

EPIDEMIC CONTROL THROUGH INCENTIVES, LOCKDOWN, AND TESTING

The government's perspective

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2. The model: SIR and principal-agent

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INTRODUCTION

Impact of the epidemic. **Cost for the society:**

- cost of care related to the disease;
 - indirect costs related to the saturation of the hospital system;
 - but also a **cost for the individual**, not necessarily monetary, linked to the infection (QALY/DALY).
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Lockdown and social distancing. While waiting for:

- an effective treatment against the disease induced by the virus;
- a vaccination campaign...

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Impact of social distancing. These restrictions have a cost for individuals: social and economic, but also in terms of (mental) health (no in-person conferences at La Grande Motte).

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- ▶ Without guidelines? **Individual point of view**, with **R. Elie and G. Turinici**.
- ▶ Modeling individual choices towards the epidemic spread, and look for a Nash – Mean-field equilibrium among the population.

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▶ Modeling individual choices towards the epidemic spread, and look for a Nash – Mean-field equilibrium among the population.

▶ Cost of anarchy: Nash equilibrium different from societal optimum...

How can we make the interests of the population converge towards the interests of the society?

▶ **Governmental point of view and incentives**, with **T. Mastrolia, D. Possamai and X. Warin**.

Noteworthy papers (in continuous-time): [Holmström and Milgrom \[5\] \(1987\)](#), [Sannikov \[6\] \(2008\)](#).

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Asymmetry of information.

Moral Hazard: the Agent's **behaviour** is not observable by the Principal.

Output process: Stochastic process X with dynamic, for $t \in [0, T]$:

$$dX_t = \alpha_t dt + \sigma_t dW_t.$$

Effort: given a contract ξ , the Agent controls X through the drift α , in order to maximise the following criteria:

$$\mathbb{E}^{\mathbb{P}^\alpha} \left[U_A(\xi) - \int_0^T c(\alpha_t) dt \right].$$

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$$U_A(\xi) = Y_0 - \int_0^T \mathcal{H}(Z_s) ds + \int_0^T Z_s dX_s, \quad (1)$$

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where

- (i) $Y_0 \in \mathbb{R}$ is chosen by the principal to ensure participation of the agent;
- (ii) Z is chosen by the Principal to encourage effort from the agent;
- (iii) \mathcal{H} is the Agent's Hamiltonian.

THE MODEL: SIR AND PRINCIPAL-AGENT

- ▶ SIR compartment model: during the epidemic, individuals go from “Susceptible” to “Infected” and then “Recovered”.

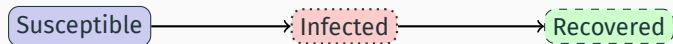


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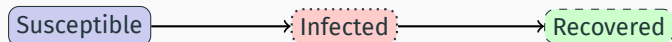


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- Dynamic of a stochastic SIR model:

$$\begin{cases} dS_t = -\beta_t \sqrt{\alpha_t} S_t I_t dt + \sigma \alpha_t S_t I_t dW_t, \\ dl_t = (\beta_t \sqrt{\alpha_t} S_t - \gamma) I_t dt - \sigma \alpha_t S_t I_t dW_t, \\ dR_t = \gamma I_t dt, \end{cases} \quad \text{for } t \in [0, T].$$

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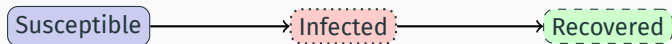


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Initial distribution (S_0, I_0, R_0) at time $t = 0$ known, s.t. $S_0 + I_0 + R_0 = 1$.

- ▶ $S_t + I_t + R_t = 1$ for all $t \geq 0$.
- ▶ 3 main parameters to describe the dynamics of the epidemic: γ , β and α .

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- ▶ **Recovery rate γ .** Exogenous, constant, assumed known, and given by

$$\gamma := \frac{1}{\text{duration of the contagious period}}.$$

- ▶ In absence of an effective treatment against the disease induced by the virus, no possibility to control γ .

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- (i) intrinsic characteristics of the disease;
- (ii) “contact rate” between individuals.

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► The population can make a **costly effort** to reduce their interactions and thus decrease the effective transmission rate of the virus.

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► We assume that β takes values in $[0, \beta_0]$, where β_0 is the “initial” transmission rate, when the population makes no social distancing effort.

► Limiting case: if individuals are fully isolated, then $\beta = 0$ and the virus does not spread.

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► “Uncertainty rate” α . By increasing testing among the population, the government can:

- (i) to reduce the uncertainty related to the spread of the epidemic, i.e. $\sigma \alpha_t$;
- (ii) to isolate those who test positive and thus reduce the effective spread rate, i.e. $\beta_t \sqrt{\alpha_t}$.

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 - (ii) to isolate those who test positive and thus reduce the effective spread rate, i.e. $\beta_t \sqrt{\alpha_t}$.
- We assume that α takes values in $[0, 1]$:
- $\alpha = 1$ means no testing policy;
 - limiting case: if all individuals are tested regularly, the spread of the epidemic is precisely known (without randomness), and isolation of positive individuals stop the propagation, i.e. $\alpha \rightarrow 0$.

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- ▶ **Stackelberg equilibrium.**
 - (i) Given a pair (α, χ) fixed by the government, solve the population's optimisation problem to find the optimal transmission rate $\beta^*(\alpha, \chi)$.

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- ▶ **Stackelberg equilibrium.**
 - (i) Given a pair (α, χ) fixed by the government, solve the population's optimisation problem to find the optimal transmission rate $\beta^*(\alpha, \chi)$.
 - (ii) Given the optimal response of the population to any couple (χ, α) , solve the government's problem to find the optimal couple (χ^*, α^*) .

RESOLUTION

- Given a testing policy α and a tax policy χ , the population's optimal control problem is:

$$V_0^A(\alpha, \chi) := \sup_{\beta \in \mathcal{B}} \mathbb{E} \left[\int_0^T u(t, \beta_t, I_t) dt + U(-\chi) \right], \quad (2)$$

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where $u : [0, T] \times [0, \beta_0] \times \mathbb{R}_+ \rightarrow \mathbb{R}$ and $U : \mathbb{R} \rightarrow \mathbb{R}$ are given by:

$$u(t, b, i) := -\frac{1}{2}(i^3 + (\beta_0 - b)^2) \text{ and } U(x) := \frac{1 - e^{-4x}}{4} + \frac{1}{2}x.$$

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► Interpretations:

- (i) the utility is zero when there is no epidemic, i.e. $I = 0$;
- (ii) the population is scared by a large number of infected – term i^3 in u ;
- (iii) an increase in the tax lowers the utility – U increasing function;
- (iv) a decrease in the level of interaction (below β_0) lowers the utility – u increasing w.r.t. $b \leq \beta_0$.

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Main theoretical result. Given an admissible contract (α, χ) , there exist a unique Y_0 and Z such that:

$$U(-\chi) = Y_0 - \int_0^T \left(\gamma Z_t l_t + u(t, \beta_t^*, l_t) - \beta_t^* \sqrt{\alpha_t} S_t l_t Z_t \right) dt - \int_0^T Z_t dl_t, \quad (3)$$

where β^* is the (unique) optimal contact rate for the population.

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Optimal effort. For all $t \in [0, T]$, $\beta_t^* := b^*(t, S_t, l_t, Z_t, \alpha_t)$ is the maximiser of:

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- ▶ Under some assumptions for existence and smoothness of the inverse of the function U , (3) gives a representation for the tax χ .

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 - the optimal form for the tax χ satisfies (3), i.e., the government only has to choose Z and Y_0 ;
 - the optimal effort of the population is given by $\beta_t^* := b^*(t, S_t, I_t, Z_t, \alpha_t)$ for all $t \in [0, T]$, and thus the epidemic spreads with the transmission rate $\beta^* \sqrt{\alpha}$.
- ▶ It remains to solve the government problem to find the optimal Y_0 , Z and α .

- The government chooses the parameter Y_0 and Z in the tax χ , as well as α to maximise her utility:

$$V_0^P := \sup_{Y_0, Z, \alpha} \mathbb{E} \left[\chi - \int_0^T (c(l_t) + k(\alpha_t)) dt \right], \quad (4)$$

where $k(a) := \kappa_g(a^{-\eta_g} - 1)$, and $c(i) := c_g(i + i^2)$.

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► Interpretations:

- (i) the utility is zero when there is no epidemic, i.e. $I_0 = 0$;
- (ii) an increase in the tax increase the utility;
- (iii) the principal pay a linear cost per infected individual, plus a quadratic cost to represent saturation of health care facilities;
- (iv) testing is costly – k increases when α decreases.

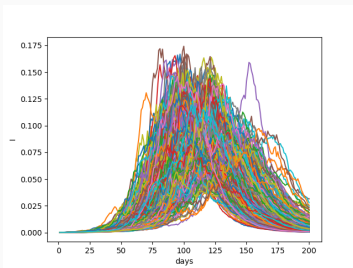
- ▶ In contrast to usual PA problems, the government implements a mandatory tax: the population cannot refuse it.
- ▶ Nevertheless, the government is “benevolent”: she will choose Y_0 in order to ensure a sufficient level \underline{y} of utility for the agent.
- ▶ In particular, \underline{y} is defined by the agent’s utility in the event of an uncontrolled epidemic, i.e., $\beta = \beta_0$, $\alpha = 1$, and $\chi = 0$.

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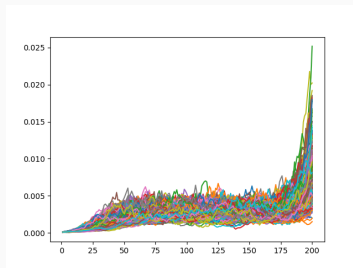
Results.

- ▶ Theoretically: PDE obtained through HJB technics.
- ▶ Numerically: semi-Lagrangian schemes, with truncated higher-order interpolators, as proposed by [Warin \[7\] \(2016\)](#).

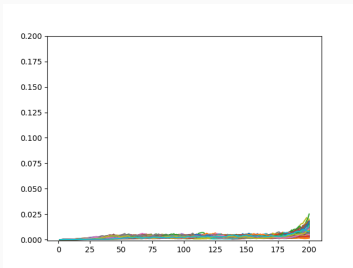
SOME NUMERICAL RESULTS: PROPORTION OF INFECTED INDIVIDUALS



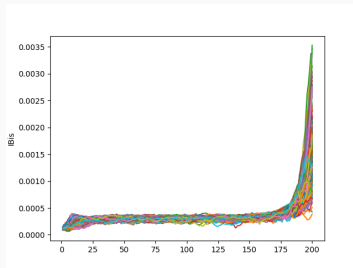
Without governmental intervention



With incentives but without testing



With incentives and testing



Without moral hazard (first-best)

LIMITS AND EXTENSIONS

On the epidemic modelling.

- ▶ Uncertainty about the parameters, especially when the epidemic is of a new kind.
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Extensions.

- Model by [Aurell et al. \[1\] \(2020\)](#) that combines MFG in the population and incentives.
- What about vaccination?

BIBLIOGRAPHY

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