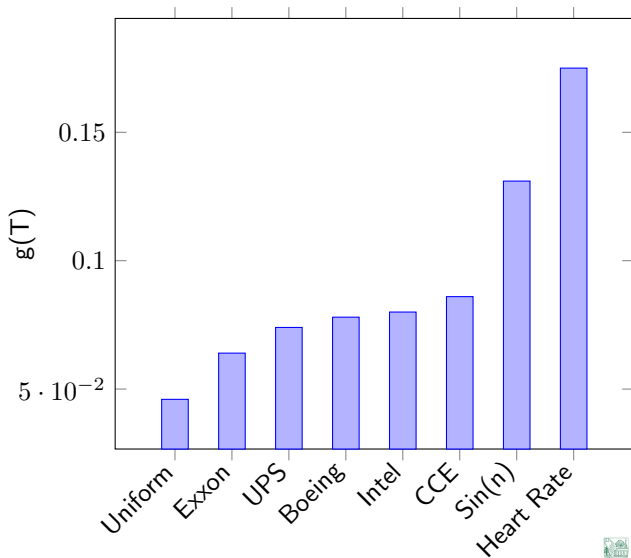
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### Corollary

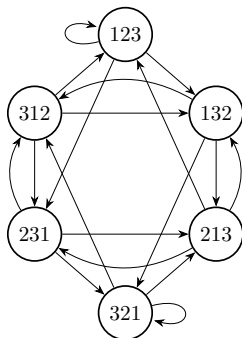
If  $Z^N = \{Z_i\}_{i=1}^N$  is a random walk, then

$$\lim_{N \rightarrow \infty} g_n(Z^N) = 0 \text{ and } \lim_{N \rightarrow \infty} h_n(Z^N) = 0.$$

# Examples



# Beyond Distributions



(a) The permutation graph for  $n = 3$  whose edges,  $\pi \rightarrow \tau$ , are weighted with probability  $P_X(\pi \rightarrow \tau)$ .

# Uniform Walks

## Lemma

Let  $\pi, \tau \in S_n$ . Then,

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## Uniform Walks

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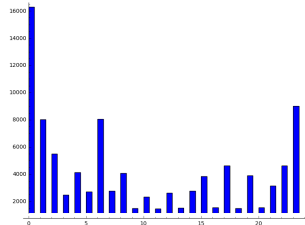
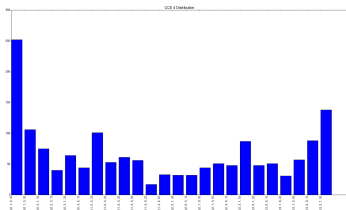
$$\mathbb{P}(\tau \rightarrow \pi) = \frac{\mathbb{P}(\tau \wedge \pi)}{\mathbb{P}(\tau)}$$

### Lemma

*Let  $\{Z_i\}$  be a random walk as above. Then, the transition probability between permutations are not the uniform distribution.*



# CCE Random Walk



## Further Reading

- M. RIEDL, A. MÜLLER, AND N. WESSEL: *Practical considerations of permutation entropy*, Eur. Phys. J. Special Topics 222, 249262 (2013).
- M. ZANIN, L. ZUNINO, O.A. ROSSO, AND D. PAPO : *Permutation Entropy and Its Main Biomedical and Econophysics Applications: A Review*, Entropy 2012, 14, 1553-1577.
- L. ZUNINO, M. ZANIN, B. M. TABAKE, D. G. PREZ, O. A. ROSSO: *Forbidden patterns, permutation entropy and stock market inefficiency*, Physica A 388 (2009) 2854-2864.

## Code (Try it yourself...)

```
https://math.dartmouth.edu/  
~ddeford/time\_series.html
```

That's all...

Thank You!