



Cellular Automata as an Alternative to Geometric Brownian Motion

Frank Cuypers

Outline



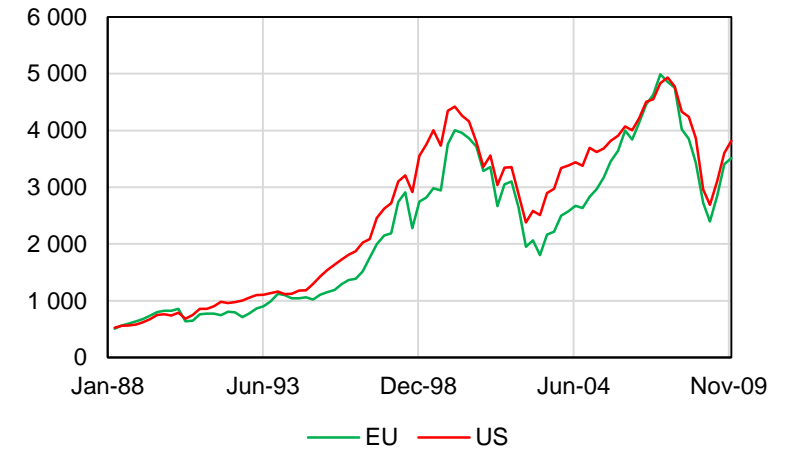
- Geometric Brownian Motion
- Cellular Automata

- Geometric Brownian Motion
- Cellular Automata

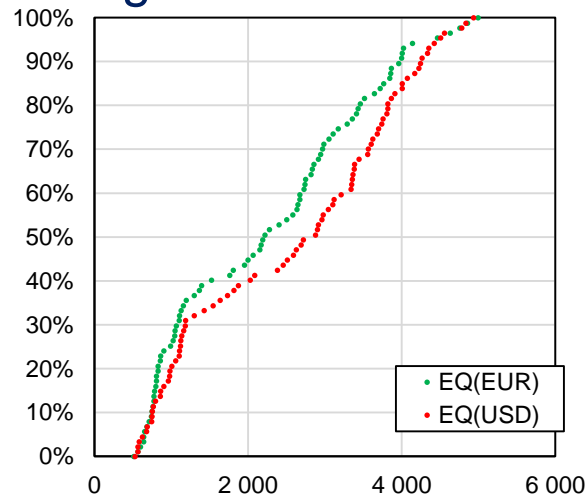
Choice of Variables

2 equity indices: EU & US

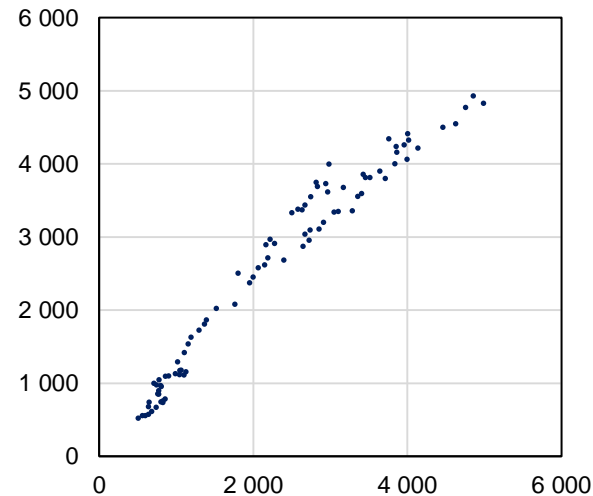
- Index values = X_t = auxiliary derived variables
⇒ Inappropriate for modelling



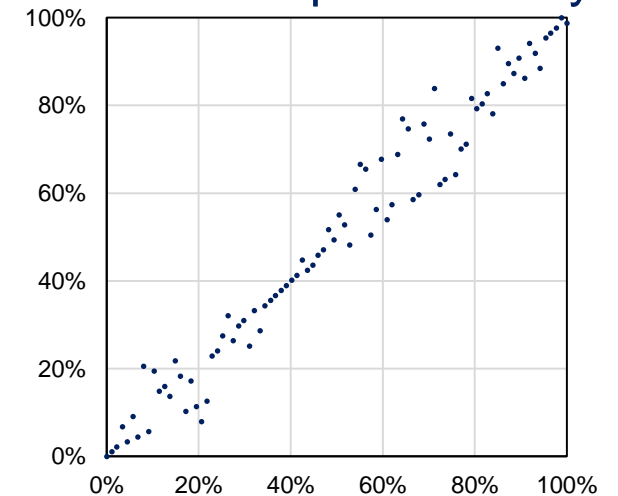
marginals CDF



linear correlation ~ 98%



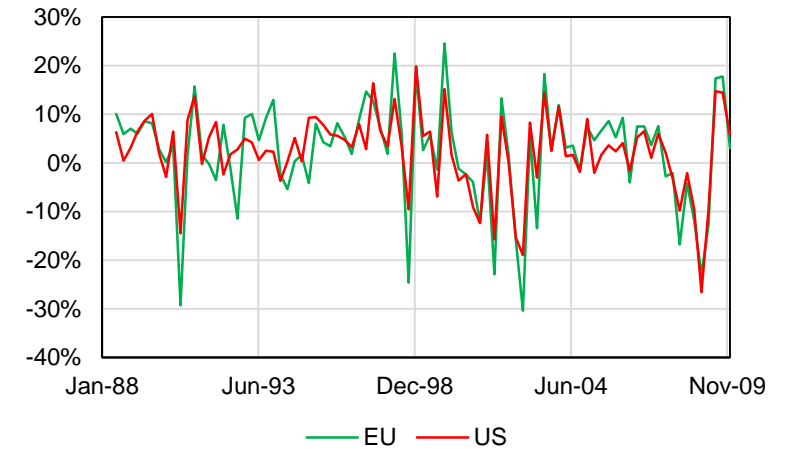
copula density



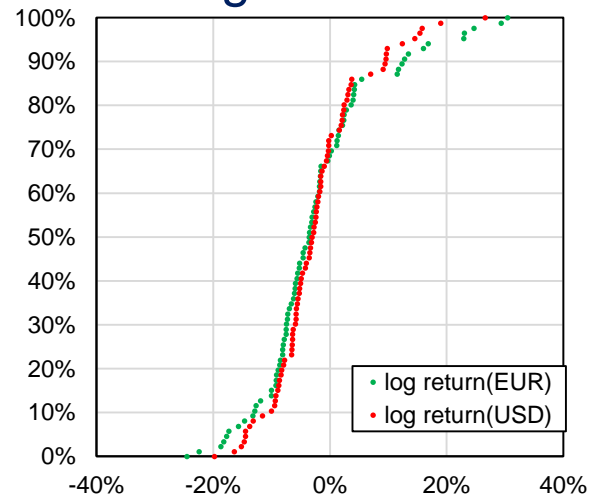
Choice of Variables

2 equity indices: EU & US

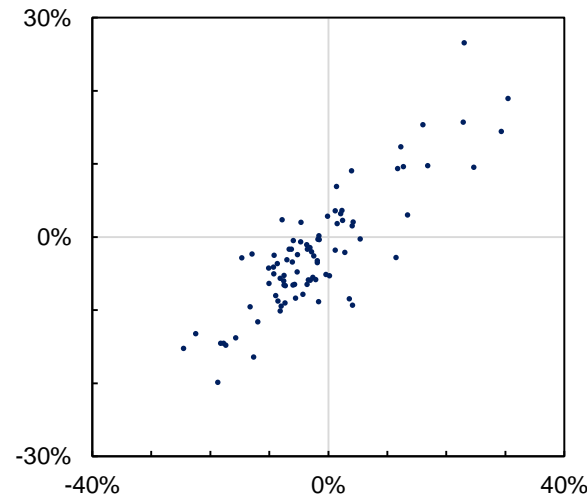
- Index values = X_t = auxiliary derived variables
⇒ Inappropriate for modelling
- Log returns $r_t = \ln \frac{X_t}{X_{t-1}}$ = fundamental stationary variables
⇒ Easier to model



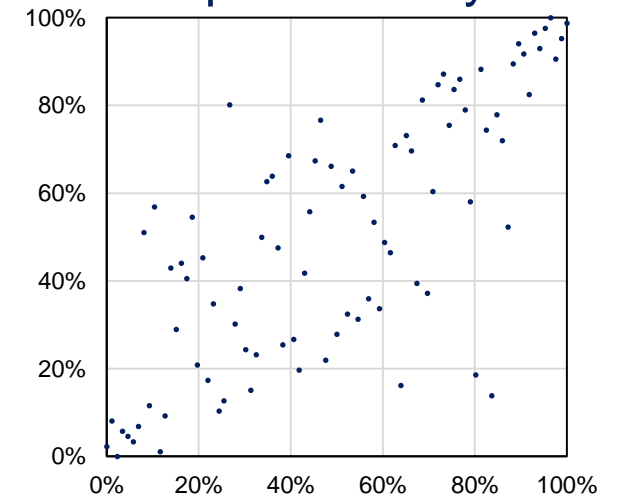
marginals CDF



linear correlation ~ 78%



copula density



Geometric Brownian Motion



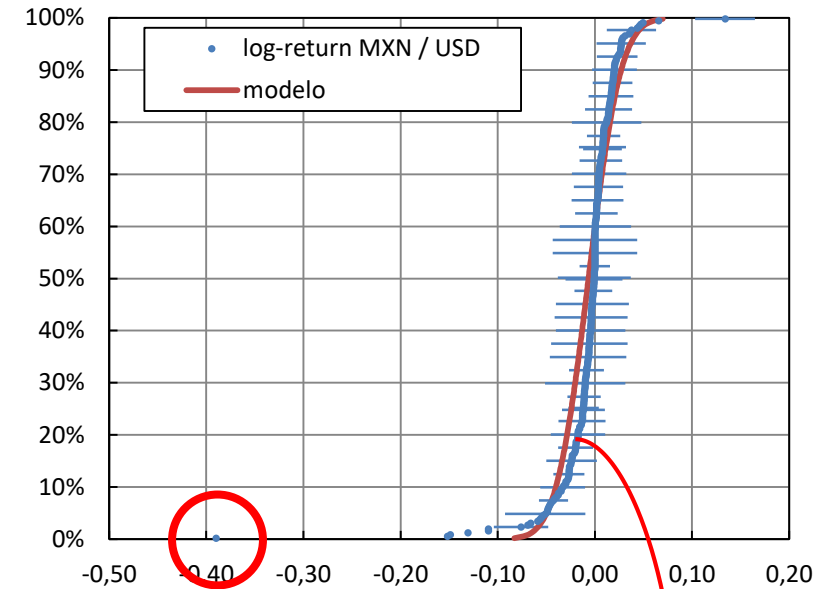
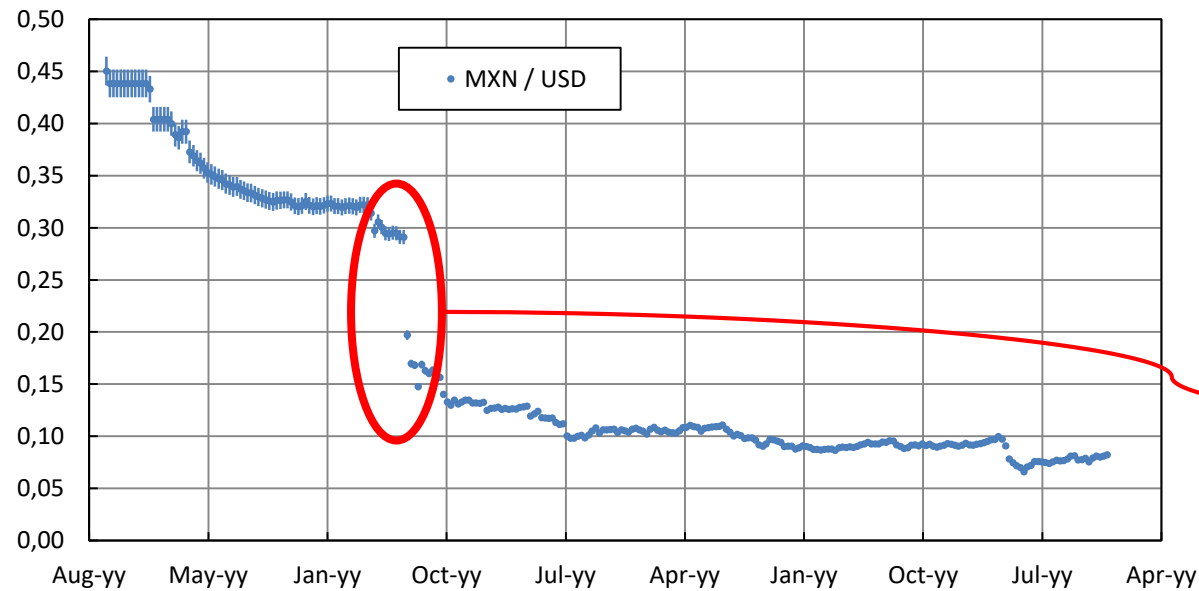
■ Risk-free $dX = \mu X dt \Rightarrow X = X_0 e^{\mu t}$

■ GBM $dX = \overbrace{\mu X dt}^{\text{drift}} + \overbrace{\sigma X dW}^{\text{diffusion}}$
Wiener process $W \sim N(0, \sqrt{t})$

■ Logreturn (Itô) $r_t = \ln \frac{X_t}{X_0} \sim N\left(\left(\mu - \frac{\sigma^2}{2}\right) t, \sigma\sqrt{t}\right) \Rightarrow X_t \sim \text{lognormal}$

■ Statistics $\mathbb{E}[X_t] = X_0 e^{\mu t}$
 $\mathbb{V}[X_t] = \mathbb{E}[X_t]^2 (e^{\sigma^2 t} - 1)$

normal models 



$\chi^2 / \text{dof} = 0.6$



Challenges with Geometric Brownian Motion

- No shocks
 - Non-normal log returns
 - Lévy process
 - Regime-switching lognormal models
 - No heteroskedasticity
 - Stochastic volatility models (Heston)
 - GARCH process
- BUT scaling
- BUT more complex
- BUT more parameters
- BUT tough to calibrate
- BUT more parameters

Outline



- Geometric Brownian Motion

- Cellular Automata

Cellular Automaton

Nearest neighbour 2 states XOR automaton:

$$Z_i(t) = Z_{i-1}(t-1) \oplus Z_{i+1}(t-1)$$

\oplus	-1	1
-1	-1	1
1	1	-1

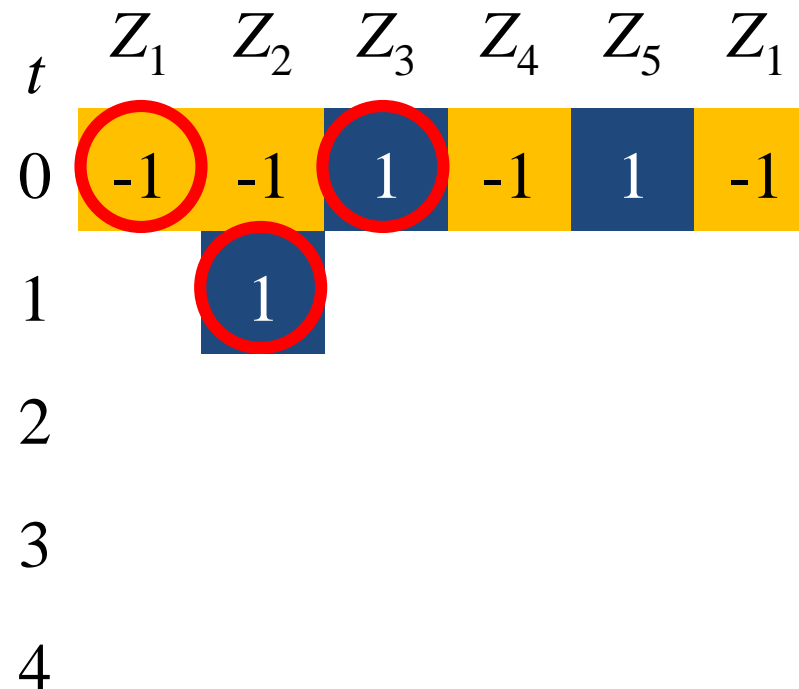
t	Z_1	Z_2	Z_3	Z_4	Z_5	Z_1
0	-1	-1	1	-1	1	-1
1						
2						
3						
4						

Cellular Automaton

Nearest neighbour 2 states XOR automaton:

$$Z_i(t) = Z_{i-1}(t-1) \oplus Z_{i+1}(t-1)$$

\oplus	-1	1
-1	-1	1
1	1	-1



Cellular Automaton

Nearest neighbour 2 states XOR automaton:

$$Z_i(t) = Z_{i-1}(t-1) \oplus Z_{i+1}(t-1)$$

\oplus	-1	1
-1	-1	1
1	1	-1

t	Z_1	Z_2	Z_3	Z_4	Z_5	Z_1
0	-1	-1	1	-1	1	-1
1	-1	1	-1	-1	-1	-1
2						
3						
4						

Cellular Automaton

Nearest neighbour 2 states XOR automaton:

$$Z_i(t) = Z_{i-1}(t-1) \oplus Z_{i+1}(t-1)$$

\oplus	-1	1
-1	-1	1
1	1	-1

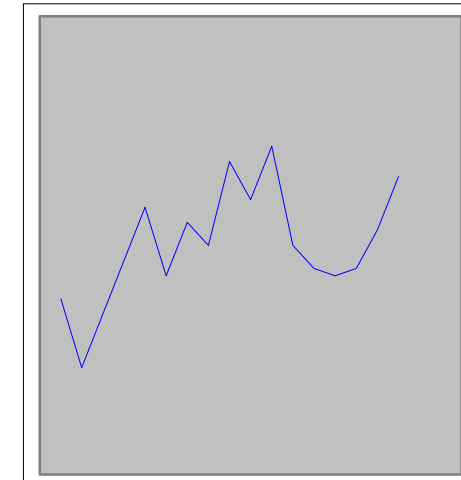
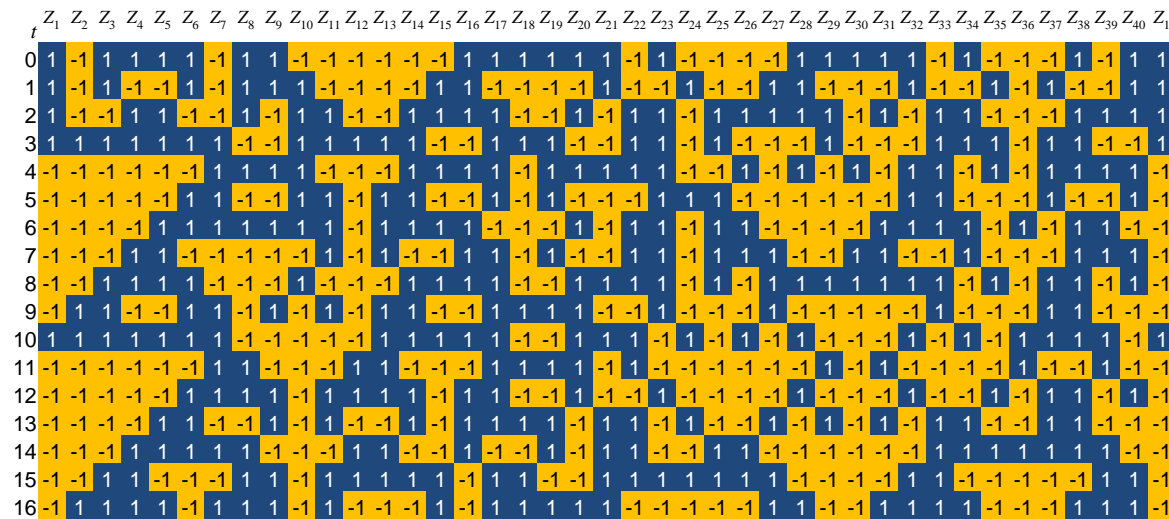
t	Z_1	Z_2	Z_3	Z_4	Z_5	Z_1
0	-1	-1	1	-1	1	-1
1	-1	1	-1	-1	-1	-1
2	1	-1	1	-1	-1	1
3	1	-1	-1	1	1	1
4	1	1	1	1	-1	1

Nearest neighbour 2 states XOR automaton:

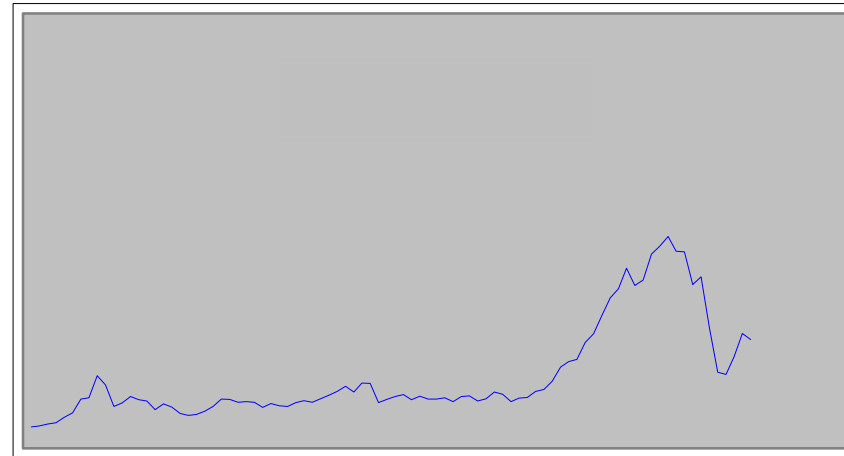
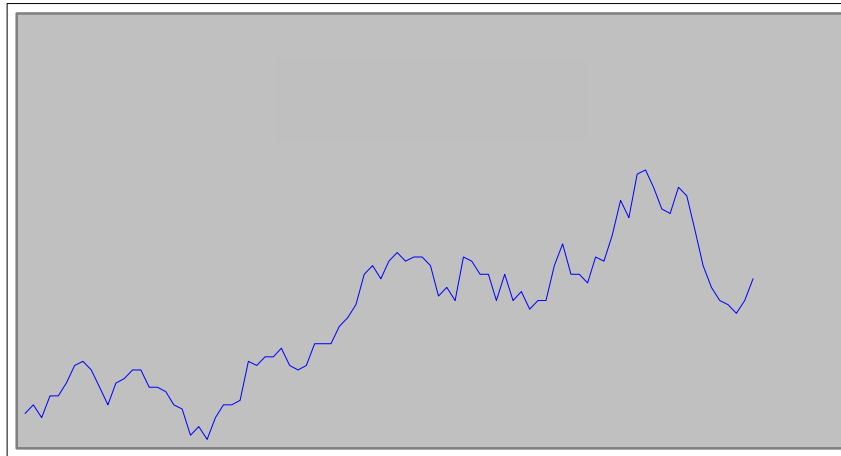
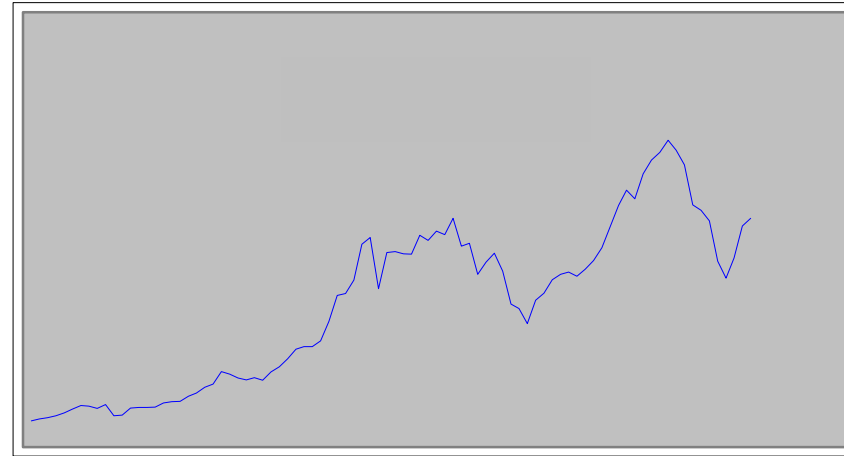
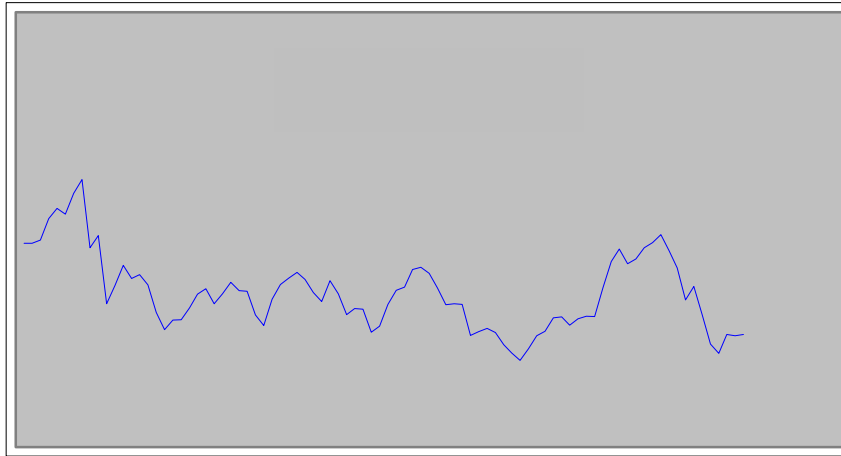
$$Z_i(t) = Z_{i-1}(t-1) \oplus Z_{i+1}(t-1)$$

$$I(t) = I(t-1) + \sum_{i=1}^N Z_i(t)$$

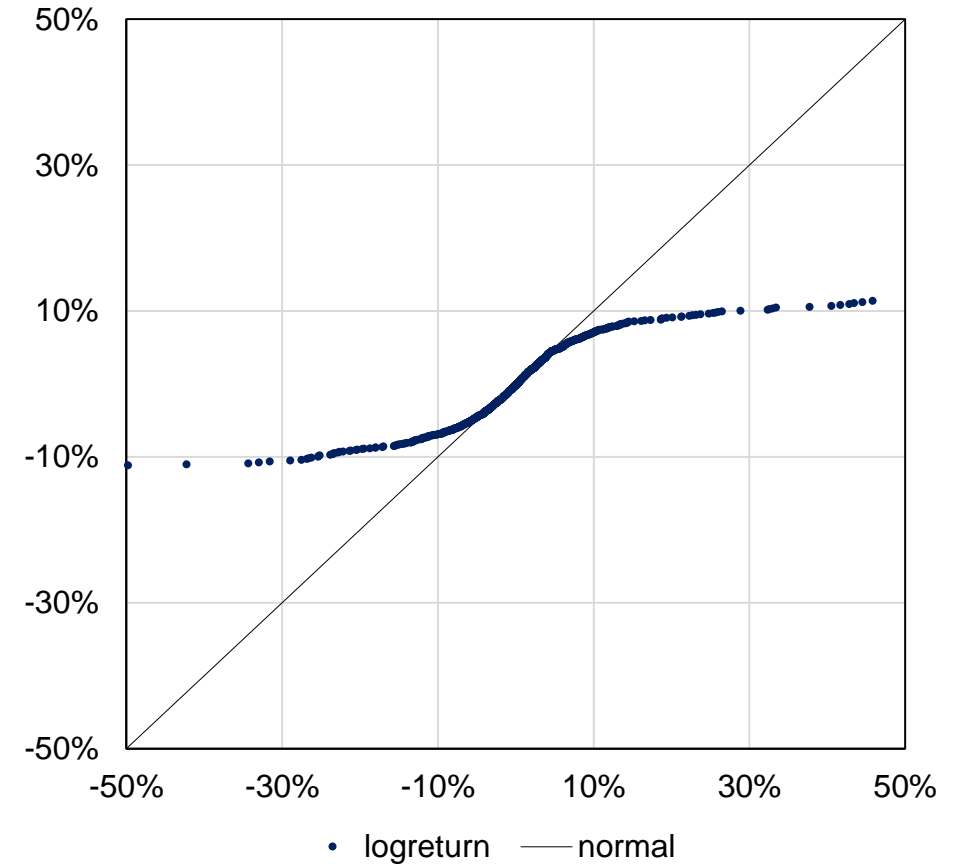
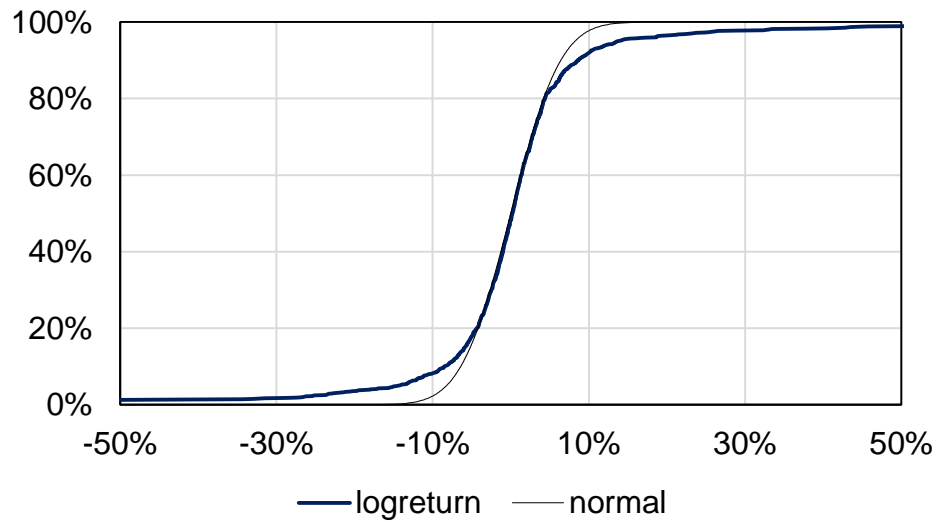
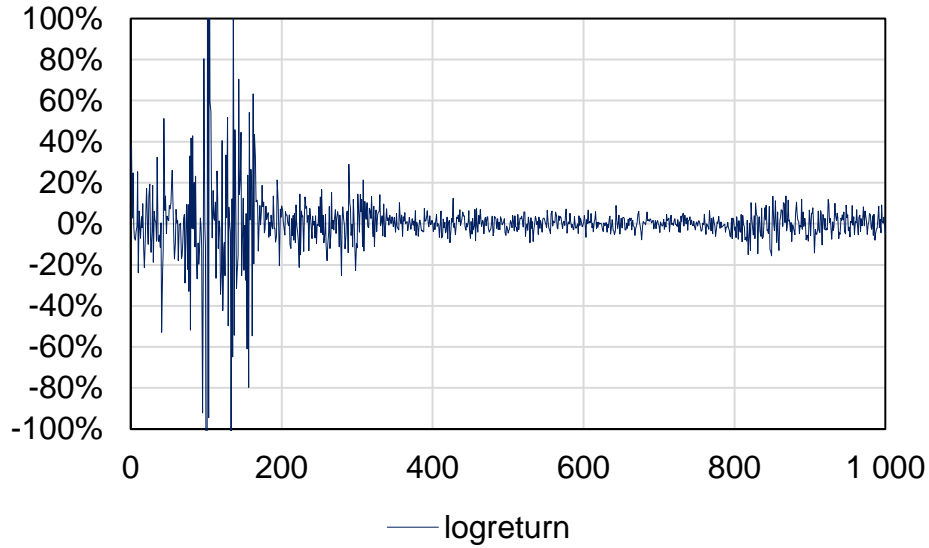
\oplus	-1	1
-1	-1	1
1	1	-1



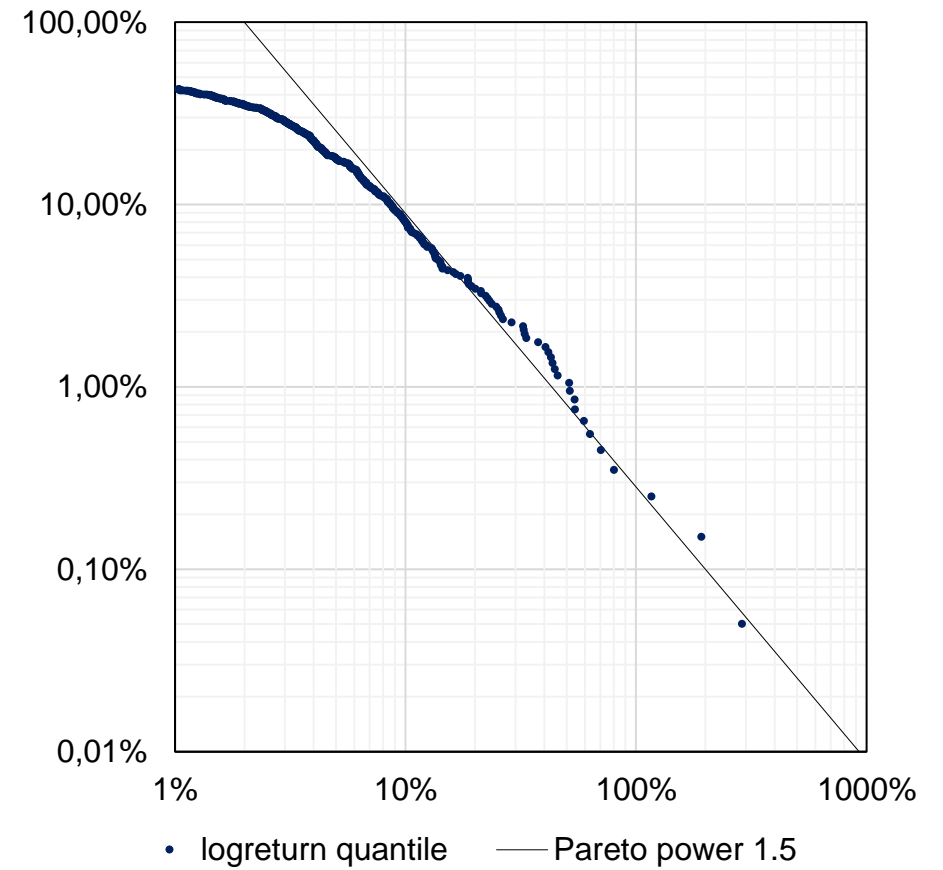
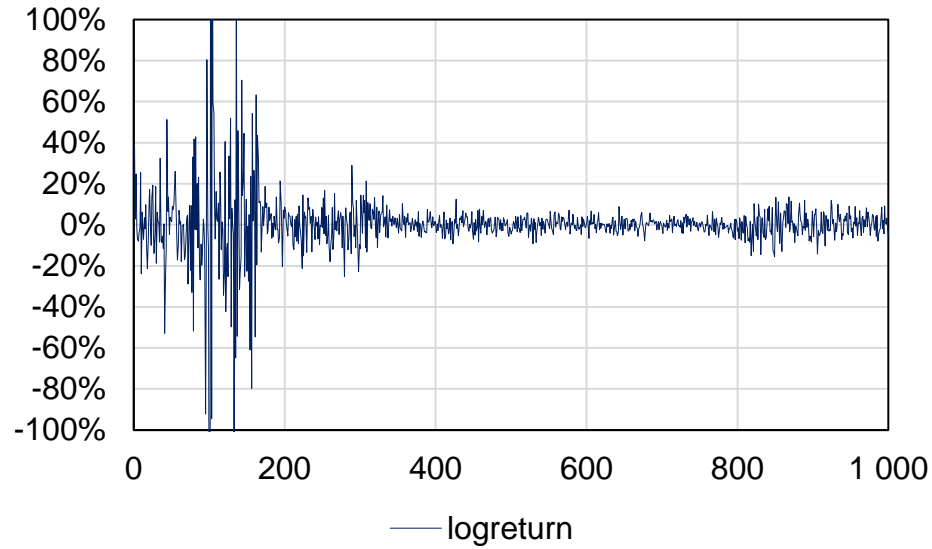
Equity Indices 1988 – 2010



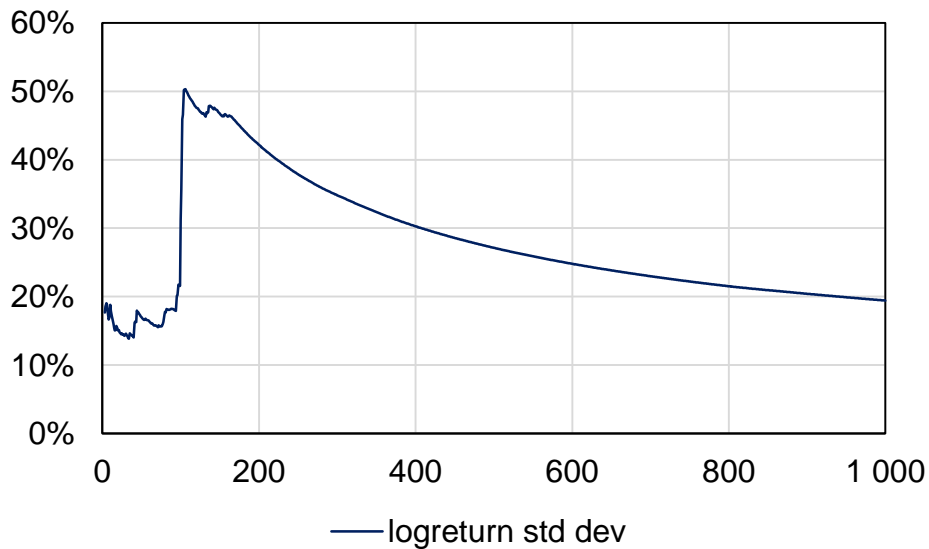
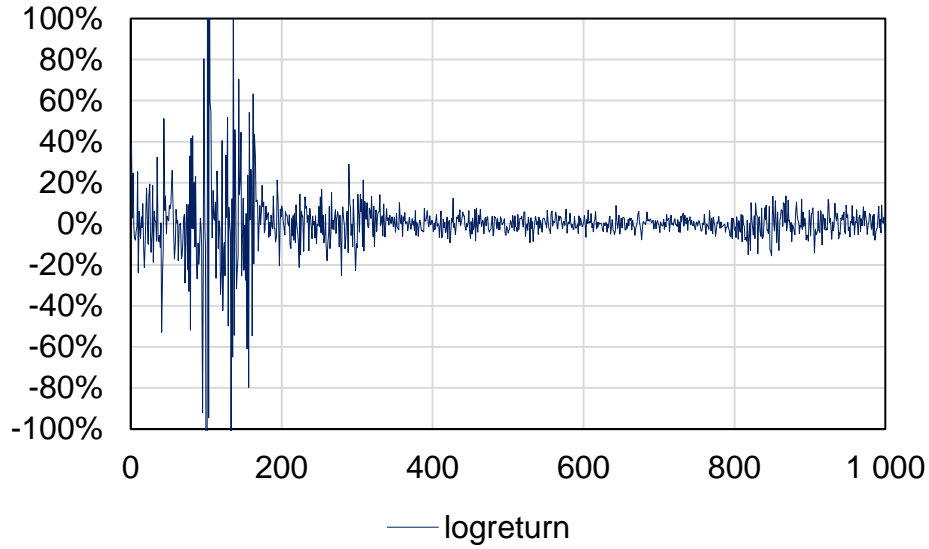
Logreturns: Distribution



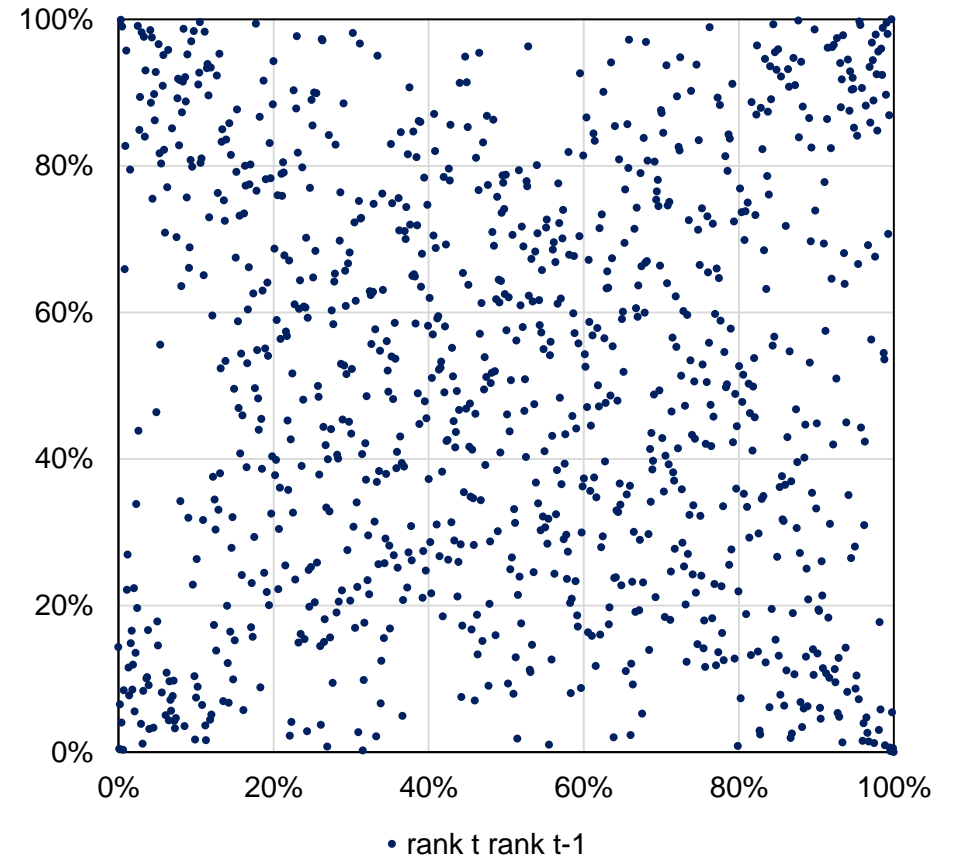
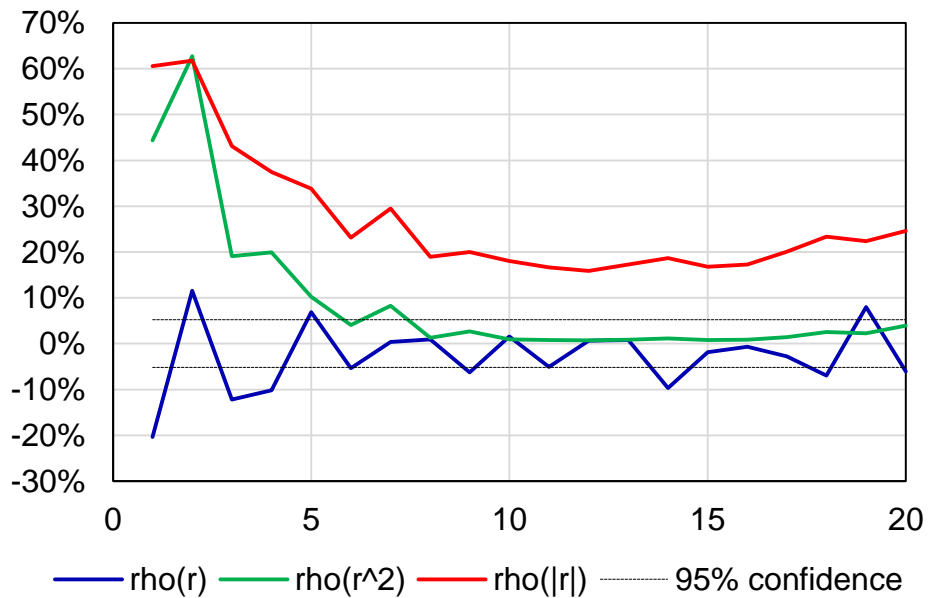
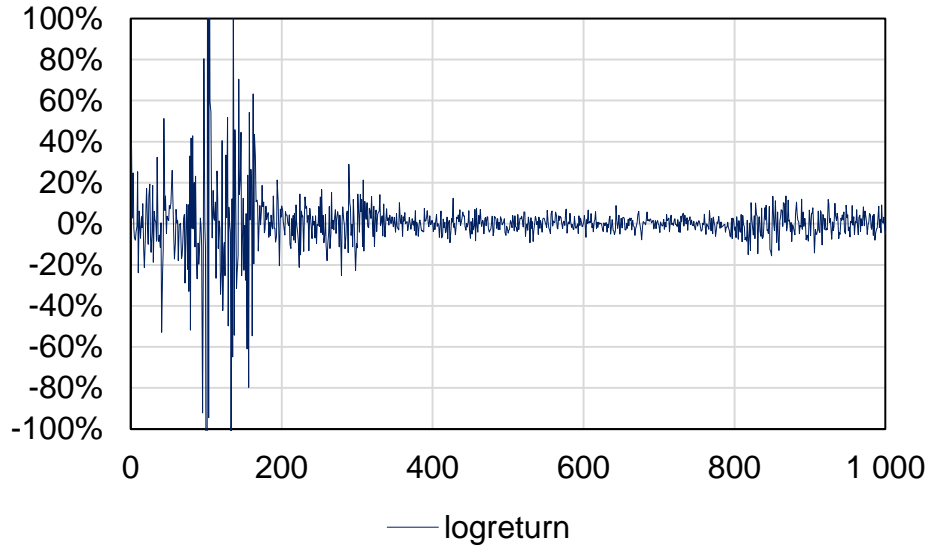
Logreturns: Tail



Logreturns: Volatility



Logreturns: Autocorrelation



- Geometric Brownian Motion
- Cellular Automata

Challenges with Geometric Brownian Motion

- No shocks
 - Non-normal log returns BUT scaling
 - Lévy process BUT more complex
 - Regime-switching lognormal models BUT more parameters

- No heteroskedasticity
 - Stochastic volatility models (Heston) BUT tough to calibrate
 - GARCH process BUT more parameters

- Rule 90 cellular automaton
 - Reproduces the rich phenomenology of financial time series
 - Requires minimal amount of parameters

Lecturer's Coordinates



Frank Cuypers

+41 (41) 725 32 94

frank.cuypers@prs-zug.com