

Limited information and diversification in the growth optimal portfolio

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The Kelly game

- 1 in one turn, a fraction f of the current wealth can be invested
 - with the probability p , the invested amount is doubled
 - with the probability $1 - p$, the invested amount is lost
- 2 repeat (infinitely) many times
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- **question:** how to find the optimal investment fraction?
 - **well-known answer:** maximise the exponential growth rate

$$G(f) := \langle \ln(1 + f R_1) \rangle$$

R_1 = game return on one-turn basis

The Kelly game

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- in real life:
 - simultaneous games
 - unknown game properties
 - ...

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- the effective portfolio size m_{ef}

optimal portfolio
constructed from
 N correlated assets



optimal portfolio
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??? uncorrelated assets

Effective portfolio size: properties

- both for the Kelly portfolio and the M-V portfolio:

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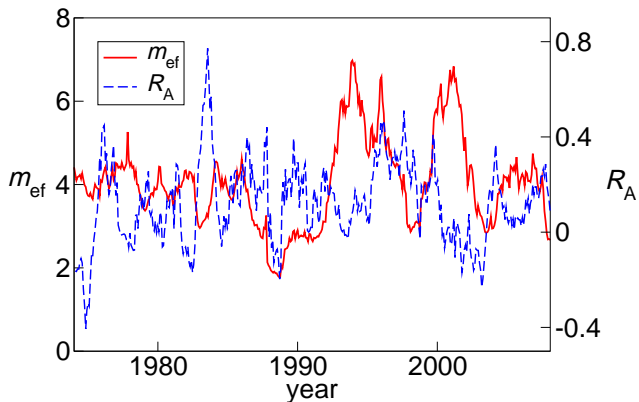
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- all correlations identical:

$$m_{\text{ef}} = \frac{M}{1 + (M-1)C} < \frac{1}{C}$$

Effective portfolio size: evolution

20 current stocks from the DJIA (Jan 1973—Apr 2008)



Limited information: framework

- even “noisy” information in the form $p \pm \Delta$ is artificial
- let's assume that we use only T past turns for learning

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T turns, w wins



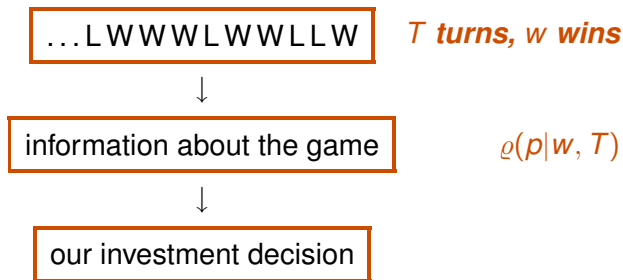
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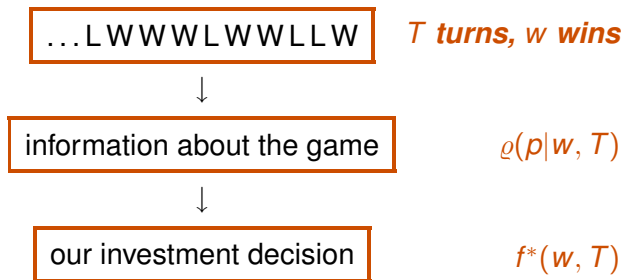
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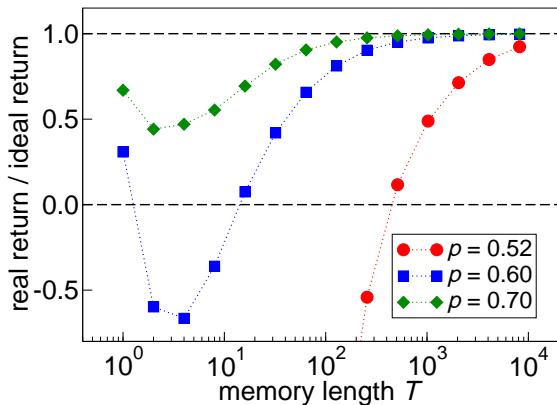
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- two interesting cases:

$$\lim_{T \rightarrow \infty} f^*(w, T) = 2 \lim_{T \rightarrow \infty} \frac{w}{T} - 1 = 2p - 1$$

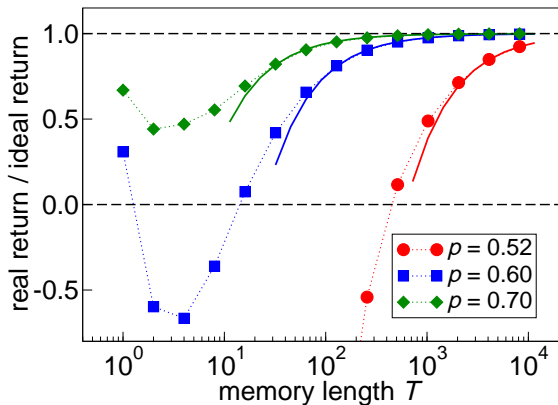
$$f^*(T, T) = \frac{T}{T + 2} < 1$$

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$$G^*(p, T) \approx \underbrace{\ln 2 + p \ln p + (1 - p) \ln(1 - p)}_{\text{perfect information}} - \underbrace{1/(2T)}_{\text{limited information}}$$



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- “There cannot be a sure-win game!”
 - set $\pi(p) = 0$ for $p > p_{\max}$
- “Great, I have my posterior $P(p|w, T)$ but what if. . .”
 - necessary because with enough data, prior beliefs are overruled!

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 - limited information in a toy system
 - simple analytical results

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- realistic risky games (log-normal returns, *etc.*)
- full capabilities of the prior information $\pi(p)$
- less frequent portfolio rebalancing
- transaction costs
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Thank you for your attention