

Functional portfolio generation

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My collaborators

- ▶ Soumik Pal (UW) - discrete time formulation of SPT and functionally generated portfolio
- ▶ Christa Cuchiero and Walter Schachermayer (Vienna) - Cover's universal portfolio and SPT
- ▶ Peter Baxendale (USC) - random concave functions

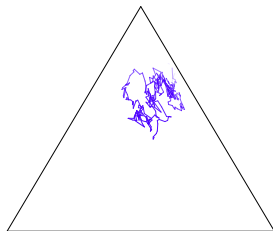
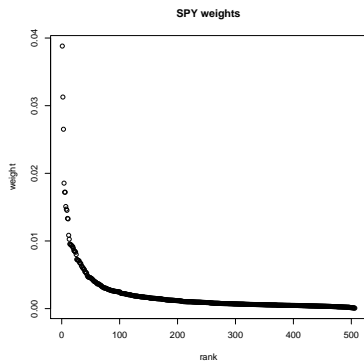
Outline

Two parts:

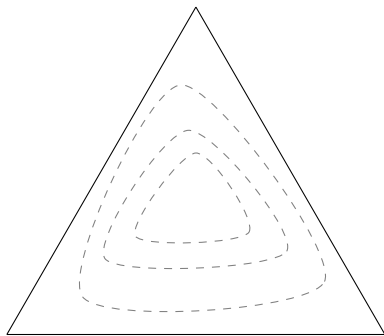
1. Concave functions on the simplex Δ_n have financial applications:
 - ▶ Measure of market diversity and volatility
 - ▶ Trading strategy
2. Models for generating *random* concave functions

Motivations

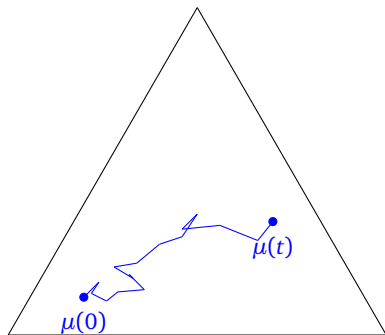
- ▶ **Stochastic portfolio theory** (Fernholz, Karatzas, ...)
- ▶ Macroscopic behaviors of stock markets, and trading strategies that exploit them.
- ▶ Mathematical framework: market weight $\mu(t) = (\mu_1(t), \dots, \mu_n(t))$ as a process in the simplex Δ_n .



Diversity and volatility



$$\varphi(\mu(t))$$



$$\sum_i \mathbf{D}[\mu(t_{i+1})|\mu(t_i)]$$

- ▶ Can we have portfolios such that
portfolio value relative to market = diversity + volatility?

Functional portfolio generation

Definition (W. 2018)

A trading strategy is functionally generated, if there exist g , φ and $\mathbf{D}[\cdot|\cdot] \geq 0$ such that the following pathwise decomposition holds:

$$g(V(t)) - g(V(0)) = \varphi(\mu(t)) - \varphi(\mu(0)) + \sum_{s < t} \mathbf{D}[\mu(s+1)|\mu(s)].$$

Here, the relative value of a self-financing strategy η is defined by

$$V(t) - V(0) := \sum_{s < t} \eta(s) \cdot \Delta\mu(s)$$

- ▶ Multiplicative generation (Fernholz 1999): $g(x) = \log x$
- ▶ Additively generation (Karatzas and Ruf 2017): $g(x) = x$

Characterization

Theorem (W. 2018)

Under suitable regularity conditions:

- (i) Up to a linear transformation g is of the form $g(x) = \frac{1}{\alpha} \log(C + x)$, where $\alpha > 0$, $C \geq 0$, or $g(x) = x$.
- (ii) $e^{\alpha\varphi} > 0$ is **concave**.
- (iii) \mathbf{D} is the $L^{(\alpha)}$ -**divergence** [Pal and W. (2016), W. (2018)]

$$\mathbf{D}[q|p] = \frac{1}{\alpha} \log(1 + \alpha \nabla \varphi(p) \cdot (q - p)) - (\varphi(q) - \varphi(p)).$$

The trading strategy is given by

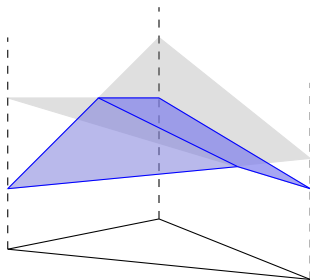
$$\eta_i(t) = \alpha(C + V(t))D_{e_i - \mu(t)}\varphi(\mu(t)) + V(t).$$

Random concave functions

- ▶ **Problem:** Construct probability measures ν on

$$\mathcal{C} := \{\psi : \Delta_n \rightarrow [0, \infty) \text{ concave}\}.$$

- ▶ **Motivation:** ν can serve as the *prior distribution* in Cover's universal portfolio and nonparametric statistics.
- ▶ **Main idea:** Let $\Psi = \min_{\alpha} \ell_{\alpha}$, the minimum of a random family of affine functions.

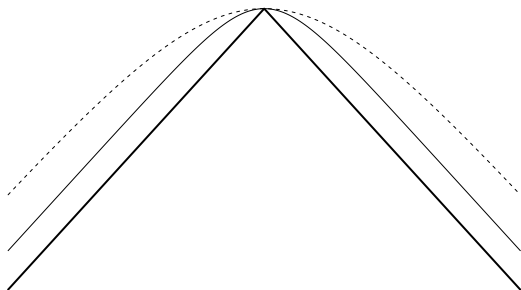


Softmin

- ▶ More generally, we consider the soft minimum

$$m_\lambda(x_1, \dots, x_K) := \frac{-1}{\lambda} \log \left(\frac{1}{K} \sum_{k=1}^K e^{-\lambda x_k} \right)$$

which is smooth and concave in x for any $\lambda > 0$.

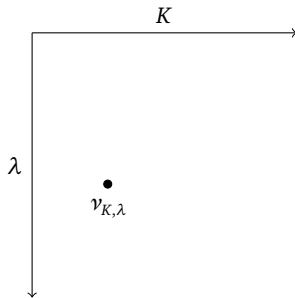
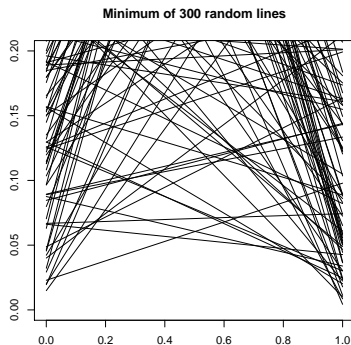


Probabilistic model

- ▶ Fix random vector $C \in (0, \infty)^n$ as coefficients of $\ell(p) = C \cdot p$.
- ▶ Let $\ell_1, \dots, \ell_K, \dots$ be i.i.d. copies of ℓ .
- ▶ Let $\Psi_K = c_K m_{\lambda_K}(\ell_1, \dots, \ell_K)$, $c_K > 0$ suitable scaling constant.

This gives a distribution $\nu = \nu_{K, \lambda_K, \text{Law}(C), c_K}$ on \mathcal{C} .

We are interested in the limiting properties as $K \rightarrow \infty$.

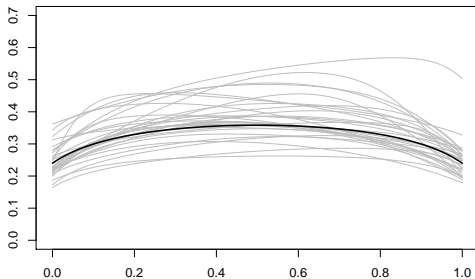


Case 1: $\lambda_K \equiv \lambda < \infty$

Theorem (Baxendale and W. (2018))

Let C be a random vector with values in $(0, \infty)^n$ which generates $\ell(p) = C \cdot p$. Let $\psi(t) = \log \mathbb{E} e^{t \cdot C}$ be its cgf. For $\lambda \in (0, \infty)$ fixed, let $\Psi_K = m_\lambda(\ell_1, \dots, \ell_K)$. Then, almost surely,

$$\Psi_K \rightarrow \Psi_\infty, \quad \Psi_\infty(p) := \frac{-1}{\lambda} \psi(-\lambda p).$$



Case 2: $\lambda_K \equiv \infty$

Conditions on the random vector $C = (C_1, \dots, C_n)$:

- ▶ C has a joint density $f(x)$ on $(0, \infty)^n$.
- ▶ f is asymptotically homogeneous of order $\alpha \geq 0$ near the origin, i.e.,

$$f(\lambda x) \sim \lambda^\alpha h(x), \quad \lambda \rightarrow 0^+.$$

(And some technical conditions.)

Theorem (Baxendale and W. (2018))

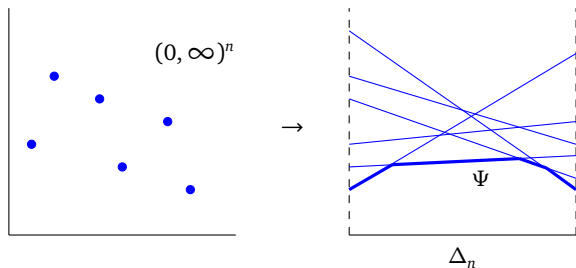
As $K \rightarrow \infty$ (with $\lambda_K \equiv \infty$) the law of the random concave function $\Psi_K = K^{1/(n+\alpha)} \min(\ell_1, \dots, \ell_K)$ converges weakly to a measure ν supported on

$$\mathcal{C}_+ = \{\psi : \Delta_n \rightarrow (0, \infty) \text{ concave}\}.$$

Realization by Poisson point process

Consider a point process $N = N(\omega)$ which is a random set in $(0, \infty)^n$. It induces

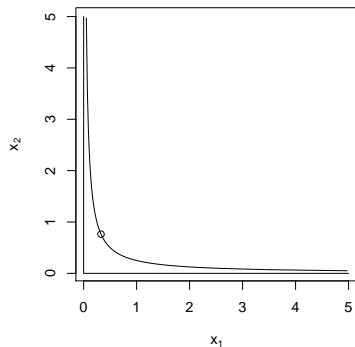
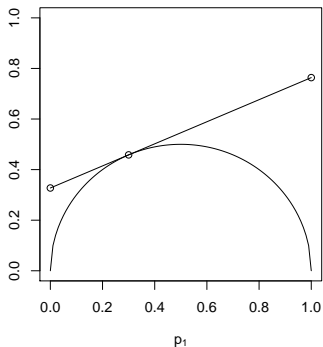
$$\Psi(p) = \inf_{x \in N(\omega)} x \cdot p.$$



Theorem (Baxendale and W. (2018))

Let N be a Poisson point process on $(0, \infty)^n$ with rate measure $m(A) = \int_A h(x) dx$. Then the law of $\Psi(p) = \inf_{x \in N} x \cdot p$ is the limiting measure ν in the previous theorem.

Convex duality



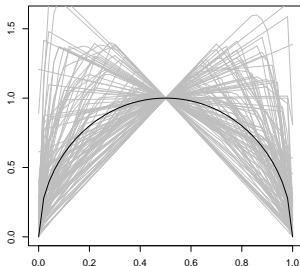
This gives a map $\psi \in \mathcal{C} \mapsto \hat{R}(\psi)$ (lower left part).

In fact, if $\Psi \sim \nu$, then

$$\mathbb{P}(\Psi \geq \psi) = \exp(-m(\hat{R}(\psi))) = \exp\left(-\int_{\hat{R}(\psi)} h(x)dx\right).$$

When intensity measure = $\gamma \times \text{Lebesgue}$

- ▶ Let $\Psi \sim \nu$, when $h = \text{const} = \gamma$. Then $\mathbb{E}\Psi(p) \propto (p_1 \cdots p_n)^{1/n}$. It generates multiplicatively the equal-weighted portfolio.



- ▶ Let π be portfolio generated multiplicatively by $\Phi \sim \nu$. Then for $p \in \Delta_n$, $\pi(p)$ is uniformly distributed in Δ_n .
- ▶ Let $\Psi \sim \nu$. For $\psi \in \mathcal{C} \cap C^2$, $\psi|_{\partial\Delta_n} \equiv 0$, we have

$$\mathbb{P}(\Psi \geq \psi) = \exp\left(\frac{-\gamma}{n} \int_{\Delta_n} \psi(p) \det(-D^2\psi(p)) dp\right).$$

Open problem

- ▶ Let $V_\psi(t)$ be value of portfolio generated by ψ .
- ▶ Let $V^*(t) = \sup_\psi V_\psi(t)$.
- ▶ Given a prior distribution ν , let

$$\hat{V}(t) = \int V_\psi(t) d\nu(\psi).$$

This is the value of *Cover's universal portfolio* in the context of SPT where portfolios are functionally generated.

- ▶ **Problem:** Prove that

$$\lim_{t \rightarrow \infty} \frac{1}{t} \log \frac{V^*(t)}{\hat{V}(t)} = 0$$

uniformly in the market path and establish convergence rate.