

Trend Switching Phenomena in Financial Markets

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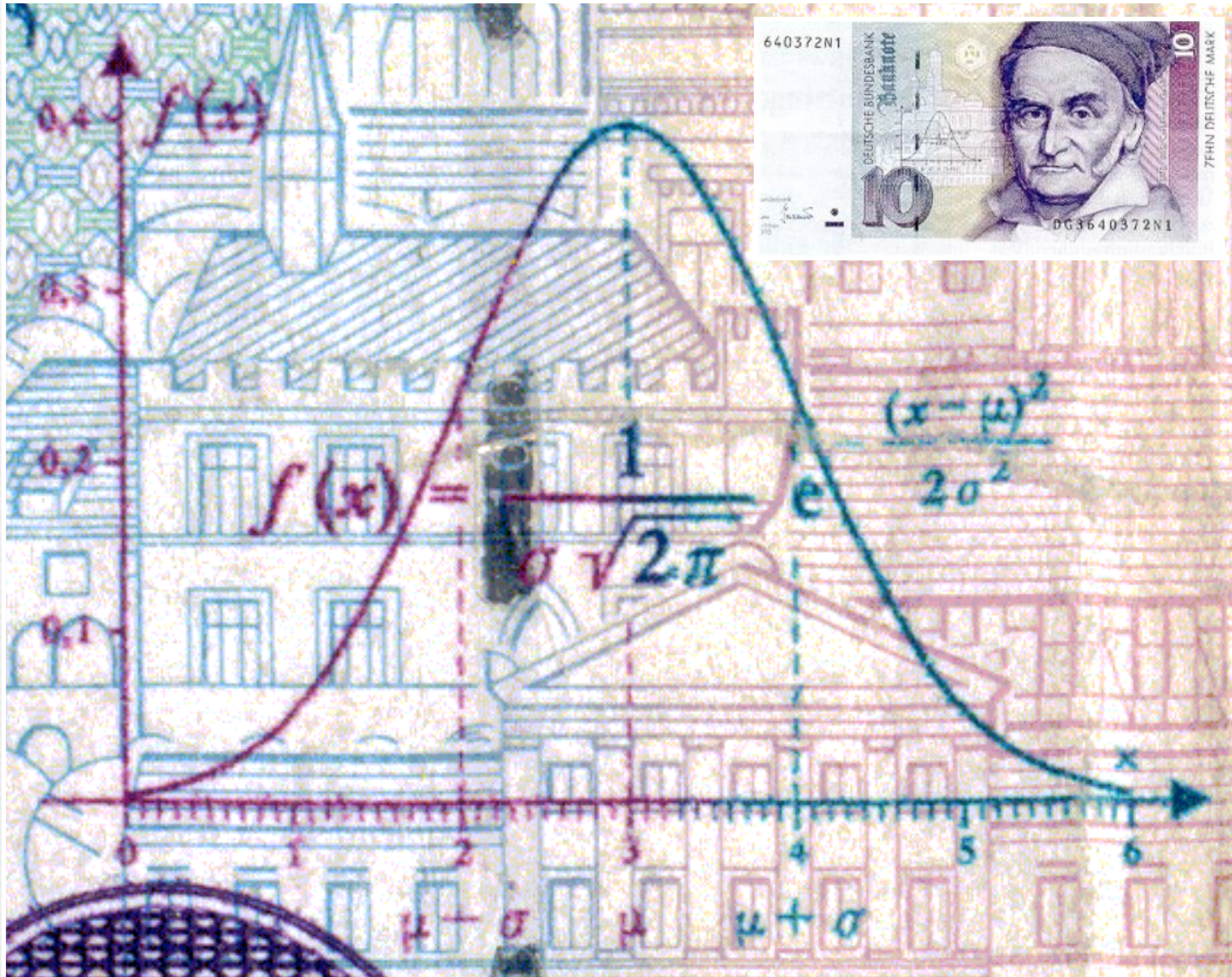
ETH

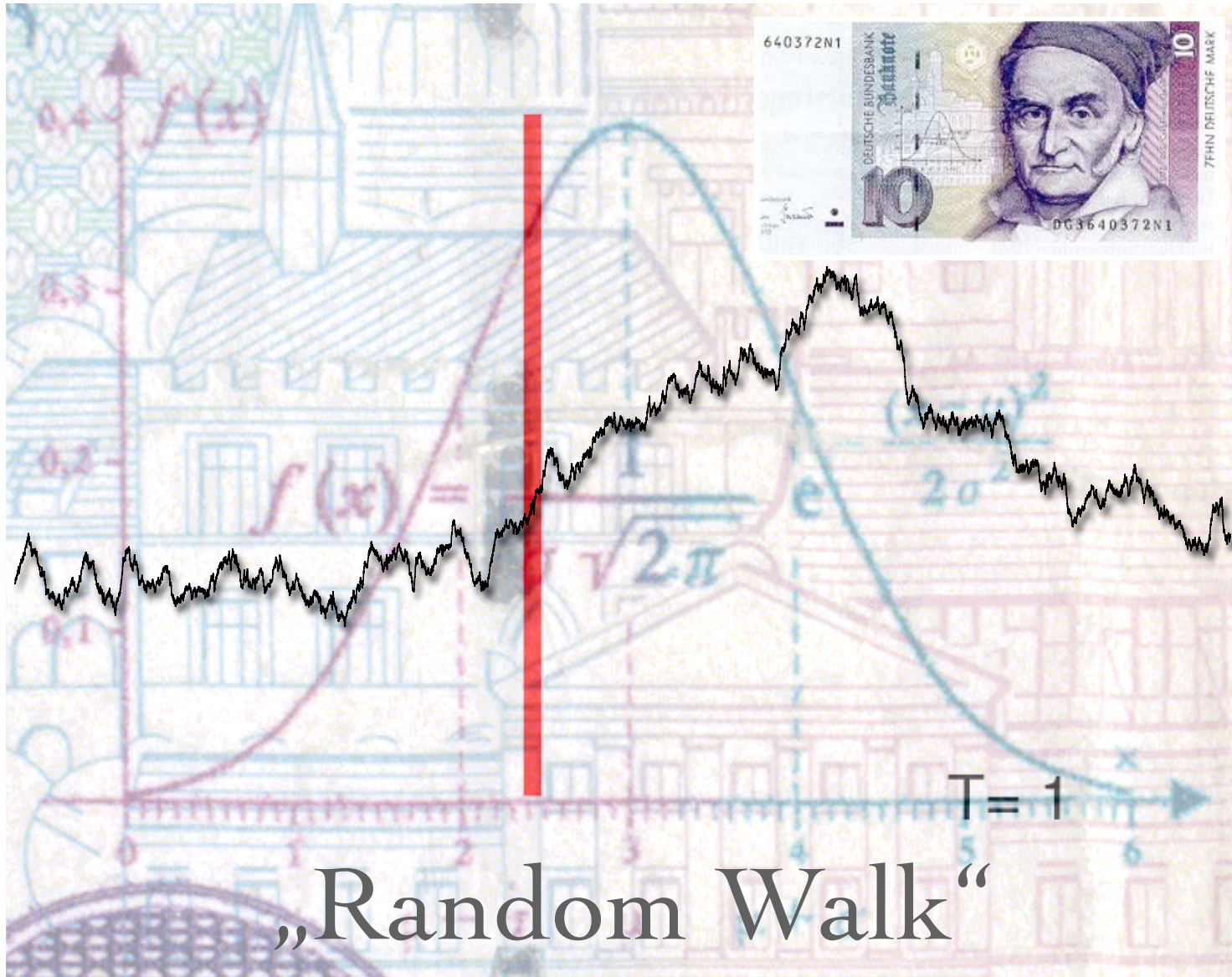
Eidgenössische Technische Hochschule Zürich
Swiss Federal Institute of Technology Zurich



Artemis







„Random Walk“

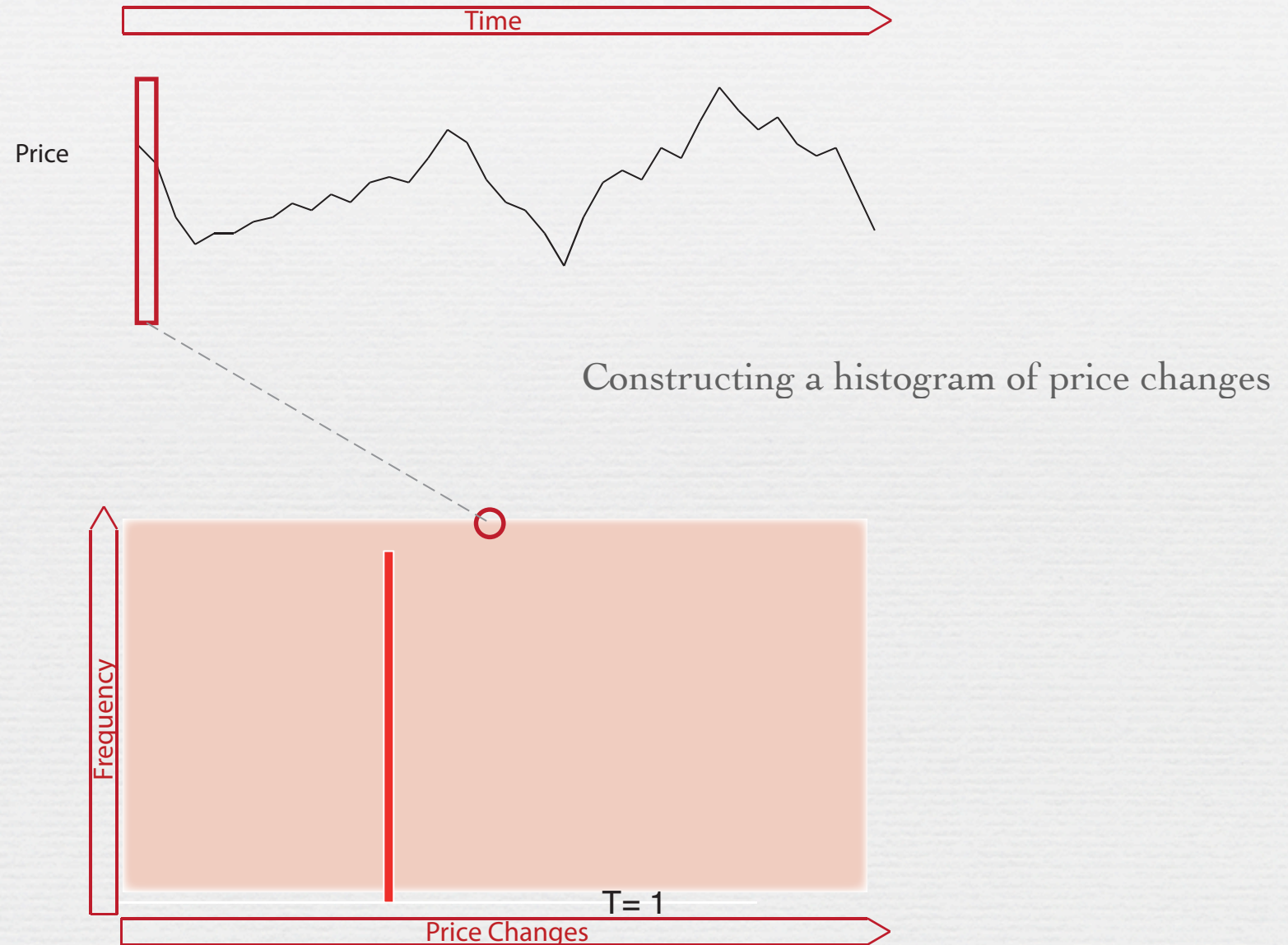
Outline

- (i) “Econostatics” vs. “Econodynamics”
- (ii) Analysis of complex pattern-based correlations in high frequency financial market data.
- (iii) Switching processes in financial markets.
- (iv) Agent-based model reproducing empirical stylized facts based on an order book structure.

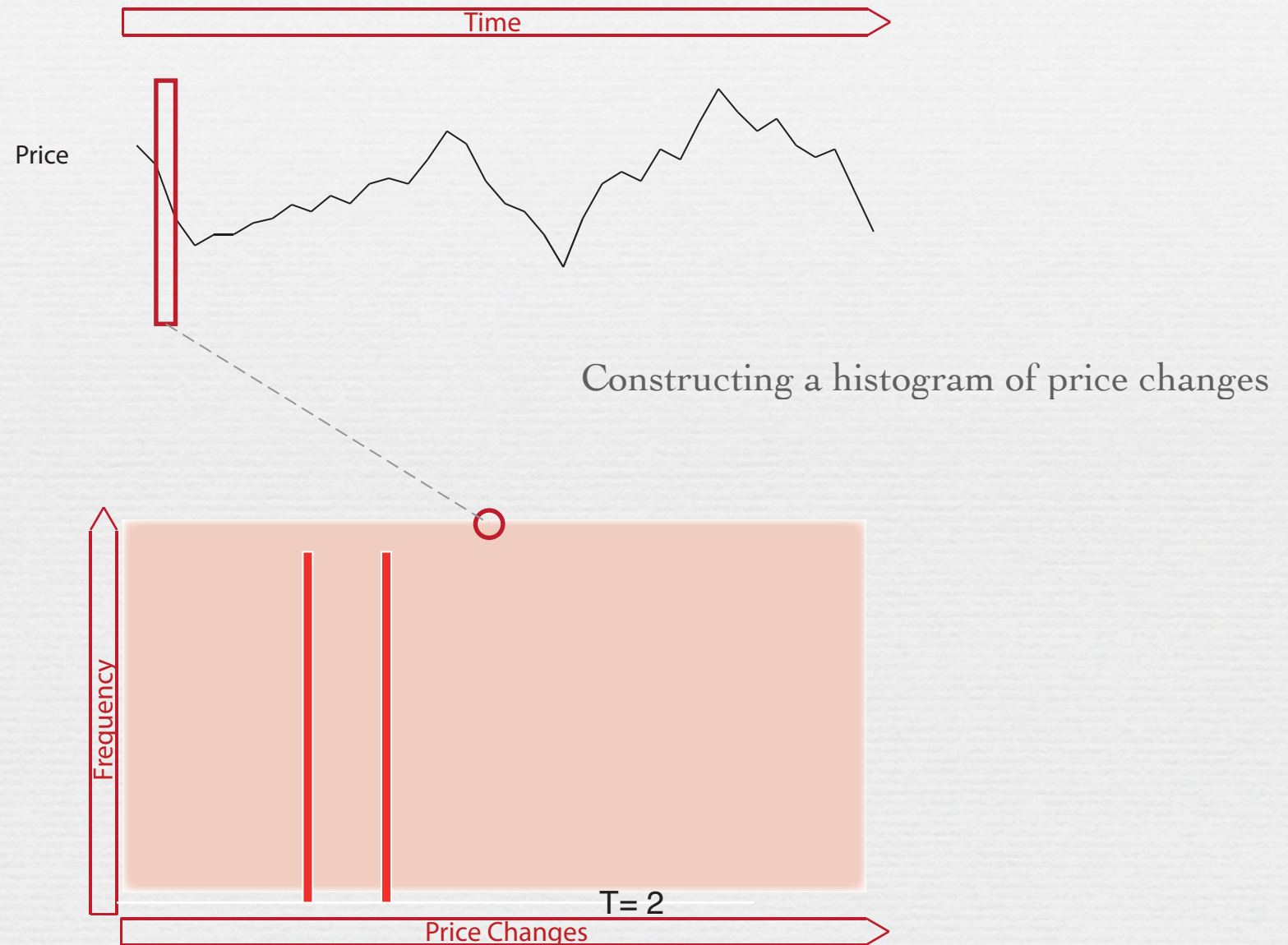
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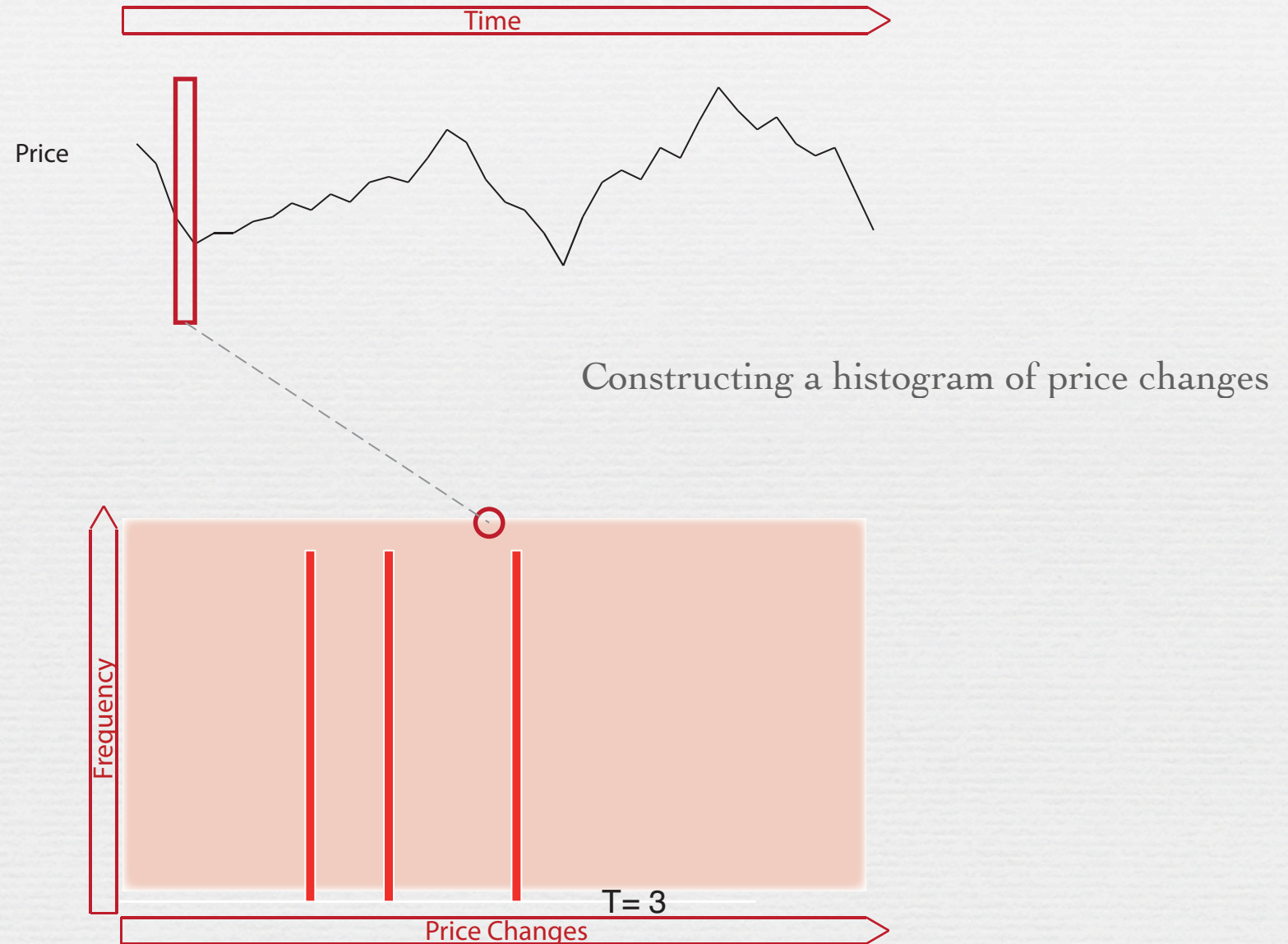
“Econostatics”



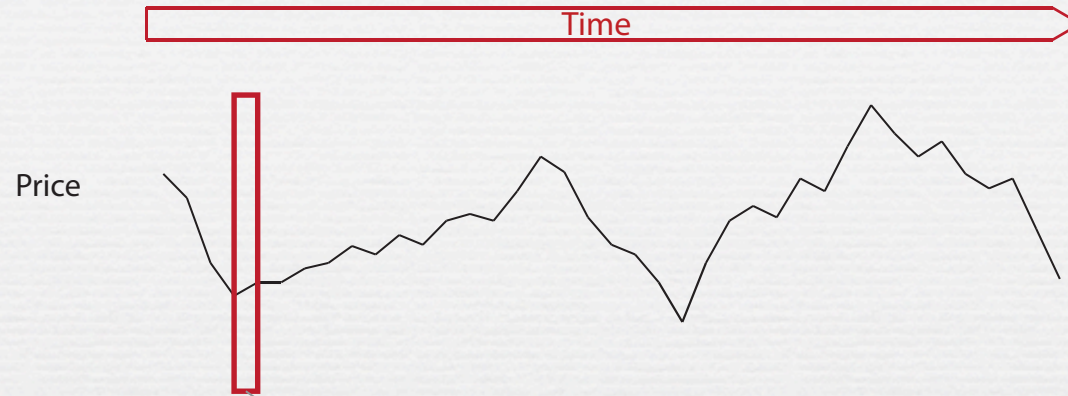
“Econostatics”



“Econostatics”



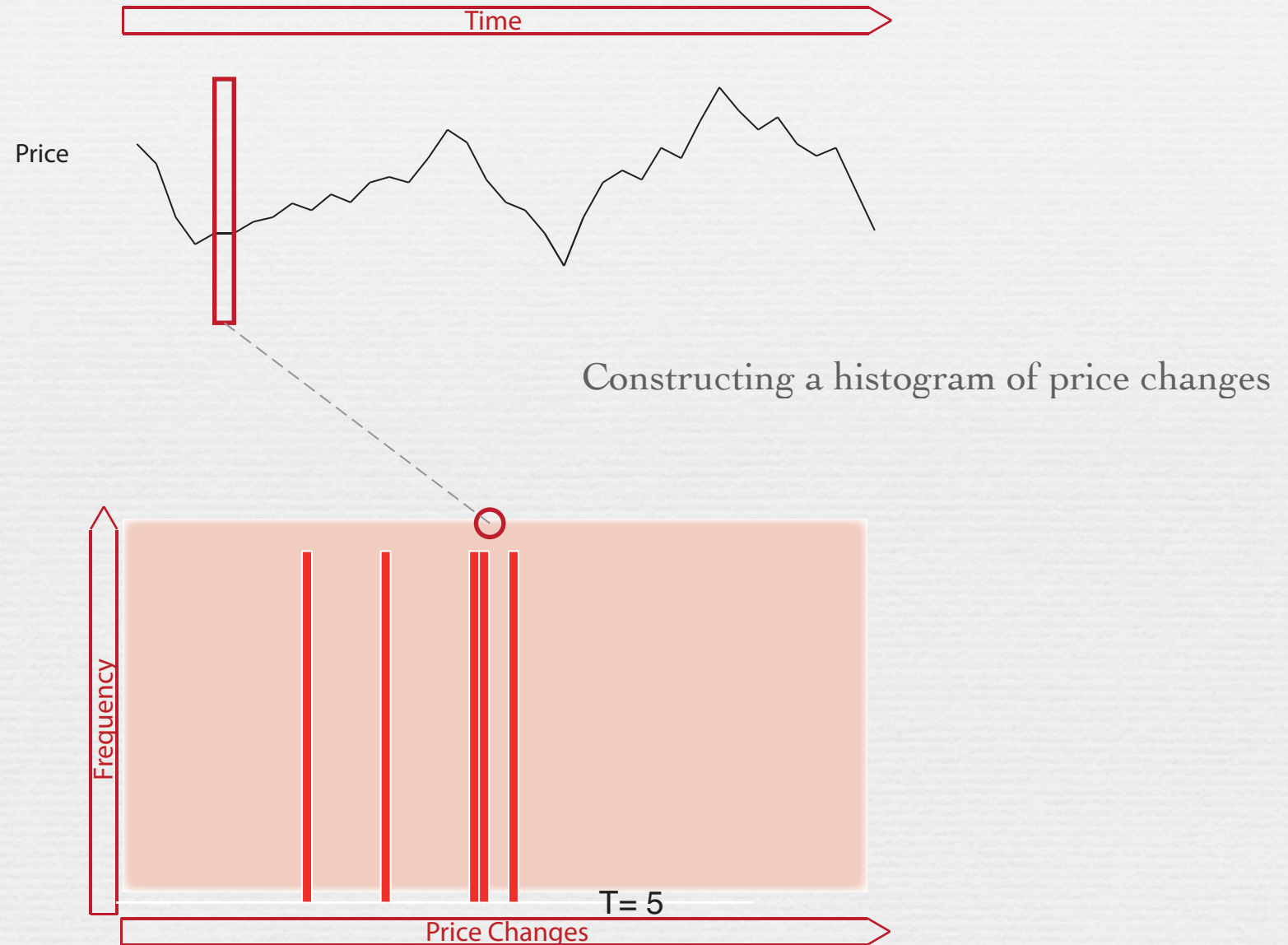
“Econostatics”



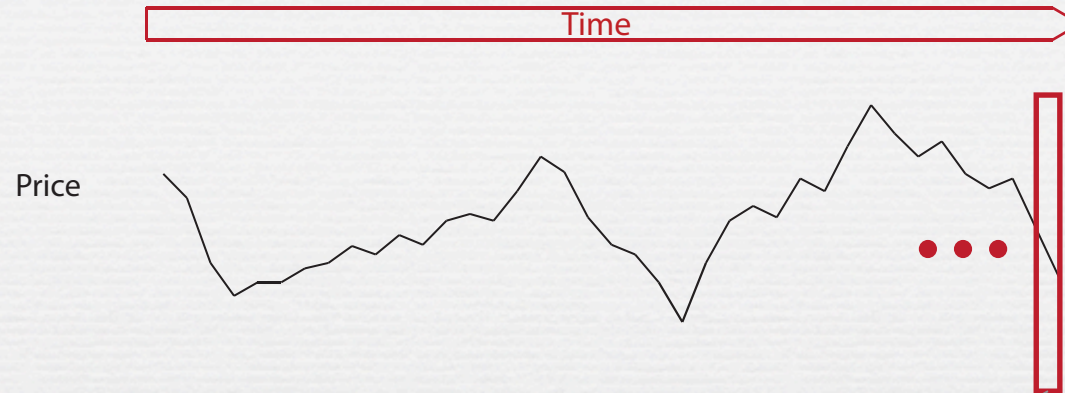
Constructing a histogram of price changes



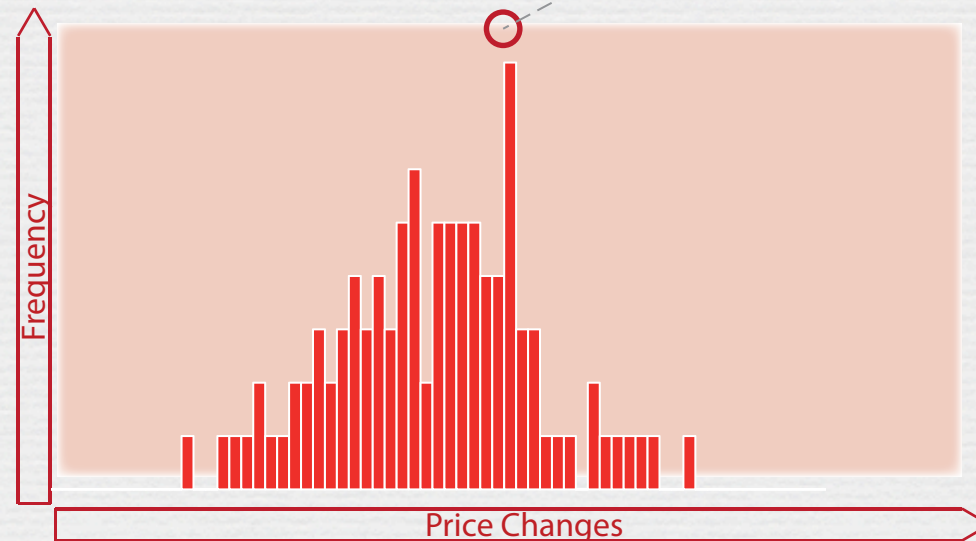
“Econostatics”



“Econostatics”

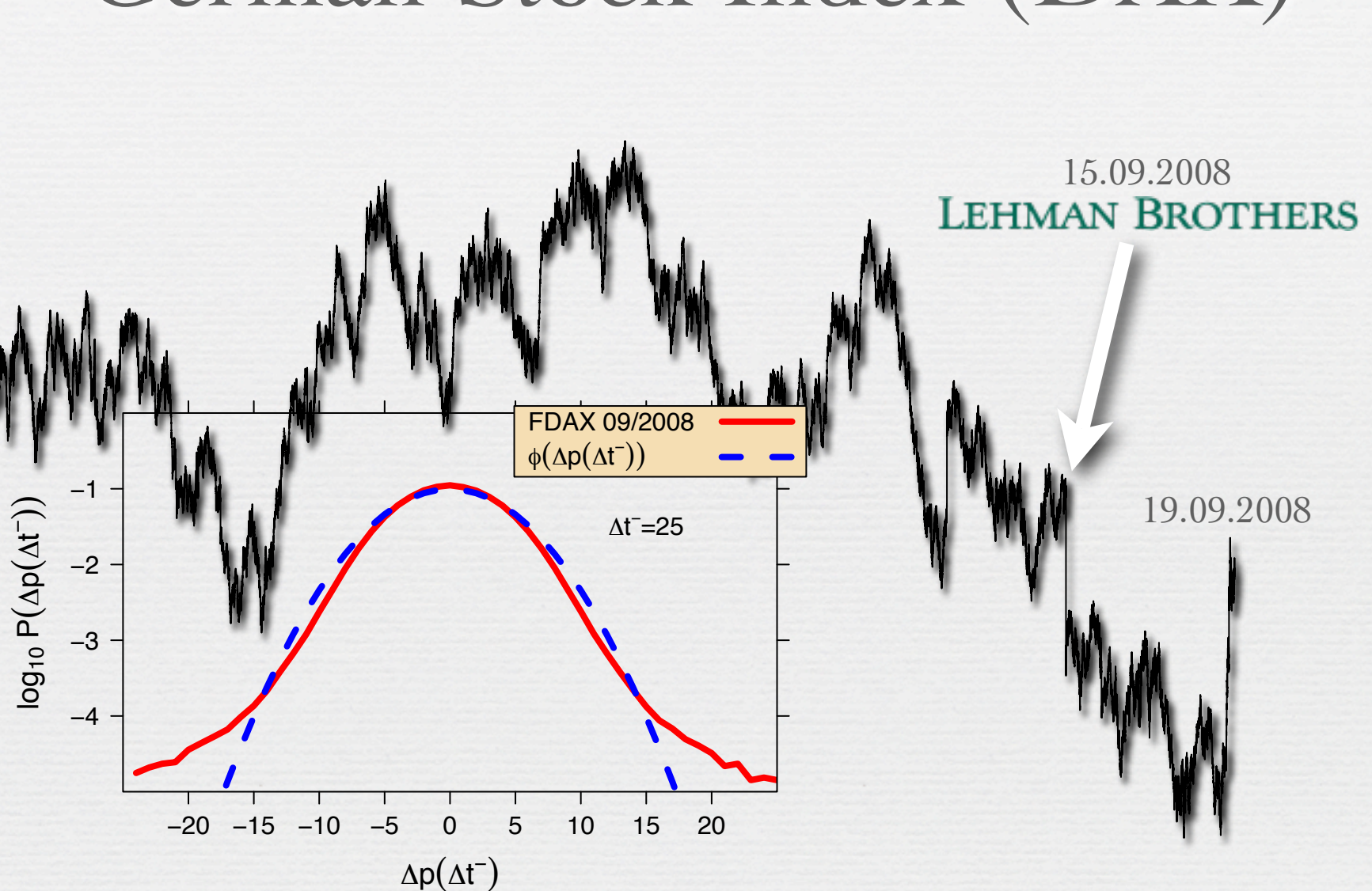


Constructing a histogram of price changes



German Stock Index (DAX)

20.06.2008

X. Gabaix et al., Nature **423**, 267-270 (2003).

Tobias Preis

$$\phi(\Delta p(\Delta t^-)) = u \exp(-v \Delta p^2)$$

April 19, 2011

Milestones of trading

CBOT trading pit, Chicago, 1993



Floor trading



Electronic trading



Deutsche Bank AG headquarter, Frankfurt, 2007

1602 Options on shares of East-Indian Company (NL)

1634-36 Tulip bulb speculation bubble (NL)

...

1848 CBOT: trading of future contracts (USA)

...

1978 CBOE: trading of option contracts (USA)

...

1990 DTB: futures and options (D)

...

1998 EUREX: DTB+SOFFEX (D, CH)

...

Sequence of Price Changes

FDAX



Random Walk



- Empirical stylized facts of financial time series include:
 - ➔ Fat-tailed return distributions / Power law (Gene Stanley et al.)
 - ➔ Log-periodic power law / Bubble detection (Didier Sornette et al.)
 - ➔ Non-trivial scaling behavior of the returns,
 - ➔ Volatility clustering
 - ➔ ...

LEHMAN BROTHERS

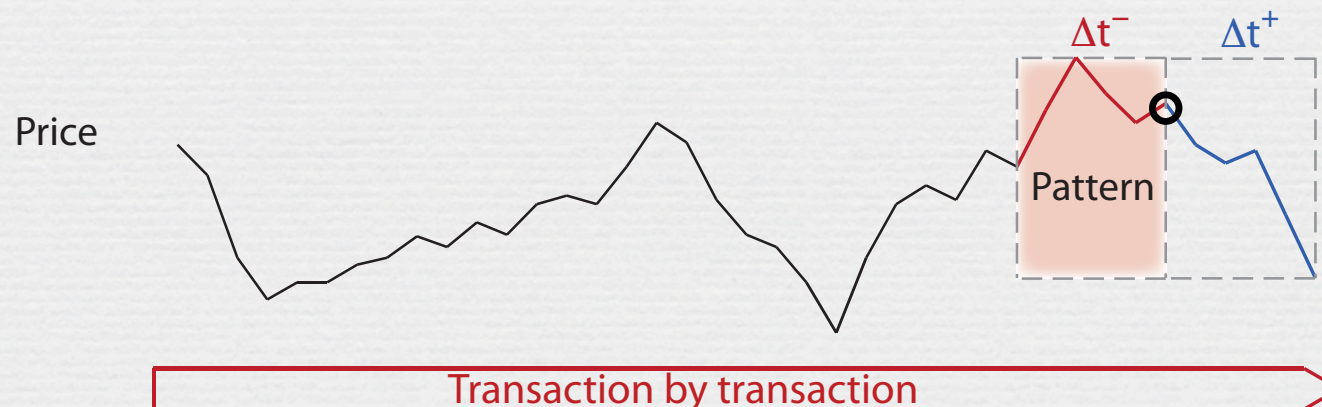
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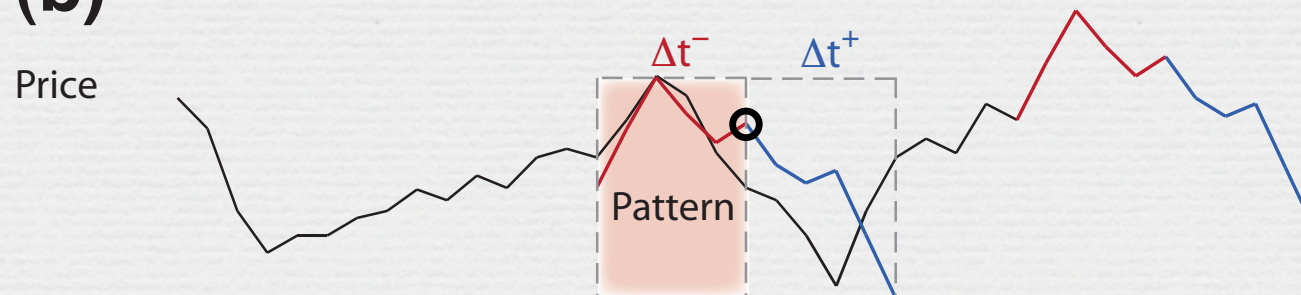
Fluctuation Patterns

„The aim is to compare the current reference pattern of time interval length Δt^- with all previous patterns in the time series.“

(a) Pattern conformity analysis of financial market fluctuations

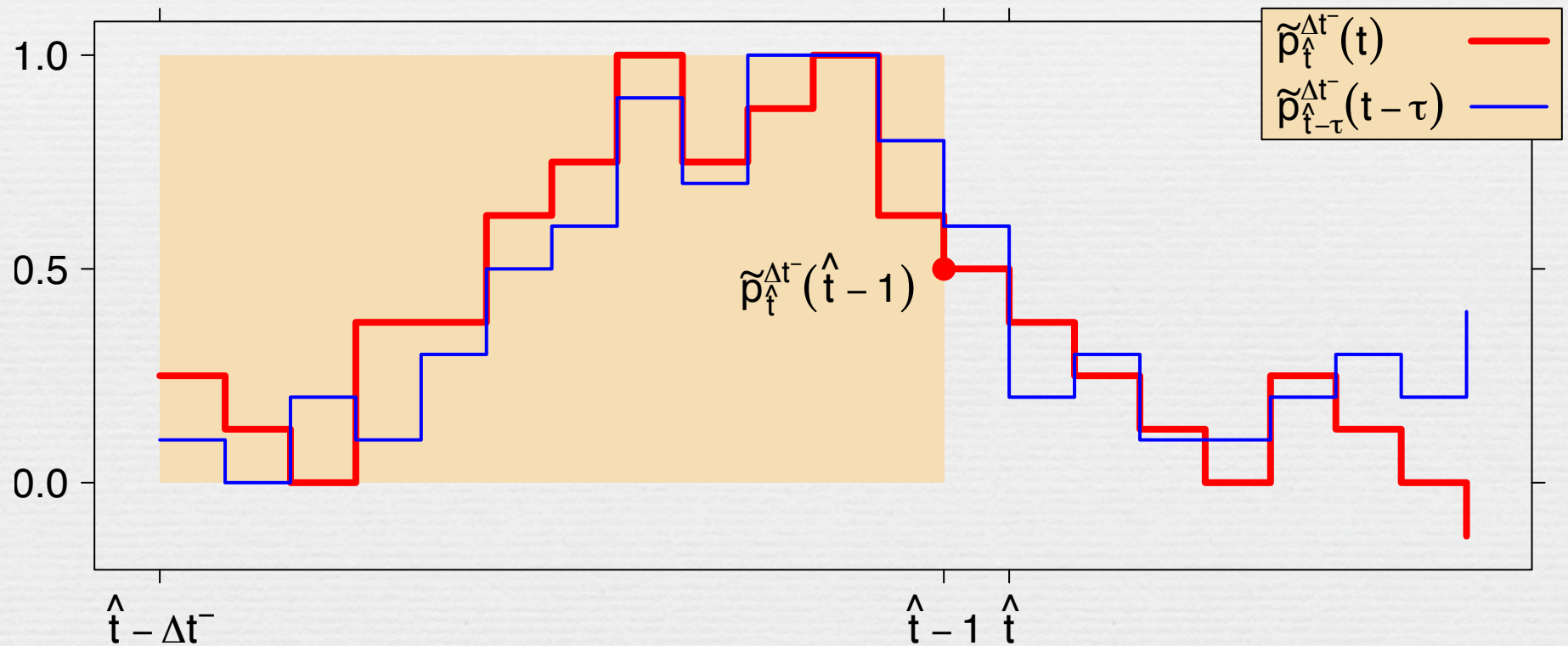


(b)



Fluctuation Patterns

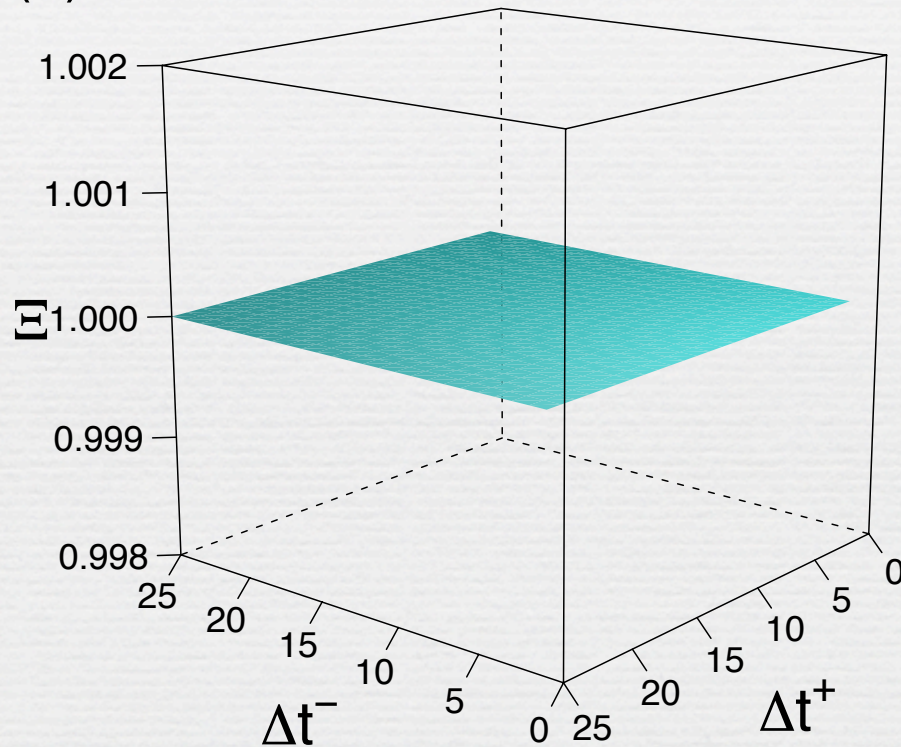
True range adapted modified time series



Pattern Conformity / Trivial Cases

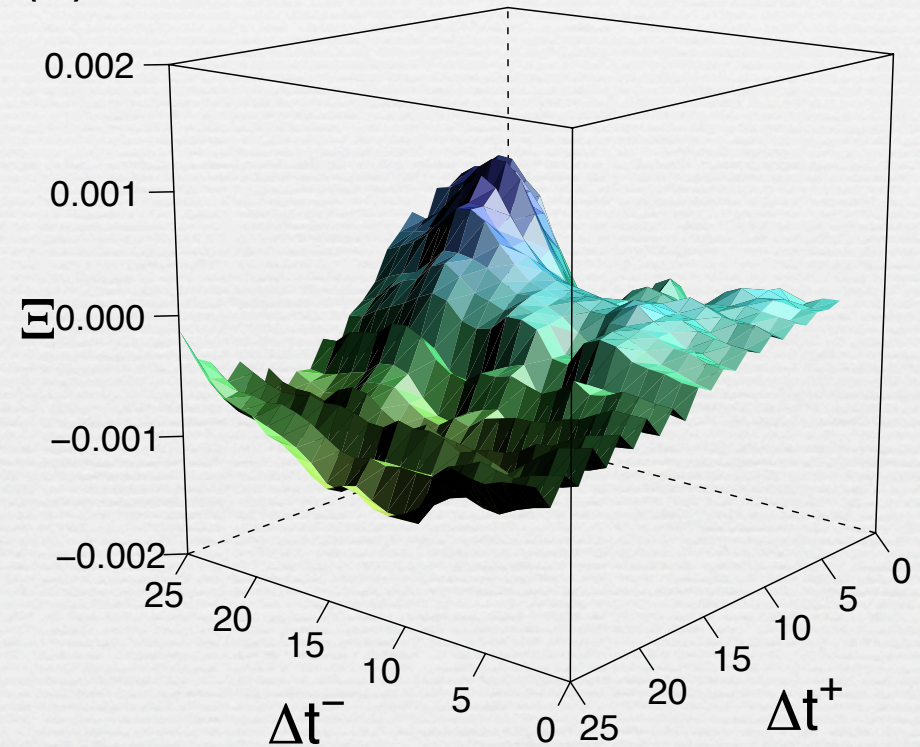
- Straight Line

(a)



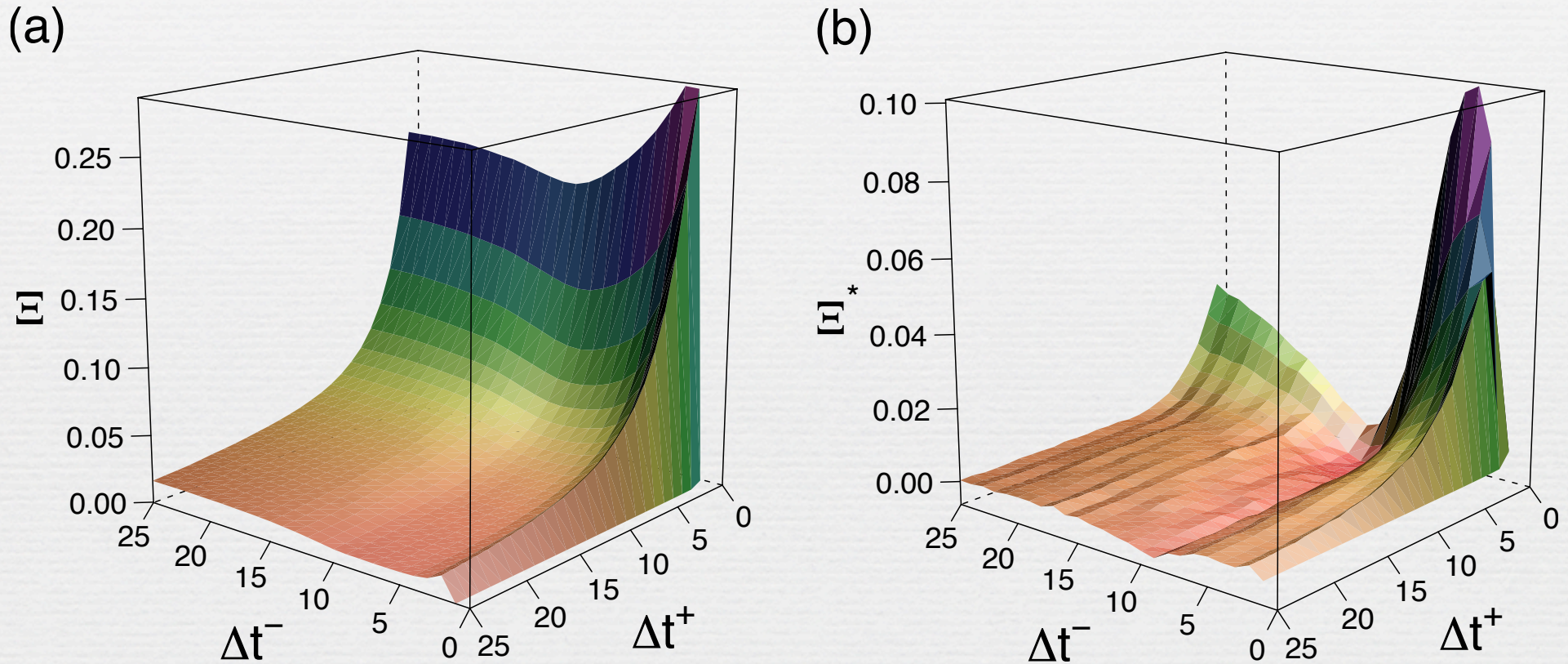
- Random Walk

(b)



$$\Xi_{\chi}(\Delta t^+, \Delta t^-) = \frac{\xi_{\chi}(\Delta t^+, \Delta t^-)}{\sum_{\hat{t}=\Delta t^-}^{T-\Delta t^+} \sum_{\tau=\tau^*}^{\hat{t}} \frac{|\text{sgn}(\omega_{\hat{t}}^{\Delta t^-}(\tau, \Delta t^+))|}{\exp(\chi Q_{\hat{t}}^{\Delta t^-}(\tau))}}$$

Pattern Conformity / FDAX



Complex correlations for financial market time series – especially for large pattern lengths.

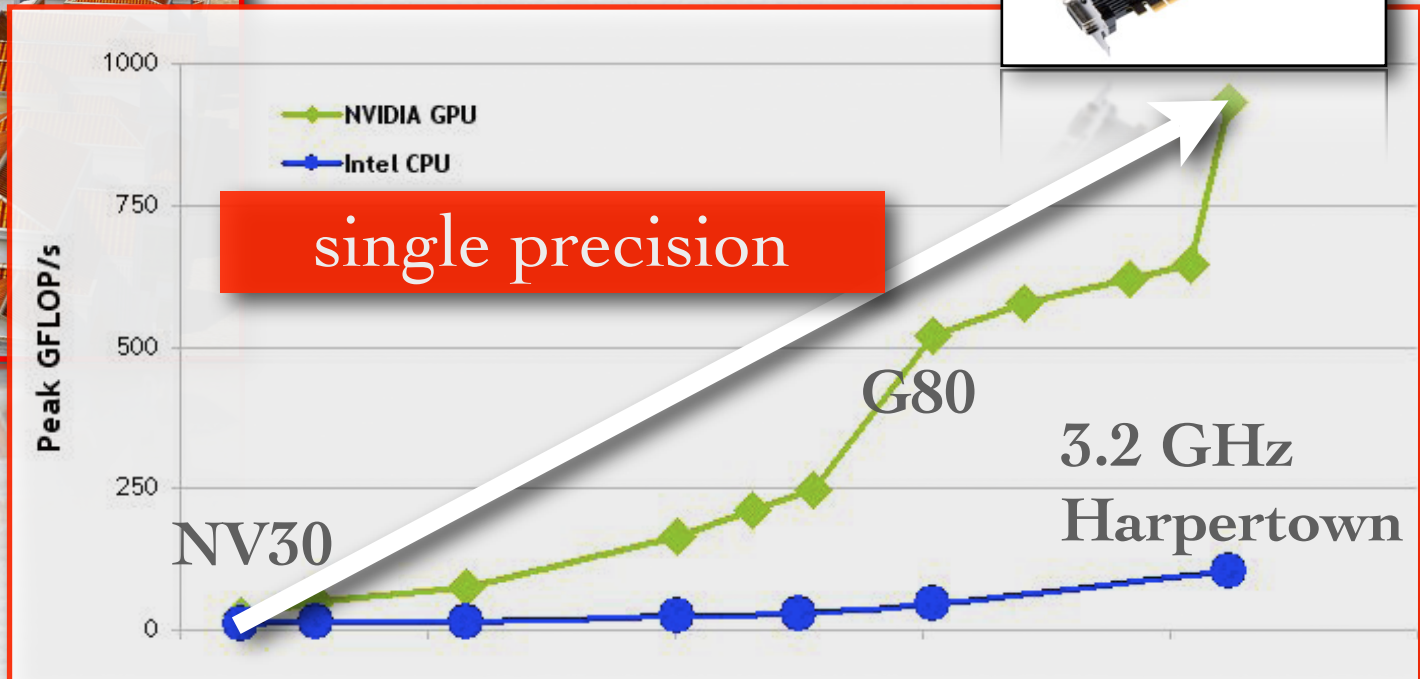
GPU-Computing



Realistic illustrations
Driving force:
computer game
industry



GT200



Source: NVIDIA CUDA programming guide

2003 2006 2008

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```
artemis-2:eurex_time_series tobias$ head -n 20 index_fdax_032007_pvt.filter
```

```
2007 1 2 8 0 9 78 6674.5 608
```

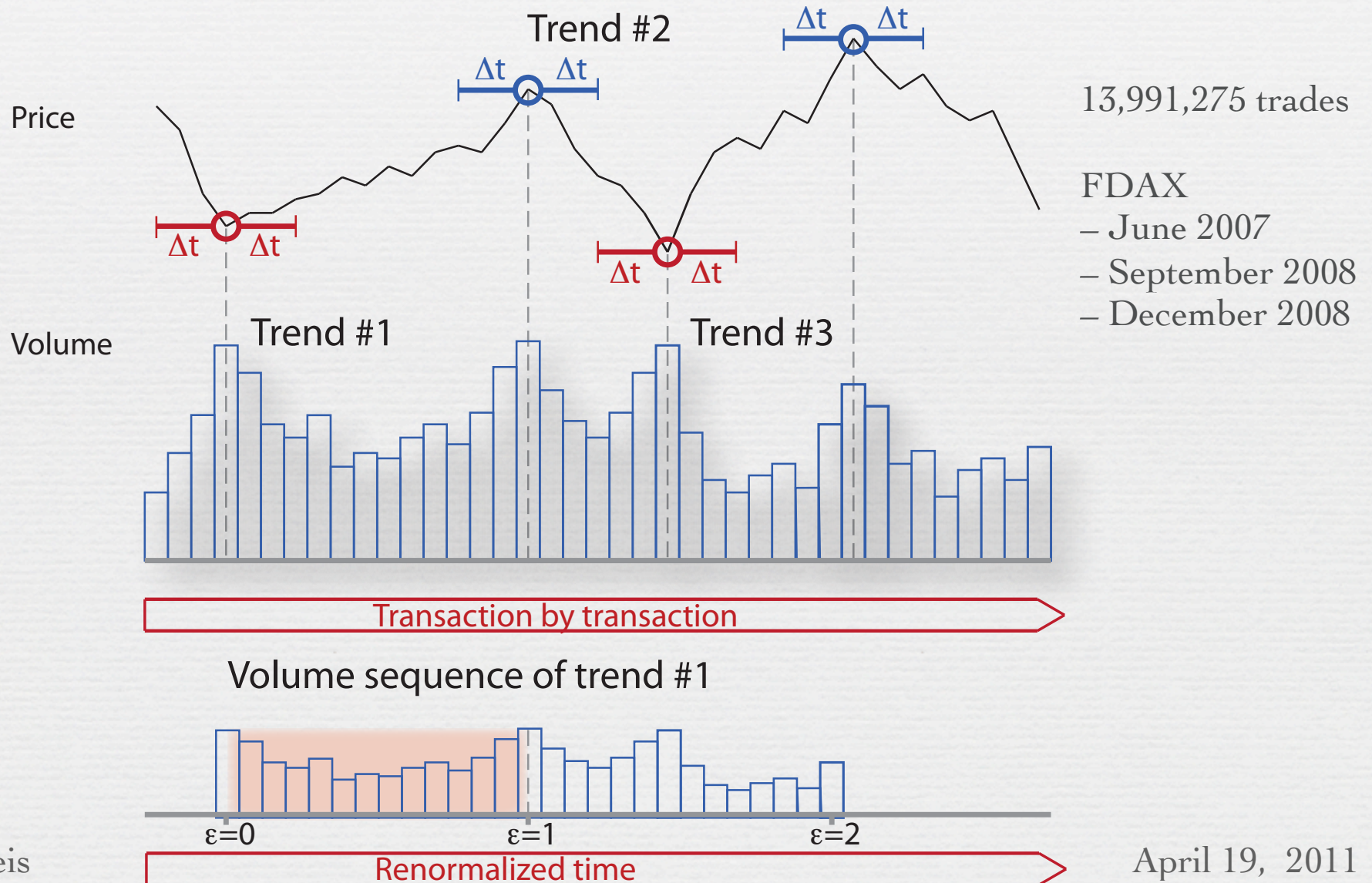


Time Series – Data Structure

```
artemis-2:eurex_time_series tobias$
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Rescaling Analysis

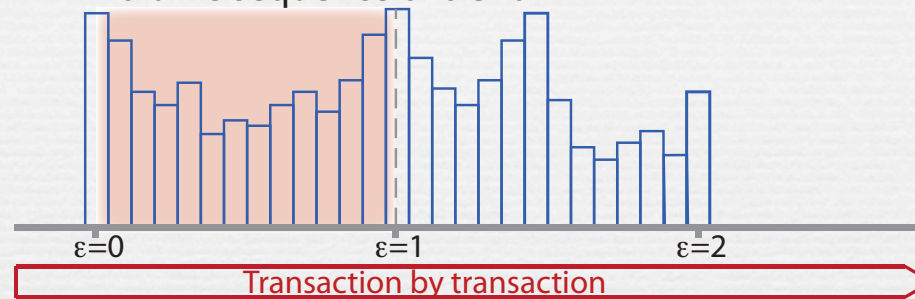
(a) Determination of local price extrema and renormalization



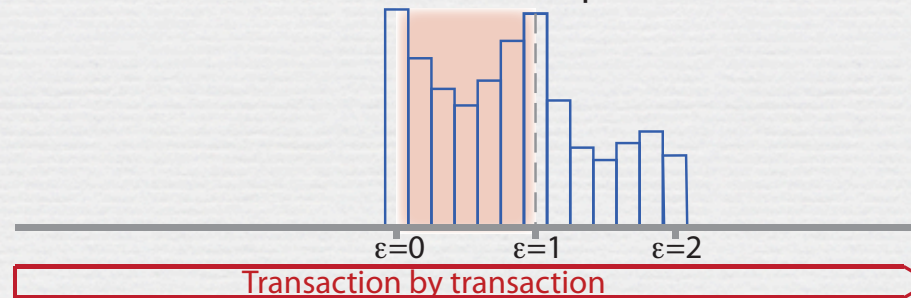
Rescaling Analysis

Extraction of trend sequences

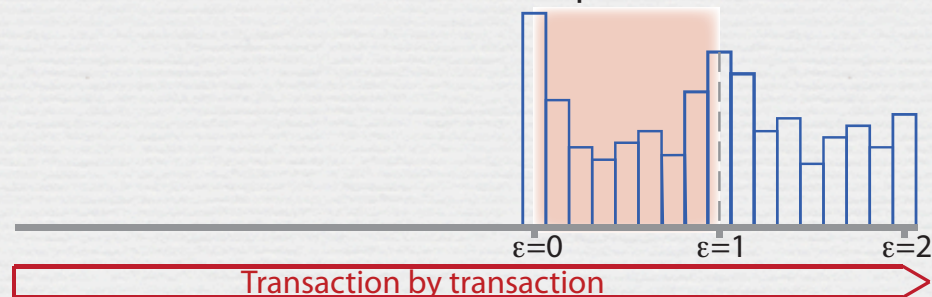
Volume sequence of trend #1



Volume sequence of trend #2



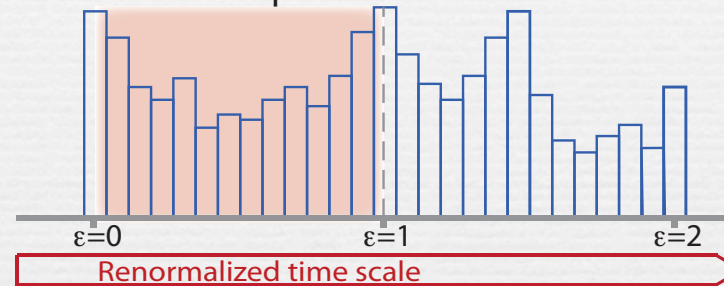
Volume sequence of trend #3



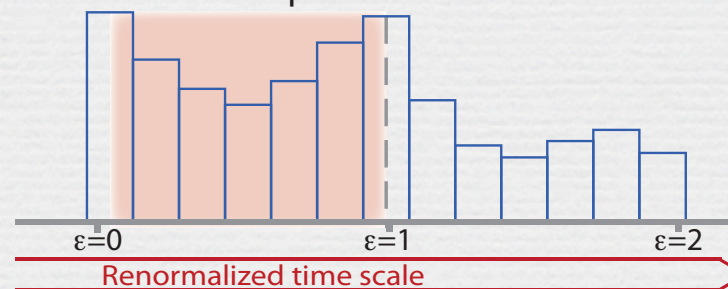
Rescaling Analysis

Renormalization of trend sequences

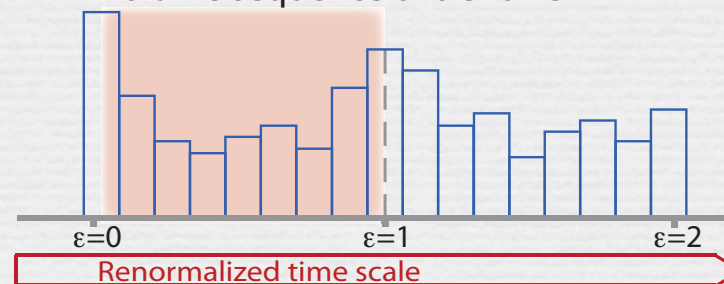
Volume sequence of trend #1



Volume sequence of trend #2

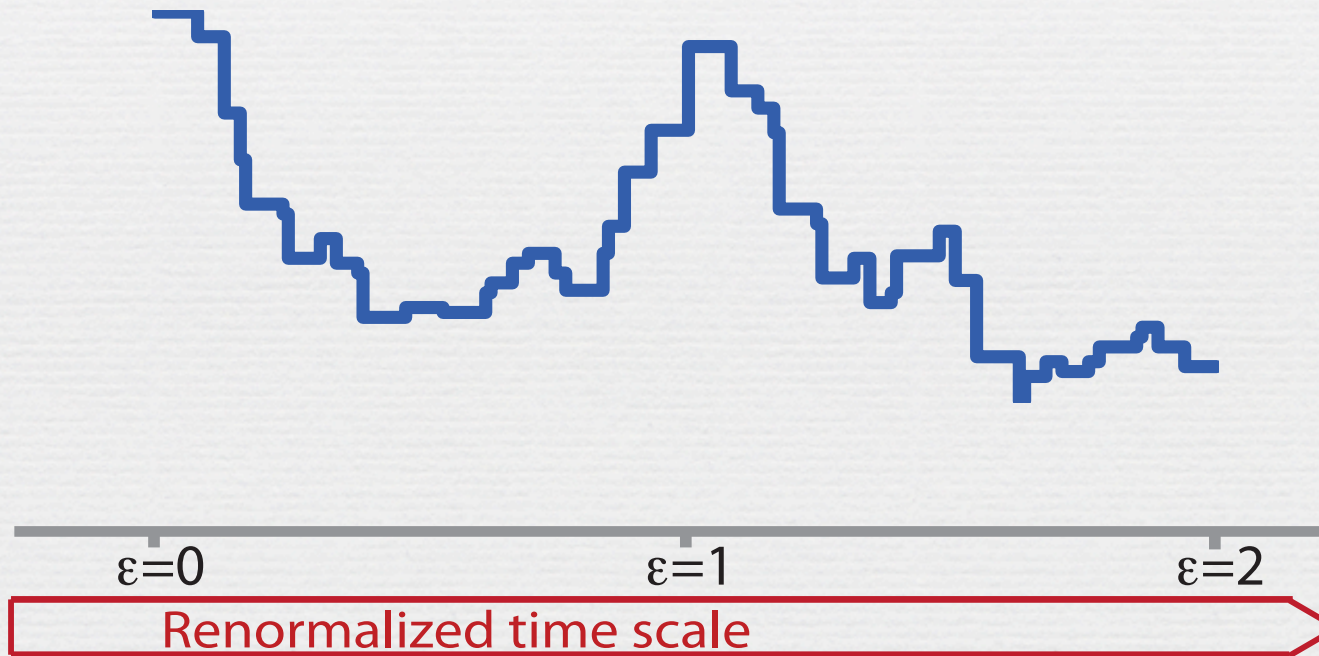


Volume sequence of trend #3



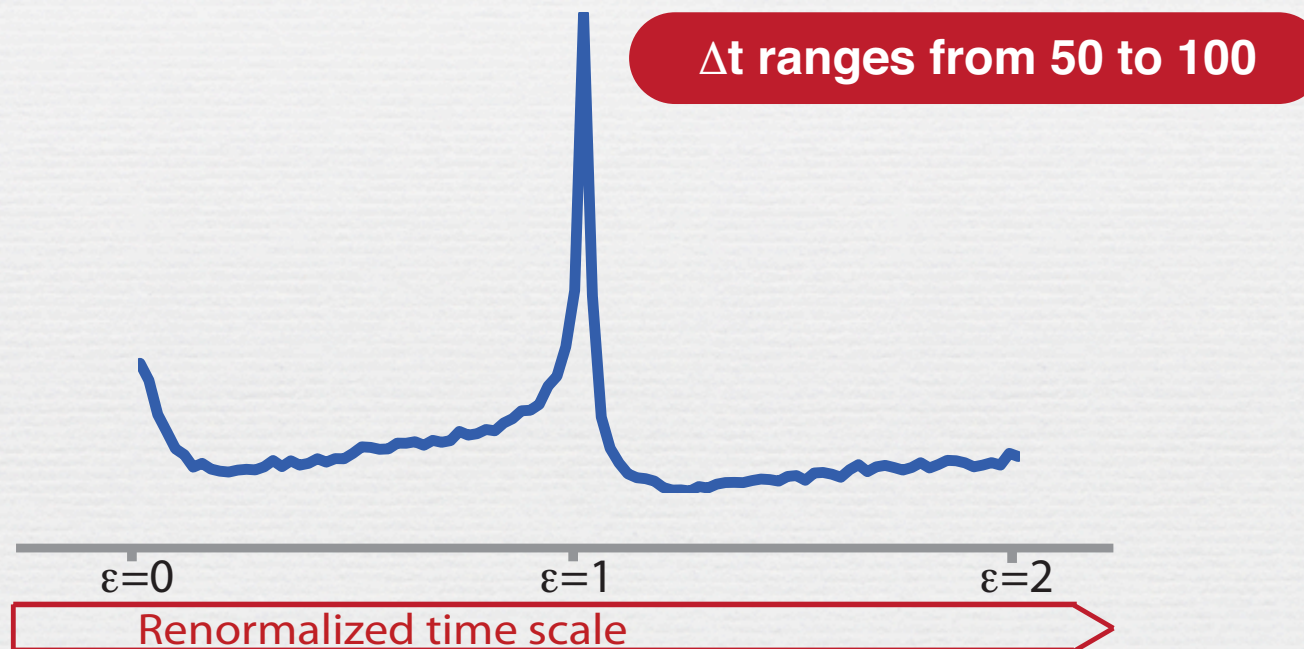
Rescaling Analysis

Averaged volume sequence

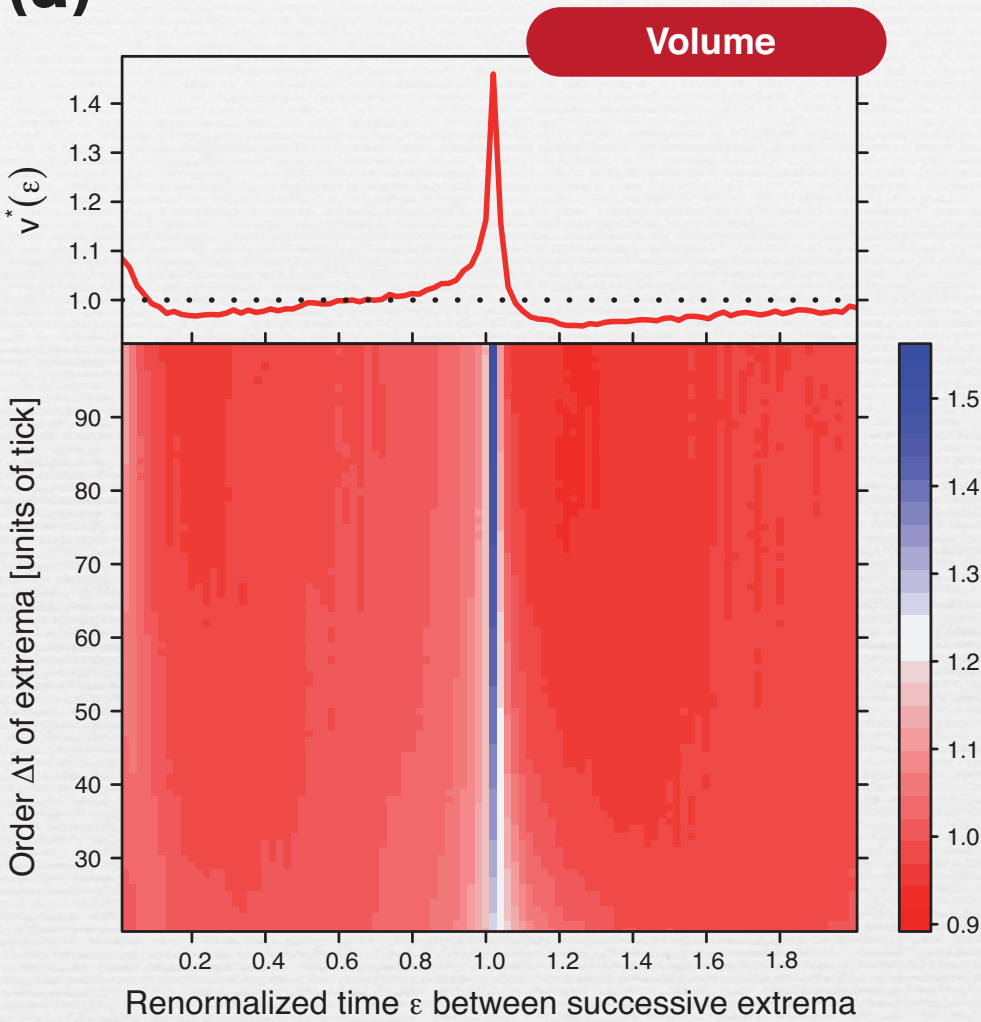


Rescaling Analysis

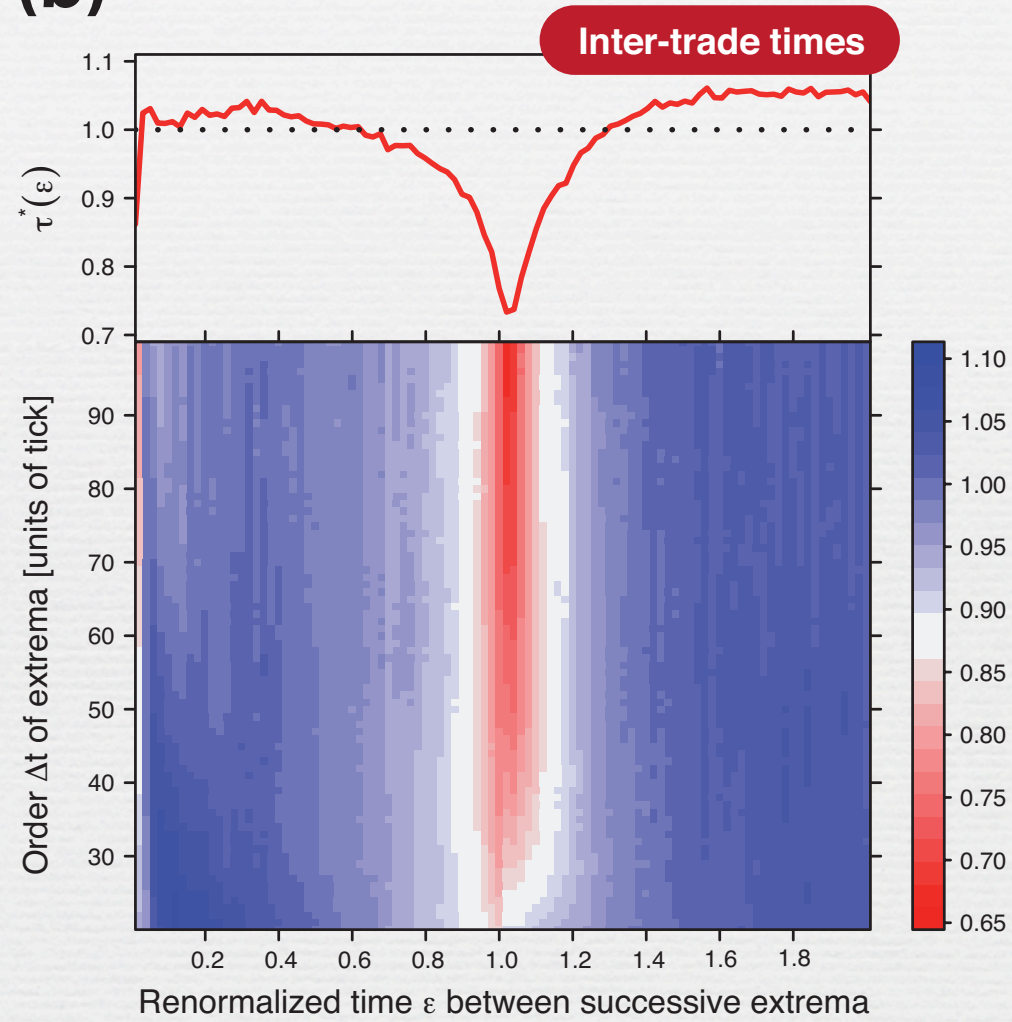
Averaged volume sequence for FDAX time series



(a)

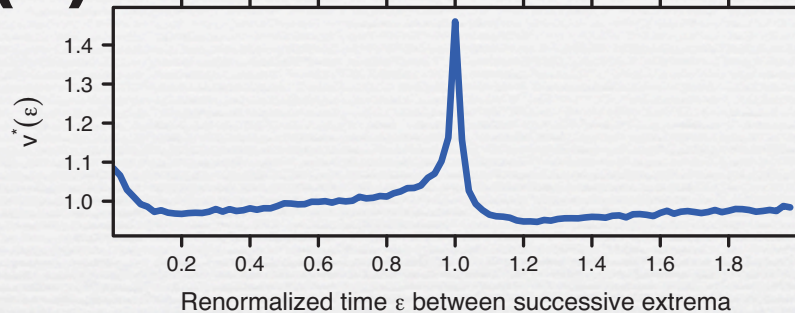


(b)

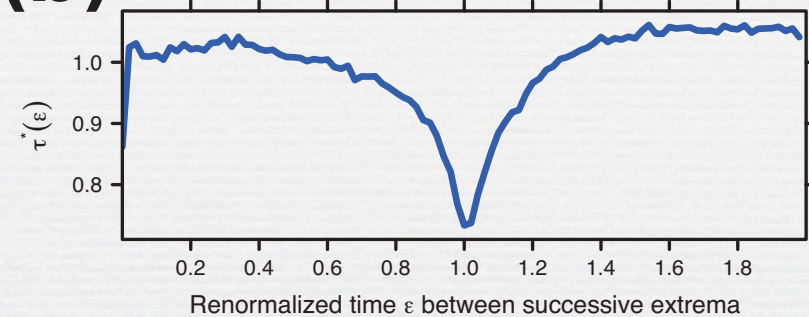


Quantities With Scale-Free Behavior

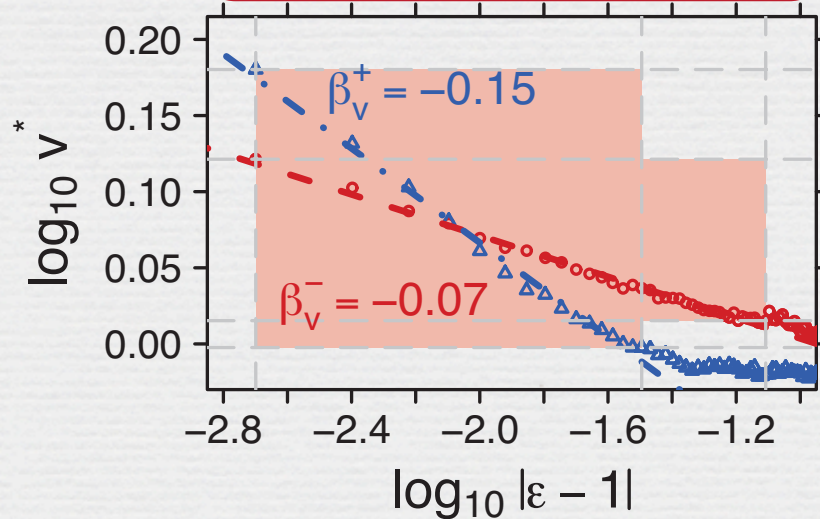
(a) Transaction volume (FDAX)



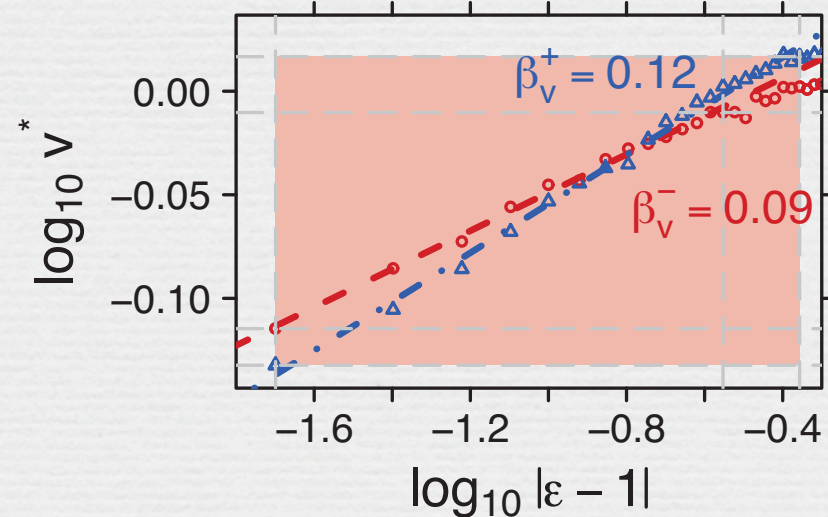
(b) Inter-trade times (FDAX)



(c) Δt range: 50 to 1000 ticks

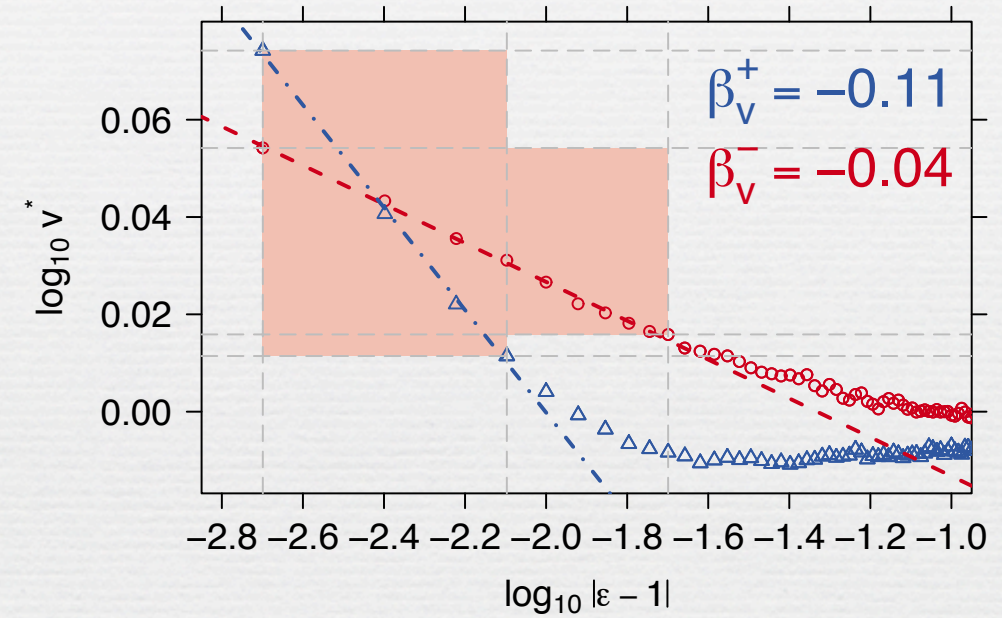
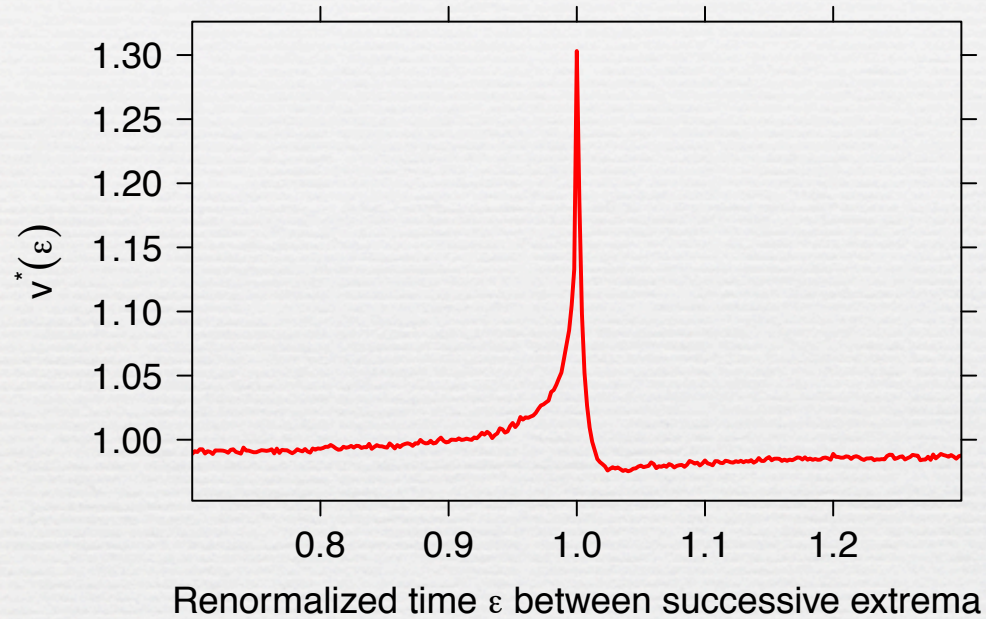


(d) Δt range: 50 to 100 ticks



- Randomly reshuffling confirms our findings.

Reproduction with a Model

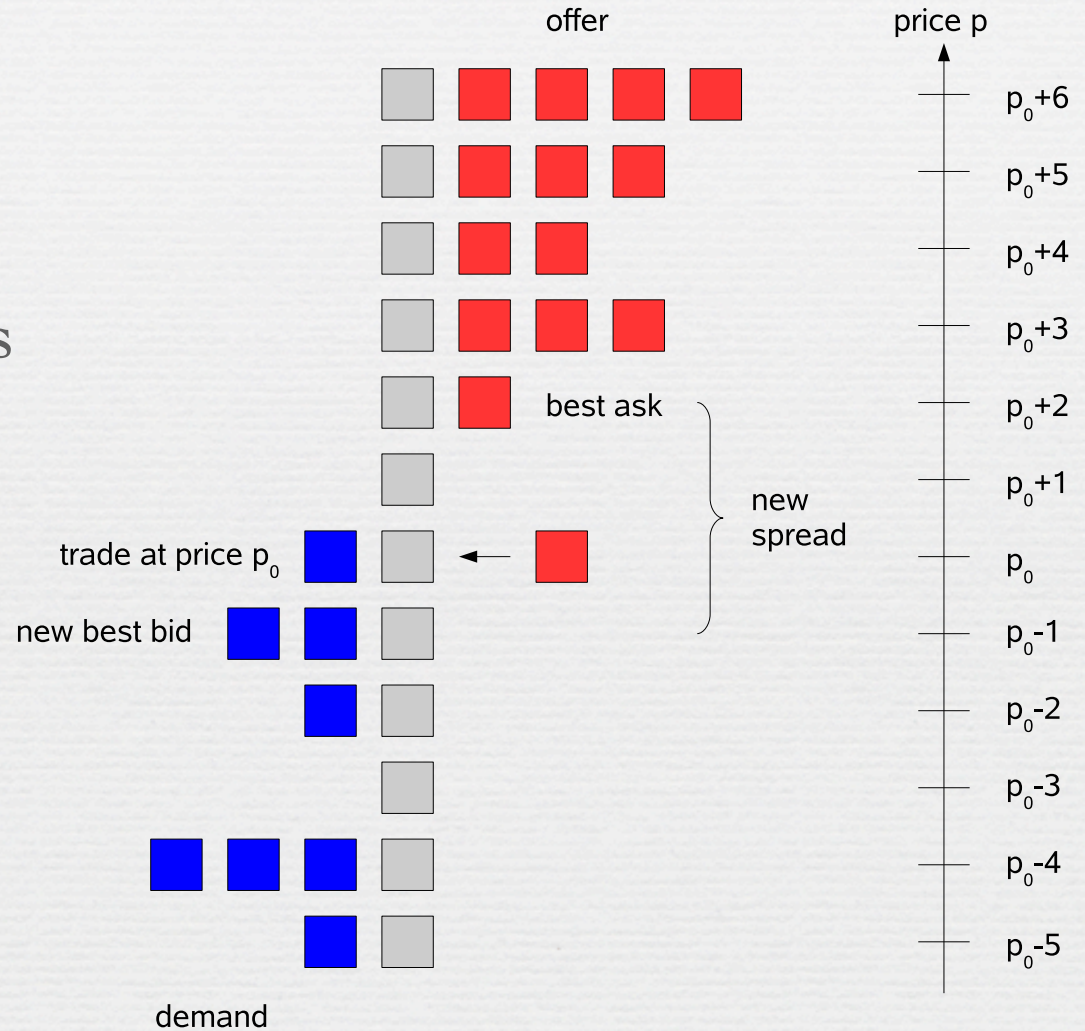


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Order Book Structure

- Price-time priority
- Discrete price levels
- Limit orders
- Market orders



T. Preis, S. Golke, W. Paul, and J. J. Schneider, Europhys. Lett. **75**, 510 (2006)

Model Definition

- Liquidity provider N_A
- Liquidity taker N_A
- Limit order rate α
- Market order rate μ
- Order cancel rate δ
- Buy/Sell probability $q_{\text{provider}} = q_{\text{taker}} = \frac{1}{2}$
- Best bid p_b
- Best ask p_a
- Midpoint $p_m = \frac{p_a + p_b}{2}$
- Spread $s = p_a - p_b$

Model Definition

- Exponentially distributed order entry depth
 - Limit price of a limit buy order

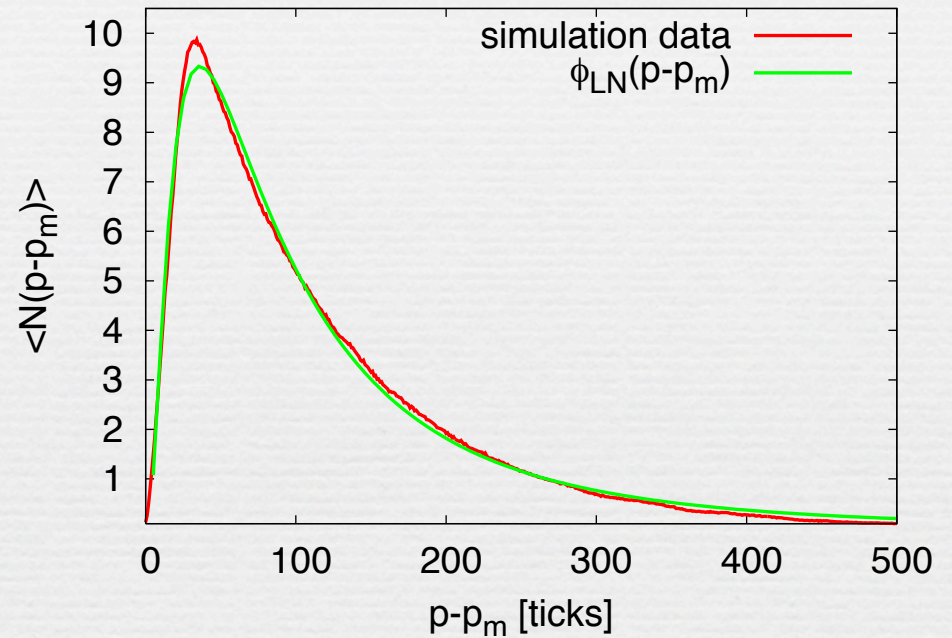
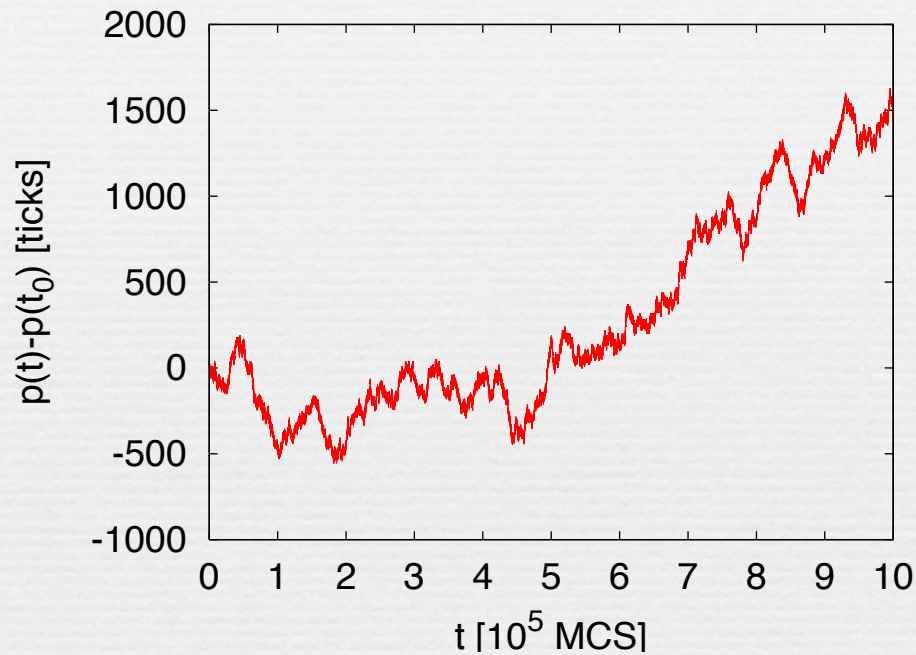
$$p_i^l = p_a - 1 - \eta$$

- Limit price of a limit sell order

$$p_i^l = p_b + 1 + \eta$$

- with stochastic variable $\eta = \lfloor -\lambda_0 \times \log(x) \rfloor$

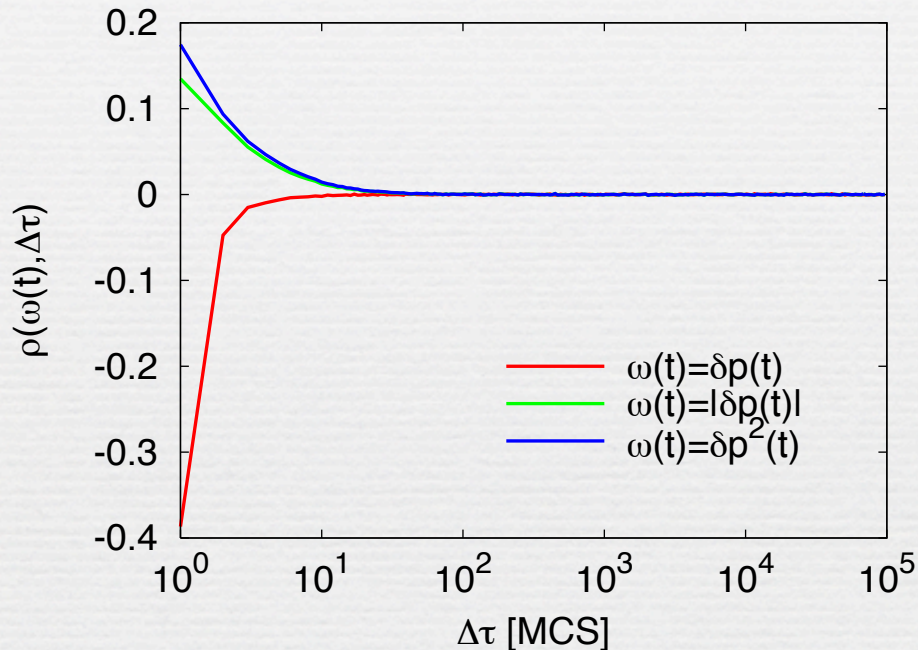
Results of the Model



$$P_{\text{LN}}(x) = A \frac{1}{Sx\sqrt{2\pi}} \exp\left(-\frac{(\ln x - M)^2}{2S^2}\right)$$

- Order book depth: log-normal distribution

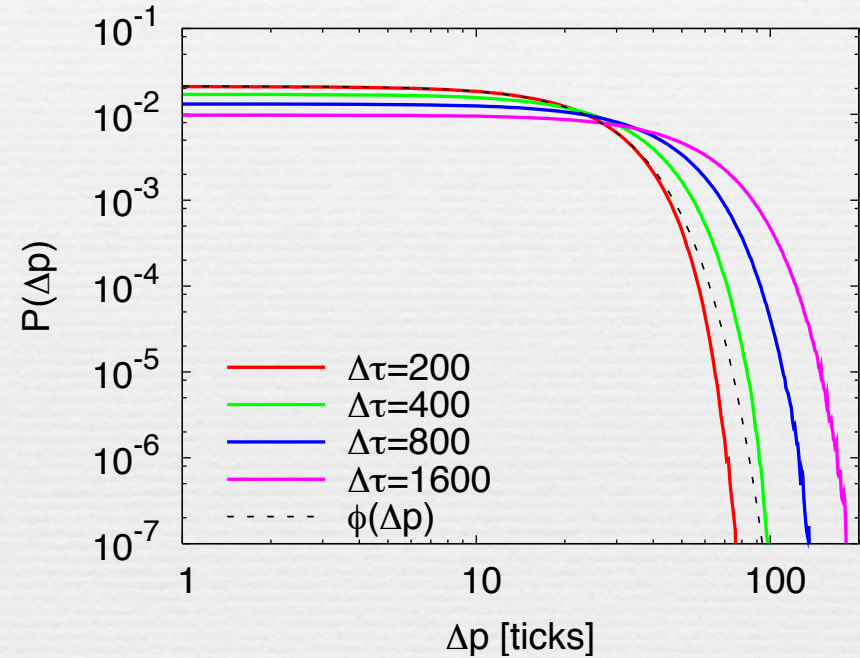
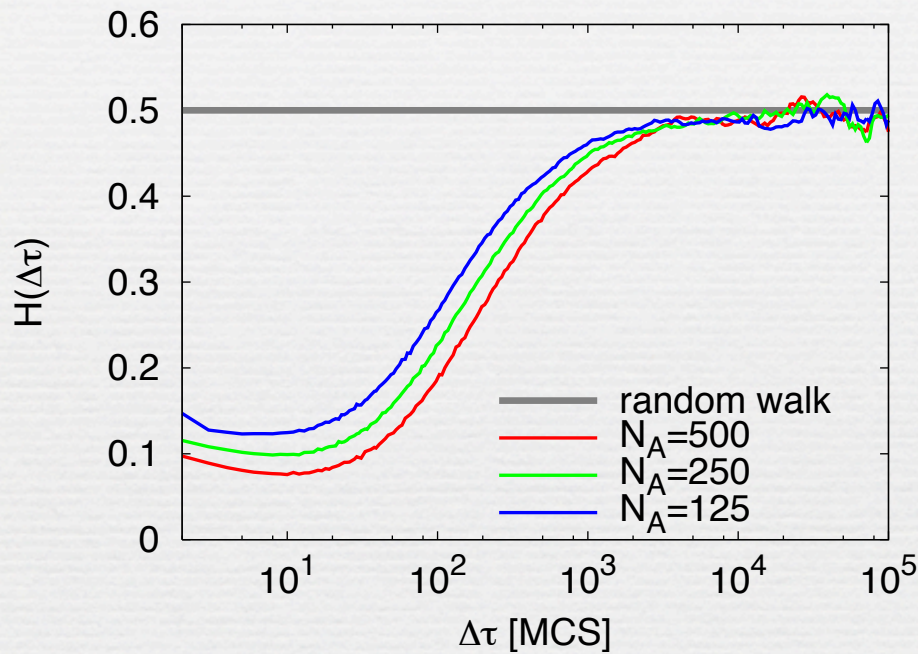
Autocorrelations



$$\rho(\omega(t)\tau) = \frac{\langle \omega(t + \tau)\omega(t) \rangle - \langle \omega(t) \rangle^2}{\langle \omega(t)^2 \rangle - \langle \omega(t) \rangle^2}$$

- Negative autocorrelation on short time scales
- Volatility clustering

Results of the Model



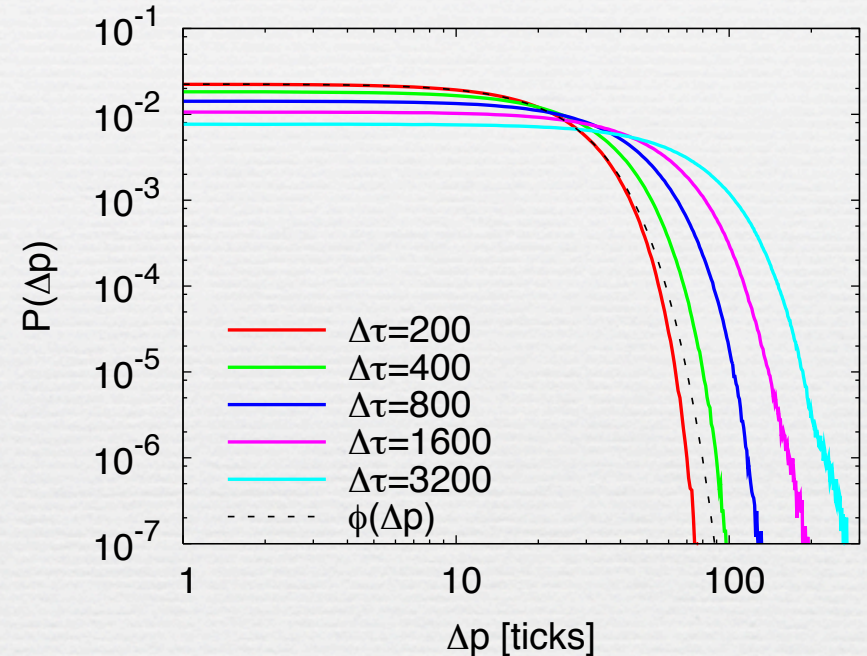
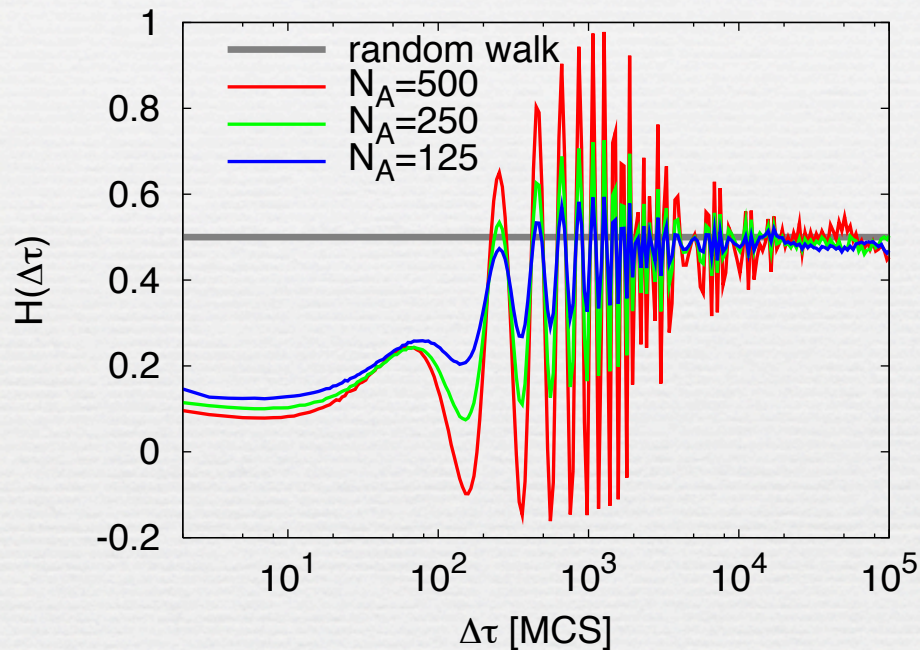
- Hurst exponent $H(\Delta\tau) : \langle (\Delta p)^2 \rangle^{1/2}(\Delta\tau) \propto \Delta\tau^H$
- Random walk: $H = 1/2$

Asymmetric Order Flow

- Time dependent buy/sell probability (micro trends):
 - Deterministic perturbation: Sawtooth
 - Stochastic perturbation: Feedback random walk (FRW): RW with mean reversion tendency

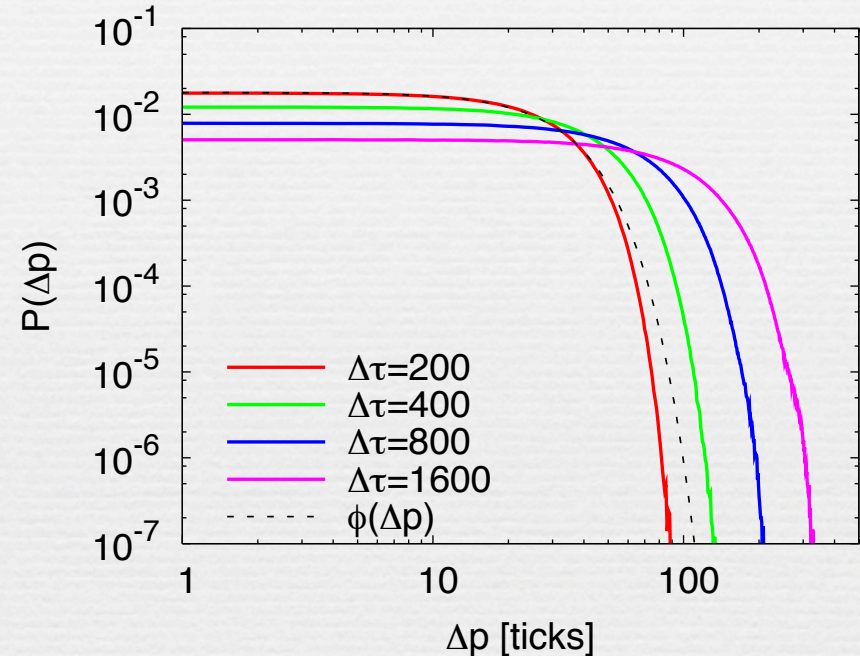
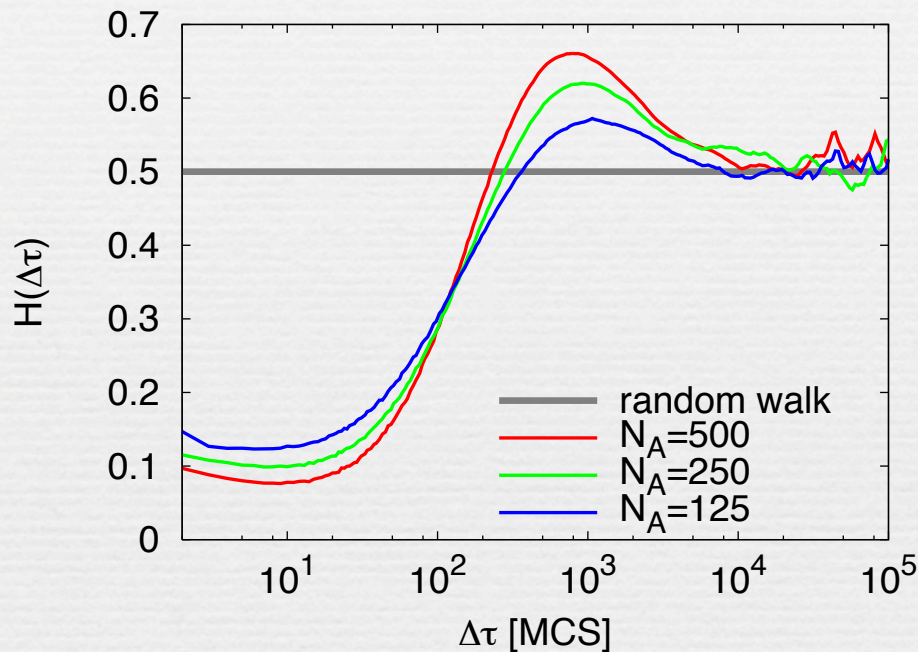
T. Preis, S. Golke, W. Paul, and J. J. Schneider, Europhys. Lett. **75**, 510 (2006)

Sawtooth



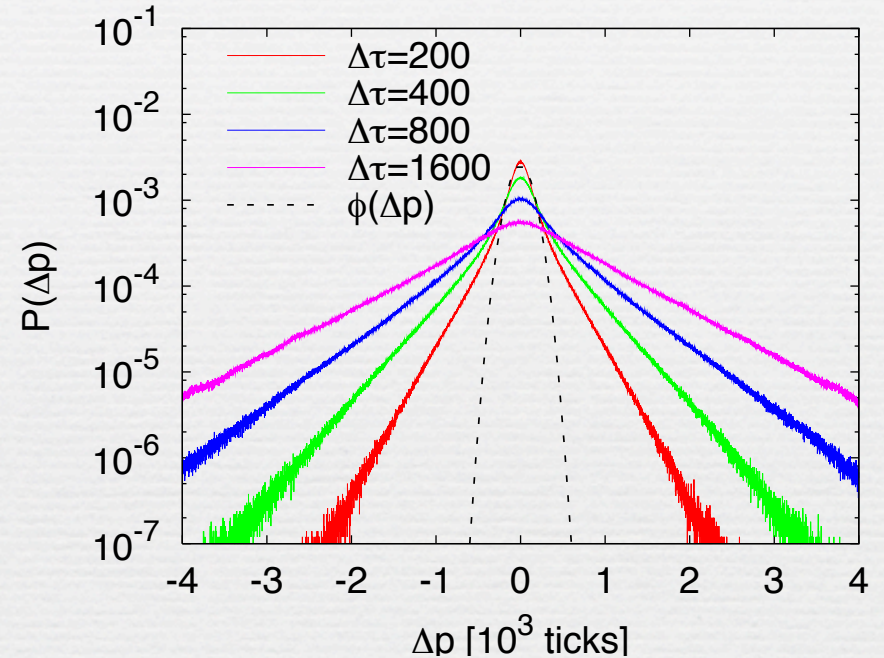
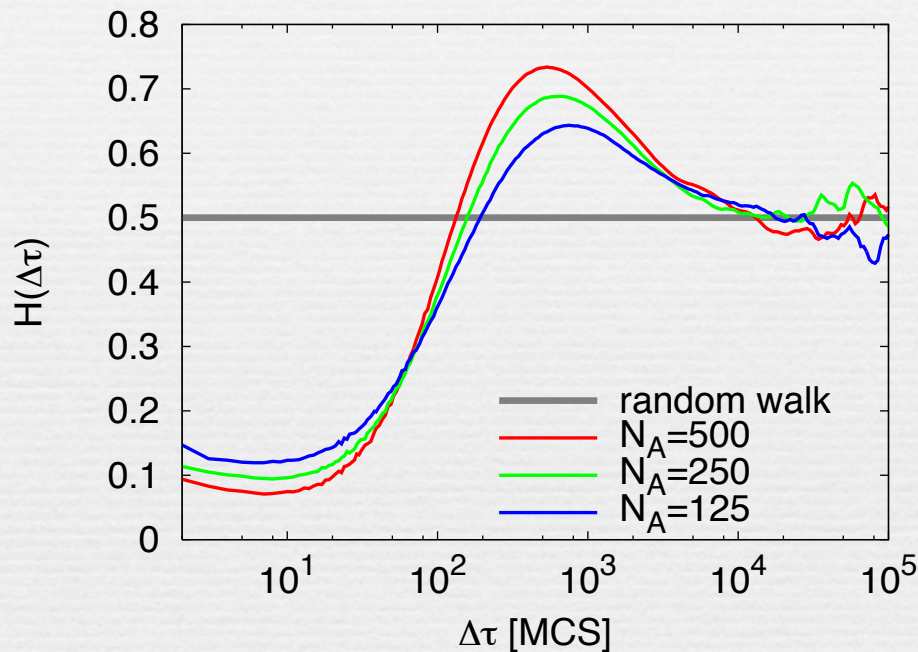
- Deterministic modulation with a period of 200 MCS is reflected in the mean square displacement, leading to quasi-periodic oscillations of the Hurst exponent.

Feedback Random Walk



- Again an anti-persistent behavior on short time scales, but a persistent behavior on medium time scales, and a diffusive regime on long time scales.

Dynamic Order Depth



- Hurst exponent H

$$\langle |p(t + \Delta t) - p(t)|^q \rangle^{1/q} \propto \Delta t^{H_q(\Delta t)}$$

- Dynamic order entry depth leads to fat-tailed distribution of price changes

T. Preis, S. Golke, W. Paul, and J. J. Schneider, *Europhys. Lett.* **75**, 510 (2006)

T. Preis, S. Golke, W. Paul, and J. J. Schneider, *Phys. Rev. E* **76**, 016108 (2007)

Conclusions

- **Complex pattern-based correlations in high frequency financial market data.**
- Switching processes in financial markets: Power-law behavior close to switching points.
- Order Book Model: Reproduction of anti-persistent scaling behavior on short time scales, persistent on medium time scales, and diffusive behavior on long time scales; fat-tailed price change distributions.
- Reproduction of switching point behavior.

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- **Reproduction of switching point behavior.**

Thank you ...

EPJ ST: Issue 194



Tobias Preis
Progress in Econophysics

TOPICS IN PHYSICS

How to Characterize Trend Switching Processes in Financial Markets

Tobias Preis^{1,2} and H. Eugene Stanley¹

Center for Polymer Studies, Department of Physics, 550 Commonwealth Avenue, Boston, Massachusetts 02215, USA
¹Mathematical Physics, Johannes Gutenberg University Mainz, Duesbergweg 10-15, D-55128 Mainz, Germany
²Volterra Capital Asset Management GmbH, Gernerstr. 14, D-60528 Frankfurt, Germany

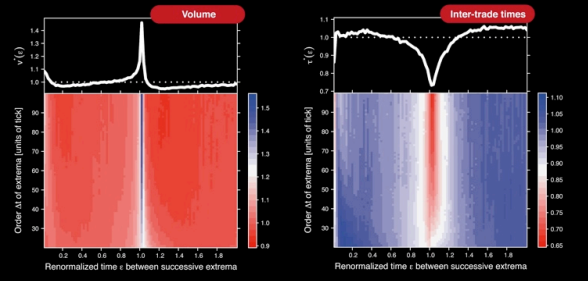



One of the key concepts in modern statistical physics is the concept of scale invariance, called self-similarity. In the scaling hypothesis that focuses on the critical behavior of systems near phase transitions, the scaling hypothesis has two versions of predictions, both of which have been verified by a wealth of experimental data on diverse systems. The first category is a set of relations, called scaling laws, that describe the universal critical-point exponent behavior.

H. Eugene Stanley is a Distinguished Professor in the Department of Physics and Director of the Center for Polymer Studies at Boston University. He has worked for 40 years as a physics professor before he moved to Boston University. He was elected an officer in the American Physical Society in 1995 and received the National Academy of Sciences Award in 2004. He was elected to the National Academy of Sciences in 2004 and received the National Medal of Science in 2005. He was elected to the National Academy of Sciences in 2005. He was elected to the National Academy of Sciences in 2005. He was elected to the National Academy of Sciences in 2005.

Tobias Preis is a Research Assistant and Doctoral Candidate in the Department of Physics at the Johannes Gutenberg University Mainz (Germany). He was elected to the Graduate School of Energy Research in 2008. He is also working for Volterra Capital Asset Management GmbH as Managing Director. <http://www.tobiaspreis.de>

18



"Blueprint" of trends in financial markets (p. 54)



Bubble trouble

When a stock market rises unsustainably, it can create a financial bubble that sooner or later will burst. **Tobias Preis** and **H Eugene Stanley** examine whether concepts from physics can be used to create a law describing exactly how such crashes occur

Wild fluctuations in the stock prices and currency exchange rates of every country have had a huge impact on the world economy and the personal fortunes of millions of us over the last few years. These instabilities have also had another, perhaps unintended, consequence – of thrusting the academic discipline of “econophysics” firmly into the limelight. But does a field that involves applying the concepts of statistical physics to economics really have anything important to contribute to discussions about the current economic crisis? Yes – absolutely – because finding laws describing fluctuations is the essence of statistical physics.

Physicists are not, of course, the first people to apply statistics to economics, with mathematicians also having contributed to the field for many years. One of their first significant breakthroughs came in 1900 when Louis Bachelier, working under the tutelage of the great French mathematician Henri Poincaré, published

a PhD thesis in which he analysed real financial data. Bachelier claimed that a histogram of the changes in share price (measured over any period of time) forms a bell-shaped curve known as a Gaussian function, with very large fluctuations essentially never occurring. In other words, he believed that the chances of a serious crash occurring are almost zero. Such serious crashes are indeed very rare, but when they do occur, their effects can be devastating.

The model associated with Bachelier is often referred to as the “random walk” or “drunkard’s walk” because he assumed that stock prices go up or down randomly by an amount that has a characteristic value. In recent years, however, econophysicists have been able to get their hands on a staggering quantity of real-time financial data, including the price, volume and timestamp of every transaction of every stock you can think of. Thanks to this information, which is now available in huge finan-

Tobias Preis and **H Eugene Stanley** are at the Center for Polymer Studies, Department of Physics, Boston University, US. Preis is also at ETH Zurich, Switzerland, [e-mail mail@tobiaspreis.de](mailto:tobiaspreis.de). Preis will be giving a free online lecture at 4 p.m. BST on 12 May – see physicsworld.com/cws/channel/multimedia/ for more details