

Stochastic dynamic optimization with multivariate stochastic dominance constraints

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References:

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Composite semi-infinite optimization,
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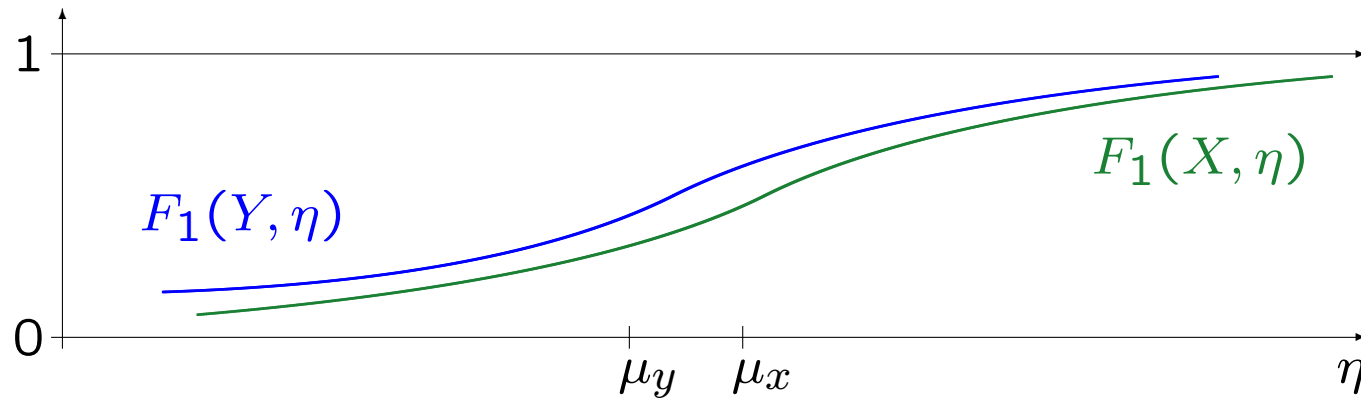
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First Order Dominance

Notation: $F_1(X, \eta) = \mathbb{P}\{X \leq \eta\}$

$$X \succeq_{(1)} Y \Leftrightarrow F_1(X, \eta) \leq F_1(Y, \eta) \quad \text{for all } \eta \in \mathbb{R}$$



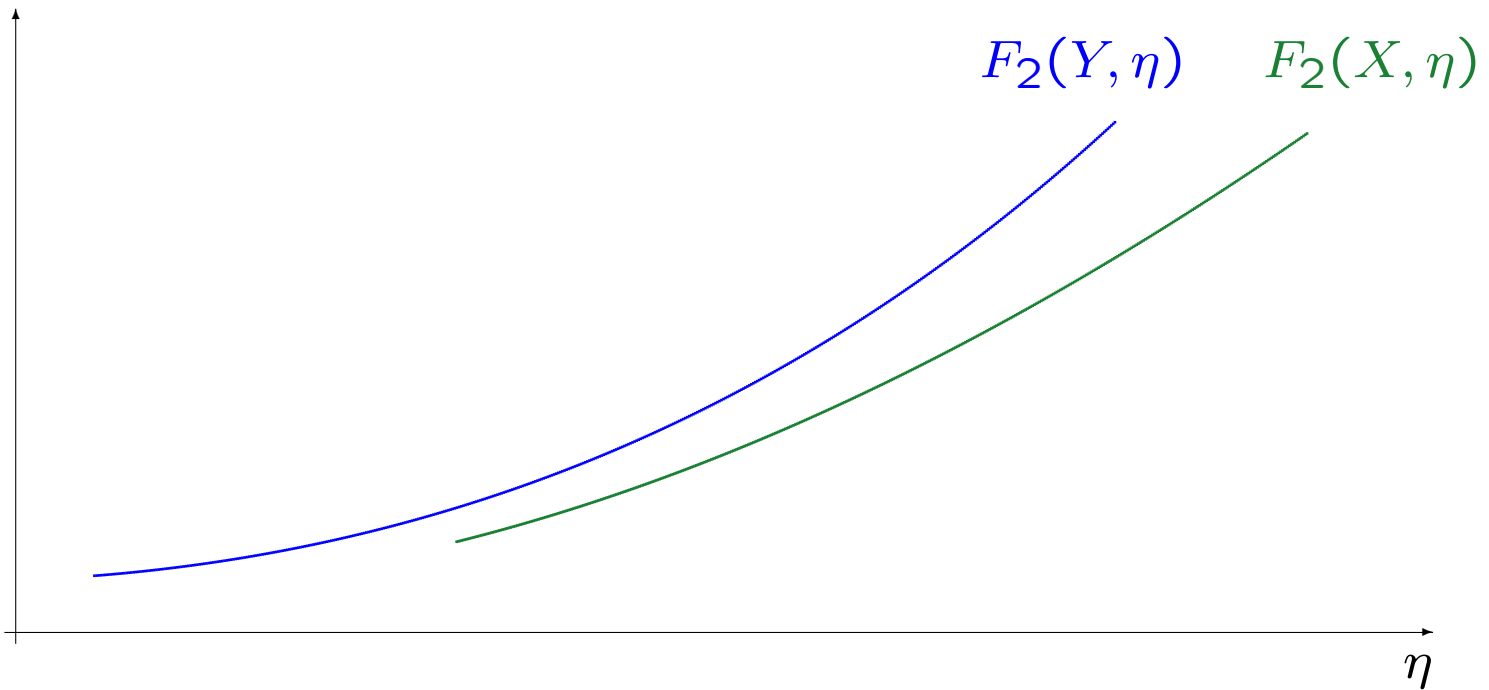
Relation to utility functions:

$$X \succeq_{(1)} Y \Leftrightarrow \mathbb{E}u(X) \geq \mathbb{E}u(Y) \quad \forall \text{ nondecreasing } u(\cdot)$$

Second Order Dominance

Notation: $F_2(X, \eta) = \int_{-\infty}^{\eta} F_1(X, \xi) d\xi = \mathbb{E}(\eta - X)_+$ for $\eta \in \mathbb{R}$

$$X \succeq_{(2)} Y \Leftrightarrow F_2(X, \eta) \leq F_2(Y, \eta) \quad \text{for all } \eta \in \mathbb{R}$$



Relation to utility functions:

$$X \succeq_{(2)} Y \Leftrightarrow \mathbb{E}u(X) \geq \mathbb{E}u(Y) \quad \forall \text{ nondecreasing concave } u(\cdot)$$

Hadar and Russell (1969)

Dominance Constrained Optimization Problems

$$\begin{aligned} & \max \mathbb{E}[H(z)] \\ & \text{subject to } G(z) \succeq_{(2)} Y \\ & z \in Z_0 \end{aligned}$$

Z_0 - convex closed subset of a Banach space \mathcal{Z}

G and H - continuous concave operators from \mathcal{Z} to $\mathcal{L}_1(\Omega, \mathcal{F}, P)$

Y - benchmark outcome in $\mathcal{L}_1(\Omega, \mathcal{F}, P)$

Convenient re-formulation of the dominance constraint:

$$F_2(G(z); \eta) \leq F_2(Y; \eta) \quad \text{for all } \eta \in [a, b]$$

How to formulate the order $\succeq_{(2)}$ if the outcome and the benchmark are multivariate?

$$\begin{aligned} G(z) &= (G_1(z), G_2(z), \dots, G_{T+1}(z)) \\ Y &= (Y_1, Y_2, \dots, Y_{T+1}) \end{aligned}$$

Important case: **dynamic with discrete time**

Multivariate Stochastic Orders

Consider $X = (X_1, \dots, X_{T+1})$ and $Y = (Y_1, \dots, Y_{T+1})$ in $\mathcal{L}_1^{T+1}(\Omega, \mathcal{F}, P)$.

Increasing Convex Order

$$X \succeq_{(\text{icx})} Y \Leftrightarrow \mathbb{E}u(X) \geq \mathbb{E}u(Y) \quad \forall u \in \mathcal{U}$$

Generator \mathcal{U} - all concave nondecreasing functions $u : \mathbb{R}^{T+1} \rightarrow \mathbb{R}$

Hard to treat analytically, the generator is too rich

Coordinate Order

$$X \succeq_{(2)}^{\text{sep}} Y \Leftrightarrow X_t \succeq_{(2)} Y_t, \quad t = 1, \dots, T+1$$

Generator \mathcal{U} - all functions $u(X) = \sum_{t=1}^{T+1} u_t(X_t)$ with concave nondecreasing $u_t : \mathbb{R} \rightarrow \mathbb{R}$. Our earlier analysis covers this case.

Ignores temporal structure and dependency

Our idea is to define the multivariate order via a family of univariate orders

Dynamic Stochastic Dominance

Consider $X = (X_1, \dots, X_{T+1})$ and $Y = (Y_1, \dots, Y_{T+1})$ in $\mathcal{L}_1^{T+1}(\Omega, \mathcal{F}, P)$.

New idea: Consider **linear scalarization** by comparing

$$\langle \varrho, X \rangle = \sum_{t=1}^{T+1} \varrho_t X_t \quad \text{and} \quad \langle \varrho, Y \rangle = \sum_{t=1}^{T+1} \varrho_t Y_t$$

for all

$$\varrho \in \mathcal{D} \subseteq \left\{ \varrho \in \mathbb{R}^{T+1} : 1 \geq \varrho_1 \geq \varrho_2 \geq \dots \geq \varrho_{T+1} \geq 0 \right\}$$

Discounted Order: $X \succeq_{(2)}^{\text{dis}} Y$ if $\langle \varrho, X \rangle \succeq_{(2)} \langle \varrho, Y \rangle$ for all $\varrho \in \mathcal{D}$

The order $\succeq_{(2)}^{\text{dis}}$ neither implies nor is implied by the coordinate order

Special case:

Finite set \mathcal{D} consisting of $T + 1$ elements ϱ^k , $k = 1, \dots, T + 1$, with

$$\varrho_t^k = \begin{cases} 1 & \text{if } t \leq k, \\ 0 & \text{if } t > k. \end{cases}$$

Then $X \succeq_{(2)}^{\text{dis}} Y$ is equivalent to the dominance of **partial sums**

$$\sum_{t=1}^k X_t \succeq_{(2)} \sum_{t=1}^k Y_t, \quad k = 1, \dots, T + 1.$$

By specifying the set \mathcal{D} we can obtain other interesting cases.

Similar ideas can be employed in the **multivariate** case, without the dynamic structure

The Generator of $\succ_{(2)}^{\text{dis}}$

Notation: $\mathcal{M}_+(S)$ - the space of nonnegative regular measures on $S \subset \mathbb{R}^m$

For every $\lambda \in \mathcal{M}_+(\mathbb{R} \times \mathcal{D})$ we define a concave nondecreasing function $\varphi_\lambda : \mathbb{R}^{T+1} \rightarrow \mathbb{R}$:

$$\varphi_\lambda(x) = - \int_{\mathbb{R} \times \mathcal{D}} \max(0, \eta - \langle \varrho, x \rangle) \lambda(d(\eta, \varrho)).$$

Theorem The class of functions

$$\Phi = \{\varphi_\lambda : \lambda \in \mathcal{M}_+(\mathbb{R} \times \mathcal{D})\}$$

is a **generator** of the order $\succ_{(2)}^{\text{dis}}$:

$$X \succ_{(2)}^{\text{dis}} Y \quad \Leftrightarrow \quad \mathbb{E}[\varphi(X)] \geq \mathbb{E}[\varphi(Y)] \quad \text{for all } \varphi \in \Phi$$

Obtained by integrating w.r.t. marginal and conditional measures of λ

Dynamic Optimization Problem

$$\begin{aligned} \max \quad & \sum_{t=1}^T \mathbb{E}G_t(s_t, v_t) + \mathbb{E}G_{T+1}(s_{T+1}) \\ \text{s.t.} \quad & s_{t+1} = A_t s_t + B_t v_t + e_t, \quad t = 1, \dots, T, \\ & (G_1(s_1, v_1), \dots, G_1(s_T, v_T), G_{T+1}(s_{T+1})) \succeq_{\binom{\text{dis}}{2}} (Y_1, \dots, Y_T, Y_{T+1}) \\ & v_t \in V_t \text{ a.s.}, \quad t = 1, \dots, T. \end{aligned}$$

Decision vector z : controls v_1, \dots, v_T , and trajectories s_2, \dots, s_{T+1}

\mathcal{F}_t - σ -algebra generated by $\{e_1, \dots, e_{t-1}\}$

Space of controls: $\mathcal{V} = \mathcal{L}_1^{n_v}(\Omega, \mathcal{F}_1, P) \times \dots \times \mathcal{L}_1^{n_v}(\Omega, \mathcal{F}_T, P)$.

Space of state trajectories: $\mathcal{S} = \mathcal{L}_1^{n_s}(\Omega, \mathcal{F}_2, P) \times \dots \times \mathcal{L}_1^{n_s}(\Omega, \mathcal{F}_{T+1}, P)$.

For technical reasons we restrict the range of $\eta \in \mathbb{R}$ to an interval $[a, b]$:

$$F_2(\langle \varrho, G(s, v) \rangle; \eta) \leq F_2(\langle \varrho, Y \rangle; \eta) \quad \text{for all } \varrho \in \mathcal{D} \text{ and all } \eta \in [a, b]$$

The Partial Lagrangian

Define the class of functions

$$\Phi([a, b], \mathcal{D}) = \{\varphi_\lambda : \lambda \in \mathcal{M}_+([a, b] \times \mathcal{D})\},$$

$$\varphi_\lambda(x) = - \int_{[a, b] \times \mathcal{D}} \max(0, \eta - \langle \varrho, x \rangle) \lambda(d(\eta, \varrho))$$

The set $\Phi([a, b], \mathcal{D})$ is a **generator** of the interval dominance order

We introduce the **partial “Lagrangian”** $L : \mathcal{S} \times \mathcal{V} \times \Phi([a, b], \mathcal{D}) \rightarrow \mathbb{R}$:

$$L(s, v, \varphi) = \mathbb{E} \left[\sum_{t=1}^T G_t(s_t, v_t) + G_{T+1}(s_{T+1}) \right. \\ \left. + \left(\varphi(G_1(s_1, v_1), \dots, G_T(s_T, v_T), G_{T+1}(s_{T+1})) - \varphi(Y_1, \dots, Y_T, Y_{T+1}) \right) \right]$$

Non-decomposable w.r.t. $t = 1, \dots, T + 1$

Auxiliary Control Problem

$$\begin{aligned} \max \mathbb{E} & \left[\sum_{t=1}^T G_t(s_t, v_t) + G_{T+1}(s_{T+1}) \right. \\ & \left. + \left(\varphi(G_1(s_1, v_1), \dots, G_T(s_T, v_T), G_{T+1}(s_{T+1})) - \varphi(Y_1, \dots, Y_T, Y_{T+1}) \right) \right] \\ \text{s.t. } & s_{t+1} = A_t s_t + B_t v_t + e_t, \quad t = 1, \dots, T \\ & v_t \in V_t \quad \text{a.s.}, \quad t = 1, \dots, T \end{aligned}$$

Uniform Dominance Condition: \exists a feasible pair (\tilde{s}, \tilde{v}) such that

$$\inf_{(\eta, \varrho) \in [a, b] \times \mathcal{D}} \left\{ F_2(\langle \varrho, Y \rangle; \eta) - F_2(\langle \varrho, G(\tilde{s}, \tilde{v}) \rangle; \eta) \right\} > 0.$$

Theorem

Assume the uniform dominance condition. If (\hat{s}, \hat{v}) is an optimal solution of the original problem then there exist $\hat{\varphi} \in \Phi([a, b], \mathcal{D})$ such that (\hat{s}, \hat{v}) is an optimal solution of the auxiliary problem with $\varphi = \hat{\varphi}$, and

$$\mathbb{E}[\hat{\varphi}(G(\hat{s}, \hat{v}))] = \mathbb{E}[\hat{\varphi}(Y)].$$

Conversely, if for some $\hat{\varphi} \in \Phi([a, b], \mathcal{D})$ an optimal solution (\hat{s}, \hat{v}) of the auxiliary problem satisfies the dominance constraint, then it is optimal.

Ideas of the Derivation

1. Write the dominance constraint as an **operator constraint** in $\mathcal{C}([a, b], \mathcal{D})$:

$$\mathbb{E}(\eta - \langle \varrho, G(s, v) \rangle)_+ \leq \mathbb{E}(\eta - \langle \varrho, Y \rangle)_+ \quad \text{for all } \varrho \in \mathcal{D} \text{ and all } \eta \in [a, b]$$

2. Associate with it a **Lagrange multiplier**

- nonnegative measure $\lambda \in \mathcal{M}_+([a, b] \times \mathcal{D})$

3. Change the order of integration in the Lagrangian term

$$\begin{aligned} \int_{[a, b] \times \mathcal{D}} \mathbb{E}(\eta - \langle \varrho, X \rangle)_+ \lambda(d(\eta, \varrho)) &= \int_{\Omega} \int_{[a, b] \times \mathcal{D}} (\eta - \langle \varrho, X(\omega) \rangle)_+ \lambda(d(\eta, \varrho)) P(d\omega) \\ &= -\mathbb{E}\varphi(X) \end{aligned}$$

with

$$\varphi(x) = - \int_{[a, b] \times \mathcal{D}} (\eta - \langle \varrho, x \rangle)_+ \lambda(d(\eta, \varrho))$$

Implied Random Discount

$$\begin{aligned}
 & \max \sum_{t=1}^T \mathbb{E}G_t(s_t, v_t) + \mathbb{E}G_{T+1}(s_{T+1}) \\
 & \text{s.t. } s_{t+1} = A_t s_t + B_t v_t + e_t, \quad t = 1, \dots, T, \\
 & \quad (G_1(s_1, v_1), \dots, G_1(s_T, v_T), G_{T+1}(s_{T+1})) \succeq_{(2)}^{\text{dis}} (Y_1, \dots, Y_T, Y_{T+1}) \\
 & \quad v_t \in V_t \text{ a.s.}, \quad t = 1, \dots, T.
 \end{aligned}$$

Theorem Assume the uniform dominance condition. If (\hat{s}, \hat{v}) is an optimal solution of the original control problem then there exist $\xi_t \in \mathcal{L}_\infty(\Omega, \mathcal{F}_t, P)$, $t = 1, \dots, T + 1$, with

$$\xi_1 \geq \xi_2 \geq \dots \geq \xi_T \geq \xi_{T+1} \geq 0 \quad \text{a.s.}$$

such that (\hat{s}, \hat{v}) is an optimal solution of the control problem

$$\begin{aligned}
 & \max \sum_{t=1}^T \mathbb{E}(1 + \xi_t)G_t(s_t, v_t) + \mathbb{E}(1 + \xi_{T+1})G_{T+1}(s_{T+1}) \\
 & \text{s.t. } s_{t+1} = A_t s_t + B_t v_t + e_t, \quad t = 1, \dots, T, \\
 & \quad v_t \in V_t \text{ a.s.}, \quad t = 1, \dots, T.
 \end{aligned}$$

Maximum Principle

Theorem Assume the uniform dominance condition. If (\hat{s}, \hat{v}) is an optimal solution of the original control problem, then there exist:

discount factors $\xi_t \in \mathcal{L}_\infty(\Omega, \mathcal{F}_t, P)$, $t = 1, \dots, T + 1$,

$$\xi_1 \geq \xi_2 \geq \dots \geq \xi_T \geq \xi_{T+1} \geq 0 \quad \text{a.s.}$$

subgradients $(\sigma_t^s, \sigma_t^v) \in \mathcal{L}_q^{n_s}(\Omega, \mathcal{F}_t, P) \times \mathcal{L}_q^{n_v}(\Omega, \mathcal{F}_t, P)$ satisfying

$$(\sigma_t^s(\omega), \sigma_t^v(\omega)) \in \partial g_t(\hat{s}_t(\omega), \hat{v}_t(\omega))$$

$$\sigma_{T+1}^s(\omega) \in \partial g_{T+1}(\hat{s}_{T+1}(\omega))$$

and dual variables $\hat{y}_t \in \mathcal{L}_q^{n_s}(\Omega, \mathcal{F}_{t+1}, P)$, $t = 1, \dots, T$, satisfying the adjoint equations

$$y_T = (1 + \xi_{T+1})\sigma_{T+1}^s$$

$$y_{t-1} = A_t' \mathbb{E}[y_t | \mathcal{F}_t] + (1 + \xi_t)\sigma_t^s, \quad t = T, \dots, 2$$

such that for almost all $\omega \in \Omega$ the control $\hat{v}_t(\omega)$ is a solution of

$$\max_{c \in V_t} \langle (1 + \xi_t(\omega))\sigma_t^v(\omega) + B_t'(\omega)\mathbb{E}[y_t | \mathcal{F}_t](\omega), c \rangle$$