

## Planning horizon affects prophylactic decision-making and epidemic dynamics

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International Congress on Agent Computing November 29, 2016

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– Infectious diseases remain a major threat to human health. For example, the recent outbreaks of SARS, H1N1 flu, and Ebola







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- Human behavior has been recognized to play a significant role in the spread of infectious diseases







Aspects of human behavior incorporated in infectious disease models:

- Individual responses to different types of public health interventions
- Influence of risk of infection and social cost on individual behavioral changes
- Effect of **awareness** or **fear spreading** on individual behavioral decisions



## **Objective**

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Understand how individuals' planning horizons influences behavioral changes (i.e., adopt or not prophylaxis) and how this in turn influences epidemic dynamics



### Definition

Planning horizon is the time in the future over which individuals consider to make a behavioral decision to adopt prophylaxis



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The SPIR model (Susceptible, Prophylactic, Infectious, Recovered) is an epidemiological agent-based model that couples individual behavioral decisions with a disease dynamics model









## Disease Dynamics Model

### Event 1: Interaction and Infection





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### Event 1: Interaction and Infection





## Disease Dynamics Model

### Event 2: Recovery





## Disease Dynamics Model

## Event 2: Recovery





### Event 3: Behavioral Decision





- Agents use a rational choice model to decide whether to adopt Susceptible or Prophylactic behavior
- $-$  Agents adopt the behavior that has the **largest utility** over the planning horizon H
- $-$  Agents have **identical** and **complete knowledge** of the disease and its prevalence
- $-$  Agents assume the disease prevalence remains at its current value during the next H time steps



#### 1. Expected Time



 $T_S = 52$ ,  $T_I = 30$ ,  $T_R = 18$   $T_P = 70$ ,  $T_I = 19$ ,  $T_R = 11$ 



#### 1. Expected Time



2. Payoff

 $T_S = 52$ ,  $T_I = 30$ ,  $T_R = 18$   $T_P = 70$ ,  $T_I = 19$ ,  $T_R = 11$ Ex:  $u_S = 1.0$ ,  $u_P = 0.9$ ,  $u_I = 0.1$ ,  $u_R = 0.95$ 



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#### 3. Utility Calculation

 $U_S = u_S T_S + u_I T_I + u_R T_R$  $U_P = u_P T_P + u_I T_I + u_R T_R$ 

#### Ex:

 $U_S = 1.0 \times 52 + 0.1 \times 30 + 0.95 \times 18 = 72.1$  $U_{\rm P}$  = 0.9  $\times$  70 + 0.1  $\times$  19 + 0.95  $\times$  11 = 75.35



#### 1. Expected Time



2. Payoff



 $T_S = 52$ ,  $T_I = 30$ ,  $T_R = 18$   $T_P = 70$ ,  $T_I = 19$ ,  $T_R = 11$ 

Ex:  $u_S = 1.0$ ,  $u_P = 0.9$ ,  $u_I = 0.1$ ,  $u_R = 0.95$ 

#### 3. Utility Calculation

$$
US = uSTS + uITI + uRTR
$$
  

$$
UP = uPTP + uITI + uRTR
$$

#### Ex:

 $U_S = 1.0 \times 52 + 0.1 \times 30 + 0.95 \times 18 = 72.1$  $U_{\rm P}$  = 0.9  $\times$  70 + 0.1  $\times$  19 + 0.95  $\times$  11 = 75.35

#### 4. Decision Making

 $\text{Decision} = \left\{ \begin{matrix} \text{Prophylactic} & \text{for } U_{\text{S}} < U_{\text{P}} \\ \text{Susceptible} & \text{otherwise.} \end{matrix} \right.$ 

**Ex:** 
$$
(U_S = 72.1) < (U_P = 75.35)
$$
  
\nDecision = Prophylastic

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## Behavioral Decision Analysis

### Switching Point

What is the proportion of infectious agents beyond which it would be advantageous for an agent to switch from Susceptible (i.e. non-prophylactic) to Prophylactic behavior or vice-versa?



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## Influence on Epidemic Dynamics

### Research Question

How does planning horizon affect prophylactic decision-making and epidemic dynamics?



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### Table: Experiment simulation input parameters







### Baseline

- Never consider adopting prophylactic behavior
- No impact on the epidemic dynamic





### Short

- Consider adopting prophylactic behavior
- Never adopt prophylactic behavior
- No impact on the epidemic dynamic





### Intermediate

- Adopt prophylactic behavior
- Impact on the epidemic dynamic





### Long

- Never consider adopting prophylactic behavior
- No impact on the epidemic dynamic
- "Get it over with"



## Influence of the Decision Frequency



### Increasing d

- Reduces the peak size
- Prolongs the epidemic
- Generates secondary infection waves

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- $-$  Agents do not engage in prophylactic behavior for short and long planning horizons; for intermediate planning horizon agents adopt prophylactic behavior depending on the disease parameters.
- $-$  The adoption of prophylactic behavior is not always monotonically associated with the **prevalence of the disease**.
- $-$  Adoption of prophylactic behavior reduces the epidemic peak size while **prolonging the epidemic** and potentially **generates** secondary waves of infection.
- **Increasing decision frequency** makes the effects of adopting prophylactic behavior stronger.



- Evaluate scenarios composed of heterogeneous agents (e.g., different risk perception, payoff structure, etc.) and different topological structures (e.g., scale-free network).
- Perform comparative studies with different behavioral decision models. For example, relax some of the assumptions of rationality.
- Incorporate social influence aspects into the behavioral decision model.



# Questions?



Thank you





NIGMS of the NIH award P20GM104420

Research reported in this presentation was supported by the NIGMS of the NIH under award P20GM104420. The content is solely the responsibility of the authors and does not necessarily represent the official views of the NIH. We acknowledge the support of the Institute for Bioinformatics and Evolutionary Studies Computational Resources Core sponsored by the NIH grant P30GM103324 that provided us computer resources to perform this study.