

Why do asset prices move?

Impact and Second Generation models

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Why ask?

- **Crucial question in theoretical Economics and Finance:** what is the information reflected by prices & what are markets good for?
- **Crucial question for investment strategies:** is there any way to predict how prices will move?
- **Crucial question for risk control/regulation:** understanding why and how prices move allows one to devise efficient risk models and useful regulation (?)

The Sacred Lore of Theoretical Economics

- **Efficient market theory:** Agents are rational and Markets are in **equilibrium**
- Prices reflect faithfully the **Fundamental Value** of assets and **only move** because of **exogeneous** unpredictable news.
- Platonian markets which merely **reveal** fundamental values **without influencing them** – or is it a mere tautology??
- Crashes can only be **exogenous**, not induced by markets dynamics itself – oh really??

By the way...

- Agents (us humans) **do make errors and have regrets**, (cognitive or sensorial biases, imperfect or superabundant information, urgency, negligence, etc.)
- Problems can be algorithmically so complex that we have to make **suboptimal decisions**
- **Agents are deeply influenced by the behaviour of others** – who might have more information (??)
- → **Even silly trades do impact prices** and may create positive **feedback loops**

First generation models of markets

- Rooted in the idea that dynamics is exogenous and markets are efficient, [Financial Engineering](#):
- (1) **postulate** any process that
 - is tractable
 - looks vaguely similar to real data
- (2) **brute force calibrate**, on “liquid” markets (supposed to be efficient) and price options or more exotic derivatives
- **Examples**: Brownian motion (Black-Scholes), GARCH, Heston, Local vol., Lévy, Multifractal, etc. , etc., etc.

BUT

- **NONE** of these models are justified by “first principles”, or agent based models, such that parameters can be (at least in principle) *computed*
- **Inspiration from physics:** macroscopic (or hydrodynamic) laws from microscopic elements
 - Navier-Stokes from molecular collisions
 - Magnetic properties from individual spins
 - Phase diagram of bodies from individual atoms, etc. etc.

BUT

- Uncontrolled brute force calibration are often based on **absurd models** (e.g. local volatility models) and can be **extremely dangerous**:
 - **Even liquid markets are in fact not liquid and not efficient**
 - e.g. plain vanilla equity option markets
 - **Errors and biases are amplified in a non-linear way** – e.g. using plain vanillas to price exotic options using local vol.
 - **Self fulfilling prophecies and feedback loops** – e.g. portfolio insurance and the 87 Black-Scholes induced crash, etc.

BUT

- To calibrate does not mean to understand
- A perfect fit is not a theory – often a red-herring
- Let's try to understand what's going on at the micro level

Some empirical facts

- Financial markets offer Terabytes of information (weekly) to try to investigate why and how prices move

- A) Are news really the main determinant of volatility?

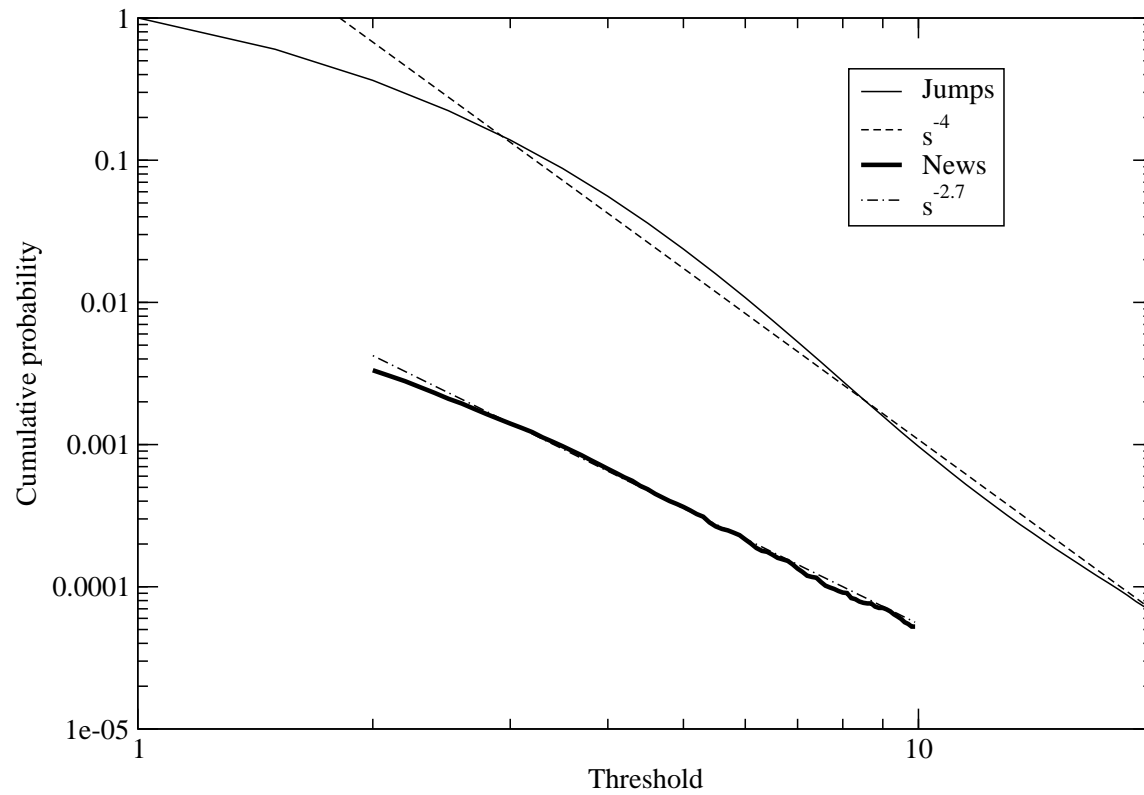
Exogenous vs. endogenous dynamics

- B) Are price really such that supply instantaneously equals demands? How fast information is included in prices?

A) Exogenous or endogenous dynamics?

- Yes, **some** news make prices jump, sometimes a lot, but **jump freq. is much larger than news freq.**
- On stocks, **only $\sim 5\%$** of $4 - \sigma$ jumps can be attributed to news, most jumps appear to be **endogeneous**
- Different statistics: **return distributions** and **'aftershocks'** (volatility relaxation)

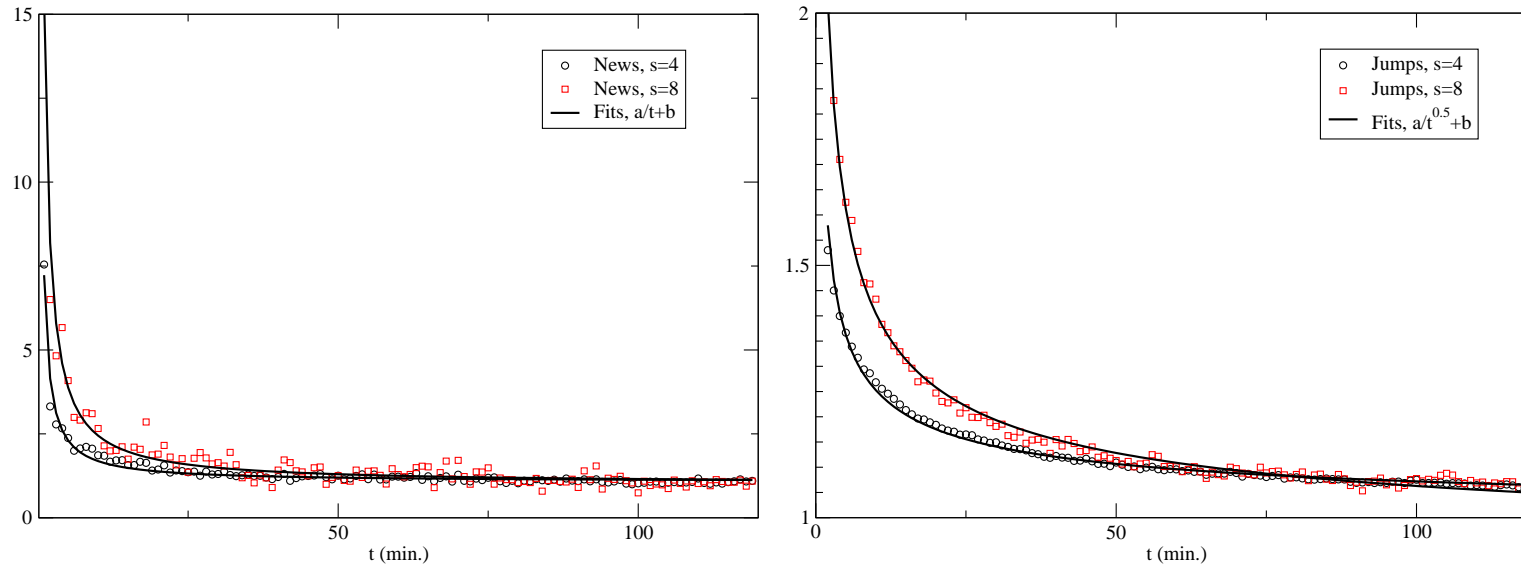
Jumps



Power-law distribution of news jumps and no-news jumps. With

A. Joulin, D. Grunberg, A. Lefevre

Two jump types: Aftershocks

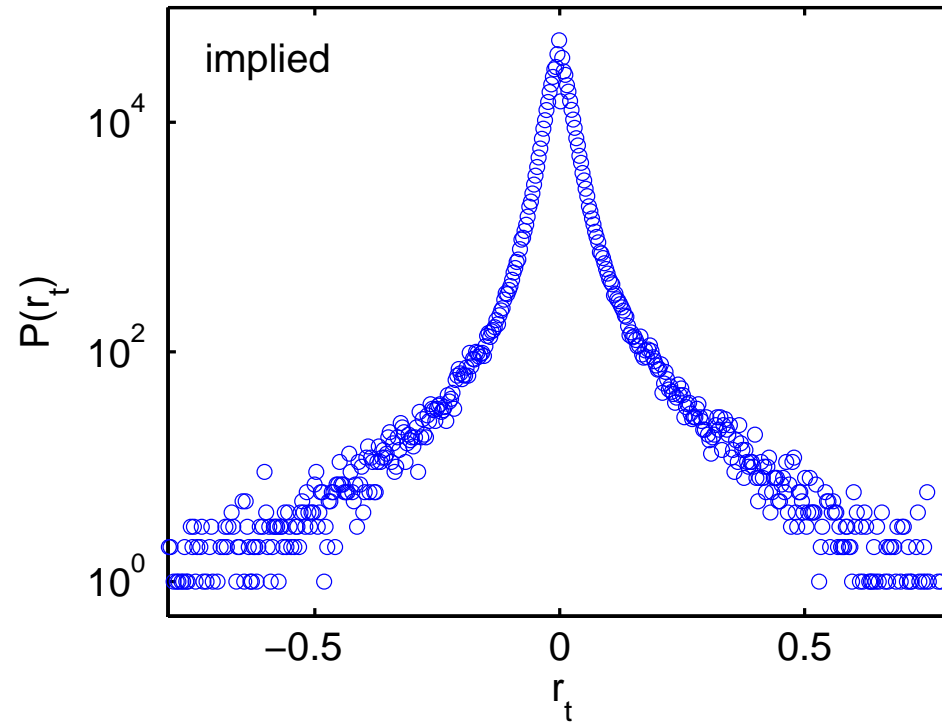


Volatility relaxation after news (t^{-1} , left) and endogenous jumps ($t^{-1/2}$, right). With [A. Joulin](#), [D. Grunberg](#), [A. Lefevre](#)

A) Exogenous or endogenous dynamics?

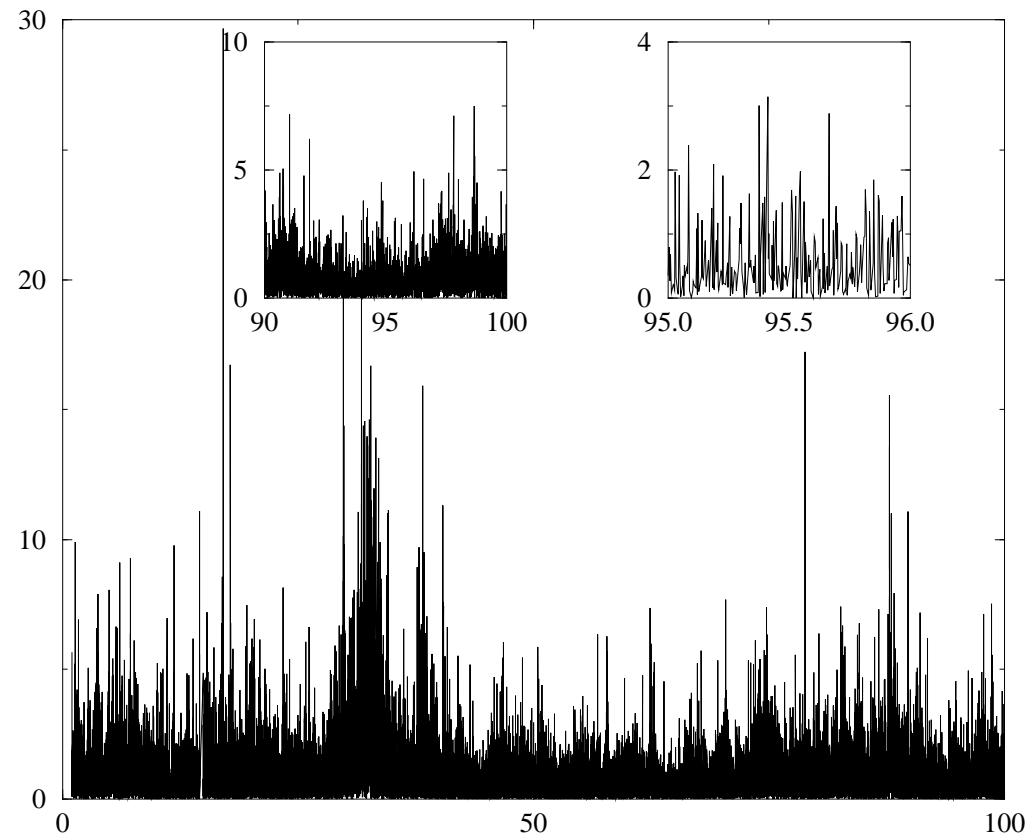
- Power-law distribution of price changes for anything that is traded
- Excess volatility, with long range memory – looks like endogenous intermittent noise in complex systems (turbulence, Barkhausen noise, etc.)
- Universal observations !!

Power-law tails



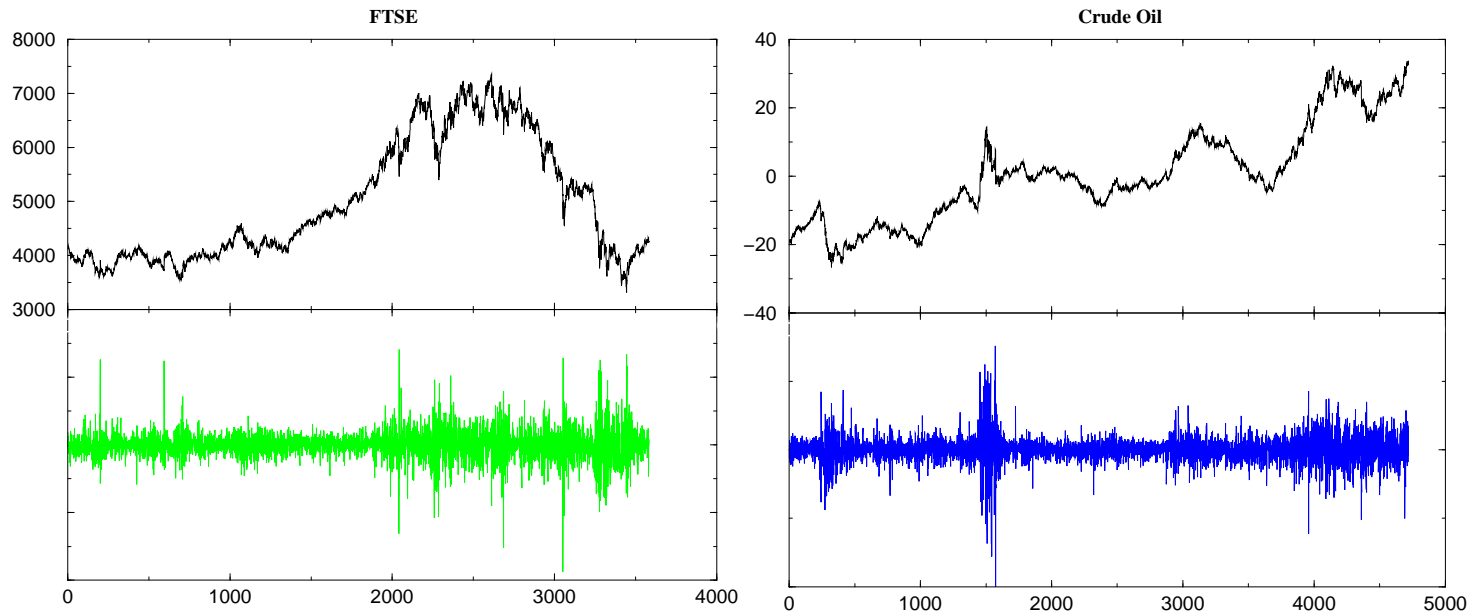
Distribution of daily volatility moves on option markets or *any other traded stuff*: **inverse cubic law**

Multiscale intermittency



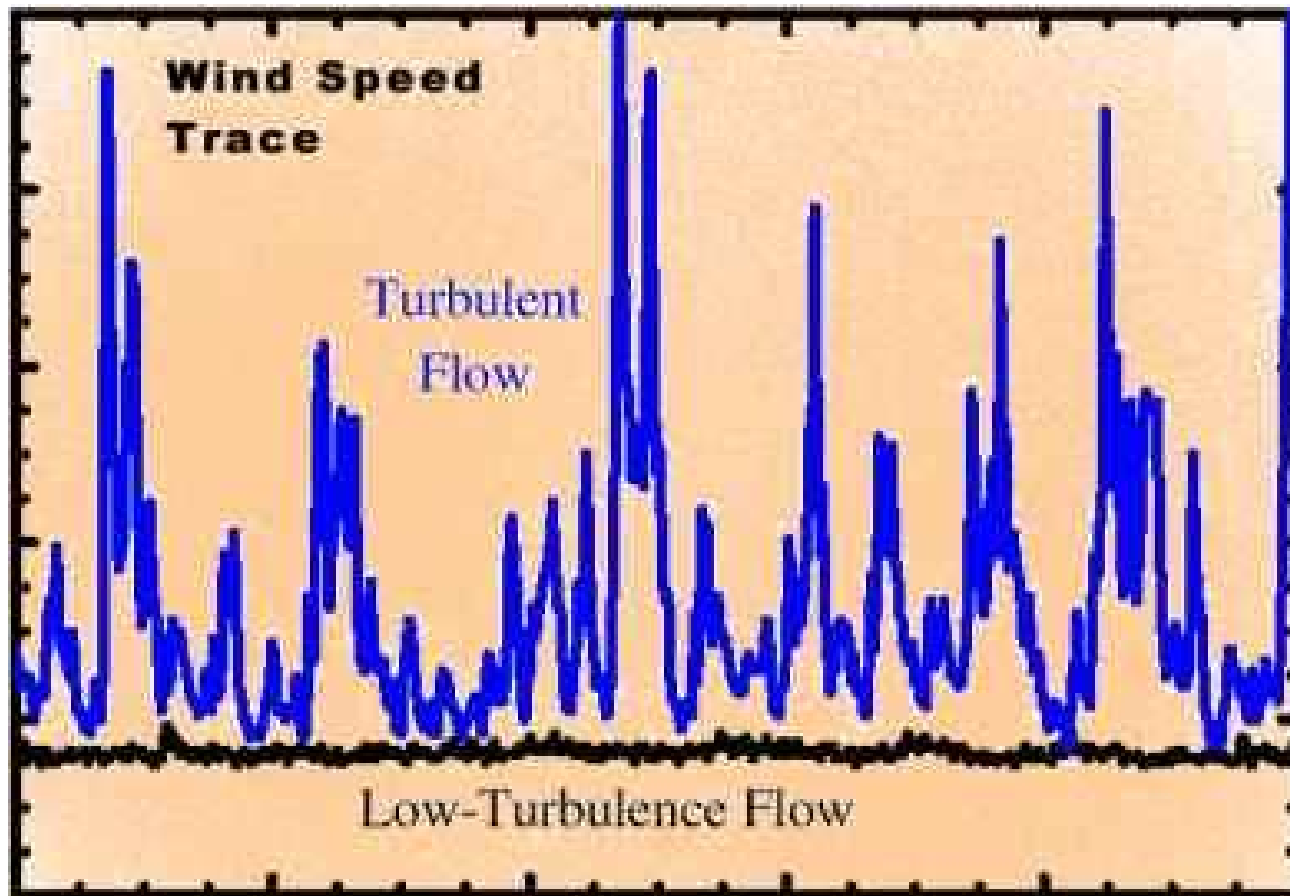
Excess volatility, with long range memory – a “multifractal” process, see Mandelbrot-Calvet-Fisher and Bacry-Muzy

Multiscale intermittency



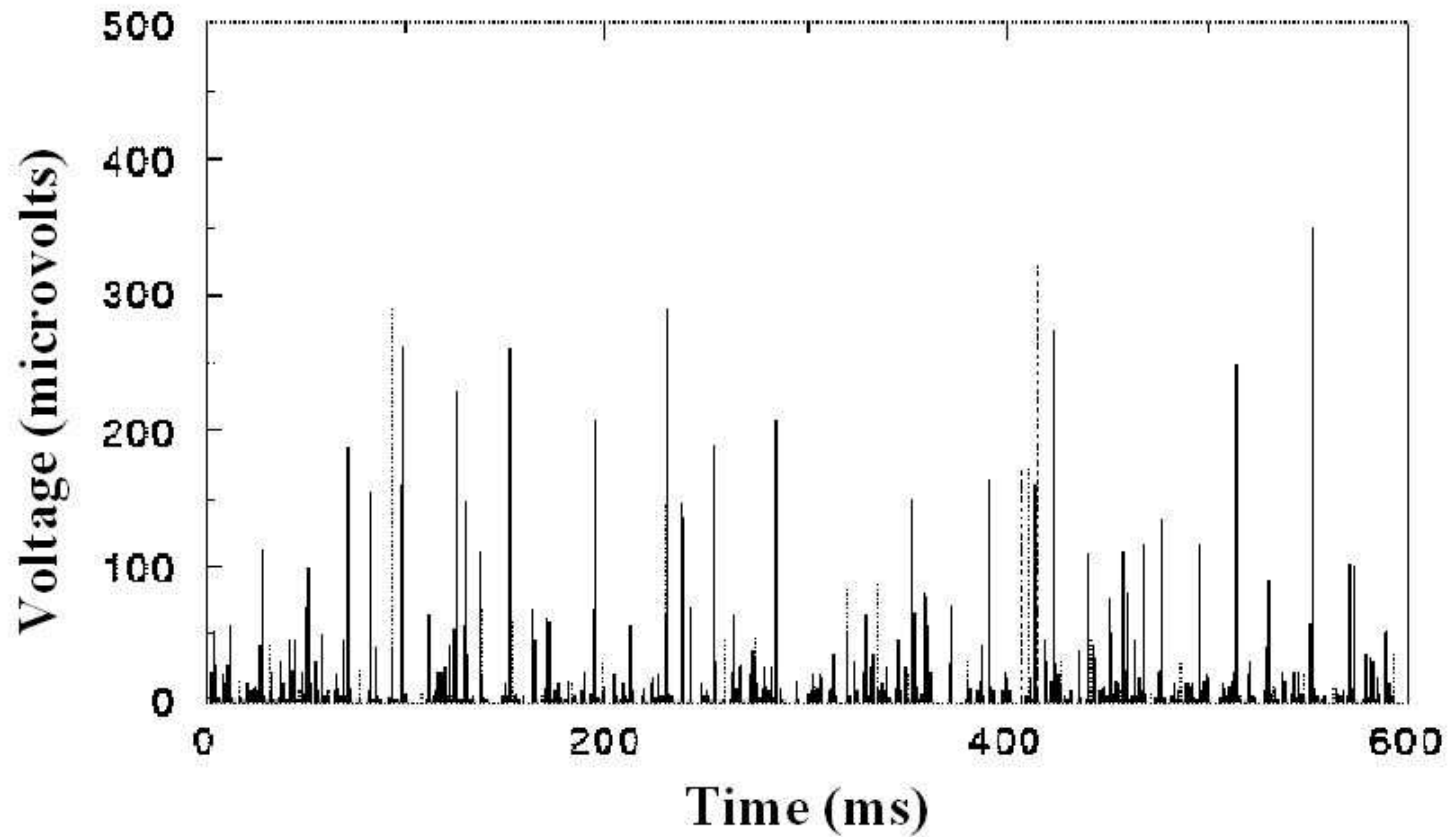
Excess volatility, with long range memory— looks a lot like
endogeneous noise in complex systems

Turbulence: intermittency



Slow, regular and featureless exogeneous drive but intermittent endogeneous dynamics

Barkhausen noise



Slow, regular and featureless exogenous drive but intermittent endogeneous dynamics

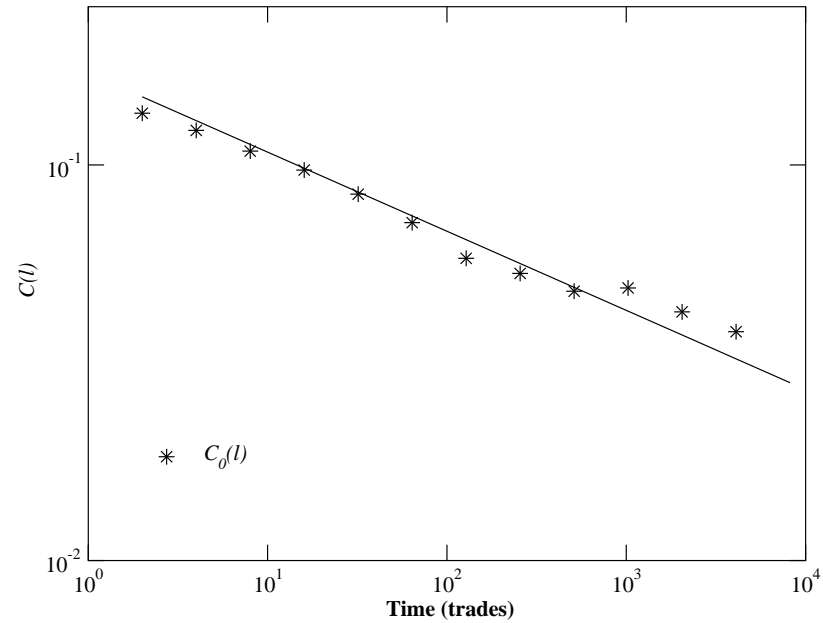
B) Are markets in “equilibrium?”

- UHF data allows one to understand the microscopics of order flow and price formation
- One can distinguish buy orders from sell orders
- Surprise: the autocorrelation of the sign of trades is very long-range correlated over several days or weeks (see also Lillo-Farmer)

$$C(l) \propto l^{-\gamma} \quad \gamma < 1$$

- A beautiful paradox: Sign of order flow very predictable and orders impact the price – but no predictability in the sign of price changes ?? – see below

Trade correlations



Correlations extend to several days! – $p_s(10000) = 53\%$

B) Are markets in “equilibrium?”

- Even “liquid” markets offer a *very small immediate liquidity* (10^{-5} for stocks) – buyers/sellers have to fragment their trades over days, weeks or even months
- “Information” can only be *slowly incorporated into prices*, latent demand does not match latent supply
- Markets are *hide and seek games* between “icebergs” of buyers and sellers and are *not* instantaneously in equilibrium

Some empirical facts

- A) Are news really the main determinant of volatility?
 - No, endogenous dynamics more likely, markets are complex systems that generate rich endogenous dynamics
- B) Are price really such that supply instantaneously equals demands?
 - No, “information” can only be very slowly incorporated into prices

What is impact?

- **Efficient market story:** Informed agents successfully forecast short term price movements and trade accordingly. This results in correlations between trades and price changes, but **uninformed trades have no price impact** – prices stick to “Fundamental Values”
- **A more plausible story:** since there is no easy way to distinguish informed from non informed traders, **all trades statistically impact prices** since other agents believe that some of these trades **might** contain useful information – a mechanism leading to feedback loops and avalanches

Impact

- On anonymous markets, the origin of trades (“informed” vs. “non-informed”) cannot be decided

Anyway, the information contained in each trade is very small – cf below

- Trading, even uninformed and with relatively small volumes in usual market conditions, strongly influences prices and leads to measurable effects – even “liquid” markets are not that liquid

(1% of the daily volume moves the price by 5% of the daily volatility!!)

- Impact of trades is crucial to understand why prices move: the price process is not God given and we merely observe it tracking the “true” value

Impact & volatility

- Using high frequency data, one can measure impact accurately:

$$\mathcal{I}_+ = E[p_{n+1} - p_n | \epsilon_n = +1], \quad \mathcal{I}_- = -E[p_{n+1} - p_n | \epsilon_n = -1]$$

- Empirical finding (1): impact is proportional to spread

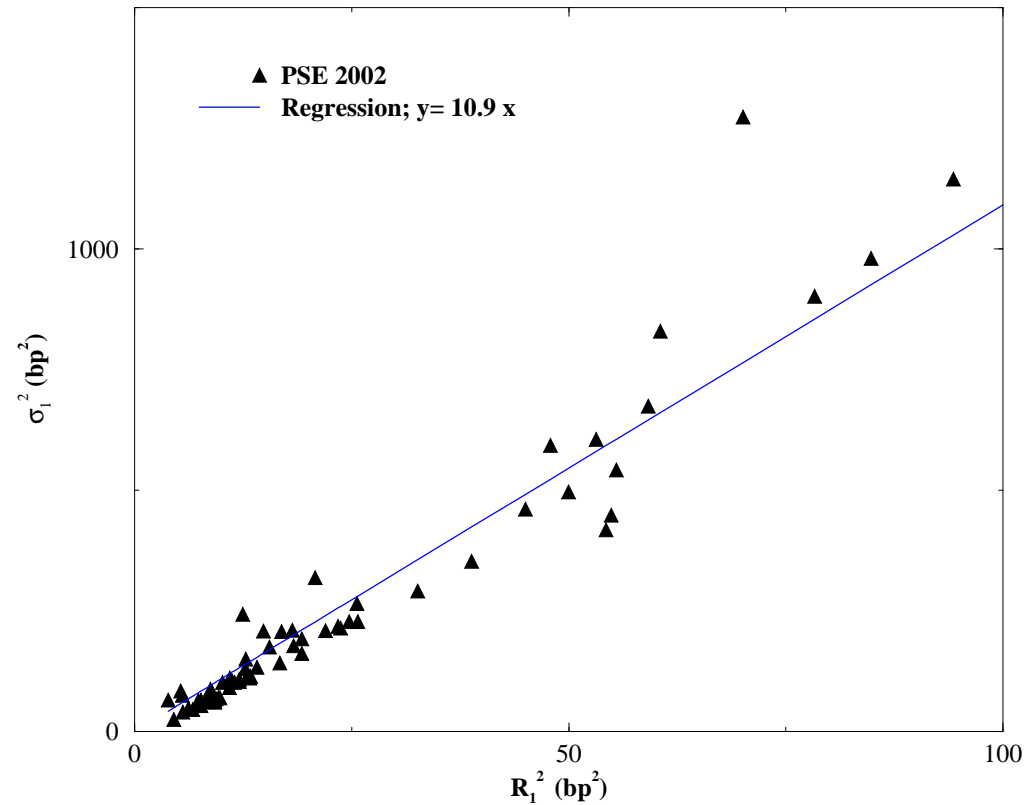
$$\mathcal{I}_+ = \mathcal{I}_- \approx 0.3 S$$

- Empirical finding (2): volatility per trade is proportional to impact

$$\sigma_1^2 = A\mathcal{I}^2 + BJ^2, \quad B \approx 0$$

- Volatility is indeed mostly due to impact of trades – very little to quote jumps J without trades (“news”)

Volatility: impact + news?



Very little contribution from quote jumps J without trades (“news”)

– with J. Kockelkoren, M. Potters, M. Wyart

Impact: non linear and transient

- Impact is **highly non trivial to model** (both non-linear and non local in time)

$$p_t = p_{-\infty} + \lambda \sum_{\ell=1}^{\infty} G(\ell) (\epsilon SV^\psi) \Big|_{t-\ell},$$

- $\psi \approx 0.2$: **very concave impact**
- The impact function $G(\ell)$ **decays as $\ell^{-\beta}$** as to exactly offset the correlation of trades and **remove predictability of returns!**

$$\beta = \frac{1-\gamma}{2}, \quad 0 \leq \gamma \leq 1$$

Impact: non linear and transient

- Bachelier's legacy: the random walk nature of prices results from a **subtle balance** between **trending order flow** and **mean-reverting impact**
- $G(\infty)$ is Hasbrouck's definition of the **information content** of a single trade, and it is very small ($G(\infty) \ll G(1)$).

Transient impact: more technicalities

- Mid-point fluctuations in trade time: diffusion

$$\mathcal{D}(\ell) = \left\langle \left(p_{n+\ell} - p_n \right)^2 \right\rangle \approx \sigma_1^2 \ell$$

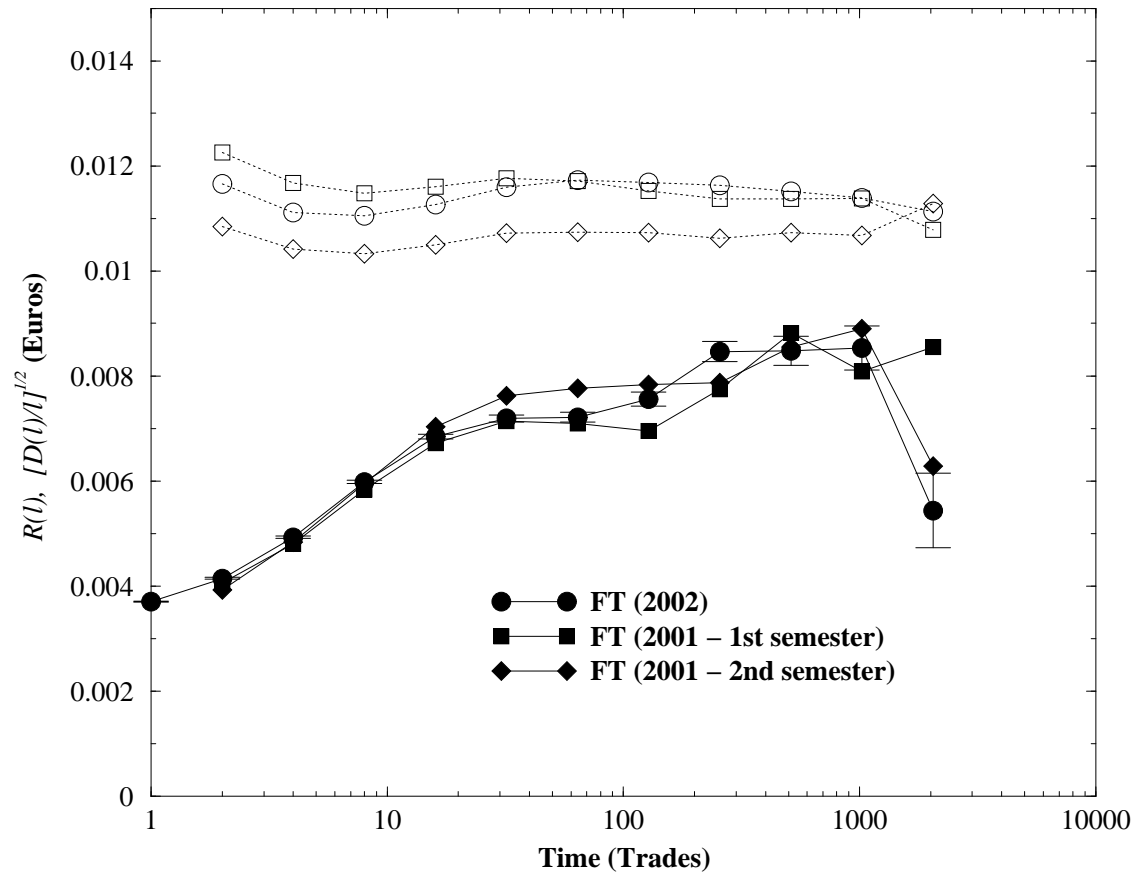
- Average response function:

$$\mathcal{I}(\ell) = \left\langle \left(p_{n+\ell} - p_n \right) \cdot \varepsilon_n \right\rangle$$

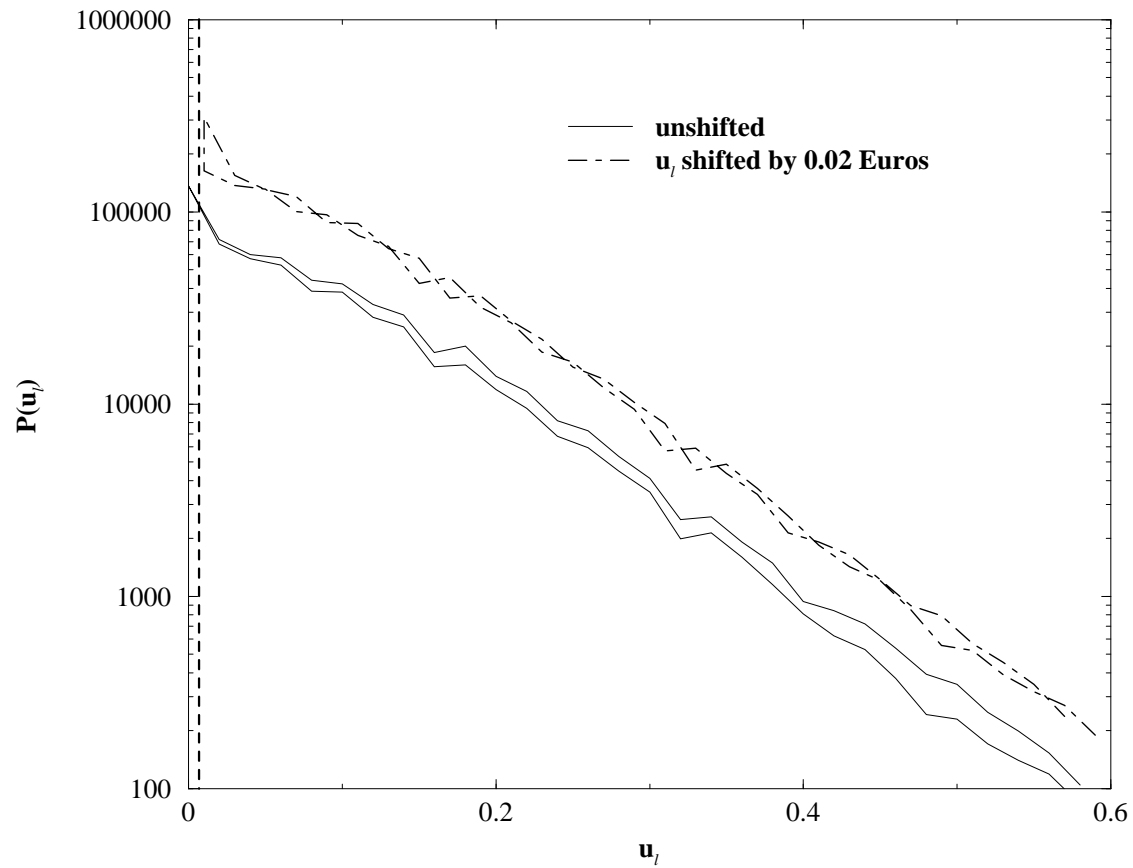
- The full distribution of $u_\ell = (p_{n+\ell} - p_n) \cdot \varepsilon_n$ is **nearly symmetrical** around its mean $\mathcal{I}(\ell) = \langle u_\ell \rangle \ll \sqrt{\mathcal{D}(\ell)}$:

→ Very few trades can be qualified as 'informed' on the short run

Average response



Impact distribution



$\ell = 128$: where are the 'informed' trades??

Transient impact: more technicalities

- An exact relation that allows to measure $G(\ell)$:

$$\mathcal{I}(\ell) = K \left[G(\ell) + \sum_{0 < n < \ell} G(\ell - n) \mathcal{C}(n) + \sum_{n > 0} [G(\ell + n) - G(n)] \mathcal{C}(n) \right]$$

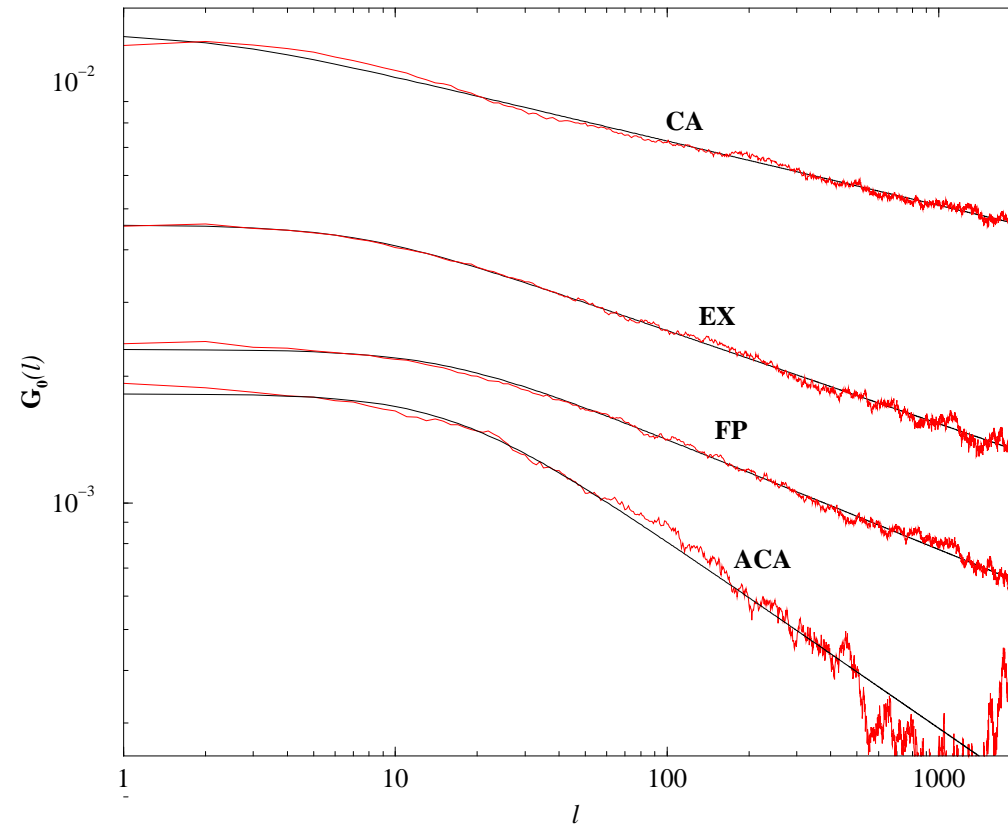
(and a more complicated equation for $\mathcal{D}(\ell)$).

- If $\mathcal{C}(\ell) \sim \ell^{-\gamma}$ and $G(\ell) \sim \ell^{-\beta}$ then:

$$\mathcal{D}(\ell) \sim \ell^{2-2\beta-\gamma}$$

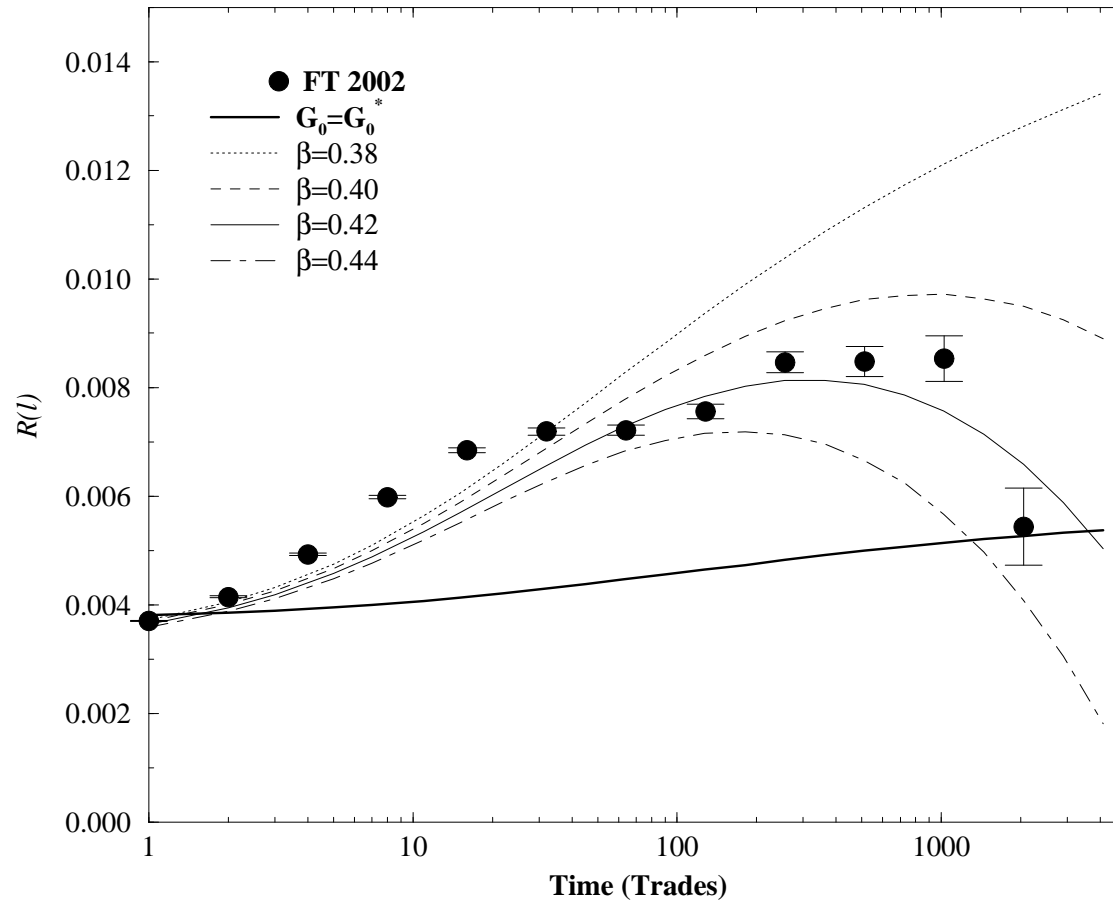
- For diffusion to be normal: $\beta = (1 - \gamma)/2$

Critically resilient markets



Decay of $G(l)$ for different stocks: impact is transient – with [J. Kockelkoren](#), [M. Potters](#)

Theoretical and empirical response function



Second generation models

- Markets are **complex systems** (i.e. made of *heterogeneous, interacting elements*) → rich **endogenous** dynamics
- “**Second generation**” models should start from:
 - agent based models (what do traders do?),
 - high frequency microstructure data,
 - a proper theory of **impact** (non-linear, transient,...)
 - identify **interactions, feedback loops and contagion mechanisms**

Second generation models

- Coarse-graining *should* lead to the emergence of some **universality, power-laws and intermittency** (but how, precisely?)
- Should allow to **predict** (at least qualitatively) the **value and dynamics of the parameters** (volatility, correlations, etc.)
- Help identify **systemic instabilities**
(e.g. spread \rightarrow vol. \rightarrow spread and May 6th “flash crash”)
- Think about **rules and regulations** that **endogenize stabilisation mechanisms**
(e.g. mark-to-market with liquidity discount, dynamic make/take fees, etc.)