

# Self-isolation under uncertainty

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and Lucie Ménager.

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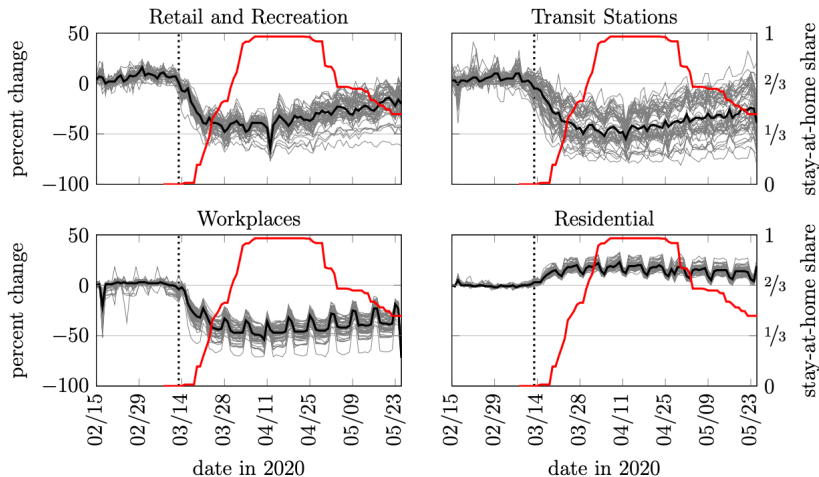
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# Introduction

Epidemiological models have revealed their usefulness to the general public during the COVID-19 epidemic.

- ▶ precious tools to predict the number of cases, to assess the risk of hospital saturation, to declare a lockdown...
- ▶ Difficulties:  
*"These scenarios are made on the basis of incomplete data and uncertain assumptions.[...] changes in behavior are difficult to predict".* (october 2021, Pasteur Institute)

Farboodi & al (2021) show that people responded to the WHO announcement in March 2020:



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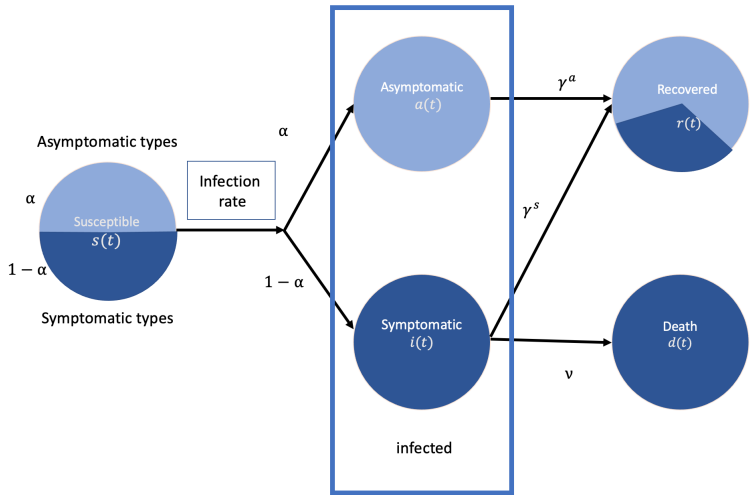


# What we do in this paper

We amend the classical epidemiological model (SIR) :

1. individuals choose a level of social distancing.
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2. individuals are uncertain about the prevalence rate.

The model : Epidemiological part.



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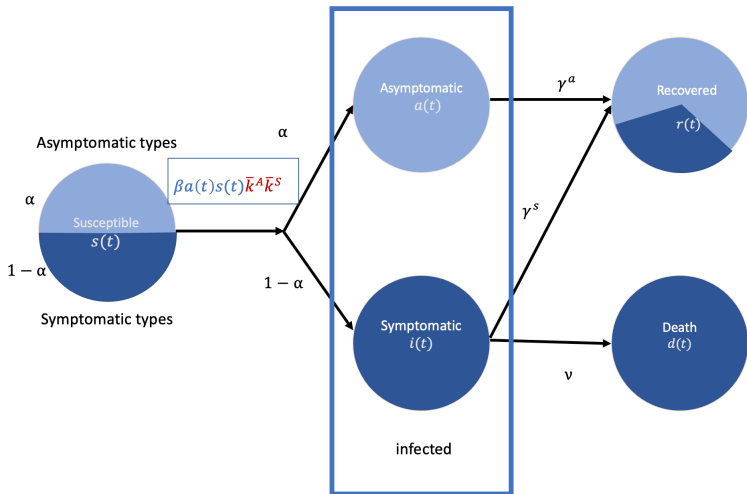
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The average social interaction level of those who are Infected Asymptomatic is

$$\bar{k}^A(t) = \frac{1}{a(t)} \int_{i \in A(t)} k_i di$$



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$$\dot{s}(t) = -\beta s(t)a(t)\bar{k}_S(t)\bar{k}_A(t) \quad (1)$$

$$\dot{a}(t) = \alpha\beta s(t)a(t)\bar{k}_S(t)\bar{k}_A(t) - \gamma^a a(t) \quad (2)$$

$$\dot{i}(t) = (1 - \alpha)\beta s(t)a(t)\bar{k}_S(t)\bar{k}_A(t) - (\gamma^s + \nu)i(t) \quad (3)$$

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A state is a vector  $\omega$  of initial conditions  $C(0)$ .

The model : Economic part.

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2. **The epidemiological state  $\omega$**  :

→  $\mu_i(t; \omega)$  is the belief at  $t$  that the true state is  $\omega$

## What is observed?

Individuals only observe whether they have symptoms or not.



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Bayesian updating gives:

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$$\text{▶ } \mu_i(t; \omega) = \frac{\mu^0(\omega)/(1 - p_i(t|\omega))}{\sum_{\omega'} \mu^0(\omega')/(1 - p_i(t|\omega'))}$$

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## Discounted payoffs

Given a discount factor  $r > 0$  and a state  $\omega$ , player  $i$  discounted payoff, conditional on  $\omega$  is:

$$v_i(k_i, (k_j)_{j \neq i}, \omega) = \int_0^T e^{-rt} e^{-\int_0^t p_i(s, \omega) k_i(s) \beta \bar{k}_A(s) a(s, \omega) ds} \times$$

$$- \underbrace{(1 - k_i(t)) c_{SD}}_{\text{cost of social distancing}} + \underbrace{p_i(t, \omega) k_i(t) \beta \bar{k}_A(t) a(t)}_{\text{proba of developing symptoms}} \underbrace{v_i}_{\text{continuation if symptoms}} dt$$

The **expected** discounted payoff is :

$$\sum_{\omega \in \Omega} \mu^0(\omega) v_i(k_i, (k_j)_{j \neq i}; \omega) \equiv E_{\mu^0}[v_i(k_i, (k_j)_{j \neq i}, \cdot)]$$

s.t.

$$\dot{p}_i(t|\omega) = -(1 - p_i(t|\omega))p_i(t|\omega)k_i(t)\beta\bar{k}_A(t)a(t|\omega), \quad \forall \omega$$

# Equilibrium

# Bellman equation

Let  $p = \{p(\cdot|\omega)\}_\omega$ .



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$$rV(t, p) = V_t(t, p) - c_{SD} + \max_{k \in [0,1]} k \times$$

$$\underbrace{(c_{SD} - \beta \sum_{\omega} \mu(t, \omega) p(t|\omega) \bar{k}_A(t) a(t|\omega) (V(t, p) - v_I + \frac{1 - p(t|\omega)}{\mu(t, \omega)} V_{p(t|\omega)}(t, p)))}_{\text{expected net cost of social activity}}$$

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## Proposition (Best response)

$k_i^*$  is a best response  $\iff k_i^*(t)$

$$\left\{ \begin{array}{ll} = 1 & \text{if } c_{SD} > \text{expected cost of social interaction} \\ \in [0, 1] & \text{if } c_{SD} = \text{expected cost of social interaction,} \\ = 0 & \text{if } c_{SD} < \text{expected cost of social interaction,} \end{array} \right. \quad (6)$$

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## Proposition (The symmetric equilibrium)

Let

$$\hat{k}(t) = \frac{c_{SD}}{\sum_{\omega} \beta a(t|\omega) \mu(t|\omega) p(t|\omega) (\psi(t|\omega) - v_I)}$$

where:

$$\left\{ \begin{array}{l} \dot{p}(t|\omega) = -p(t|\omega)(1 - p(t|\omega))\beta \hat{k}^2(t)a(t|\omega), \\ \dot{s}(t|\omega) = -\beta \hat{k}^2(t)s(t|\omega)a(t|\omega), \\ \dot{a}(t|\omega) = -\alpha \dot{s}(t|\omega) - \gamma^a(\omega)a(t|\omega), \\ \dot{\psi}(t|\omega) - r\psi(t|\omega) = \hat{k}^2(t)\beta a(t|\omega)(\psi(t|\omega) - v_I) + (1 - \hat{k}(t))c_{SD}, \\ \psi(T|\omega) = 0 \end{array} \right. \quad (7)$$

If  $\hat{k}(t) \in (0, 1)$ ,  $\forall t$  then the strategy profile  $k_i(t) = \hat{k}(t)$ ,  $\forall i, t$  is an equilibrium.

# Simulations

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We assume that at date 20, individuals are informed that a pandemic has started.

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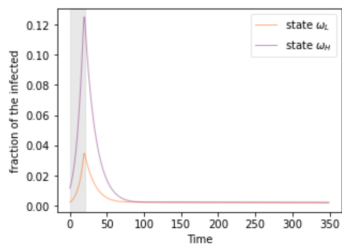
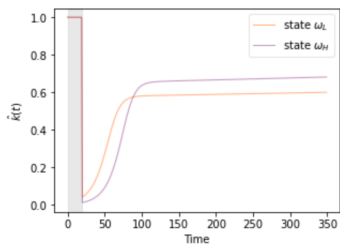
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- ▶ State  $\omega_L$  :  $a(0) = 0.1/100$
- ▶ State  $\omega_H$  :  $a(0) = 0.5/100$



# Simulations

Without uncertainty:

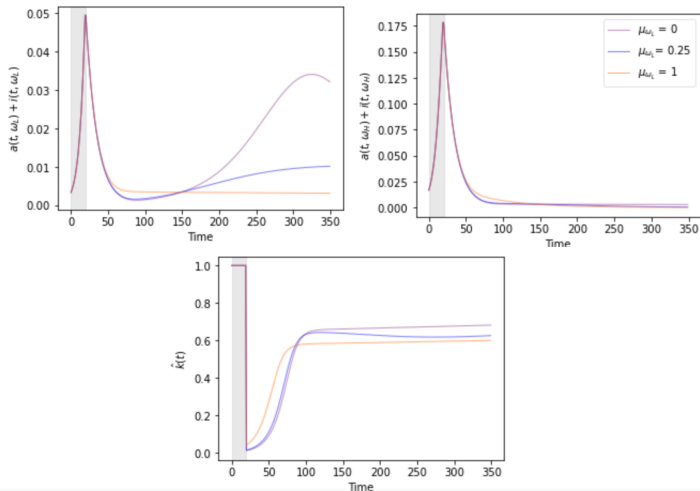


## Simulations

Under uncertainty:  $\mu_0(\omega_L) \in \{0, 0.25, 1\}$ .

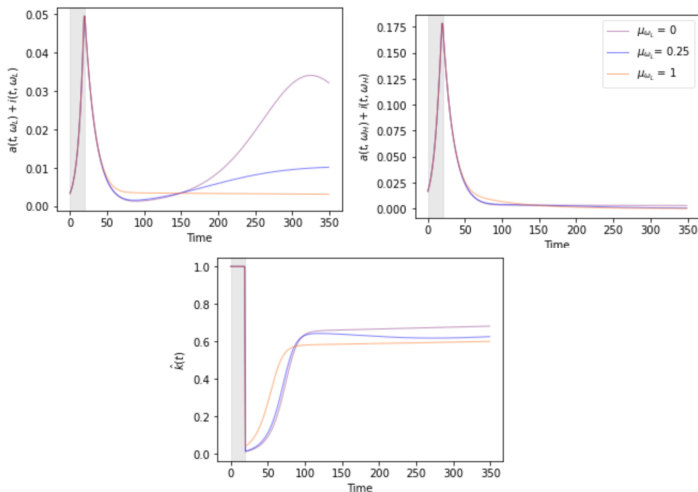
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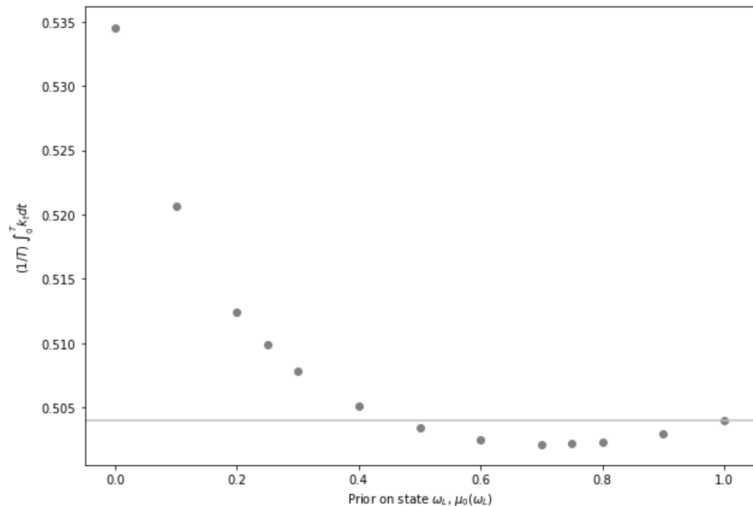
Under uncertainty:  $\mu_0(\omega_L) \in \{0, 0.25, 1\}$ .



Result 1: a second wave may arise under uncertainty.

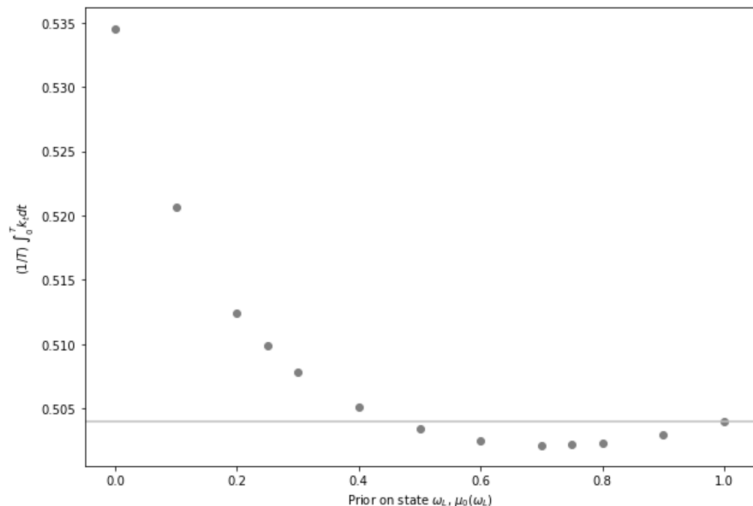
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Average social activity level:  $(1/T) \int_0^T \hat{k}_t dt$ .



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Result 2: average social activity may be lower under uncertainty.

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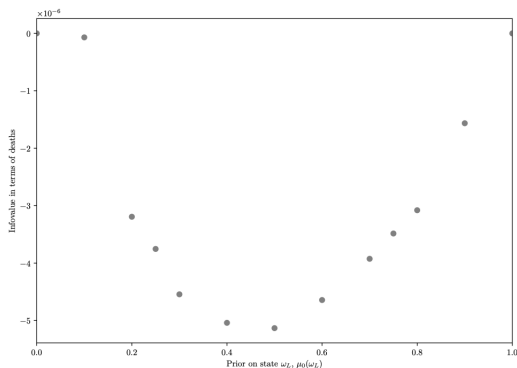
Information value in terms of deaths.

$$IVD_{\bar{\mu}} = - \underbrace{\bar{\mu}TD_1(\omega_L) + (1 - \bar{\mu})TD_0(\omega_H)}_{\text{Ex-ante fraction of deaths without uncertainty}} + \underbrace{\bar{\mu}TD_{\bar{\mu}}(\omega_L) + (1 - \bar{\mu})TD_{\bar{\mu}}(\omega_H)}_{\text{Expected fraction of deaths with uncertainty}}$$

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Result 3: Uncertainty can reduce the *ex-ante* number of deaths.



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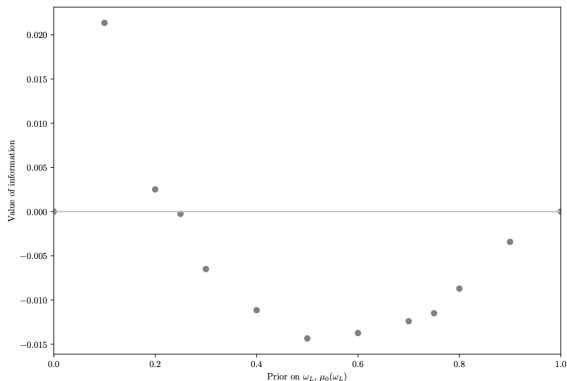
Information value in terms of payoffs.

$$IVP_{\bar{\mu}} = \underbrace{\bar{\mu}v(\hat{k}_1 | \omega_L) + (1 - \bar{\mu})v(\hat{k}_0 | \omega_H)}_{\text{Ex-ante payoff without uncertainty}} - \underbrace{\bar{\mu}v(\hat{k}_{\bar{\mu}} | \omega_L) - (1 - \bar{\mu})v(\hat{k}_{\bar{\mu}} | \omega_H)}_{\text{Expected payoff with uncertainty}}$$

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Result 4: Uncertainty can be welfare improving.

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  - ▶ Information value is negative for some priors.

Many thanks for your attention!