

Dynamic Efficiency With Private Information

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Motivation

- Contribute to Macro models with **frictions** (output **diversion**) and *many heterogeneous* agents (**different participation constraint**).
- Rely on continuous-time contract theory (Sannikov 2008, Cvitanic-Possamai-Touzi 2018) and mean-field techniques to study with many agents.
- Mathematically speaking, we are dealing with a mean-field control problem with **Value-Dependent Constraints**.
- *Road Map*:
 - Focus on the *continuum limit* ($N \rightarrow \infty$) for tractability.
 - Characterize optimal allocations via a functional HJB equation and a *fixed point argument*.
 - Effect of the Common noise

Endogenous growth model: AK model

Agents: N infinitely lived agents with discount rate ρ and same utility u .

- Maximize continuation utility: $\mathbb{E} \left[\int_0^\infty \rho e^{-\rho t} u(c_t^i) dt \right]$
- Participation constraint: $\mathbb{E} \left[\int_0^\infty \rho e^{-\rho t} u(c_t^i) dt \right] \geq \omega^i$, ω^i exogenous.

Output of agent i $dY_t^i = \frac{k_t^i}{N} [\mu dt + \sigma dZ_t^i + \sigma_A dZ_t^A]$

Controls for the Social planner

- 1 c_t^i consumption for agent i .
- 2 k_t^i amount of capital managed by agent i .
- 3 μ, σ and σ_A exogenous.

Capital and the Resource Constraint

- **Average Capital:** $K_t := \frac{1}{N} \sum_{i=1}^N k_t^i$
- **Evolution of Capital:**

$$dK_t = (\mu K_t - \bar{c}_t) dt + \sigma_A K_t dZ_t^A + \frac{\sigma}{N} \sum_{i=1}^N k_t^i dZ_t^i, \quad \bar{c}_t = \frac{1}{N} \sum_{i=1}^N c_t^i$$

- If we have $\mathbb{E}(e^{-\mu t} K_t) \rightarrow 0$ when $t \rightarrow \infty$, then

$$K = \mathbb{E} \left[\int_0^{\infty} e^{-\mu t} \bar{c}_t dt \right].$$

The Promise-Keeping Condition

The planner tracks the agent's expected continuation utility

$$\omega_t^i = \mathbb{E}_t \left[\int_t^\infty \rho e^{-\rho t} u(c_t^i) dt \right]$$

By the representation theorem of Brownian martingales, the "promise keeping" condition

$$d\omega_t^i = \rho [\omega_t^i - u(c_t^i)] dt + \frac{\sigma}{N} \sum_{j=1}^N \beta_t^{ij} dZ_t^j + \sigma_A \beta_t^{A,i} dZ_t^A$$

Observations: At this stage, all β are not **independent** control variables but depend on the consumption stream chosen by the social planner.

The Principal-Agent Problem as a Stackelberg game

- Let us assume $\sigma_A = 0$.
- **The Social Planner (Principal):** Signs bilateral contracts with firms (c_t^i, k_t^i) and does not consume.
- **Each individual Firm (Agent):** Reports output $d\hat{Y}_t = dY_t - \delta_t dt$ (output diversion).

$$\begin{aligned}d\hat{Y}_t^i &= dY_t^i - \delta_t dt \\ &= \mu k_t^i dt + \sigma k_t^i (dZ_t^i - \frac{\delta_t}{\sigma k_t^i} dt) \\ &= \mu k_t^i dt + \sigma k_t^i d\hat{Z}_t^i,\end{aligned}$$

- **Firm's objective:**

$$\omega_0^i = \sup_{\delta} \mathbb{E} \left[\int_0^{\infty} e^{-\rho s} u(c_s^i + \delta_s) ds \right]$$

Martingale Optimality Principle

Step 1: Construct a process R_t^δ to apply the martingale optimality principle:

$$R_t^\delta = e^{-\rho t} \omega_t^\beta + \int_0^t e^{-\rho s} u(c_s + \delta_s) ds$$

$$d\omega_t^\beta = \rho(\omega_t^\beta - u(c_t^i)) dt + \beta_t d\hat{Z}_t^i, \text{ with } \beta_t \geq \rho \sigma k_t^i u'(c_t^i).$$

Step 2: The drift of dR_t^δ is proportional to:

$$\rho(u(c_t + \delta_t) - u(c_t)) - \frac{\beta_t \delta_t}{\sigma k_t}$$

Step 3: Using the concavity of u :

$$u(c_t^i + \delta_t) - u(c_t^i) - \frac{\beta_t \delta_t}{\rho \sigma k_t^i} \leq \delta \left(u'(c_t^i) - \frac{\beta_t}{\rho \sigma k_t^i} \right) \leq 0$$

$(R_t^\delta)_t$ is a **supermartingale** for $\delta > 0$ and a **martingale** for $\delta = 0$.

To sum up

The social planner has to select capital and consumption flow such that

- $\beta_t^{ii} \geq \rho \sigma k_t^i u'(c_t^i)$.
- $\omega_0^i \geq \omega^i$.

A simple benchmark: $N=1$

With one agent, how much capital is needed to obtain utility ω ?

- Inverse of Merton problem

$$V(k) = \sup_c \mathbb{E} \left[\int_0^\infty \rho e^{-\rho t} u(c_t) dt \right]$$

under

$$dk_t = (\mu k_t - c_t) dt + \sigma k_t dZ_t$$

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- When $u(c) = \log c$ we have a closed form solution:

$$k(\omega) = \exp(\omega - a_1), c(\omega) = \rho k(\omega)$$

where

$$a_1 = \log \rho + \frac{\mu - \rho - \frac{\sigma^2}{2}}{\rho}.$$

Characterizing Optimal Allocations when $N \rightarrow \infty$

- Agents are indexed by their participation constraint ω , where \mathbb{P} represents the initial distribution across the population.
- We assume **Markov** controls, i.e. $c(\mathbb{P}, \omega)$ and $k(\mathbb{P}, \omega)$.
- Control approach (Sannikov 2008, CPT 2018), starting from a random variable ω_0 with distribution \mathbb{P} , the social planner controls the continuation utility process

$$d\omega_t = \rho(\omega_t - u(c(\mathbb{P}_t, \omega_t))) dt + \sigma \beta_t dZ_t.$$

- β_t is now a second control that must satisfy the IC condition $\beta_t \geq \rho k u'(c)$.

Social Planner Problem

$$K(\mathbb{P}) = \min_{\{k, c, \beta\}} \int_0^{\infty} e^{-\mu t} \left[\int_{\mathbb{R}} c(\mathbb{P}_t, \omega) d\mathbb{P}_t(\omega) \right] dt \quad (1)$$

subject to the following constraints for all t :

$$d\omega_t = \rho [\omega_t - u(c_t)] dt + \sigma \beta_t dZ_t \quad (\text{Promise Keeping})$$

$$\beta_t \geq \rho k_t u'(c_t) \quad (\text{Incentive Compatibility})$$

$$K(\mathbb{P}_t) = \int_{\mathbb{R}} k(\mathbb{P}_t, \omega) d\mathbb{P}_t(\omega) \quad (\text{Capital Allocation})$$

$$\mathbb{P}_0 = \mathbb{P} \quad (\text{Initial Condition}).$$

Differential Calculus in $\mathcal{P}_2(\mathbb{R})$ and Chain Rule

Definition: First Variation and L-Differentiability

First variation exists at \mathbb{P} if there is a continuous $\nabla F[\mathbb{P}] : \mathbb{R} \rightarrow \mathbb{R}$ such that:

$$\lim_{\varepsilon \rightarrow 0} \frac{F((1 - \varepsilon)\mathbb{P} + \varepsilon\mathbb{Q}) - F(\mathbb{P})}{\varepsilon} = \int_{\mathbb{R}} \nabla F[\mathbb{P}](\omega) d(\mathbb{Q} - \mathbb{P})(\omega)$$

- F is L-differentiable if the first variation $\nabla F[\mathbb{P}](\cdot)$ is twice differentiable in \mathbb{R} . Notation: $\partial_{\omega} \nabla F[\mathbb{P}]$ and $\partial_{\omega\omega} \nabla F[\mathbb{P}]$.

Itô's Lemma (Carmona & Delarue, Vol I,II)

$$\begin{aligned} F(\mathbb{P}_t) &= F(\mathbb{P}_0) + \int_0^t \mathbb{E} [\partial_{\omega} \nabla F(\mathbb{P}_s)(\omega_s) \rho(\omega_s - u(c(\mathbb{P}_s, \omega_s)))] ds \\ &\quad + \frac{1}{2} \int_0^t \mathbb{E} [\partial_{\omega\omega}^2 \nabla F(\mathbb{P}_s)(\omega_s) \sigma^2 \beta^2(\mathbb{P}_s, \omega_s)] ds \end{aligned}$$

Constrained Functional HJB

Taking $\lambda(\mathbb{P})$ as the Lagrange multiplier for $\int k(\mathbb{P}, \omega) d\mathbb{P} = K(\mathbb{P})$:

$$(\mu - \lambda(\mathbb{P}))K(\mathbb{P}) = \inf_{c, k} \int_{\mathbb{R}} \left[c + \rho(\omega - u(c))\partial_{\omega}\nabla K + \frac{(\rho\sigma k u'(c))^2}{2}\partial_{\omega\omega}^2\nabla K - \lambda(\mathbb{P})k \right] d\mathbb{P}$$

- **Binding IC:** Assuming $\partial_{\omega\omega}^2\nabla K > 0$ (convexity), the volatility constraint binds: $\beta = \rho k u'(c)$.
- **Structure:** This is a second-order PDE in the space of measures $\mathcal{P}_2(\mathbb{R})$. **Existence, Uniqueness**

Verification Theorem

Existence and Uniqueness

Let $\lambda(\cdot)$ be a given Lagrange multiplier and let $K^\lambda(\mathbb{P})$ be a L-differentiable function such that:

- 1 **HJB & Transversality:** K^λ solves the constrained HJB equation and satisfies:

$$\lim_{t \rightarrow \infty} e^{-\mu t} K^\lambda(\mathbb{P}_t) = 0$$

- 2 **Admissibility:** There exists an optimal control pair $(c_\lambda^*, k_\lambda^*)$ attaining the infimum.
- 3 **Consistency:** There is some λ^* such that:

$$\int k_{\lambda^*}^*(\omega) d\mathbb{P}_t(\omega) = K^{\lambda^*}(\mathbb{P}_t), \quad \forall t \geq 0$$

Then, $K = K^{\lambda^*}$

Explicit Solution: The CRRA Utility Case

Assume a Constant Relative Risk Aversion (CRRA) utility function:

$$u(c) = \frac{c^{1-\alpha}}{1-\alpha}, \quad \alpha > 1$$

Main result

Under CRRA utility, the optimal controls (k^*, c^*) are **independent** of the probability measure \mathbb{P} . They are parameterized by two constants (a^*, y^*) :

- **Capital Allocation:** $k(\omega) = a^* u^{-1}(\omega)$
- **Consumption:** $c(\omega) = \frac{\rho k(\omega)}{y^*}$

Fixed Point Equation for a^*

The constant a^* is the unique solution to the scalar optimization:

$$\frac{\mu}{\rho} = \inf_y \left[\frac{1}{y} + \frac{\left(\frac{\rho a^*}{y}\right)^{1-\alpha} - 1}{\alpha - 1} + \frac{\alpha \sigma^2}{2\rho} (\rho a^*)^{2-2\alpha} y^{2\alpha} \right]$$

Sketch of the Proof I

A standard Guess and Verify approach and a fixed point argument.

- Ansatz: $K(\mathbb{P}) = \int g(\omega) d\mathbb{P}(\omega)$ with $g(\omega) = a^* u^{-1}(\omega)$.
- The first variation ∇K simply coincides with $g(\cdot)$.
- The functional equation must hold for *all* probability measures \mathbb{P}
- **Pointwise Identity:** Since the integrand is independent of \mathbb{P} , the following relation must hold for every state ω .

Individual HJB Equation

For any given λ , we seek for a function $g_\lambda(\omega)$ that satisfies:

$$(\mu - \lambda)g(\omega) = \inf_{c,k} \left[c + \rho(\omega - u(c))g'(\omega) + \frac{(\rho\sigma k u'(c))^2}{2} g''(\omega) - \lambda k \right]$$

First-Order Conditions

We minimize over c and k the expression:

$$c + \rho(\omega - u(c))g'(\omega) + \frac{(\rho\sigma ku'(c))^2}{2}g''(\omega) - \lambda k$$

FOC with respect to c :

$$1 - \rho u'(c) g''(\omega) [(\omega - u(c)) - \rho\sigma^2 k^2 u''(c)] = 0$$

FOC with respect to k :

$$k^* = \frac{\lambda}{(\rho\sigma)^2 (u'(c))^2 g''(\omega)}$$

Guess and Change of Variables

With $u(c) = \frac{c^{1-\alpha}}{1-\alpha}$, so that $u^{-1}(\omega) = ((1-\alpha)\omega)^{\frac{1}{1-\alpha}}$.

Guess: $g(\omega) = a^* u^{-1}(\omega) \equiv a^* \phi(\omega)$, with derivatives:

$$\phi'(\omega) = \frac{\phi(\omega)}{(1-\alpha)\omega}, \quad \phi''(\omega) = \frac{\alpha \phi(\omega)}{(1-\alpha)^2 \omega^2}$$

Change of controls:

$$c = \gamma u^{-1}(\omega), \quad k = a u^{-1}(\omega)$$

which gives:

$$u(c) = \gamma^{1-\alpha} \omega \implies \omega - u(c) = (1 - \gamma^{1-\alpha}) \omega$$

$$u'(c) = \gamma^{-\alpha} \phi(\omega)^{-\alpha}$$

Static Optimisation Problem

Substituting into the HJB, all ω -dependence cancels. The problem reduces to:

$$(\mu - \lambda) a^* = \min_{a, \gamma} \left\{ \gamma + \frac{\rho a^*}{1 - \alpha} (1 - \gamma^{1-\alpha}) + \frac{\alpha (\rho \sigma)^2 a^2 \gamma^{-2\alpha}}{2} a^* - \lambda a \right\}$$

This is a **static problem** in (a, γ) :

- a^* appears *linearly* \Rightarrow FOCs in a and γ pin down the optimal controls independently.
- The full equation then determines a^* as a **fixed point**.

First-Order Conditions for a and γ

FOC with respect to a :

$$\alpha (\rho\sigma)^2 a \gamma^{-2\alpha} a^* = \lambda \implies a = \frac{\lambda}{\alpha (\rho\sigma)^2 \gamma^{-2\alpha} a^*}$$

There is a unique λ^* such that $a = a^*$. Define $\gamma = \frac{\rho a^*}{y}$ and divide by ρa^* to get

$$\frac{\mu}{\rho} = \inf_y \left\{ \frac{1}{y} + \frac{\left(\frac{\rho a^*}{y}\right)^{1-\alpha} - 1}{\alpha - 1} + \frac{\alpha \sigma^2}{2\rho} (\rho a^*)^{2-2\alpha} y^{2\alpha} \right\}.$$

Let us denote $I(a^*)$ the right-hand side. There is exactly one value a^* such that $I(a^*) = \frac{\mu}{\rho}$. Finally, y^* is the unique value of y for which the inf is reached for this particular value of a^* .

Social Planner Problem with common noise

The planner minimizes the expected discounted cost over a flow of random measures \mathbb{P}_t :

$$K(\mathbb{P}) = \min_{\{k, c, \beta, \beta^A\}} \mathbb{E} \left[\int_0^\infty e^{-\mu t} \left(\int_{\mathbb{R}} c_t(\omega) d\mathbb{P}_t(\omega) \right) dt \right] \quad (2)$$

where:

- \mathbb{P}_t is the conditional law of ω_t given the common noise \mathcal{F}_t^A .
- \mathbb{E} is the expectation over the realizations of the common noise Z^A .

- **Promise keeping**

$$d\omega_t = \rho [\omega_t - u(c_t)] dt + \sigma \beta_t dZ_t + \sigma^A \beta_t^A dZ_t^A$$

- **Incentive Compatibility:**

$$\beta_t \geq \rho k_t u'(c_t) \quad (3)$$

- **Stochastic Capital Allocation:**

$$K(\mathbb{P}_t) = \int_{\mathbb{R}} k_t(\omega) d\mathbb{P}_t(\omega) \quad (4)$$

- **Initial Condition:** $\mathbb{P}_0 = \mathbb{P}$.

The HJB Equation (Master Equation)

Without the capital allocation constraint, the value function $K(\mathbb{P})$ would satisfy the second-order HJB equation in the Wasserstein space:

$$\begin{aligned} \mu K(\mathbb{P}) = & \inf_{\{c, \beta, \beta^A\}} \left\{ \int_{\mathbb{R}} c(\omega) d\mathbb{P}(\omega) + \int_{\mathbb{R}} \rho(\omega - u(c)) \partial_{\omega} \nabla K d\mathbb{P} \right. \\ & + \frac{1}{2} \int_{\mathbb{R}} (\sigma^2 \beta^2 + (\sigma^A)^2 (\beta^A)^2) \partial_{\omega\omega}^2 \nabla K d\mathbb{P} \\ & \left. + \frac{(\sigma^A)^2}{2} \int_{\mathbb{R}^2} \beta^A(\omega) \beta^A(\omega') \partial_{\omega\omega'}^2 \nabla^2 K d\mathbb{P} d\mathbb{P}' \right\} \end{aligned}$$

The Primal Problem

Assuming (IC) binding, we consider the following saddle-point problem over the conditional law \mathbb{P} of ω given the common noise:

$$\mu K(\mathbb{P}) = \sup_{\lambda \in \mathbb{R}} \left\{ \inf_{c(\cdot), k(\cdot), \beta^A} \mathcal{L}(c, k, \beta^A, \lambda; \mathbb{P}) \right\}$$

Where the Lagrangian \mathcal{L} is defined as:

$$\begin{aligned} \mathcal{L} = & \int_{\mathbb{R}} \left[c(\omega) + \rho(\omega - u(c(\omega))) \partial_{\omega} \nabla K(\mathbb{P})(\omega) + \frac{(\sigma^2 + (\beta^A)^2)(\rho k(\omega) u'(c(\omega)))^2}{2} \partial_{\omega\omega}^2 \nabla K(\mathbb{P})(\omega) + \lambda k(\omega) \right] d\mathbb{P}(\omega) \\ & + \frac{(\beta^A)^2}{2} \int_{\mathbb{R}} \int_{\mathbb{R}} \partial_{\omega\omega'}^2 \nabla^2 K(\mathbb{P})(\omega, \omega') d\mathbb{P}(\omega) d\mathbb{P}(\omega') - \lambda K(\mathbb{P}) \end{aligned}$$

Assume $K(\mathbb{P})$ is a linear functional: $K(\mathbb{P}) = \int_{\mathbb{R}} g(\omega) d\mathbb{P}$.

The constraint $K(\mathbb{P}) = \int k(\omega) d\mathbb{P}$ must hold **path-by-path**, not just in expectation, which yields two fundamental conditions:

- Condition 1:

$$\int_{\mathbb{R}} (k(\omega) - g(\omega)) d\mathbb{P}(\omega) = 0$$

- Condition 2:

$$\int_{\mathbb{R}} (\beta^A(\omega)g'(\omega) - g(\omega)) d\mathbb{P}(\omega) = 0$$

Lagrange Multipliers and Point-wise HJB

Since the HJB must hold for any distribution \mathbb{P} , it reduces to a **point-wise ODE** for $g(\omega)$:

$$\begin{aligned} \mu g(\omega) = \sup_{\lambda_1, \lambda_2} \inf_{c, k, \beta^A} & \left\{ c + \rho(\omega - u(c))g'(\omega) + \frac{(\sigma \rho k u'(c))^2 + (\sigma^A \beta^A)^2}{2} g''(\omega) \right. \\ & \left. + \lambda_1(k - g(\omega)) + \lambda_2(\beta^A g'(\omega) - g(\omega)) \right\} \end{aligned}$$

The Case of Log Utilities

Assume log utility $u(c) = \log c$ and the functional guess $g(\omega) = a^* \exp(\omega)$. The solution is $g(\omega) = a^* \exp(\omega)$, where a^* is uniquely defined by:

$$\mu + \rho \log a^* - \frac{\sigma_A^2}{2} = \min_{\gamma} \left[\gamma - \rho \log \gamma + \frac{\sigma^2 \rho^2}{2\gamma^2} \right] \quad (5)$$

- **Consumption:** Agents consume a constant fraction γ^* of their capital: $c_t = \gamma^* k_t$.
- γ^* is the value that realizes the minimum in (5).

- **Aggregate Growth:** The economy grows at a random rate:

$$\frac{dK_t}{K_t} = (\mu - \gamma^*)dt + \sigma_A dZ_t^A$$

- **Individual Reallocation:** Capital evolves relative to the aggregate:

$$\frac{dk_t}{k_t} = \frac{dK_t}{K_t} + \sigma x^* dZ_t, \quad \text{where } x^* = \frac{\rho}{\gamma^*}$$

- **Theoretical Framework:** We established a Mean Field Control approach to characterize **Pareto optima** in economies with a continuum of heterogeneous agents.
- **Novelty:** Successfully integrated **value-dependent constraints** within a functional HJB equation on the **Wasserstein space**.
- Derived explicit solutions for the *AK growth model* by leveraging the properties of CRRA utility functions.
- **Future Work:** Extension to non-CRRA utilities.