3) On *Practice (Impact) of Research*

Jiming Liu

Chair Professor in Computer Science

Hong Kong Baptist University

- "AI is thought to be an *impossible* dream by many. But ***not*** to us in AI. …the future promises to generate orders of magnitude greater impact than its progress to date."

Raj Reddy
(1994 Turing Award)

Q1: Why Computer Science and AI?

Emerging Diseases: Past *30+ Years*

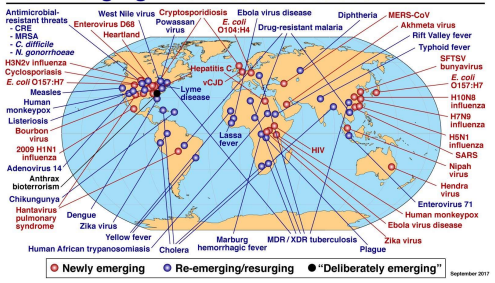
NIAID Director Anthony S. Fauci, M.D., Highlights Lessons

NIH National Institute of Allergy and Infectious Diseases

"*Three Decades* of Responding to Infectious Disease Outbreaks" (NIH/NIAID, 2017)

1985:

2017: Global Examples of Emerging and Re-Emerging Infectious Diseases



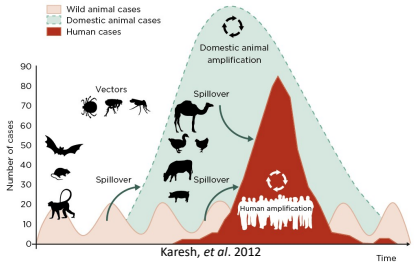
NIH/National Institute of Allergy and Infectious Diseases (NIAID), 14 Nov. 2017

Covid-19

Covid-19 ←…← Zoonotic Viruses?

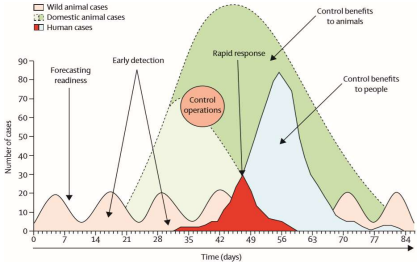
Reservoirs: *Animal, Human, & Environment*

Transmission Modeling



Karesh, *et al.* 2012

Response Planning



Neglected Tropical Diseases (NTD): Past *3000+ Years*

Source: Britannica

- According to *WHO World **Malaria** Report **2019***
  - ~50% of the world's population at risk
  - ~228 million cases (artemisinin-resistance!!!)
  - **~405 000 deaths** in 2018 (**67% under 5 years**)

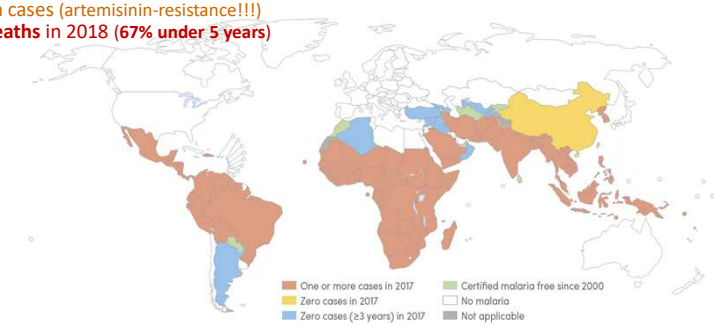
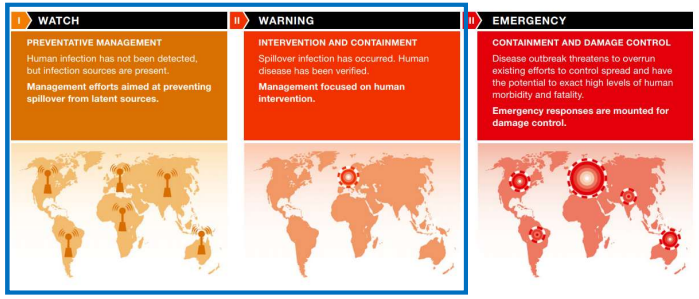


Image Credit: World Malaria Report 2018

Published online: May 11, 2016
EMBO reports

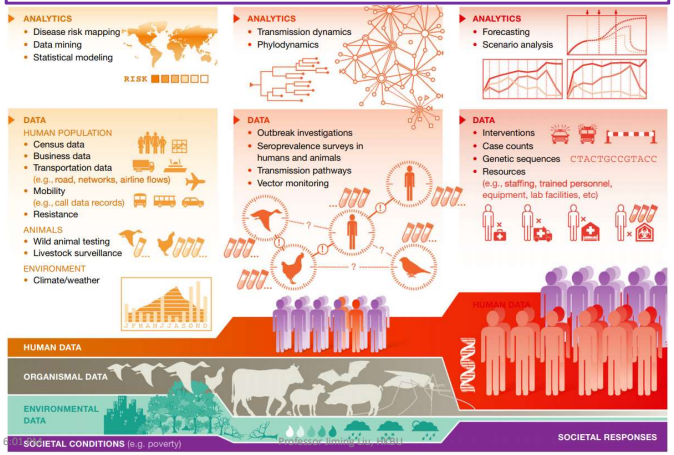
Infectious disease intelligence *Barbara A Han & John M Drake*



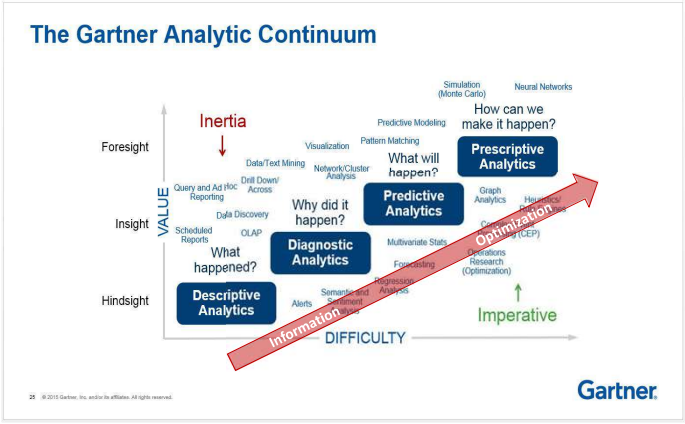
Data

- **CDC and/or clinical data**
- **Weather or climate data**
- **Census data**
- **Geospatial data** (to superimpose hundreds of feeds like terrain, land use, or transportation for situational awareness in complex emergencies)
- **Drug Administration data** (e.g., mobile app reporting, to detect adverse events and medication errors)
- **Tweets** (to extract the number of people who were hospitalized or sick from air/vector/water/food-borne illnesses

Using big data is not just to find something interesting for the sake of discovery, but to find something interesting that is **actionable** at scale.



The Gartner Analytic Continuum



Computer Science and AI

- Abstractions & representations
- Theoretical / computational characterization
- Algorithmic thinking & problem solving
- Data
- Realities

To See the Unseen

Q1: *Why Computer Science and AI?*

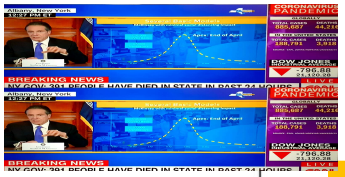
To Dream the Possible Dream

Q1: *Why Computer Science and AI?*

Q2: What and How?

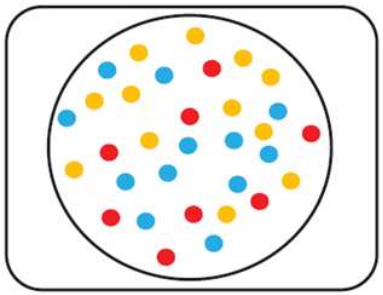
To See the Unseen:
How Disease Transmits

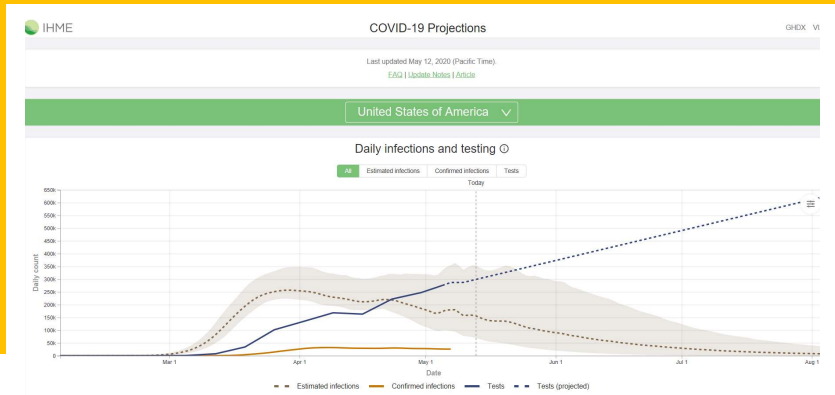
Infection Dynamics



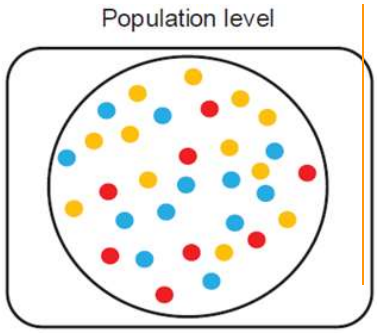
Predicting the **Future**

Population level





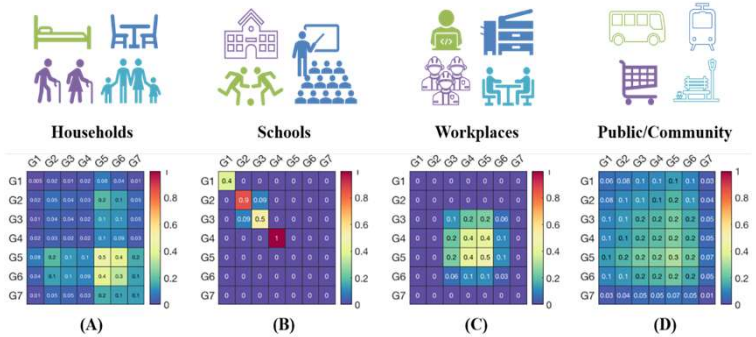
Limits of Existing Models



Unsolved Puzzle

- How disease transmits among **different age-groups**, in major **social settings**
  - How **future risks** and trends may evolve
  - What are the right distancing **strategies**
  - When will be SAFE to **reopen**

Close Contacts, and *Setting*-Related Transmissions?

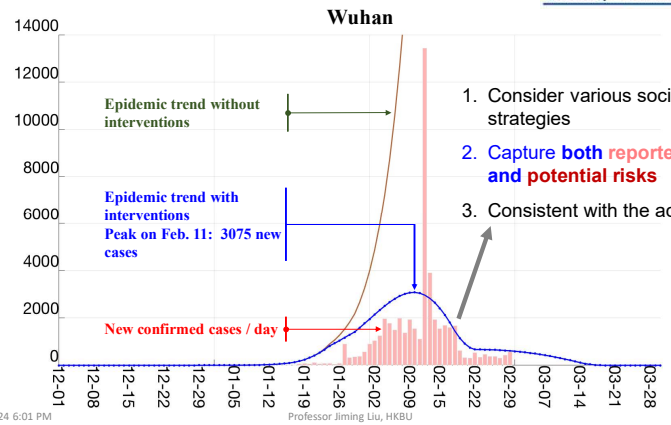


Age Groups
G1: 0-6;
G2: 7-14;
G3: 15-17;
G4: 18-22;
G5: 23-44;
G6: 45-64;
G7: >= 65

Observation: Initial COVID-19 transmissions in Wuhan took place mainly in **households** and **public places**

Prediction: Seeing the *Unseen Risks*

EClinicalMedicine
Published by THE LANCET

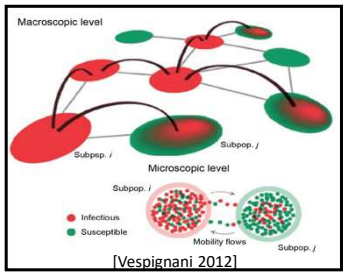


1. Consider various social distancing strategies
2. Capture **both** reported cases **and potential risks**
3. Consistent with the actual situations

Key to Unveiling *Meta-Population* Transmission

- To understand, predict, and control epidemic dynamics by characterizing **age-specific** or **spatial *sub-populations***

$$I_{t+1} = \mathbb{K}_t I_t = g(S_t B \boldsymbol{C} A) I_t$$



[Vespignani 2012]

S_t: Susceptible population
B: Infection acquiring rate
***C*: Contact matrix**
A: Infection transmission rate

Contact: Individuals' mutual exposure in the same physical environment

By Yang, Pei, Xia, & Liu, *et al.*

Unveiling Hidden Diffusion Networks

FIGURE 5. Traditional method and inverse engineering analysis method for modeling, analyzing, and inferring of infectious disease. (A) Traditional method: modeling and analysis of infectious disease by given contact network. (B) Inverse engineering: inferring dynamic social contact patterns using temporally observed incidences.

Q1: Why Computer Science and AI?

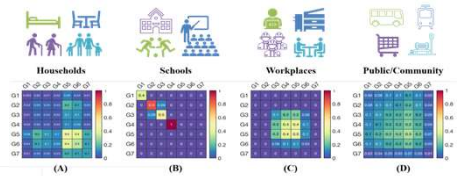
To Dream the Possible Dream

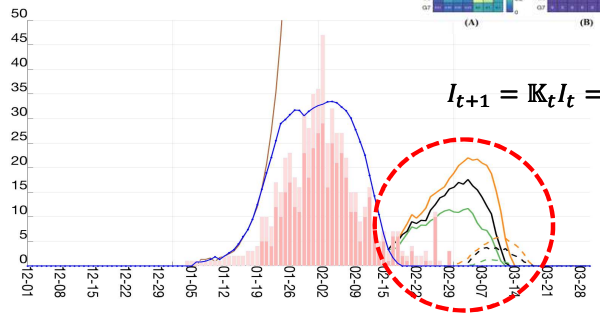
To Dream the Possible Dream: *Computer Science for Safety*

Unsolved Puzzle

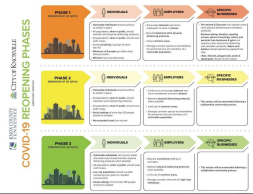
- How disease transmits among **different age-groups**, in major **social settings**
  - How **future risks** and trends may evolve
  - What are the right distancing **strategies**
  - When will be **SAFE to reopen**

Science of *Reopening*

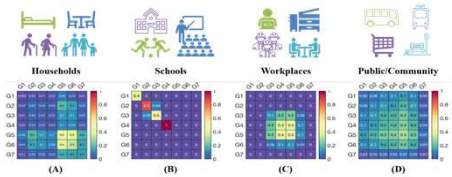


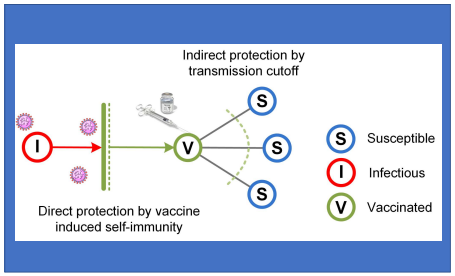


$$I_{t+1} = \mathbb{K}_t I_t = g(S_t B C A) I_t$$



On *Herd Immunity*





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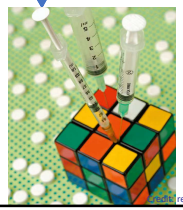
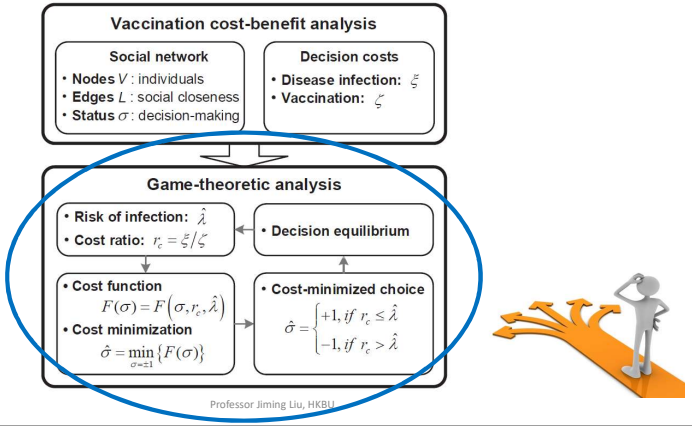


Essence of Herd Immunity: *Individual Decisions*



To Dream the Possible Dream:
AI for the "Last Mile"

Neglected Tropical Diseases (NTD): Past *3000+ Years*

Source: Britannica

- According to *WHO World **Malaria** Report **2019***
  - ~50% of the world's population at risk
  - ~228 million cases (artemisinin-resistance!!!)
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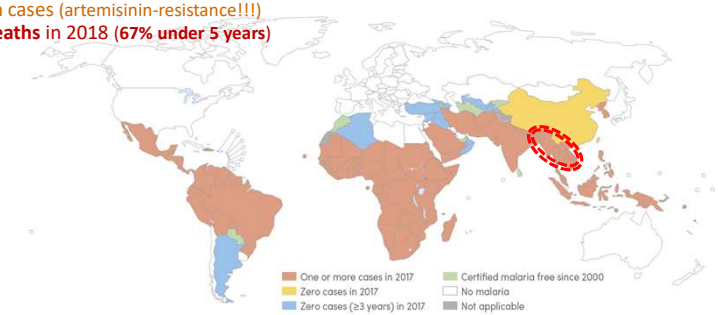
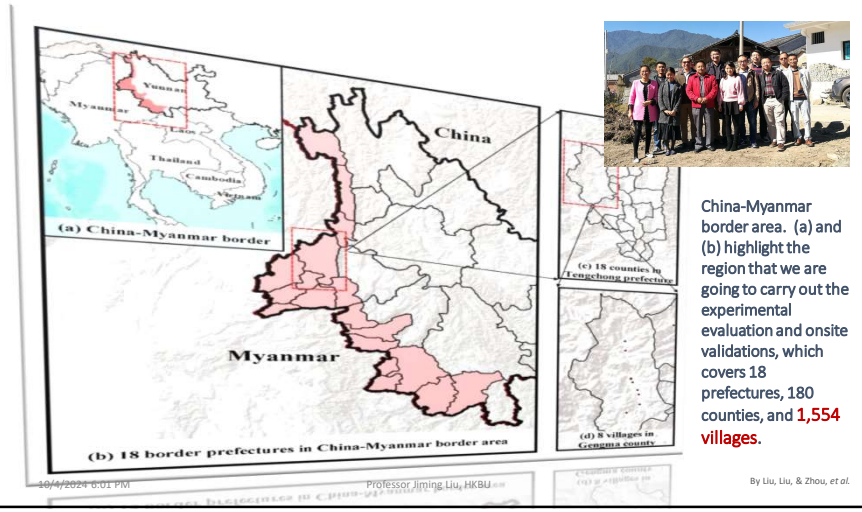


Image Credit: World Malaria Report 2018



China-Myanmar border area. (a) and (b) highlight the region that we are going to carry out the experimental evaluation and onsite validations, which covers 18 prefectures, 180 counties, and 1,554 villages.

By Liu, Liu, & Zhou, *et al.*

… in *Disease Control*

Abstractions & representations

Theoretical / computational characterization

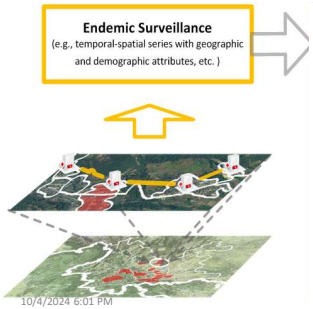
Algorithmic thinking & problem solving

Data

Realities

Machine Learning: To Reveal *What's Happening*

- **Endemic Surveillance**
  - Spatio-temporal observations of infections and environmental attributes
- **Inferring the Underlying Diffusion Network**



Abstractions & representations

Theoretical / computational characterization

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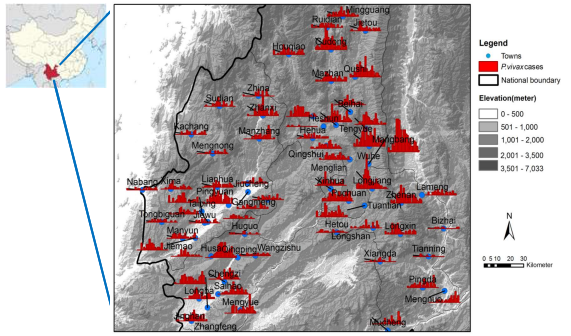
Data

Realities

Spatio-Temporal Risk Mapping

- **Malaria** in Yunnan Province, China



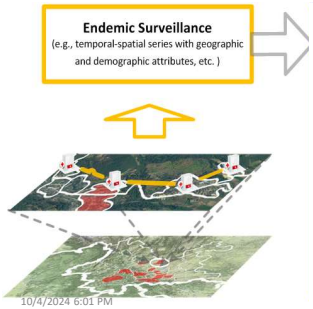
***P. vivax* incidences of 62 towns in Yunnan, China**

By Benyun Shi, Jiming Liu, and Xiao-Nong Zhou, *et al.*

… in *Machine Learning*

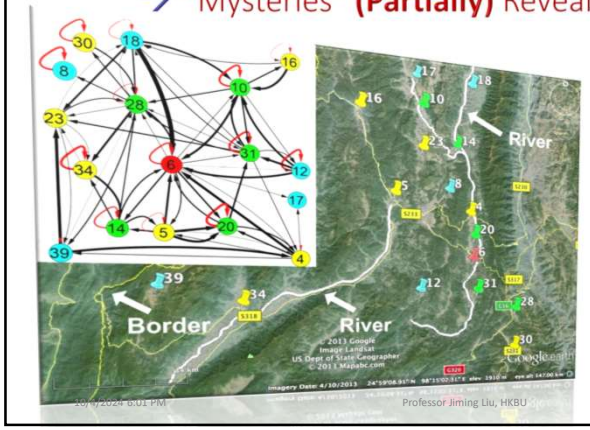
Machine Learning (ML): To Reveal *What's Happening*

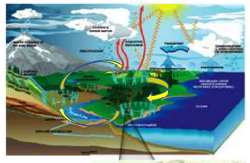
- **Endemic Surveillance**
  - Spatio-temporal observations of infections and environmental attributes
- **Inferring the Underlying Diffusion Network**



Diffusion Network

→ "Mysteries" **(Partially)** Revealed!

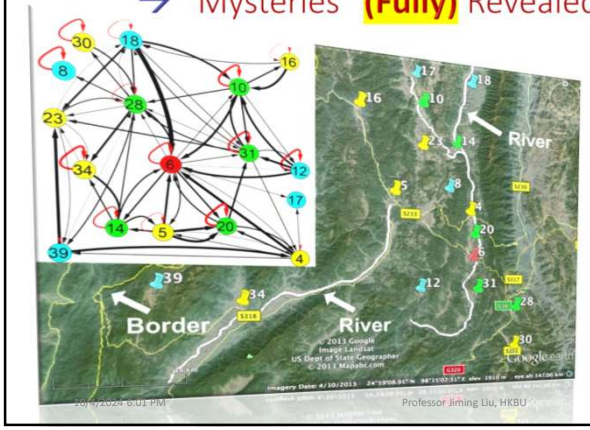


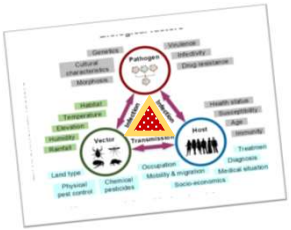


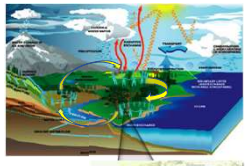
- Host population
- Vector
- Environmental factors

Diffusion Network

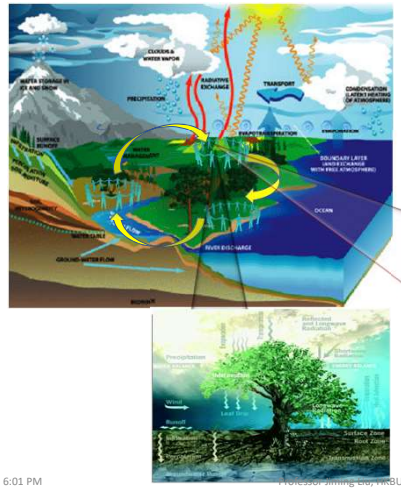
→ "Mysteries" **(Fully)** Revealed?





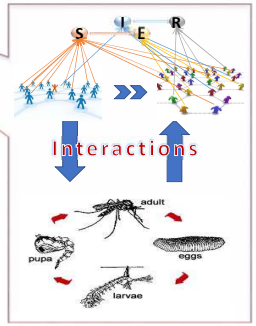


- Host population
- Vector
- Environmental factors

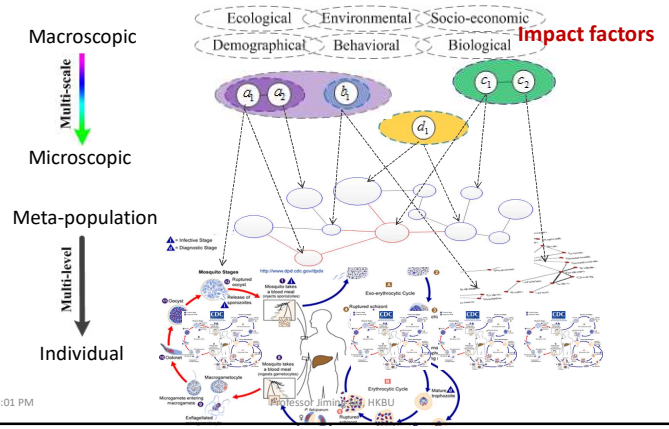


Dynamics

- Meta-population
- Vector: Extended Ross-Macdonald model
- Environmental factors
- Human mobility



Multi-Level Models & Multi-Scale Impacts



- Abstractions & representations
- Theoretical / computational characterization
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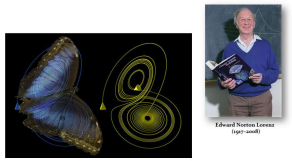
Q3: Multiple *Scales* are *Inter-Related*… How?

Butterfly Effect



Complex Spatio-Temporal *Dependencies*

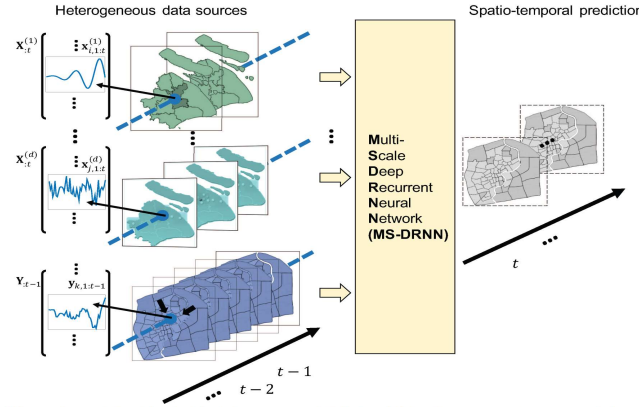
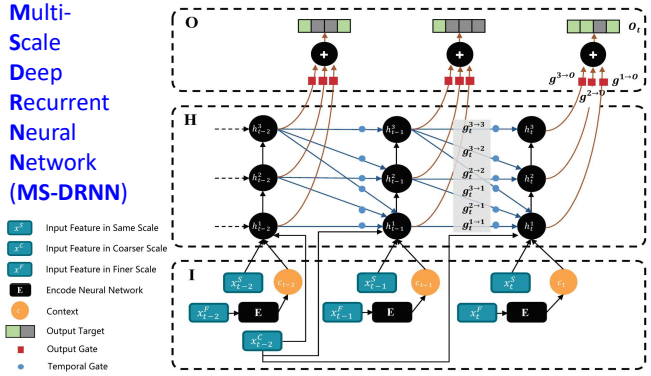


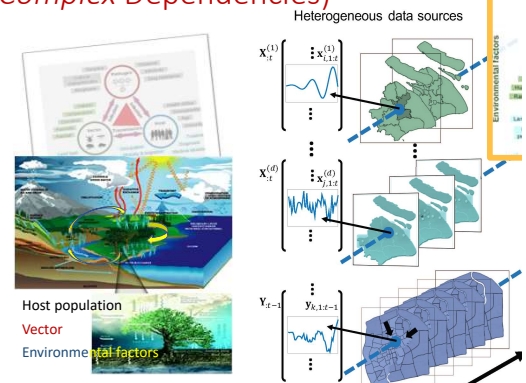
Illustration of the idea behind the proposed MS-DRNN. Given a spatio-temporal dataset with the target variable $\mathbf{Y}_{:t-1}$ and multiple covariates $\mathbf{X}_{:t}^{(1)}, \cdots, \mathbf{X}_{:t}^{(d)}$ observed from d heterogeneous data sources, MS-DRNN aims to integrate the data from various sources and capture the complex dependencies among them for making predictions on $\mathbf{y}_t$.

Multi-
Scale
Deep
Recurrent
Neural
Network
(MS-DRNN)



1) Encoder and decoder structures in module **I** integrates *heterogeneous* data.
2) Hierarchical structure in module **H** captures *multiple spatio-temporal effects* on target variable caused by covariates from different sources.
3) Integrative *effects at varying scales* in module **O** generate predictions.

ML Challenge: *Unseen Interactions*
(*Complex* Dependencies)



Abstractions & representations

Theoretical / computational characterization

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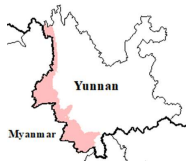
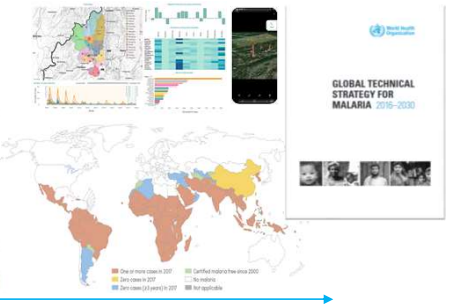
Global Malaria Eradication, by 2030



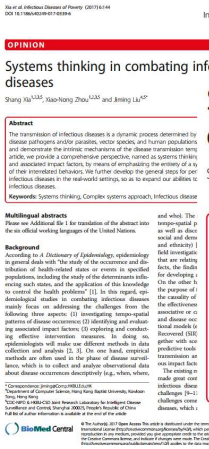


Yunnan, China

SEA and GMS Countries

Global Eradication

Image Credit: Google Maps



Infectious Diseases of Poverty

OPINION Open Access

Systems thinking in combating infectious diseases

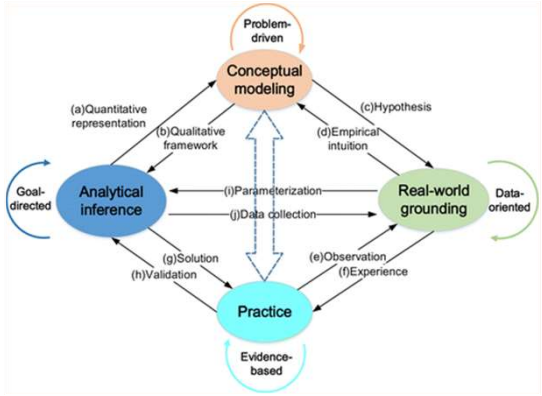
Shang Xia[1,2,3,5], Xiao-Nong Zhou[1,2,3,5] and Jiming Liu[4,5*]

Abstract

The transmission of infectious diseases is a dynamic process determined by multiple factors originating from disease pathogens and/or parasites, vector species, and human populations. These factors interact with each other and demonstrate the intrinsic mechanisms of the disease transmission temporally, spatially, and socially. In this article, we provide a comprehensive perspective, named as systems thinking, for investigating disease dynamics and associated impact factors, by means of emphasizing the entirety of a system's components and the complexity of their interrelated behaviors. We further develop the general steps for performing systems approach to tackling infectious diseases in the real-world settings, so as to expand our abilities to understand, predict, and mitigate infectious diseases.

Keywords: Systems thinking, Complex systems approach, Infectious disease control

Complex Systems Approaches



Problem-Oriented Modeling/Learning