# Evolutionary Vaccination Games with premature vaccines to combat ongoing deadly pandemic

#### Vartika Singh

#### Joint work with

### Khushboo Agarwal, Shubham and Veeraruna Kavitha

Industrial Engineering and Operations Research (IEOR), IIT Bombay, Mumbai, India



## 1 Motivation

- 2 Problem description
- 3 System Dynamics
- 4 Evolutionary behaviour

## **5** Conclusions

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- today, living through a pandemic, COVID-19
- disease characteristics are unknown
- pre-mature vaccines are being introduced
- lack of information, e.g., possible end of disease
- vaccination responses of others matter
- decision about vaccination needs to be made





- dynamic behavioural vaccination responses of population
- choice between two contrasting fears fear of vaccine, fear of infection
- voluntary vaccination leads to evolutionary game theoretic framework

## Objectives

- to study the dynamics and understand the equilibrium states
  - based on disease parameters
  - based on population vaccination responses
- to identify the vaccination responses that are stable against mutations
  - evolutionary stable (ES) equilibrium states

In [1], [2] seasonal variations are considered

- vaccination season: vaccinate and pay a fixed cost, or try free riding
- disease season: infected pay infection cost
- replicator dynamics based models, learn from previous seasons
- In [3], vaccinate before the outbreak or on any week of the pandemic or never
  - $\bullet$  one among the 53 weeks is chosen, 53 being never vaccinate
  - early vaccination or free-rider, 'wait and see' equilibrium
  - in majority of literature, some information about disease is assumed
  - e.g., time duration and time of occurrence of disease is known
  - we do not assume any such information

• 
$$N(t) = S(t) + V(t) + I(t)$$
  
•  $\theta(t) := \frac{I(t)}{N(t)}, \ \psi(t) := \frac{V(t)}{N(t)}, \ \text{and} \ \phi(t) := \frac{S(t)}{N(t)}$ 

•  $b > d + d_e$ 



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- vaccination response depends on  $(\theta(t),\psi(t))$
- probability of vaccination is  $q(\theta, \psi) = \min\{1, \tilde{q}(\theta, \psi)\}$
- vaccination policy:  $\pi(\beta)$

Follow-the-crowd (FC)  $\tilde{q} := \beta \psi(t)$ 

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Follow-the-crowd (FC)Free riders (FR)
$$\tilde{q} := \beta \psi(t)$$
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| Follow-the-crowd (FC)        | Free riders (FR)                           | Vigilant follow-the-crowd (VFC1) |
|------------------------------|--|----------------------------------|
| $\tilde{q} := \beta \psi(t)$ | $\tilde{q} := \beta \psi(t) (1 - \psi(t))$ | $	ilde{q}:=eta	heta(t)\psi(t)$   |

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## Dynamics

• at  $(k+1)^{th}$  transition epoch

$$I_{k+1} = I_k + \underbrace{\mathbb{I}_{k+1}}_{\text{new infection}} - \underbrace{\mathbb{R}_{k+1}}_{\text{new recovery}} - \underbrace{\mathbb{D}_{I,k+1}}_{\text{new death}},$$

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• stochastic approximation iterates for  $\theta_k = I_k/N_k$ 

$$\theta_{k+1} = \theta_k + \epsilon_k \frac{1}{\eta_{k+1}} \left[ \mathbb{I}_{k+1} - \mathbb{R}_{k+1} - \mathbb{D}_{I,k+1} - \Delta N_k \theta_k \right], \quad \epsilon_k := \frac{1}{k+1}$$

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## Dynamics

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• similarly, one can write stochastic approximation iterates for  $\psi_k$ ,  $\phi_k$  and  $\eta_k = N_k/k$ 

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# Stochastic approximation induced ODE[5]

$$\dot{\theta} = \frac{1}{\eta \varrho} \left[ \theta \phi \lambda - r\theta - b\theta - (d_e - d_e \theta) \theta \right], \ \phi = 1 - \theta - \psi$$
  
$$\dot{\psi} = \frac{1}{\eta \varrho} \left[ q(\theta, \psi) \phi \nu - (b - d_e \theta) \psi \right], \text{ and,}$$
  
$$\dot{\eta} = \frac{b - d - d_e \theta}{\varrho} - \eta, \ \varrho = (b + d + d_e \theta + \lambda \theta \phi + \nu \phi + r\theta).$$
(1)

#### Theorem

Under some assumptions:

• the sequence  $(\theta_k, \psi_k, \eta_k)$  converges to attractors of the ODE

**(1)** dynamics of the process can be approximated using the ODE solutions

With no excess deaths  $d_e = 0$ :  $(\rho := \lambda/(r + b + d_e), \mu := b/\nu)$ 

- self-eradicating disease  $(\rho < 1)^1$  gets eradicated without vaccination,  $(\theta, \psi) = (0, 0)$
- endemic disease  $(\rho > 1)$  :
  - vaccination hesitancy  $\rightarrow$  non-vaccinated disease fraction, NVDF =  $\left(1 \frac{1}{a}, 0\right)$
  - vaccination urgency at equilibrium  $\rightarrow$  coexisting equilibrium

$$(\theta_E, \psi_E) := \left(1 - \frac{1}{\rho} - \frac{1}{\mu\rho}, \frac{1}{\mu\rho}\right)$$

- vaccination urgency from beginning  $\rightarrow (0, \psi_f)$  disease is eradicated!
- not possible to eradicate the disease with vigilant agents

 $^{1}\lambda - r - d_e < b$ 

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## Numerical simulations: ODE attractors versus actual system



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$$d_e = 0$$

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- conjectured attractors for follow the crowd and free-riders
- self eradicating disease leads to (0,0)
- for endemic disease, let  $\rho_e := \lambda d_e/r + b$ ,  $\mu_e := b d_e/\hat{\beta}\nu d_e$ 
  - vaccination hesitancy  $\rightarrow$  NVDF  $\left(1 \frac{1}{\rho_e}, 0\right)$
  - vaccination urgency at equilibrium  $\rightarrow$  coexisting equilibrium  $(\theta_E^{d_e}, \psi_E^{d_e})$

$$(\theta_E^{d_e}, \psi_E^{d_e}) \quad \approx \quad \left(1 - \frac{1}{\rho_e} - \frac{o^{d_e}}{\mu \rho_e}, \ \frac{o^{d_e}}{\mu \rho_e} \frac{\lambda - d_e}{\lambda}\right) \text{ with } o^{d_e} := \frac{1}{1 + \frac{d_e(r + d_e - \lambda - \nu)}{\mu \lambda \nu}}$$

# Evolutionary behaviour

- for any  $\pi(\hat{\beta})$ , system reaches an equilibrium
- what if mutants invade a system in equilibrium?
  - is original vaccination response still better?
  - does the equilibrium drift away?
  - which equilibrium states are evolutionary stable?
  - which policies lead to evolutionary stable (ES) states?

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## ESS-AS : Evolutionary Stable Strategies Against Static mutations

A policy  $\pi(\beta)$  is said to be ESS-AS if

- if  $\{q_{\pi(\beta)}\} = \arg\min_{q \in [0,1]} u(q, \pi(\beta)),$
- **(b)** there exists an  $\bar{\epsilon}$  such that  $\{q_{\pi(\beta)}\} = \arg\min_{q \in [0,1]} u(q, \pi_{\epsilon}(\beta, q'))$ , for any  $\epsilon \leq \bar{\epsilon}$

# Anticipated user utility at equilibrium



cost of infection  $c_{I_2} d_e \hat{\theta}$  $p_I(\hat{\theta}$  $c_{I_1}$ P(infection) infection cost cost of death

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# Anticipated user utility at equilibrium





utility of policy q against population profile  $\pi(\hat{\beta})$ :

 $u(q; \pi(\hat{\beta})) := q \times \text{cost of vaccination} + (1 - q) \times \text{cost of infection}$  $= q \times \Delta + \text{cost of infection, where}$  $\Delta = \text{cost of vaccination} - \text{cost of infection}$ 

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## Non-vaccinating ESS

- if the disease is self eradicating  $\rightarrow$  (0,0), or
- if anticipated cost of vaccination ( $\Delta > 0$ ) is more at NVDF  $\rightarrow (1 1/\rho, 0)/(1 1/\rho_e, 0)$
- in some cases, there is no ESS at all
- ESS is either possible in all vaccination responses, or in none!!!

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## Vaccinating ESS

- probability of vaccination is 1 at equilibrium
- vaccination response leads to co-existing equilibrium  $(\theta_E, \psi_E) / (\theta_E^{d_e}, \psi_E^{d_e})$
- anticipated cost of infection at corresponding equilibrium is high  $(\Delta < 0)$

- with excess deaths,
  - higher vaccinated fraction
  - lower infected fraction
  - pattern is similar
- for small birth-rate
  - higher infection rate per birth
  - higher vaccinated fraction
  - lower infected fraction as ESS
- for large birth-rate, self eradicating disease





- infected  $\downarrow$  and vaccinated  $\uparrow$  initially
- might expect high vaccinated with more vaccine availability
- the converse is true !!
- perception towards infection cost changes with abundance of vaccine
- non-vaccinating ESS leads to NVDF

- motivated by COVID-19 pandemic, prematurely introduced vaccines
- derived infected/vaccinated proportions for various behavioural patterns
- many vaccination responses eradicate the disease
- disease is never eradicated under ES policies, unless self eradicating
- different dynamics for different response, but ES state is same
- ESS either exists in all behaviours or in none
- abundance of vaccine leads to lower vaccinated proportion
- ironically, disease can be better curbed with excess deaths

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