

The Importance of Unlikely Events

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Long-Run Investment

how should you choose a pension fund?

- enormous number of possible scenarios

popular advice: assume the **typical** shock distribution

- based on the Law of Large Numbers
- but incorrect for generic EU maximizers
- rare contingencies can matter enormously

we:

- only one (or two) **atypical** shock distributions matter
- all other contingencies are too rare/inconsequential

Our Contribution

a systematic study of investment problems with long horizons

the most standard DM: Bayes-rational expected-utility maximizer

generically, choice is driven by rare-event considerations

tools from large deviations theory

information has no value in many settings

a simple representation via a fear-of-ruin constraint

Literature

Kelly'56 vs Samuelson'71,'79

Rietz'88, Barro'06

Robson, Samuelson & Steiner'23, Samuelson & Steiner'25, [Millner'25](#)

Weitzman'98 Gollier & Weitzman'10

Hansen & Sargent'01, Maccheroni, Marinacci & Rustichini'06,
Strzalecki'11

1 Model

2 Two Tools

3 CRRA Utility

4 Information and Hedging

5 Fear Constraint

6 A Normative Remark

7 Conclusion

Decision Problem

$$t = 0, \dots, T$$

$a \in A$ chosen at $t = 0$; A compact

a known stochastic process generates $(\theta_1, \dots, \theta_T)$ on finite Θ

- $q \in \Delta(\Theta)$ an empirical distribution of the sequence (a r.v.)

$$w_T(a) = \prod_{t=1}^T R(a, \theta_t) = \exp \left[\sum_{t=1}^T r(a, \theta_t) \right] = \exp [E_q r(a, \theta) T] = \exp [r(a, q) T]$$

the DM solves

$$\max_{a \in A} E u(w_T(a))$$

$u(w)$ increasing, continuous; $r(a, \theta)$ continuous and concave in a

Examples

finance

- set of assets J
- $a \in A \equiv \Delta(J)$ a stationary rebalanced portfolio
- $r(a, \theta)$ the portfolio's log-return

net-present values

- $R(a, \theta)$ is a stochastic discount factor

demographics

- w_T is the population size

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Rate Function

assumption (rate function)

The sequence $(q_T)_{T=1}^{\infty}$ satisfies a large-deviation principle with a **rate function** $I(q)$.

informally,

$$\Pr(q_T \approx q) \approx e^{-I(q)T}$$

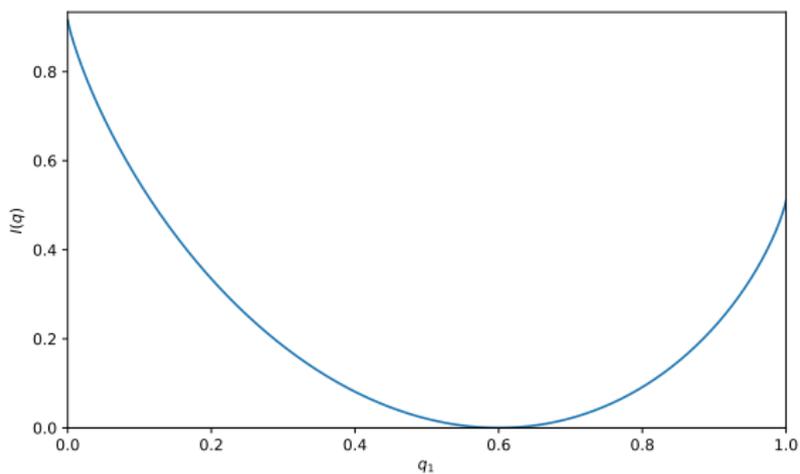
rate functions exist for many processes: iid, Markov, ...

regularity condition $I(q)$ continuous, convex

IID Example

θ_t iid from p

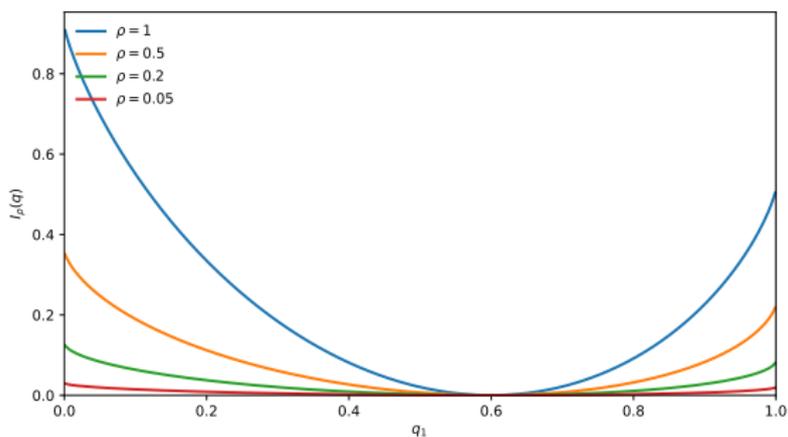
Sanov: $I(q) = \text{KL}(q||p)$



Persistence Example

Markov transition matrix

$$\begin{bmatrix} 1 - \rho p_2 & \rho p_2 \\ \rho p_1 & 1 - \rho p_1 \end{bmatrix}$$



Certainty Equivalents

definition

Certainty equivalent $C_T(a)$ is the deterministic growth rate that makes the DM indifferent to a :

$$E u(w_T(a)) = u(\exp[C_T(a) T]).$$

expected-utility maximization is equivalent to

$$\max_{a \in A} C_T(a)$$

asymptotic certainty equivalent:

$$C(a) := \lim_{T \rightarrow \infty} C_T(a)$$

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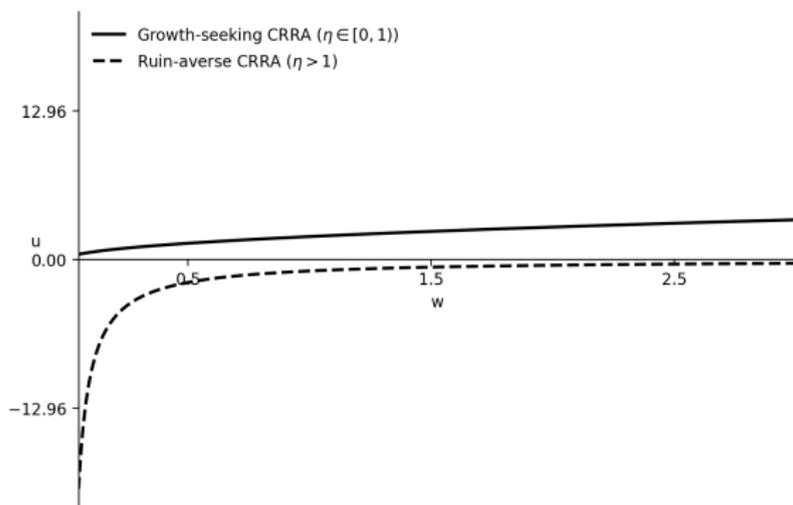
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CRRA Utility Function

$$u(w) = \frac{w^{1-\eta}}{1-\eta} \quad (\eta \neq 1).$$



Limit Certainty Equivalents

lemma

growth-seeking DM

$$C_{\text{CRRA}}(a) = \sup_q \left\{ r(a, q) - \frac{1}{1-\eta} I(q) \right\}$$

ruin-averse DM

$$C_{\text{CRRA}}(a) = \inf_q \left\{ r(a, q) + \frac{1}{\eta-1} I(q) \right\}$$

let

$$C_{\text{CRRA}} := \max_{a \in A} C(a)$$

assume a unique maximizer a_{CRRA}^*

proposition

$$a_T^* \rightarrow a_{\text{CRRA}}^* \text{ and } \max_a C_T(a) \rightarrow C_{\text{CRRA}}.$$

Intuition

consider a risk-neutral DM

$$\begin{aligned} EU_T &= \int_{\Delta} \exp[r(a, q)T] d\pi_T(q) \\ &\approx \int_{\Delta} \exp[(r(a, q) - I(q))T] dq \\ &\approx \exp\left[\underbrace{\max_{q \in \Delta} \{r(a, q) - I(q)\}}_{C_{\text{CRRA}}(a;0)} T\right] \end{aligned}$$

- growth-seeking: selection on the growth of **utility**
- ruin-aversion: selection on the growth of **disutility**

proof via Varadhan's lemma

Example: Welfare Cost of Business Cycles

let's revisit the binary Markov chain

- stationary distribution is independent of persistence
- persistence doesn't affect welfare on the typical path

assume, realistically, $\eta > 1$

- welfare decreases with persistence
- the job of the adversarial Nature is getting easier

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Perfect Foresight

the DM with **perfect foresight** knows q_T before choosing a

asymptotic perfect-foresight certainty equivalent

$$C_{\text{pf}} = \lim_{T \rightarrow \infty} \sup_{a(\cdot)} C_{\text{pf}, T}(a(\cdot));$$

the deterministic growth rate as good as perfect foresight

Proposition

*At the exponential rate, the uninformed **CRRA** DM achieves the same payoff as with perfect foresight:*

$$C_{\text{CRRA}} = C_{\text{pf}}.$$

a CRRA DM won't pay any fraction of return for perfect foresight

Proof (ruin aversion)

zero-sum game between the DM and Nature with payoff

$$r(a, q) + \frac{1}{\eta - 1} l(q)$$

uninformed sequential game:

- DM chooses a
- Nature chooses q

perfect-foresight sequential game:

- Nature chooses q
- DM chooses a

Minimax theorem: order of moves doesn't matter

Intuition

the perfect-foresight DM achieves growth rate

$$v(q) = \max_a r(a, q) \text{ in each contingency } q$$

but only one such contingency, q^* , dominates the EU aggregation

uninformed DM best responds q^* and matches the perfect-foresight DM

Example: Kelly Meets Samuelson

Kelly maximizes the typical growth rate

$$a_{\text{Kelly}}^*(p) \in \arg \max_{a \in \Delta(J)} r(a, p), \text{ achieving } v(p)$$

a CRRA investor chooses a Kelly portfolio

$$a_{\text{CRRA}}^* = a_{\text{Kelly}}^*(q^*),$$

for a distorted belief

$$\arg \max_q \left\{ v(q) - \frac{1}{1-\eta} I(q) \right\} \text{ (growth-seeking)}$$

$$\arg \min_q \left\{ v(q) + \frac{1}{\eta-1} I(q) \right\} \text{ (ruin-aversion)}$$

Value of Information

for each T , the DM observes x generated by $\mu_T(x | q_T)$ and chooses $a(x)$

Value of Information

$$\text{Vol} = \lim_{T \rightarrow \infty} \left(\sup_{a(\cdot)} C_{\text{info}, T}(a(\cdot)) - \sup_{a \in A} C_T(a) \right)$$

Corollary

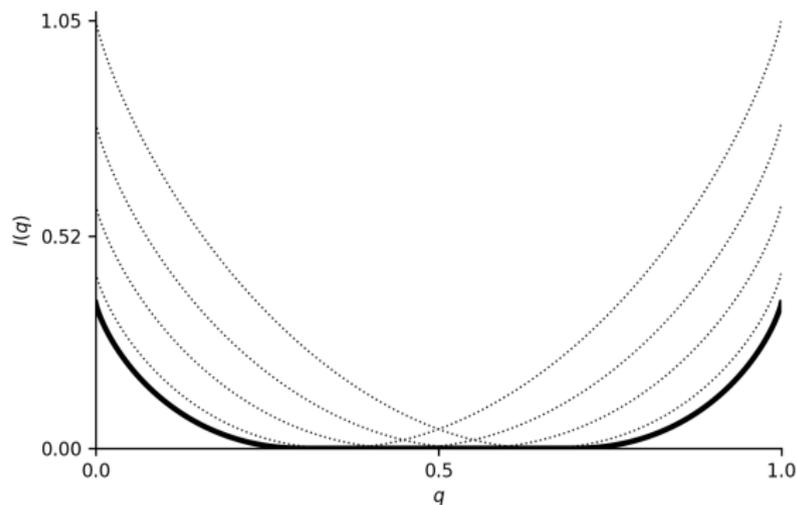
Information has no value for a CRRA DM at the exponential rate

$$\text{Vol} = 0.$$

Example: Hierarchical Process

x drawn from $X \subset \Delta(\Theta)$

θ_t iid from x



Hedging

the DM splits initial wealth into subfunds $k = 1, \dots, K$ with shares λ_k

$$w_T(\mathbf{a}, q_T) = \sum_k \lambda_k w_T(a_k, q_T)$$

hedging is powerful:

- asymptotically, it achieves the best response to q within $\{a_1, \dots, a_K\}$

Value of Hedging

Value of Hedging

$$\text{VoH} = \lim_{T \rightarrow \infty} \left(\sup_{\mathbf{a}} C_{\text{hedge}, T}(\mathbf{a}) - \max_{a \in A} C_T(a) \right);$$

Corollary

*Hedging has no value for a **CRR**A DM at the exponential rate*

$$\text{VoH} = 0.$$

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Two-Tailed Utilities

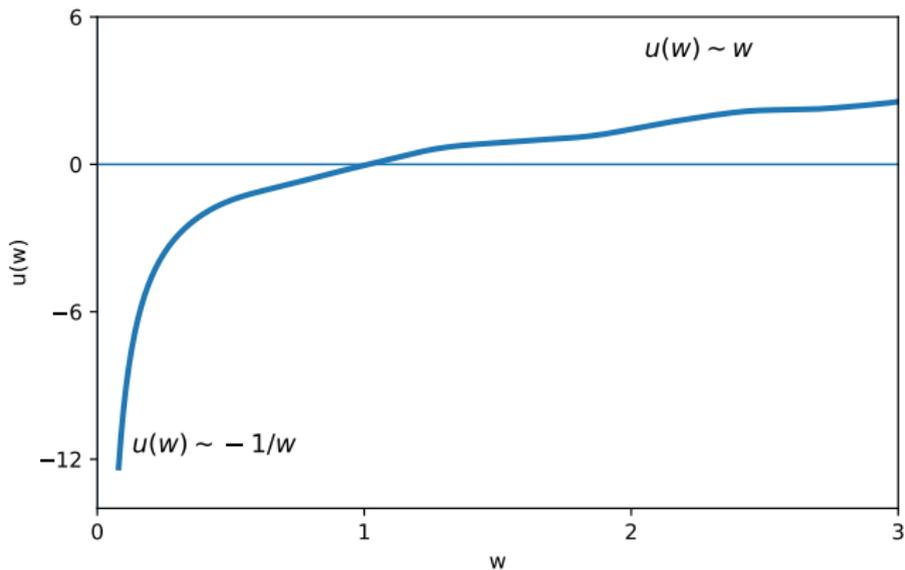
$u : \mathbb{R}_{++} \rightarrow \mathbb{R}$ increasing, continuous, and

$$\lim_{w \rightarrow \infty} \frac{\ln u(w)}{\ln w} = 1 - \eta^+, \quad \lim_{w \downarrow 0} \frac{\ln(-u(w))}{\ln w} = 1 - \eta^-.$$

with $\eta^+ \in [0, 1)$ and $\eta^- > 1$

both growth-seeking and ruin aversion

In This Presentation



Certainty Equivalents

recall from the CRRA setting:

$$C^+(a) = \max_q \{r(a, q) - I(q)\}$$

$$C^-(a) = \min_q \{r(a, q) + I(q)\}$$

Lemma

The asymptotic certainty equivalent for the two-tailed utility is

$$C(a) = \begin{cases} C^+(a) & \text{if } C^+(a) + C^-(a) > 0, \\ C^-(a) & \text{if } C^+(a) + C^-(a) < 0. \end{cases}$$

intuition:

- $EU \approx e^{C^+(a)T} - e^{-C^-(a)T}$
- the larger growth coefficient prevails

Solution

definition

Ruin-robust growth program is

$$\begin{array}{ll} \max_{a \in A} & C^+(a) \\ \text{s.t.} & C^+(a) + C^-(a) \geq 0, \end{array}$$

with a convention that if no a satisfies the constraint, then the DM solves

$$\max_{a \in A} C^-(a).$$

Let C_{tt} be the value and assume a unique maximizer a_{tt}^* .

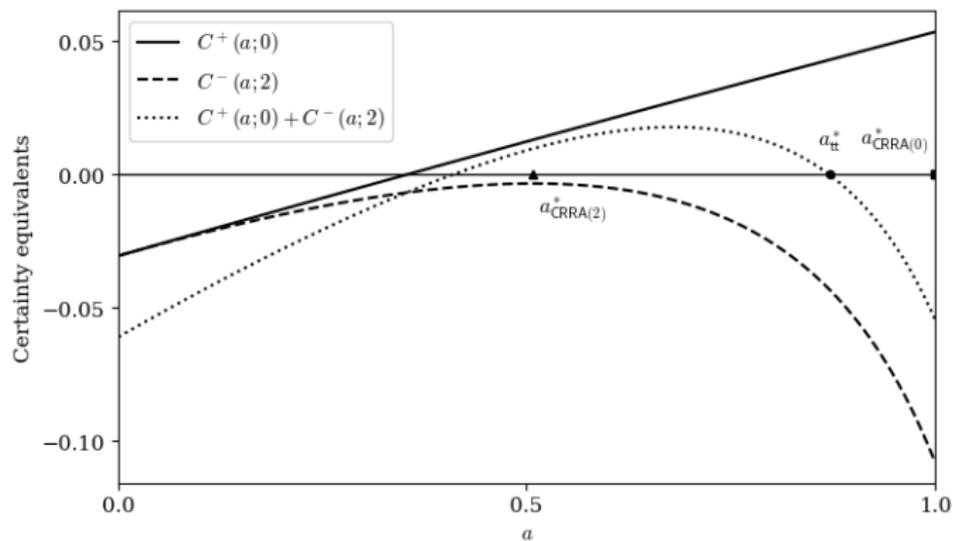
regularity condition: any a that generates a tie can be approximated by actions at which the upper tail is selected.

Proposition

$$a_T^* \rightarrow a_{tt}^* \text{ and } \max_a C_T(a) \rightarrow C_{tt}.$$

Example: Equity Premium Puzzle

$$R(a, \theta) = s \begin{array}{cc} H & L \\ 1.10 & 0.2 \\ c & 0.97 \quad 0.97 \end{array}$$



Valuable Information

Lemma

The perfect-foresight certainty equivalent for the two-tailed utility is

$$C_{\text{pf}} = \begin{cases} C_{\text{CRRA}}^+ & \text{if } C_{\text{CRRA}}^+ + C_{\text{CRRA}}^- > 0, \\ C_{\text{CRRA}}^- & \text{if } C_{\text{CRRA}}^+ + C_{\text{CRRA}}^- < 0. \end{cases}$$

⇒ information can generate value for two-tailed DMs

Hedging Achieves Perfect Foresight

Proposition

The DM with two-tailed utility who can hedge achieves the same payoff at the exponential rate as if she had perfect foresight:

$$C_{\text{hedge}} = C_{\text{pf}}.$$

constructive proof: hedge with two subfunds, with $a_{\text{CRRA}(+)}^*$ and $a_{\text{CRRA}(-)}^*$ matches the perfect foresight, asymptotically

Kelly Meets Samuelson Again

A two-tailed investor achieves the upper bound, \mathbb{C}_{pf} , on her long-run performance by dividing her initial wealth into two distinct Kelly portfolios,

$$a_{\text{Kelly}}^*(q_+^*) \text{ and } a_{\text{Kelly}}^*(q_-^*),$$

optimized for Nature's benevolent and malevolent large deviations:

$$q_+^* \in \arg \max_{q \in \Delta(\Theta)} \{v(q) - I(q)\},$$

$$q_-^* \in \arg \min_{q \in \Delta(\Theta)} \{v(q) + I(q)\}.$$

Hedging Makes Information Redundant

Define the *marginal value of information (given hedging)*,

$$\text{mVol} = \lim_{T \rightarrow \infty} \left(\sup_{\mathbf{a}(\cdot)} C_{\text{hedge, info}, T}(\mathbf{a}(\cdot)) - \sup_{\mathbf{a}} C_{\text{hedge}, T}(\mathbf{a}) \right),$$

as the asymptotic increase of the certainty equivalent due to information relative to that of the DM who can hedge but has no information.

Corollary

$$\text{mVol} = 0.$$

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Utilitarian Social Planner

continuous homogenous population with private shocks

utilitarian welfare = expected utility

but welfare evaluation is driven by a vanishing fraction of population

Truncated Utilitarian Welfare

assume the SP disregards extreme wealth tails

truncated utilitarian welfare:

$$W_T^\varepsilon(a) = \int_\varepsilon^{1-\varepsilon} u(w) dP(w; a).$$

certainty equivalent

$$(1 - 2\varepsilon) u(\exp[C_T^\varepsilon(a) T]) = W_T^\varepsilon(a);$$

delivers the same welfare once the ε wealth tails are truncated

Proposition

For any $\varepsilon \in (0, 1/2)$,

$$\lim_{T \rightarrow \infty} C_T^\varepsilon(a) = r(a, p).$$

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Conclusion

EUT for long-horizon growth control is

- driven by rare events
- tractable

Is it normatively appealing?

Details

The family of probability measures $(\pi_T)_{T=1}^\infty$ over q_T satisfies the **large deviation principle** with a rate function I :

For any closed set $F \subseteq \Delta(\Theta)$ and any open set $G \subseteq \Delta(\Theta)$,

$$\limsup_{T \rightarrow \infty} \frac{1}{T} \ln \pi_T(F) \leq - \inf_{q \in F} I(q),$$

$$\limsup_{T \rightarrow \infty} \frac{1}{T} \ln \pi_T(G) \geq - \inf_{q \in G} I(q).$$