

Why System Science Methods?

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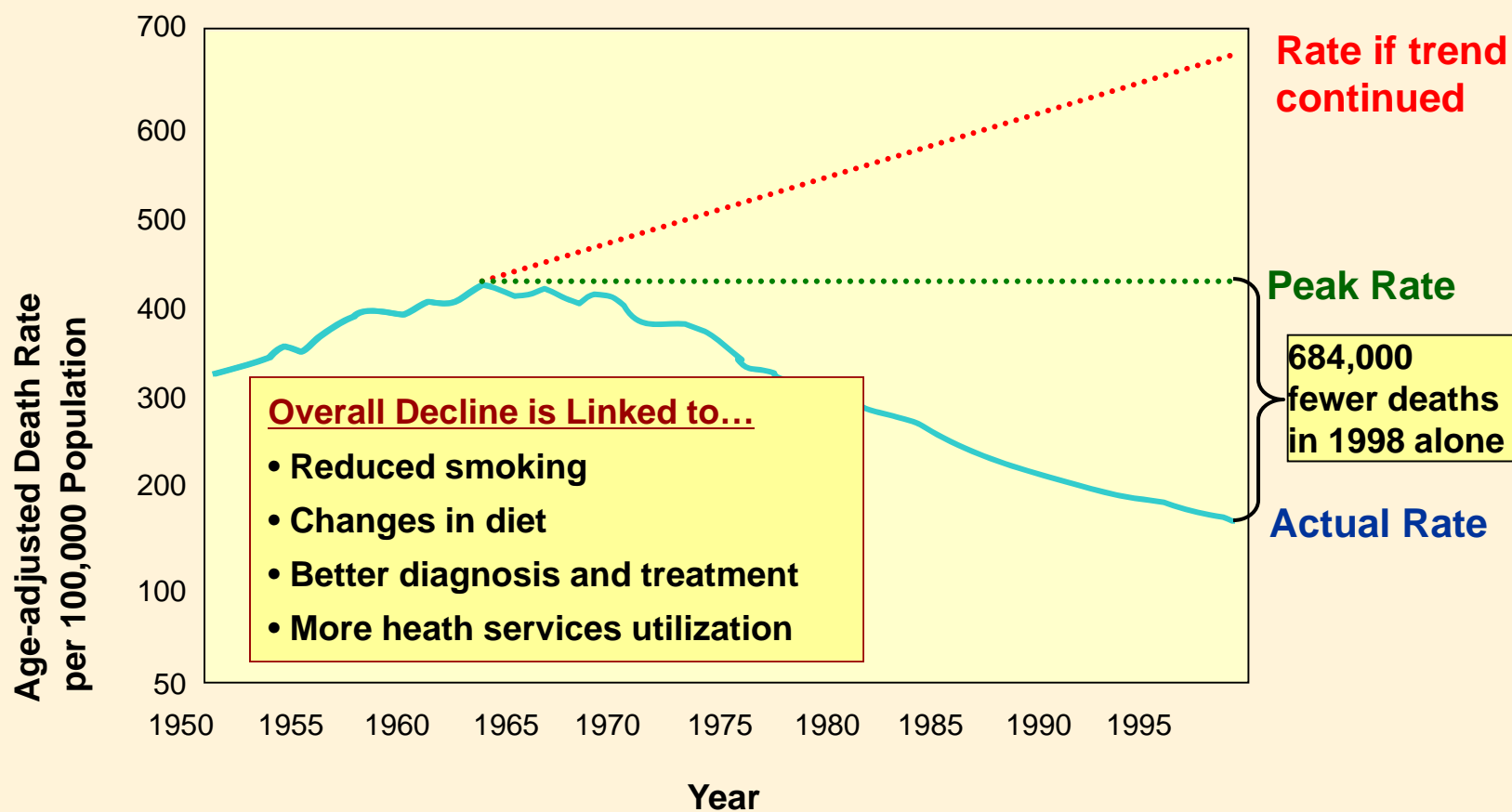
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7-23-2010



Public Health as “Redirecting the Course of Change”

Actual and Expected Death Rates for Coronary Heart Disease, 1950–1998

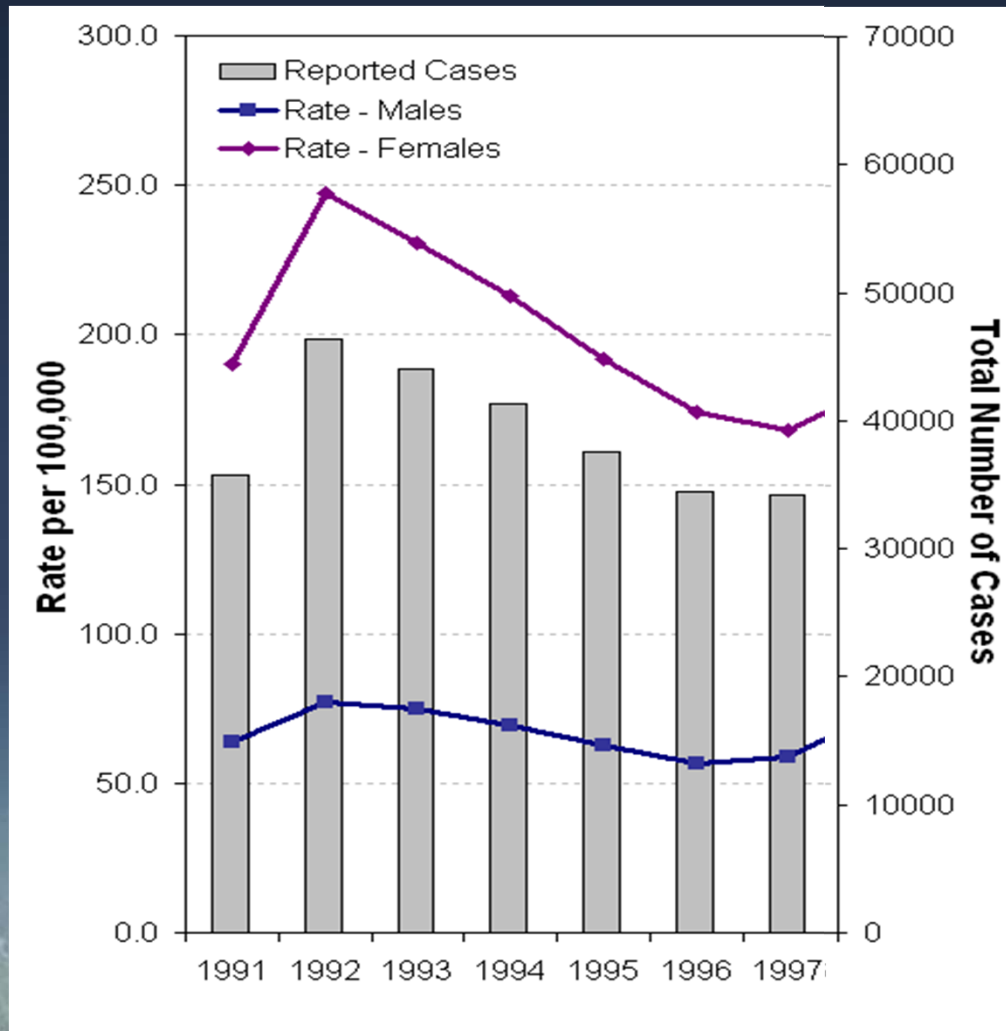


Adapted from Homer & Millstein 2007

Marks JS. The burden of chronic disease and the future of public health. CDC Information Sharing Meeting. Atlanta, GA: National Center for Chronic Disease Prevention and Health Promotion; 2003.

Centers for Disease Control and Prevention. Achievements in public health, 1900-1999: decline in deaths from heart disease and stroke -- United States, 1900-1999. MMWR 1999;48(30):649-656. Available at <<http://www.cdc.gov/mmwr/preview/mmwrhtml/mm4830a1.htm>>

Public Health as “Redirecting the Course of Change”

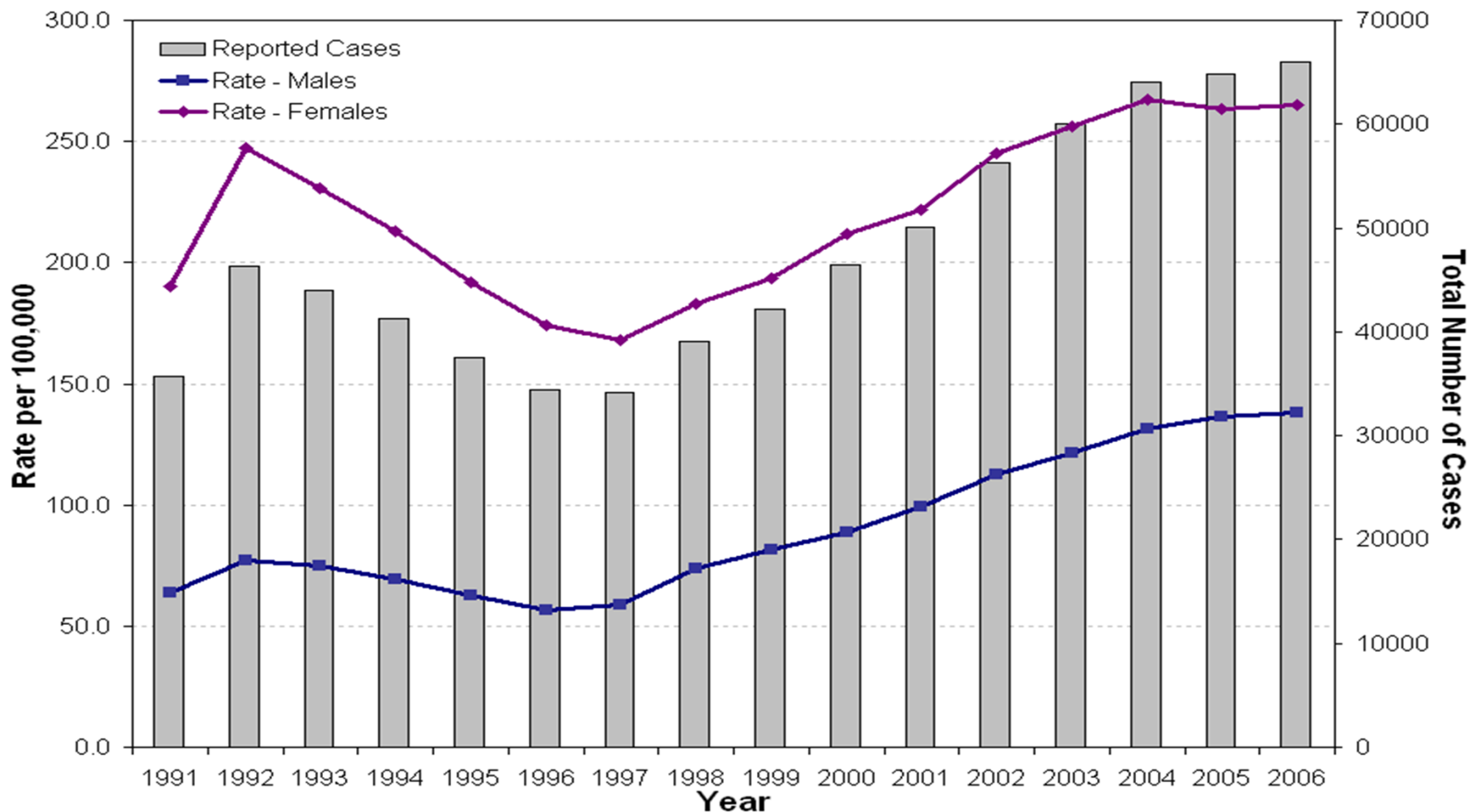


•Adapted from Tom Wong, 2007

•Data for 2005 and 2006 are preliminary and are anticipated to change

•Source: Surveillance and Epidemiology Unit, Community Acquired Infections Division, PHAC

Public Health as “Redirecting the Course of Change”



•Tom Wong, 2007

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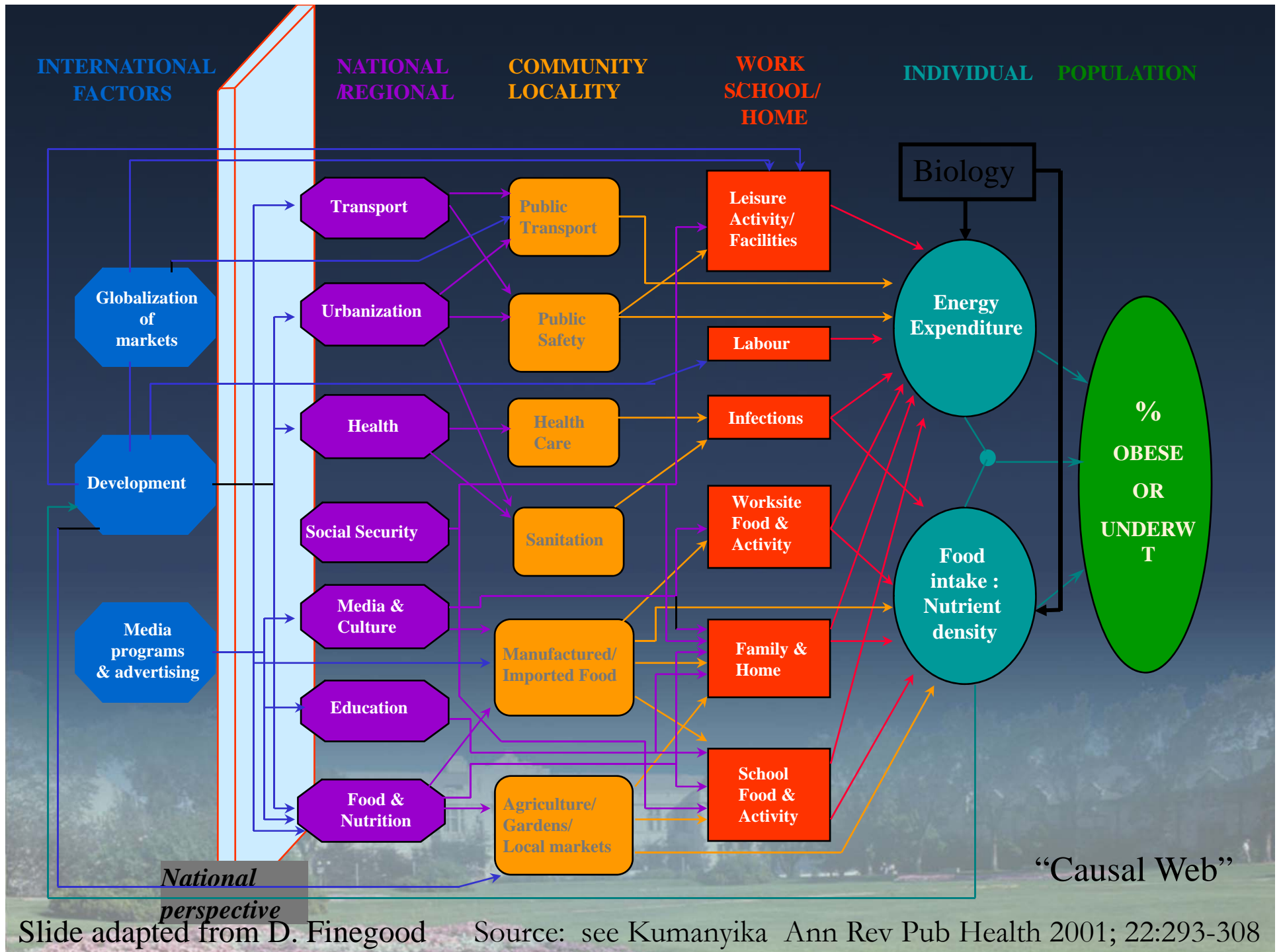
•Source: Surveillance and Epidemiology Unit, Community Acquired Infections Division, PHAC

Securing Public Health Requires Grappling with Great Complexity

- **Structural Complexity**
- **Dynamic Complexity**

Anderson & May
Infectious Diseases of Humans

A background image showing a large, multi-story building with a lawn and trees. The image is faded and serves as a backdrop for the text.



Slide adapted from D. Finegood

Source: see Kumanyika Ann Rev Pub Health 2001; 22:293-308

Obesity System Map

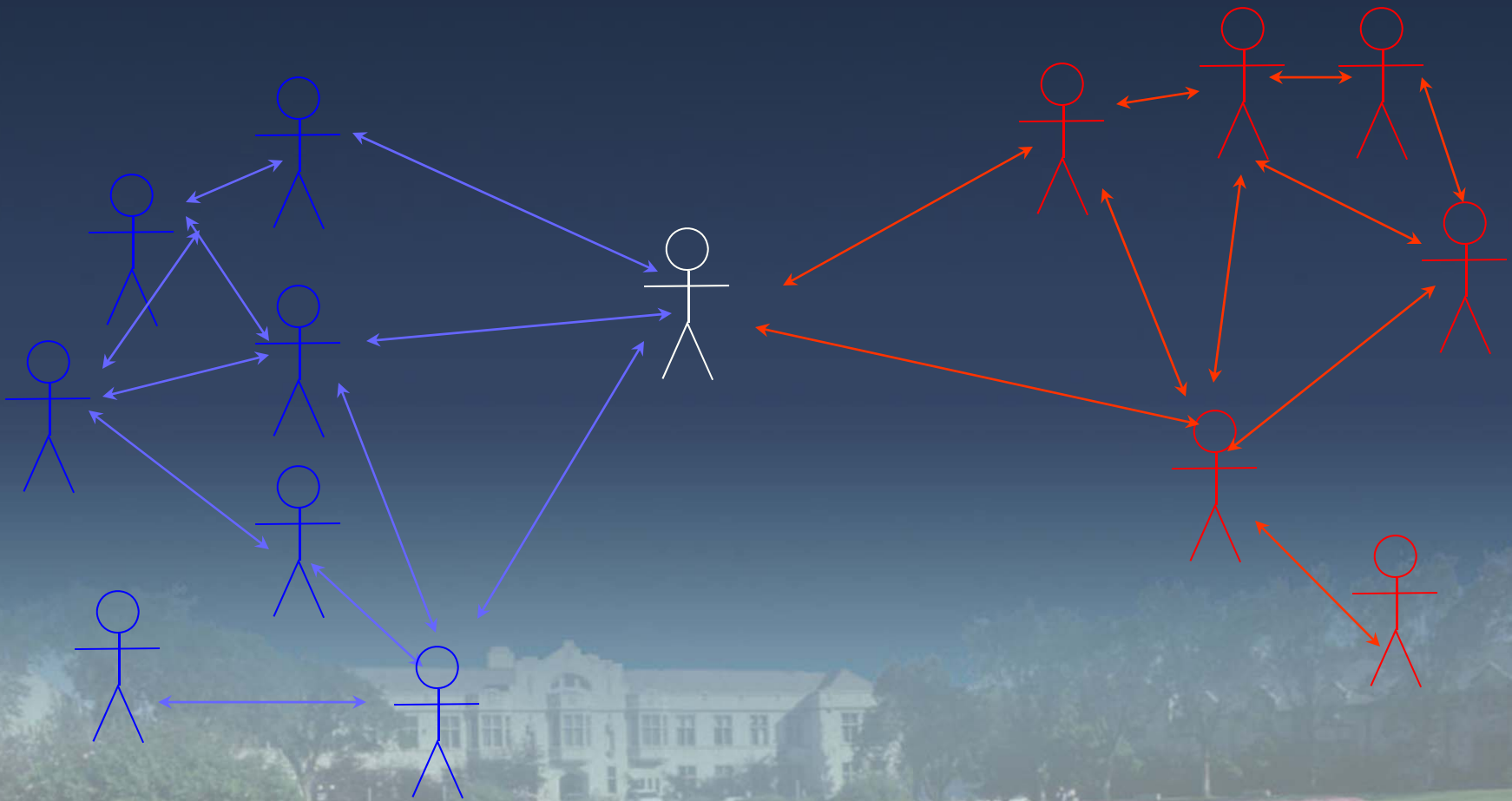
Map 27

Weighted
Causal Linkages

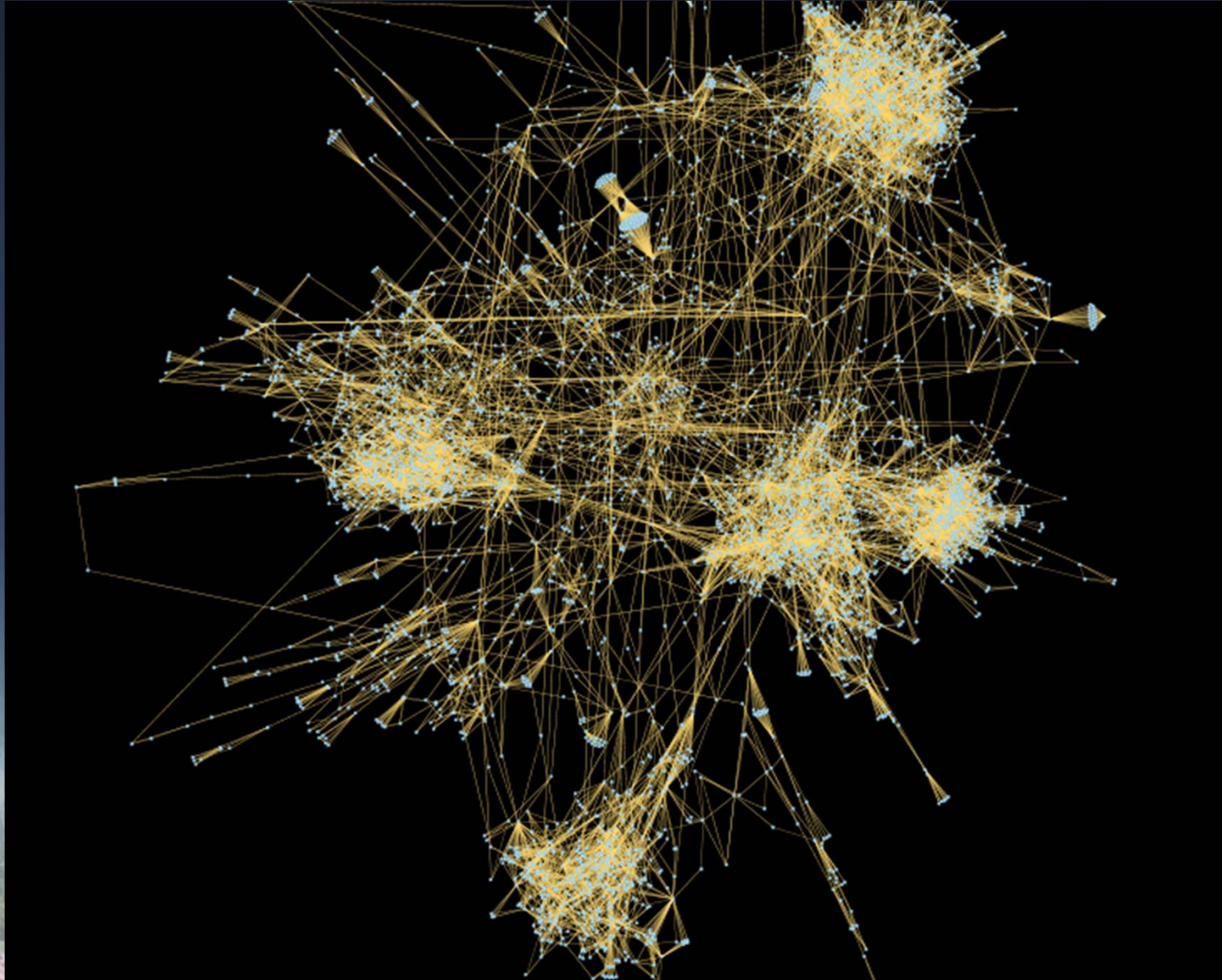
Obesity System Map
Version 1.0 - 29 November 2005



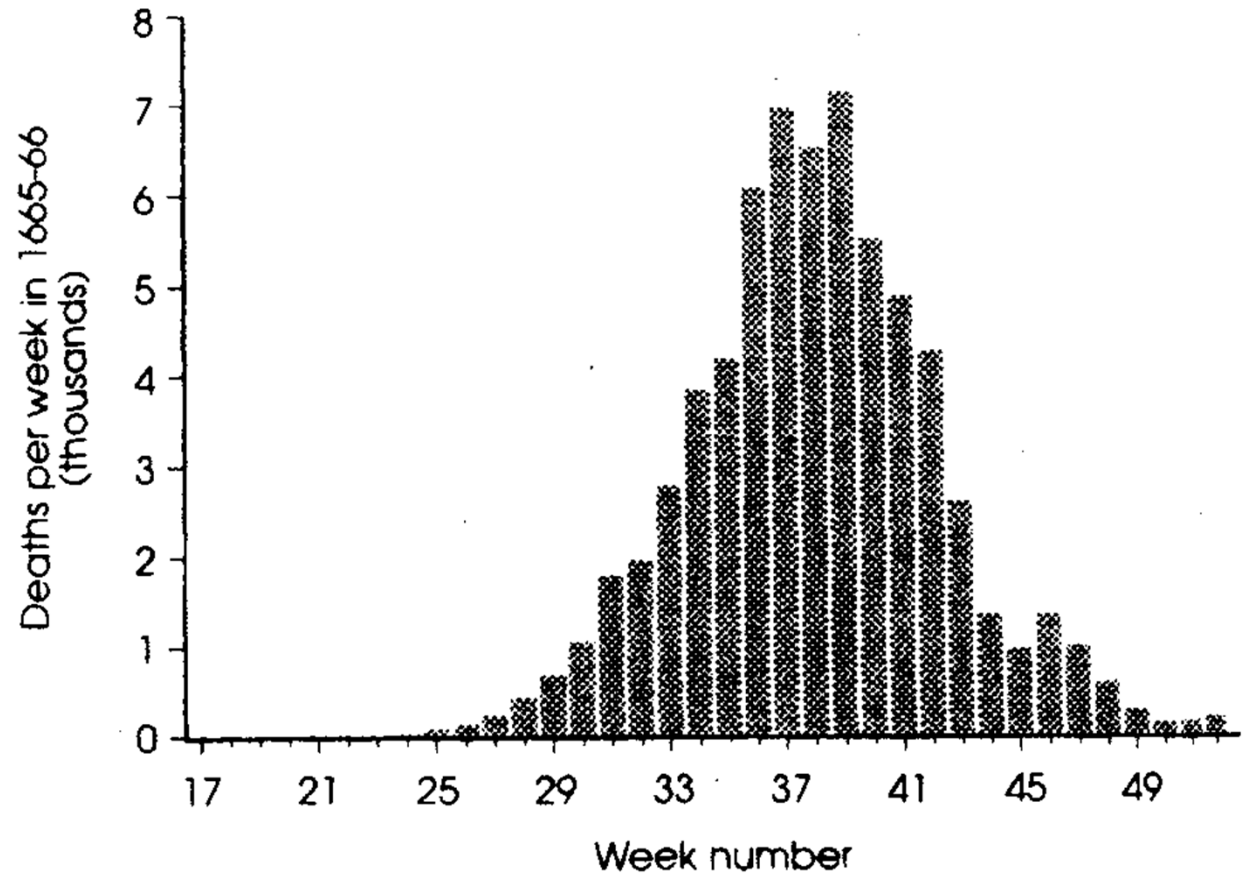
Importance of Bridging Individuals



TB Network Substructure



Dynamic Complexity: Exponential Growth & Decay



Anderson & May

Infectious Diseases of Humans

Fig. 3.11. Recorded deaths from the bubonic plague in London during the year (data from Brayley 1722).

Dynamic Complexity: Oscillations (Damped & Otherwise)

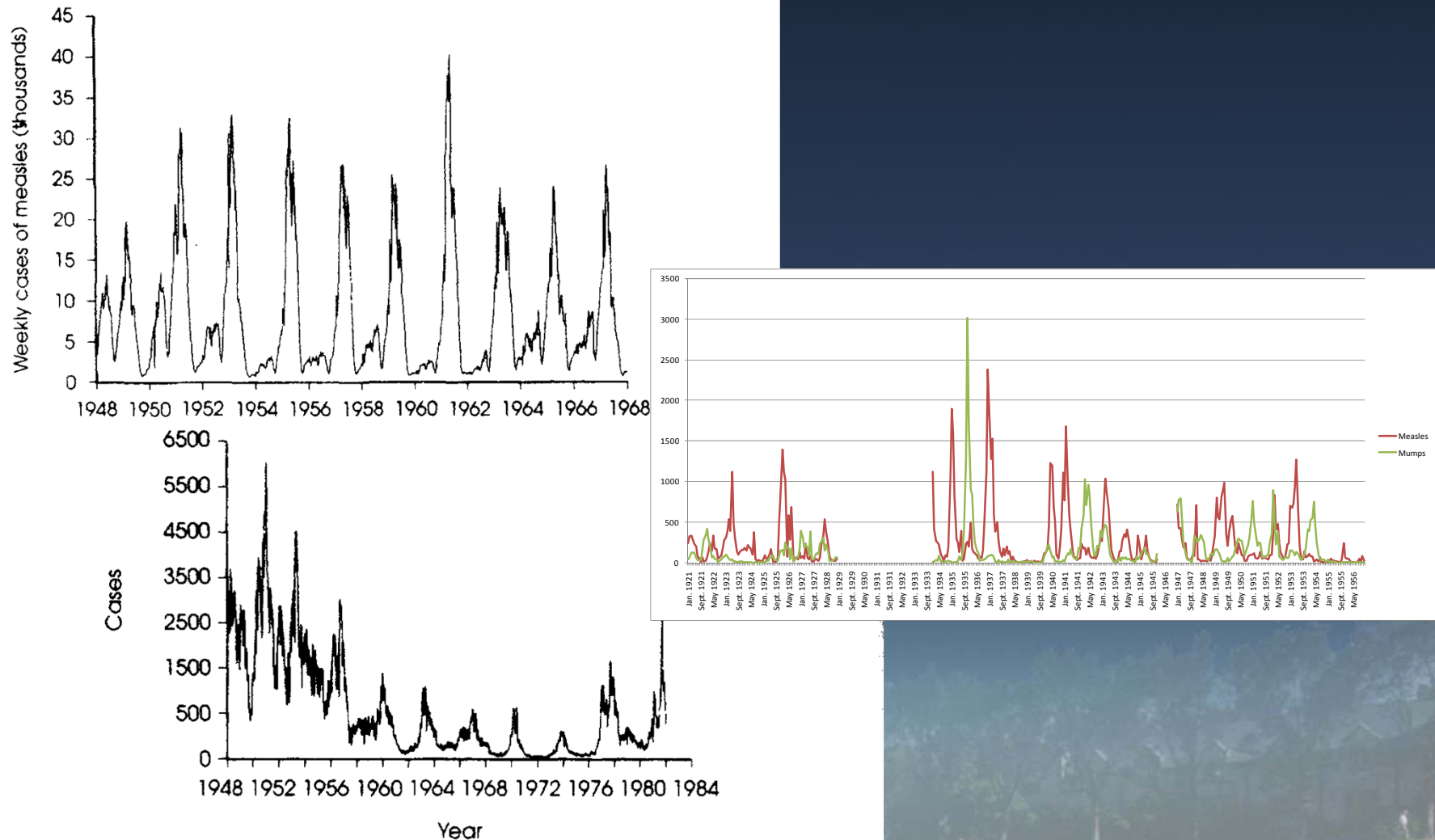
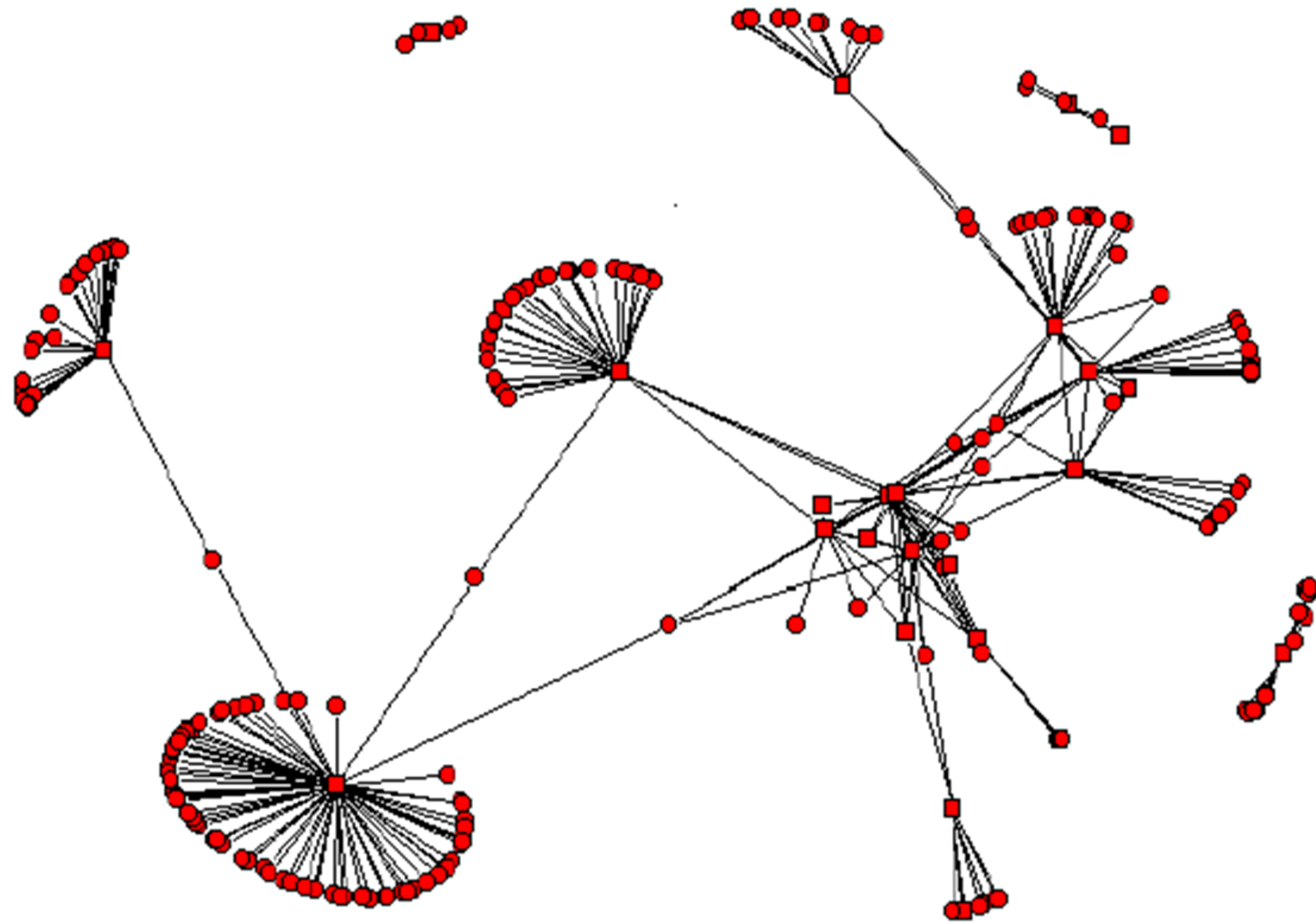


Fig. 6.8. Weekly case notifications of pertussis (whooping cough) in England and Wales for the time period 1948–82. Mass vaccination was introduced in 1956.

Dynamic Complexity: Tipping Points

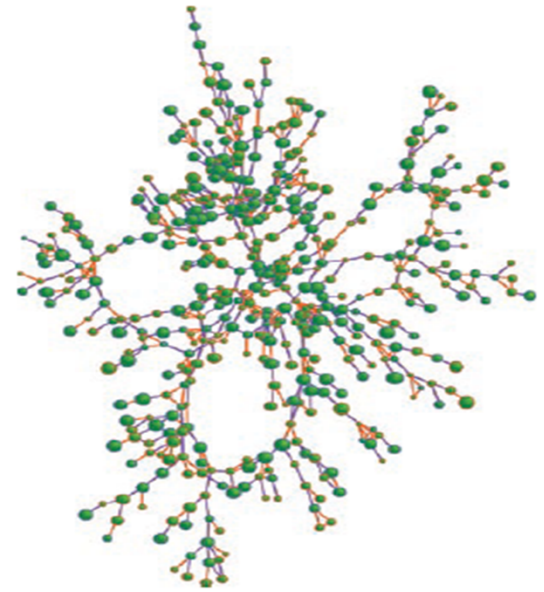
- **Sufficiently fast delivery of treatment or high enough vaccination rates can prevent an infection from being able to establish itself**
- **While the components of the system are the same (most individuals remain susceptible), the population as a whole is protected**
 - **This “herd immunity” is a feature of the system as a whole, not of its individual pieces**

Persistence of Endemic Infection in Network “Cores”

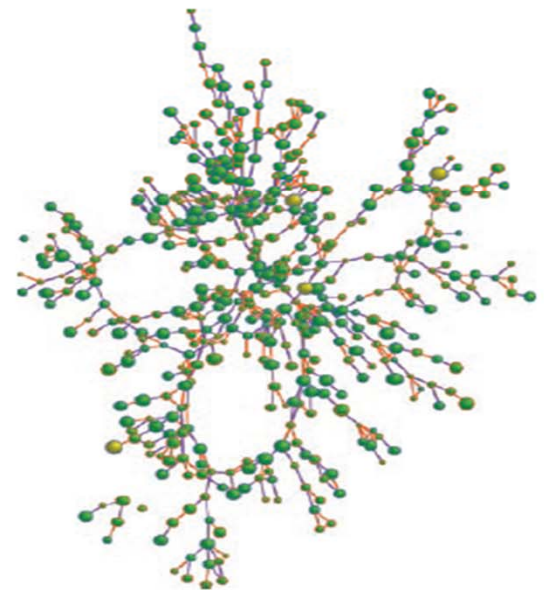


Network Spread of Obesity

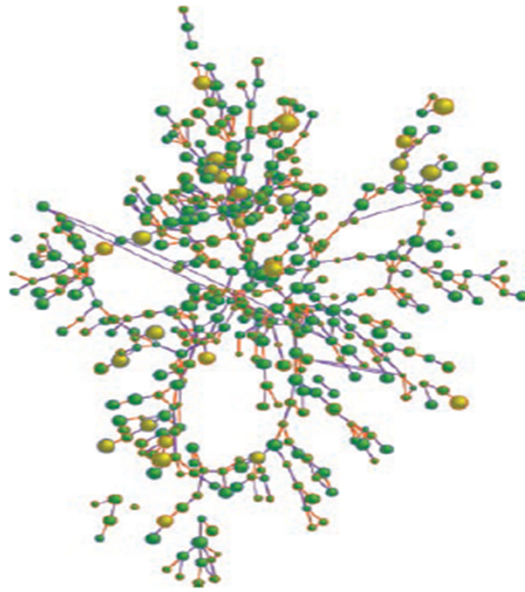
A 1975



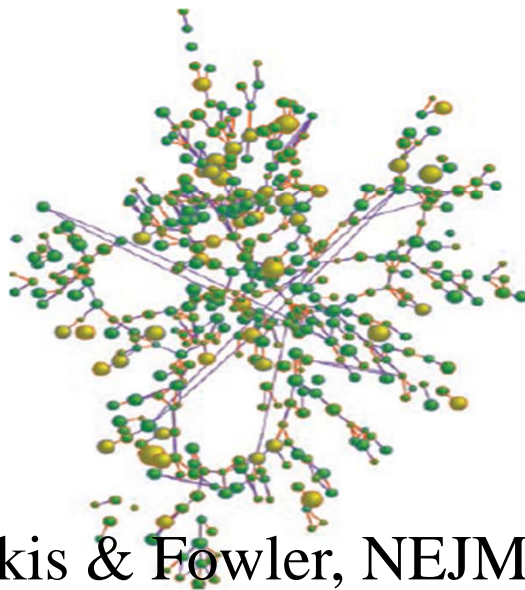
B 1980



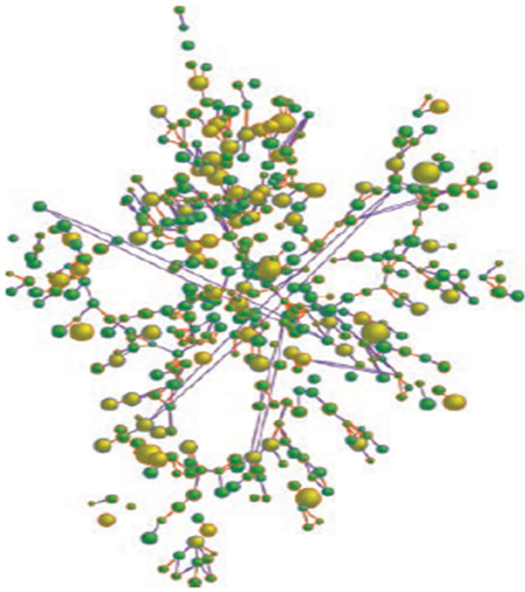
C 1985



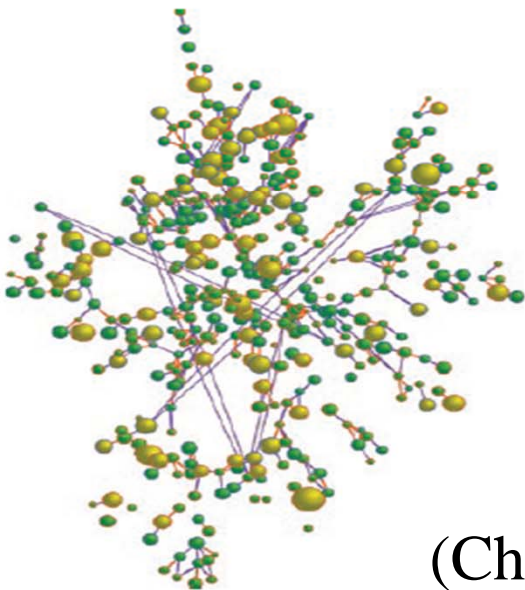
D 1990



E 1995



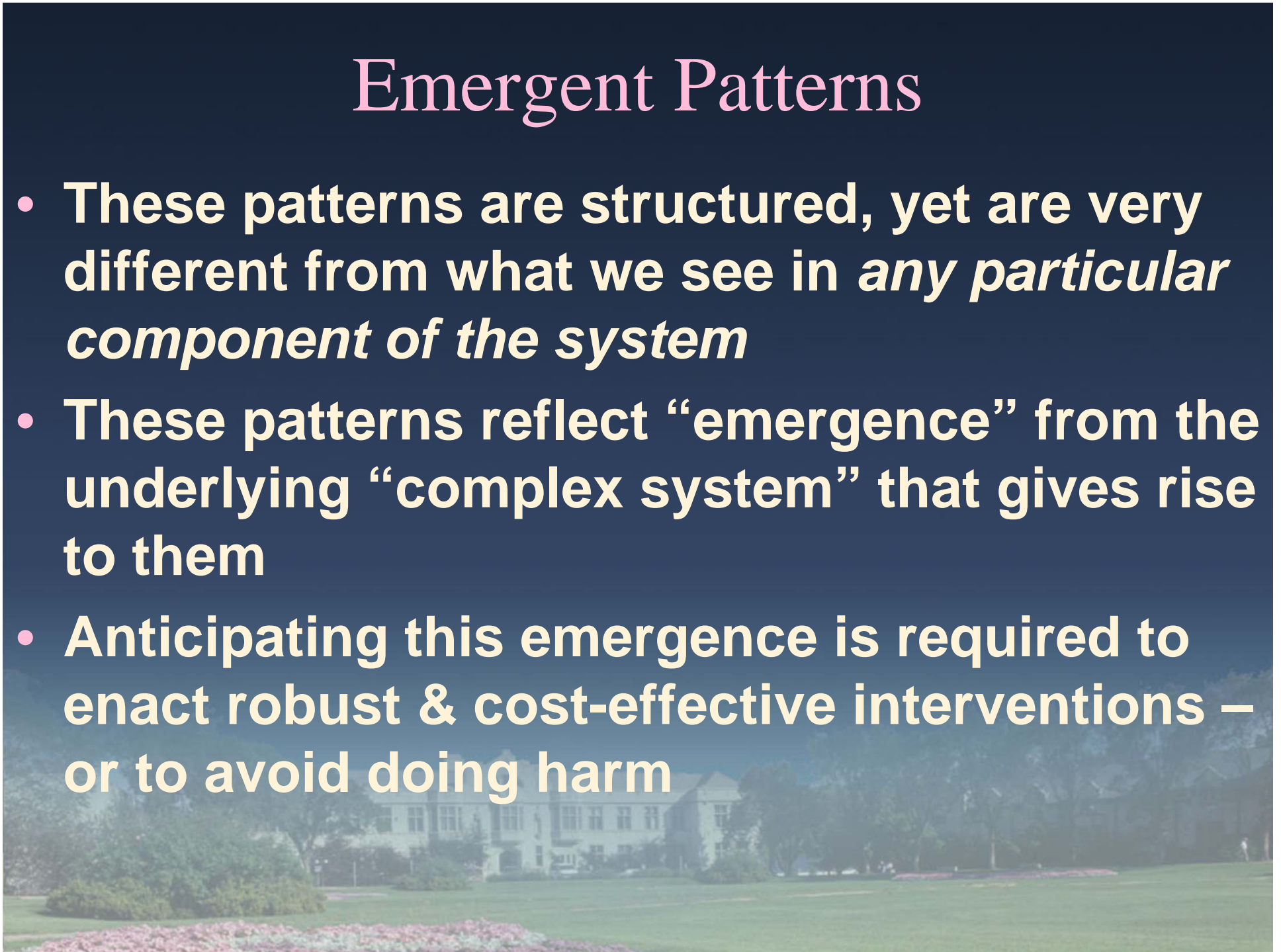
F 2000



(Christakis & Fowler, NEJM)

Emergent Patterns

- **These patterns are structured, yet are very different from what we see in *any particular component of the system***
- **These patterns reflect “emergence” from the underlying “complex system” that gives rise to them**
- **Anticipating this emergence is required to enact robust & cost-effective interventions – or to avoid doing harm**



Regularities Arise from Underlying *Processes*

- **The components of the system are tightly interrelated, not independent**
- **Many of the features of time series from a system are driven by the same underlying processes – interactions between**
 - **Natural history of infection**
 - **Demographic change of the population**
 - **Mechanisms of infection transmission**
 - **Risk behaviour & risk perception**
 - **Health system response**
- **The emergent behaviour of the system comes not from one component in isolation, but from combination of all**

Complexities & Systems Effects Matter for Intervention Selection

- Blowback, multiplier effects
- Presence of “tipping points”
- Tradeoffs of prevention vs. screening vs. contact tracing & treatment
- Interaction between infections and with chronic diseases
- Evaluation of focused intervention on different parts of system or classes of individuals
- Evaluation of intervention portfolios

Challenges of Complex Systems

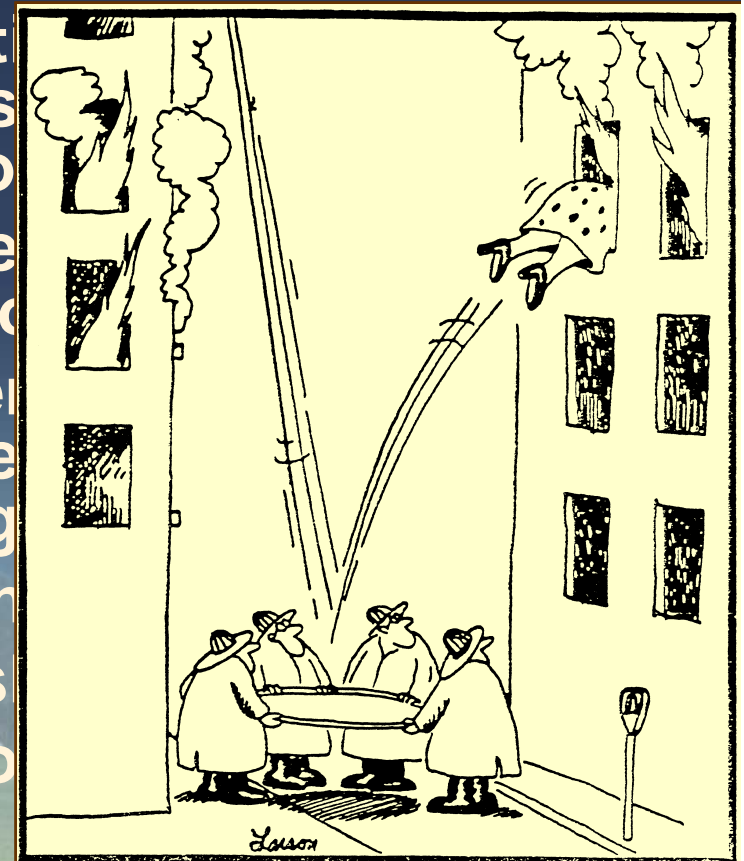
- **Counter-intuitive behavior**
- **Misperceptions**
- **Policy resistance**
- **These phenomena pose problems for**
 - **Learning from experience:** Painful & slow
 - **Coordinating:** Actors in 1 area of the system often have poor sense as to how actions of actors in other areas of the system affect them
⇒ risk of working at cross purposes
 - **Deciding:** Unclear tradeoffs between choices
 - **Designing:** Not clear how to best structure the roles/responsibilities of the actors, reporting, etc

Policy Resistance

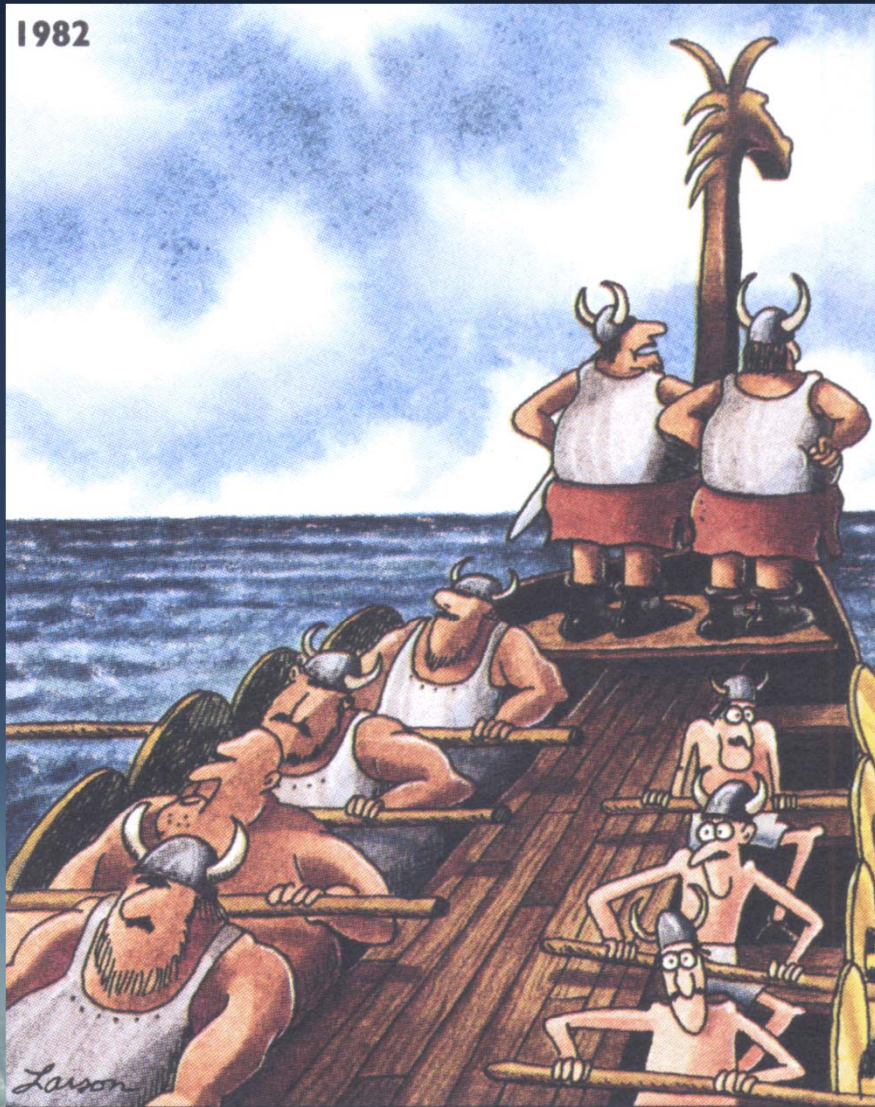
- Development of pathogen drug resistance
- Cutting cigarette tar levels reduces cessation
- Cutting cigarette nicotine levels leads to compensatory smoking
- Targeted anti-tobacco interventions lead to equally targeted coupon programs by tobacco industry
- Charging for supplies for diabetics leads to higher overall costs by increases costs due to reduced self-management, faster disease progression
- ARVs prolong lives of HIV carriers, but lead to resurgent HIV epidemic due to lower risk perception
- Attempts to economize by understaffing STI clinics leads to long treatment wait, greater risk of transmission by infectives & bigger epidemics
- Antibiotic overuse worsens pathogen resistance
- Antilock breaks lead to more risky driving
- Natural feedback: Intergenerational “Vicious Cycles”

Policy Resistance

- Development of pathogen drug resistance
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- Charging for supplies for diabetes increases overall costs by increasing costs of management, faster disease progression
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- Attempts to economize by underfunding health care leads to long treatment wait, greater disease transmission by infectives & bigger outbreaks
- Antibiotic overuse worsens pathogen drug resistance
- Antilock brakes lead to more risk-taking
- Natural feedback: Intergenerational



A Systems Problem



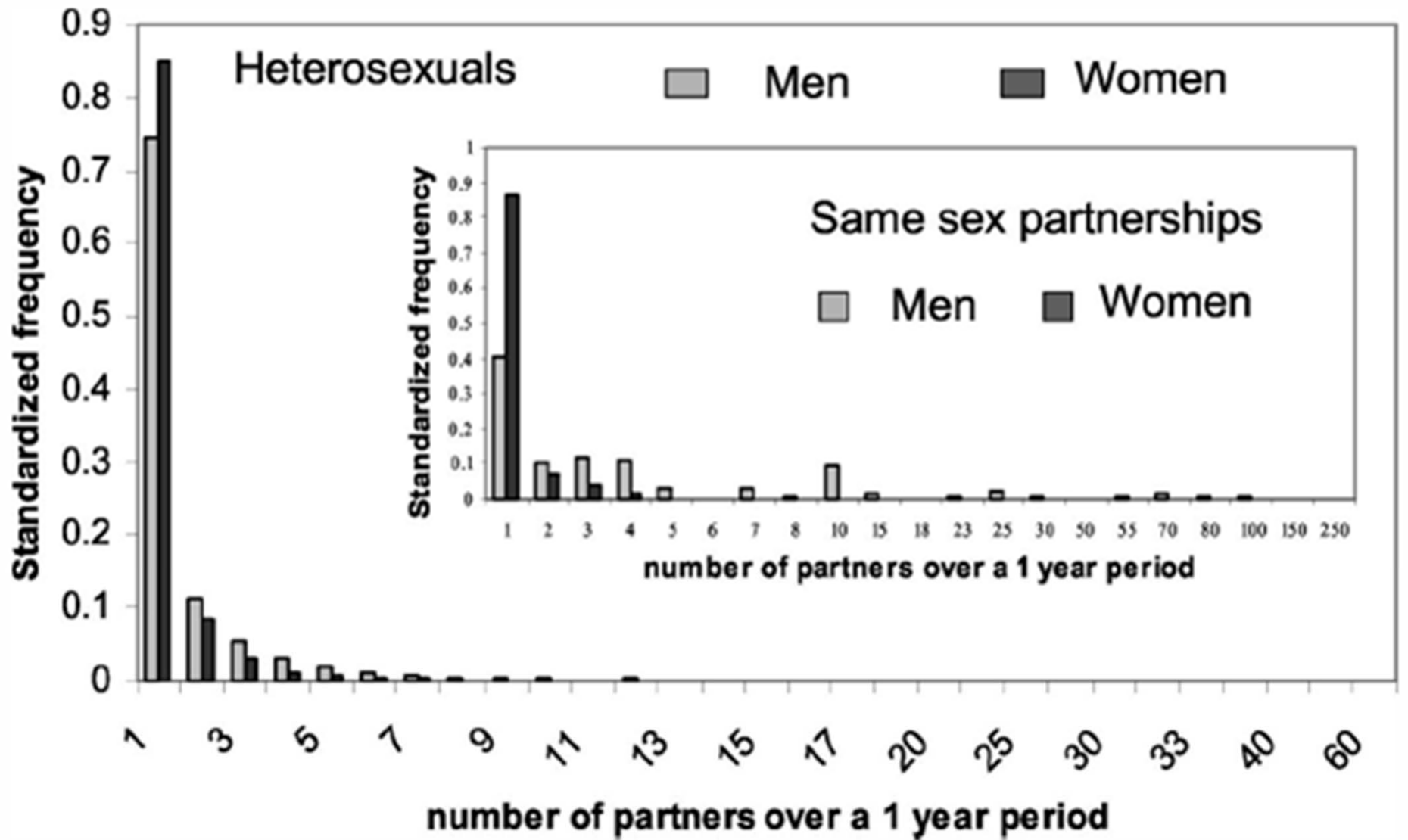
"I've got it, too, Omar ... a strange feeling like we've just been going in circles."

Larson, The Far Side

Emergence Reflects Complexity of Underlying System

- **Interactions**
 - Interpersonal
 - Between conditions (e.g. STIs & HIV, HCV&HIV, Chronic & Infectious illness)
- **Delays**
 - Presentation of symptoms/Contact tracing/Identification of asymptomatic
- **Feedbacks**
 - Intergenerational/social network mediated
 - Immune system
 - With healthcare system
 - Behavior change after knowledge of health status
 - Risk perceptions
- **Nonlinear: Risk, cost, intervention synergies**
- **Heterogeneity in progression, behaviour**

Heterogeneities (Contact Rates)



Imperatives for Protecting Health



Typical Current State

“Static view of problems that are studied in isolation”



Proposed Future State

“Dynamic systems and syndemic approaches”



“An imperative to shift from the typical view of static problems in isolation toward a future state in which the focus is on dynamic systems and syndemic approaches.”

Currently, application of complex systems theories or syndemic science to health protection challenges is in its infancy.”

Former CDC Director Julie Gerberding

Adapted from Millstein, 2007

Gerberding JL. Protecting health: the new research imperative. Journal of the American Medical Association 2005;294(11):1403-1406.

Strengths & Weaknesses of Reductionist Approaches

- **Traditional scientific approaches have pursued a primarily reductionist strategy**
- **This strategy has offered profound insights into how mechanisms work in isolation, but limited understanding how the connections between mechanisms combine to yield overall behaviour**
- **Much observed behaviour is emergent – results from the collective interaction of a set of components, rather than any component in isolation**

Missing the Forest for the Trees

- **By itself, traditional reductionist scientific approach to understanding health problems is via specialized study**
 - **Fragmented understanding**
 - **“Silos”**
- **Many complexities concerning are missed**
 - **Interactions/relationships**
 - **Context**
- **We often get a good understanding of the parts – but little understanding of the whole**

A More Holistic “Systems” Approach

- **Systems science can help us visualize understand implications of connections between model components**
- **A key way in which system science aids this is through the use of systems models**
 - **These models are simplified representations of a hypothesized situation that obtains in reality**
 - **The models help us reason about the implications of our understanding**

A Metaphor for Scientific Exploration



Complementary Model Types

- **Static Models**
 - Models help us understand connections between system components, but don't explicitly represent time
 - Aid reasoning about structure of system
- **Dynamic models**
 - Aid in understanding dynamic implications (consequences over time) of system structure & choices

Social Network Analysis

- **Understanding structural relationship between parties**
- **Understanding how network position influences patterns of health**
- **Identifying highly influential or critical parties**
- **An important enabler for and “synergizer” with dynamic modeling**
- **(Dynamic extensions are possible)**

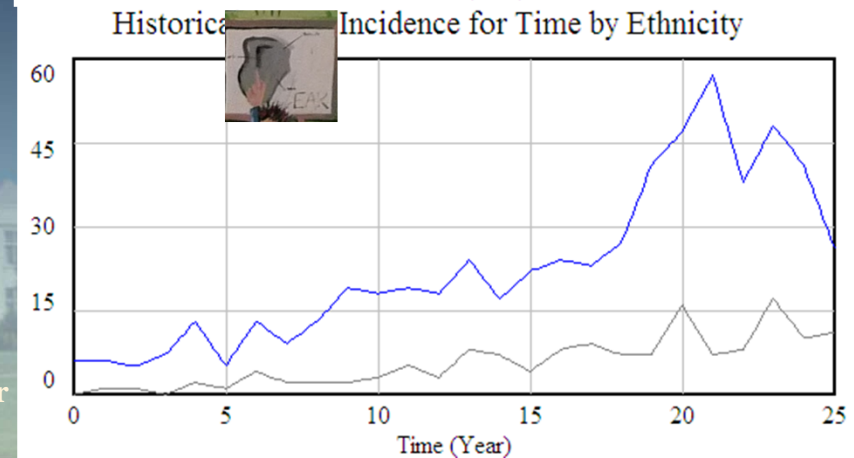
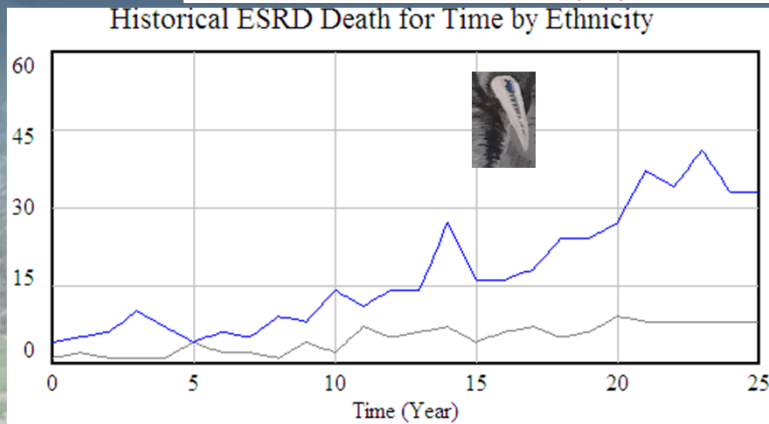
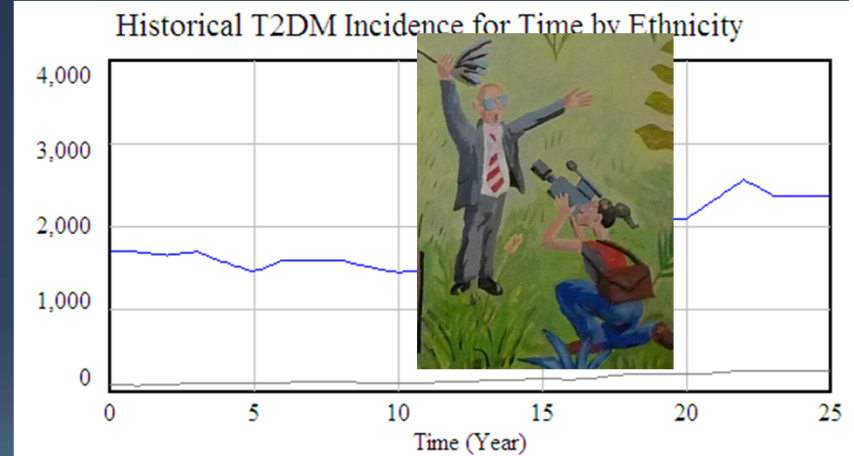
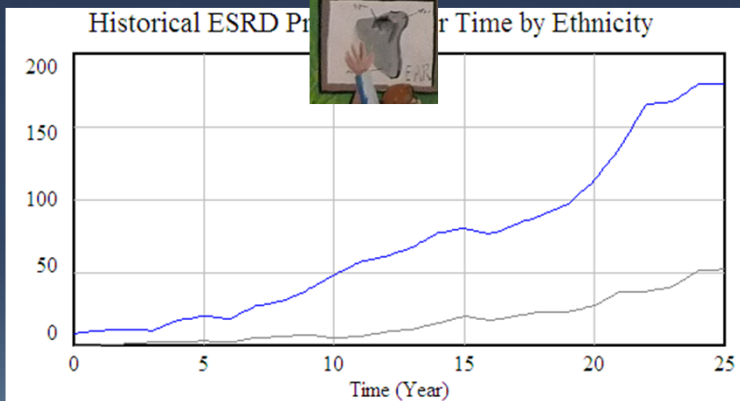
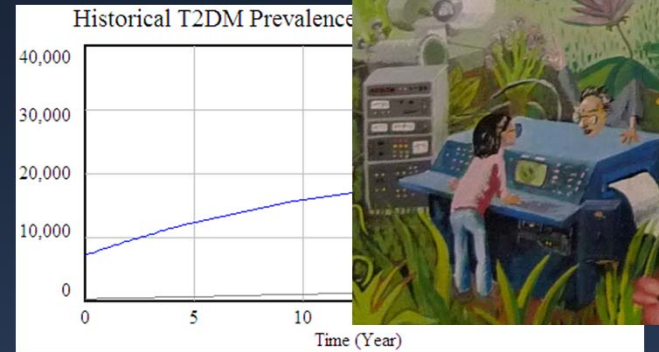
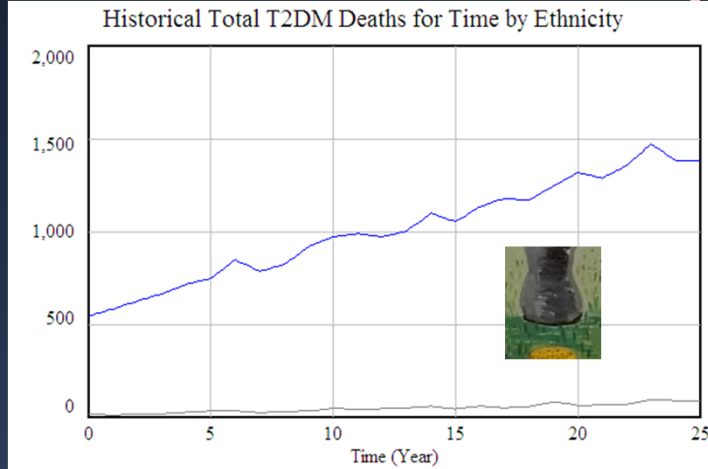
Dynamic Models

- **Simulation models represent hypothesized *causal relationships* between diverse factors**
- **Models provide a provide a way to examine diverse consequences of changes in one area of the system to the whole system**
- **Models help us and system actors to understand**
 - **System vulnerabilities, leverage points**
 - **Ways of fruitfully changing system structure**
 - **Improved ways of working together**

Analogy: Other Simulators to Improve Performance & Lower Risk

- **Pilot decision making: Flight simulators**
- **Climate policy: Climate simulators**
- **Process & power plants: Plant simulators**
- **Driver training: Vehicular simulators**
- **Street design & traffic flow regulation: Traffic simulators**
- **Construction coordination: Construction process simulators**

Pieces of the Elephant: T2DM/ESRD



Department of Computer

Regularities Arise from Underlying *Processes*

- The time series shown are tightly interrelated, not independent
- Many of the features of the time series are driven by the same underlying processes
 - Natural history of infection
 - Demographic change of the population
 - Mechanisms of infection transmission
 - Risk behaviour & risk perception
 - Health system response
- Simulation seeks insight from characterizing *causal structure* of those processes

Understanding Intervention Impact⇒
Seek to Understand *Causal* Relationships

- **Progression of infection**
- **Immunity**
- **Response to treatment**
- **Mixing patterns (e.g. between communities)**
- **Intergenerational/social network mediated effects**
 - **Role modelling & Behavior change emulation**
- **Diversity in contact rates**
- **Strain interaction**
- **Diversity in symptoms**

Simulation Models: Some Uses

- **Make explicit mental models of causality, for discussion and collective refinement**
- **Assist in management of complex situations**
 - **Serve as “What if” tool for identifying desirable policies**
 - Cost-effective/High-leverage/Robust
 - **Understand trends & help make sense of interaction of diverse information, processes**
 - **Prioritizing research/data collection & identifying inconsistencies**
 - **Understanding commonalities between contexts, infection spread**
- **Communication (e.g. “learning labs”)**

Simulation Models as Dynamic Hypotheses

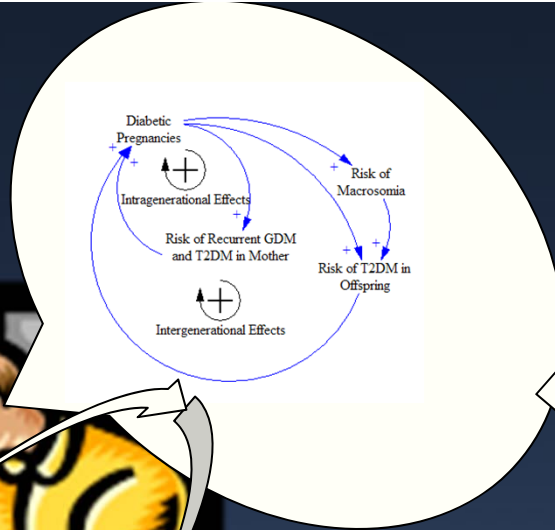
- **Explaining drivers for trends or anticipating intervention impact requires understanding processes underlying observables**
- **A model represents a hypothesis regarding the possible causal interaction of diverse factors often studied in isolation**
 - **Operationally captures a hypothesis for “how the system works” at certain level of description**
- **Model parameters: Detailed assumptions for particular epidemiological contexts**

Coevolution

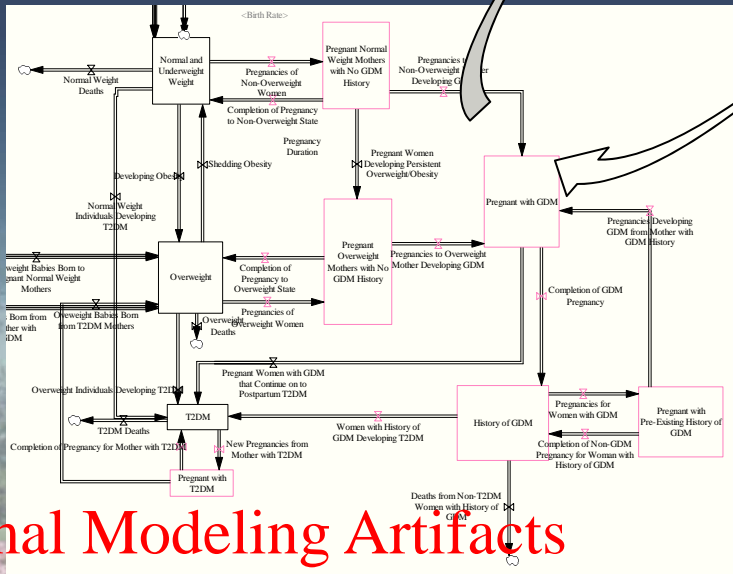
Observation
Evaluation

External World

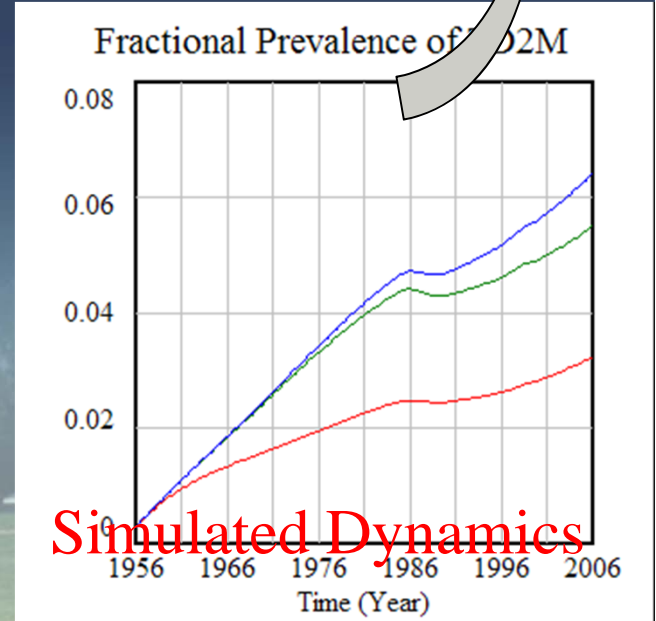
Actions & Choice of
Observations



Mental Model

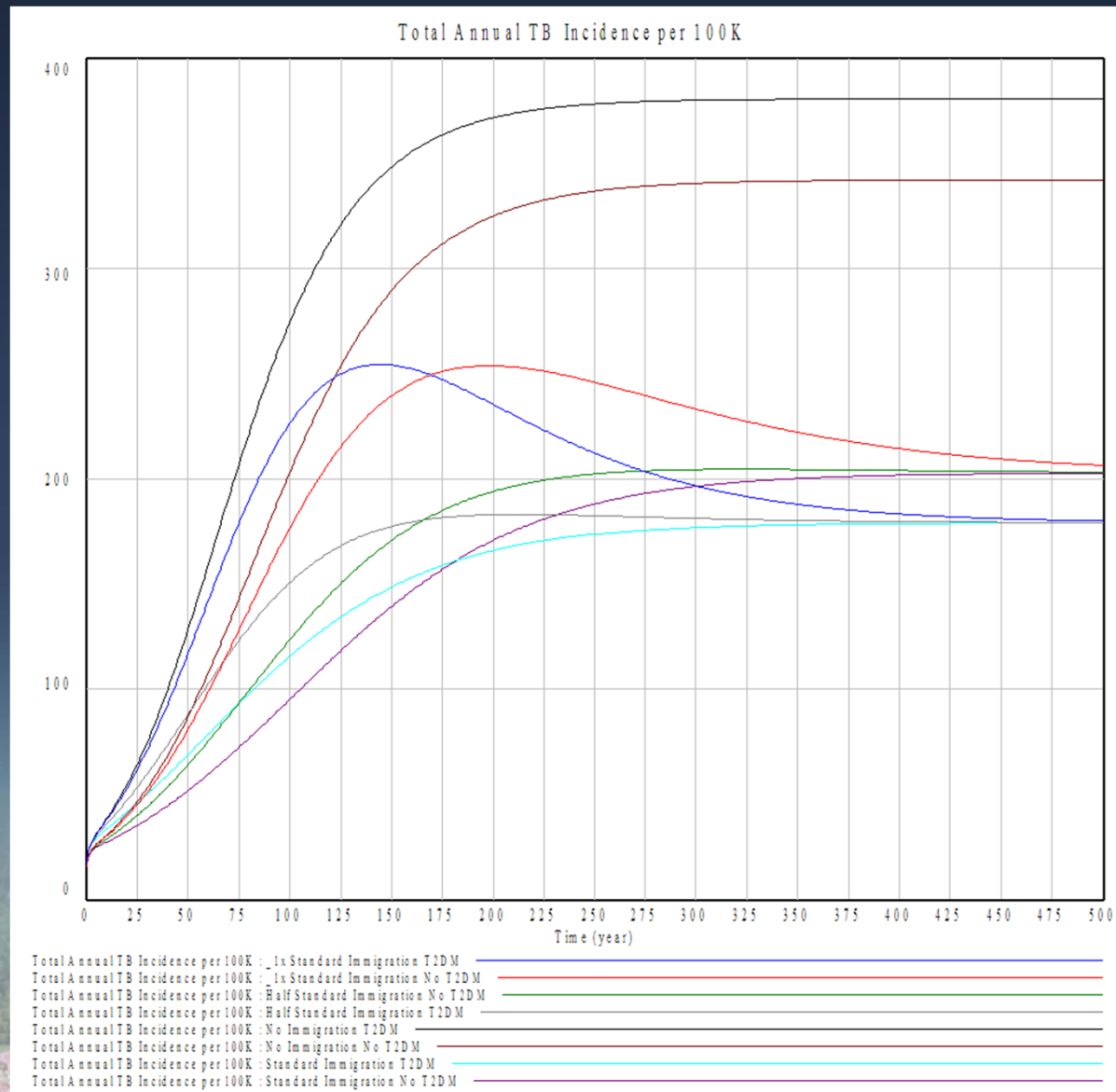


Formal Modeling Artifacts

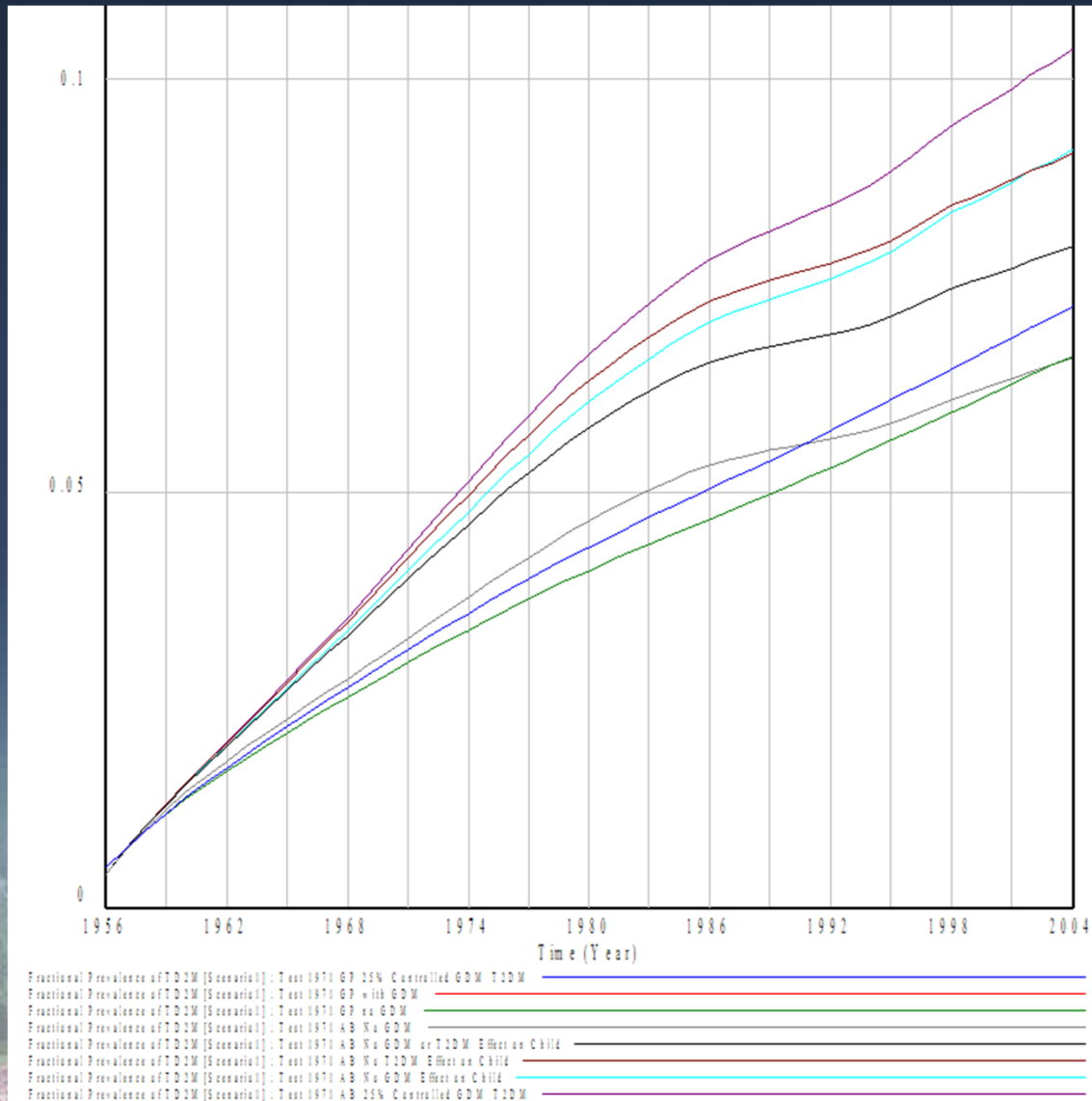


Simulated Dynamics

Scenarios for Understanding How Does X affect System

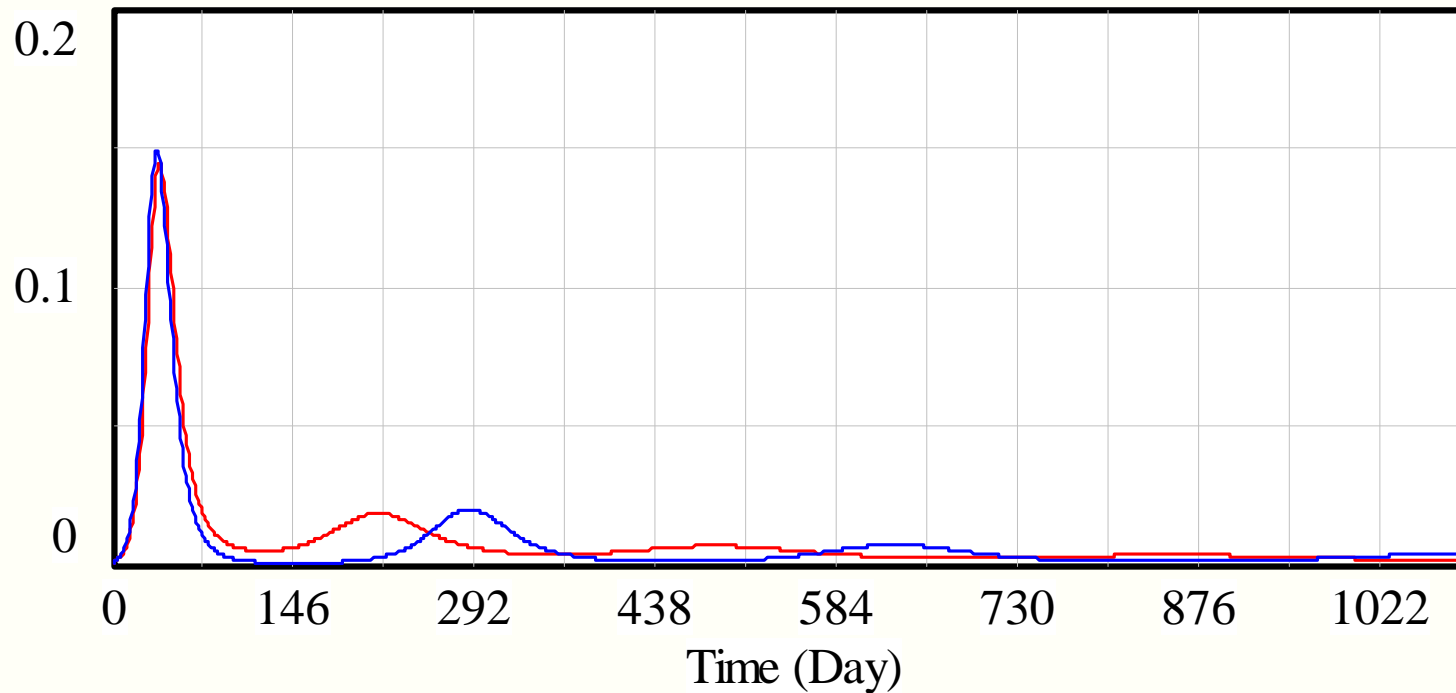


Policy Formulation & Evaluation



Model Can Be Used to Investigate Scenarios

Fractional Prevalence



Fractional Prevalence : Alternative SIR

Fractional Prevalence : Baseline SIR

Examples of Dynamic Modeling Approaches

- **System Dynamics Modeling**

- Feedback-centric modeling approach
- Focuses on feedbacks & accumulations
- Spans qualitative & quantitative methods
- Supports rich mathematical analysis
- Interactive model runs

- **Agent-Based Modeling**

- Captures interactions between individuals within populations
- Captures individual histories & trajectories
- Gracefully represents network connections
- Easier capturing of heterogeneity
- Detailed policy planning

- **Discrete Event Simulation**

Simulates flow of individuals through processes
Captures resource use

