



# **Modeling Infectious Diseases: Projecting Spread, Evaluating Interventions, and Resource Allocation**

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**Outreach**

Education  
Collaboration  
Sustainable Health & Well-Being  
Projects  
Disease Modeling  
Medical Decision-Making  
Resource Allocation



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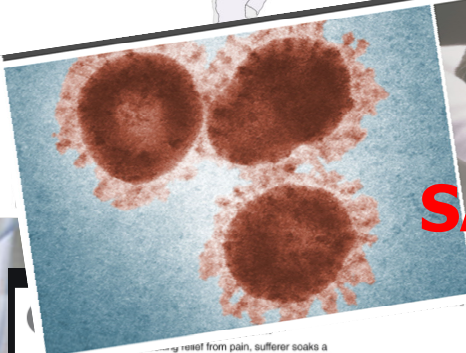
### Malaria incidence (per 1,000 population at risk), 2015

Incidence of malaria is the number of new cases of malaria in a year per 1,000 population at risk. SDG Target 3.3 is to end the epidemic of malaria.

Our World in Data

World

**Malaria**

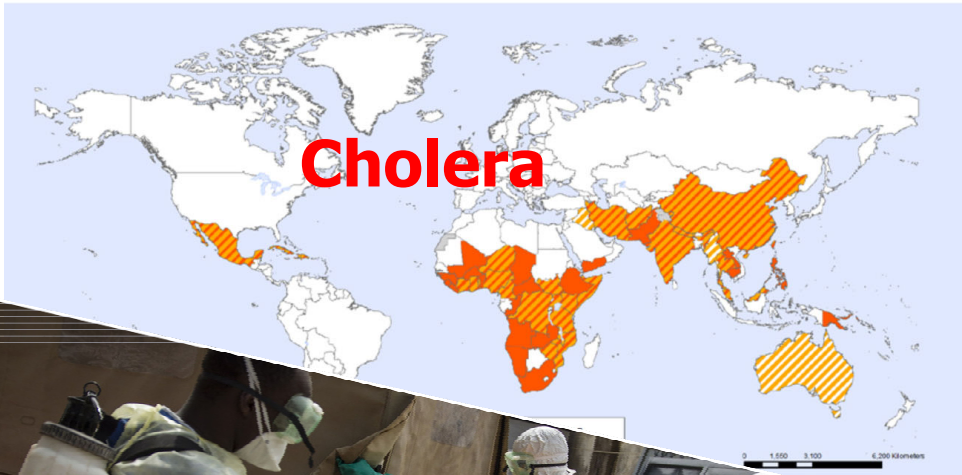


**SARS**



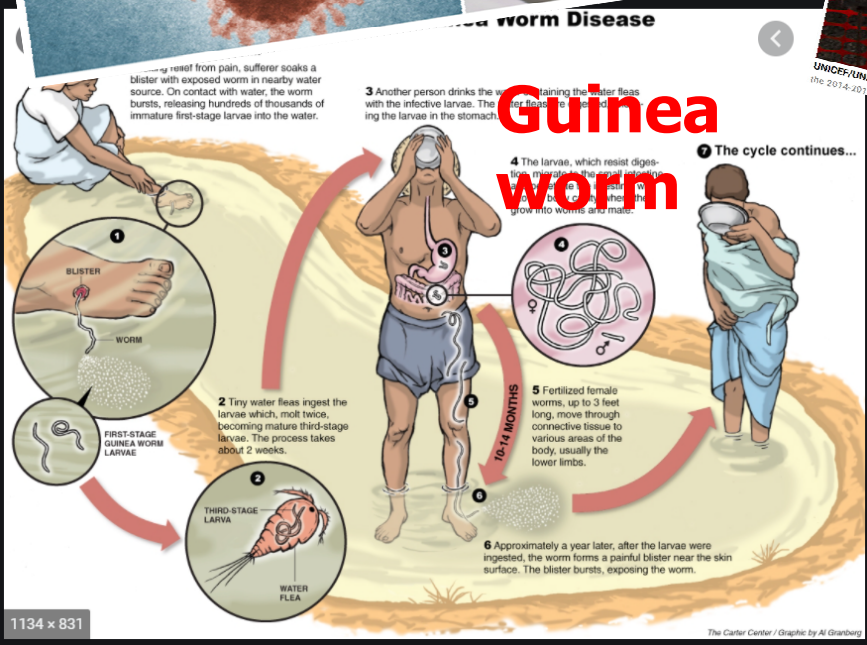
### Countries reporting cholera, 2010–2015

**Cholera**



**Ebola**

**Guinea worm**



**Pandemic flu**

New Contributed Photographs Collection/Otis Historical Archives/National Museum of Health and Medicine



*Source: Nigeria Health Online, 2016*

## Prevention

Insecticide nets  
Indoor residual spraying  
Vaccines



*Source: Medecins Sans Frontieres, 2015*

## Surveillance

Monitoring of confirmed malaria cases



*Source: Making Malaria History, 2014*

## Treatment

Rapid diagnostic tests and  
artemisinin-based combination  
therapy

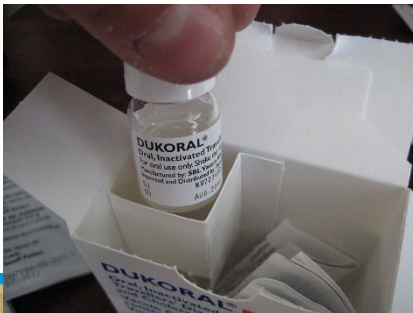




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## Prevention

Insecticide nets  
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Source: Medecins Sans Frontieres, 2015

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**STOP CHOLERA**

**WHAT IS CHOLERA?**

CHOLERA is a water-borne disease, spread by contaminated food or water.

Cholera causes acute watery diarrhoea (watery stool), and if left untreated, it can lead to DEATH within hours.

Other Symptoms Include: Vomiting.

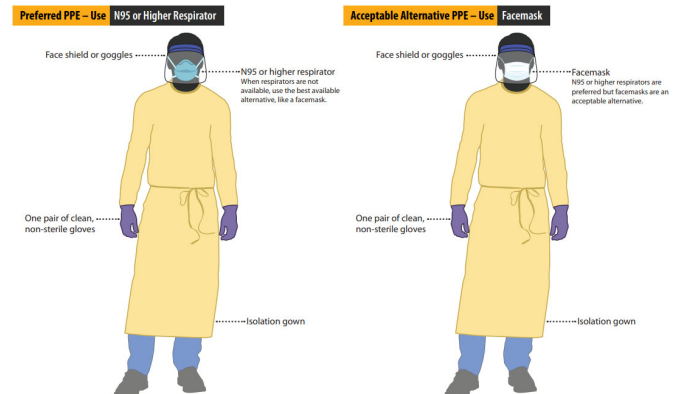
**Make Water Safe**

**TO PREVENT CHOLERA**

- Use water from reliable sources.
- Boil water before drinking.
- Store water in properly sealed containers.
- Ensure bottled water is properly sealed before you drink.

IF YOU EXPERIENCE SUDDEN DIARRHOEA (WATERY STOOL), VISIT A HEALTH CARE FACILITY IMMEDIATELY.

NGC



Credit: WHO/Marco Kokic



Source: Carter Center

With the assistance of his colleagues, Wegwa Odol Othow (yellow shirt) measures a pond for application of a safe larvicide that helps stop the Guinea worm life cycle.



[www.cdc.gov/coronavirus](http://www.cdc.gov/coronavirus)



# Disease Models → Decision-Making

- Understanding and projecting the disease spread
- Evaluating the impact of intervention strategies
- Estimating resource needs
- Resource planning and allocation

Geographically and over time,  
by sub-populations

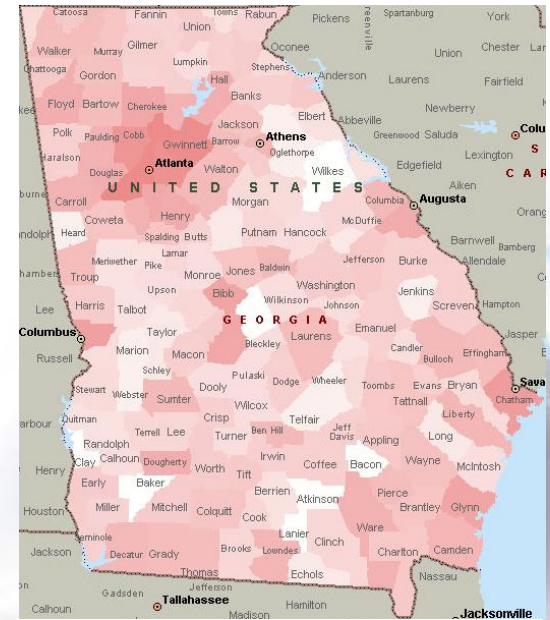


Figure shows projected disease spread of flu after 90 days (Keskinocak/Swann team)

Influenza

Cholera

Guinea  
worm

Malaria

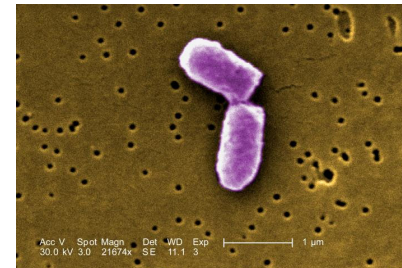
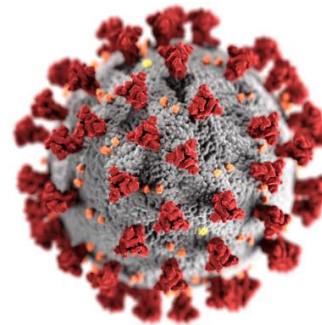
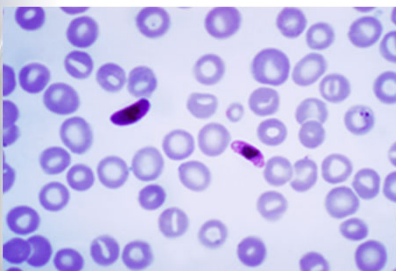
Polio

# Infectious Disease Modeling

Disease progression  
in an individual –  
Natural history

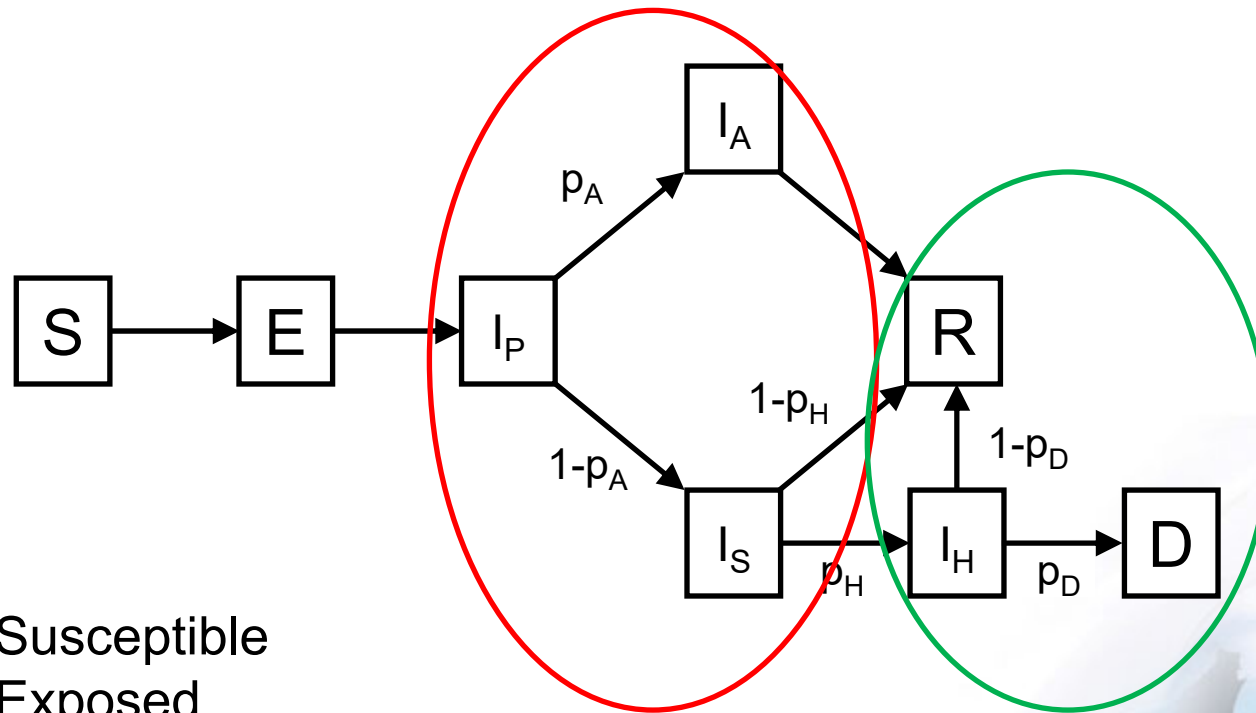


Disease spread



<https://www.cdc.gov/media/subtopic/images.htm>  
<https://www.cdc.gov/fungal/antifungal-resistance.html>  
[https://en.wikipedia.org/wiki/Plasmodium\\_falciparum](https://en.wikipedia.org/wiki/Plasmodium_falciparum)

# Natural history – Example: Covid19



S: Susceptible  
E: Exposed  
 $I_P$ : Presymptomatic  
 $I_A$ : Asymptomatic  
 $I_S$ : Symptomatic  
R: Recovered  
 $I_H$ : Hospitalized  
D: Dead

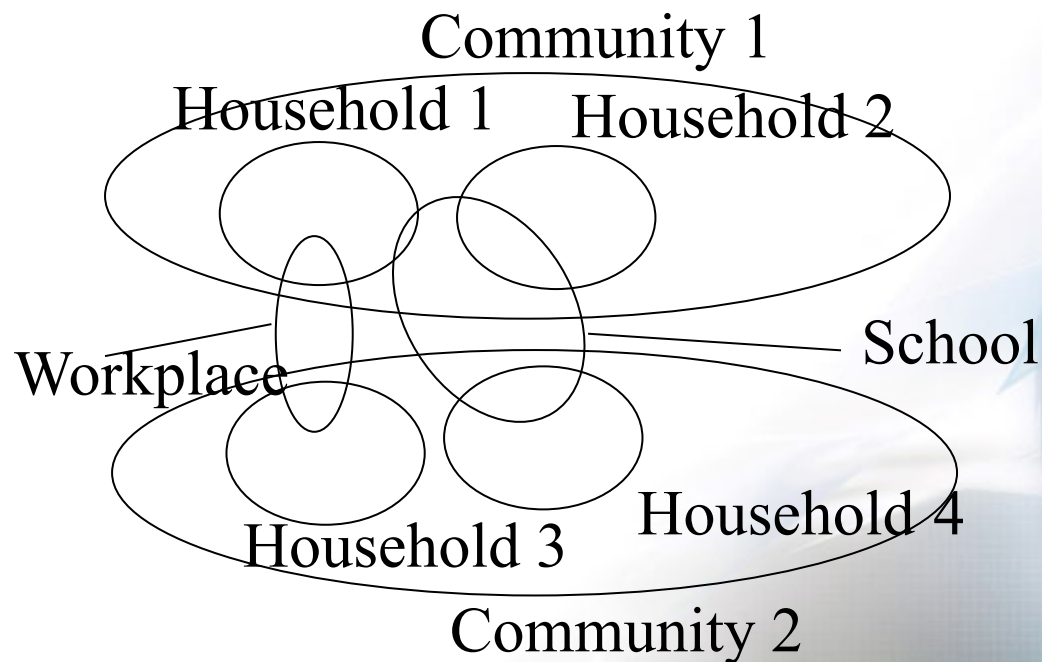
S-E-I-R

S-I-R



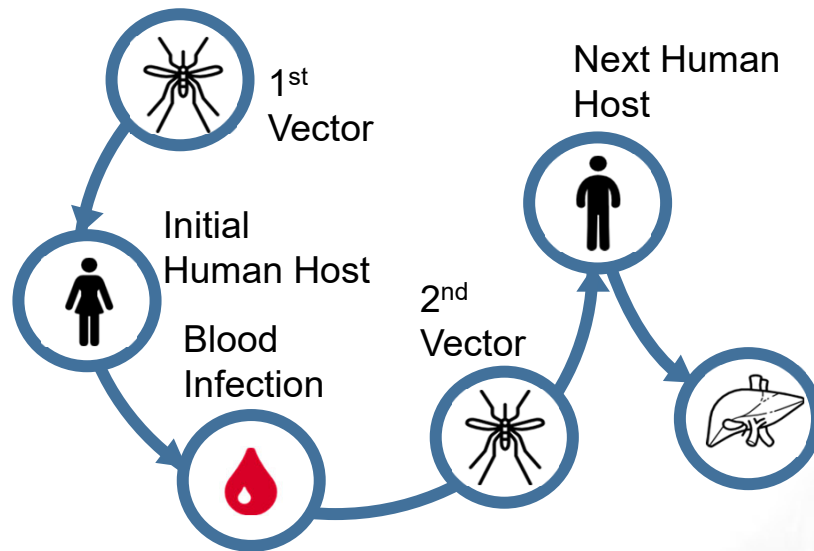
# Disease spread – Example: Covid19

- Households
- Peer groups (e.g., workplace and schools)
- Community
- Household statistics, classroom sizes, age statistics
- Mobility & interaction patterns, e.g., workflow data



# Disease spread - Transmission

- Human-to-human
  - Respiratory diseases (e.g., flu, Covid-19), STDs
- Vectors/Animals
  - Malaria, Guinea Worm



# Disease spread – Example: Malaria

## Mosquito

- Natural death rate
- Percent of the population that bites per day
- Duration in the incubation stage
- Probability of a mosquito contracting malaria from a human in the incubation and infection stage
- Probability of contracting malaria from an asymptomatic person
- ...

## Human

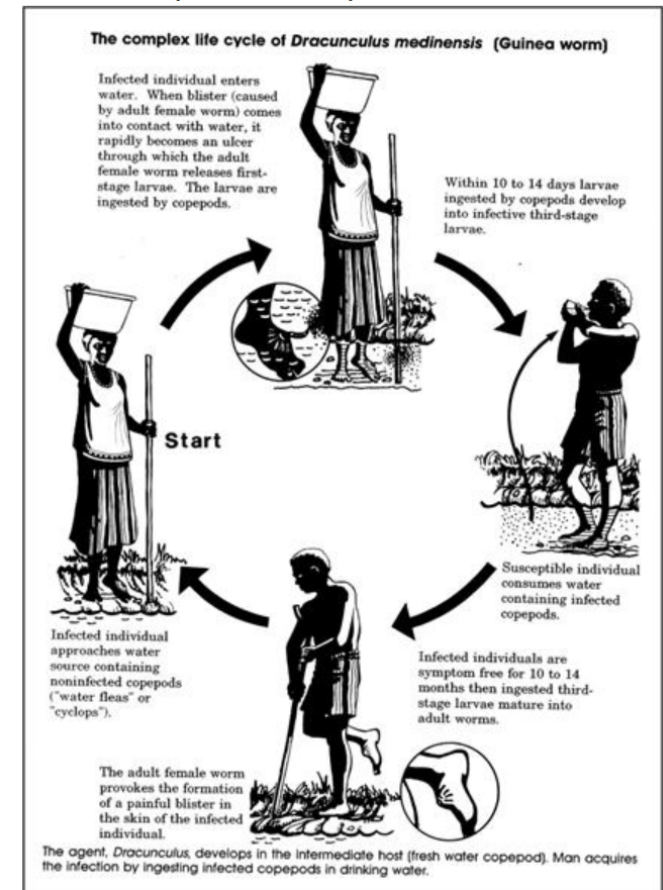
- Probability of contracting malaria from an incubating or infectious mosquito
- Duration in the incubation stage for each age group
- Probability of transitioning from slow recovery to immunity / fast recovery
- Lag from incubation to symptomatic, infectious
- Recovery rates
- ...

- Environmental or other risk factors

# Disease spread - Transmission

- Human-to-human
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Figure 1.20 Complex Life Cycle of *Dracunculus medinensis* (Guinea worm)

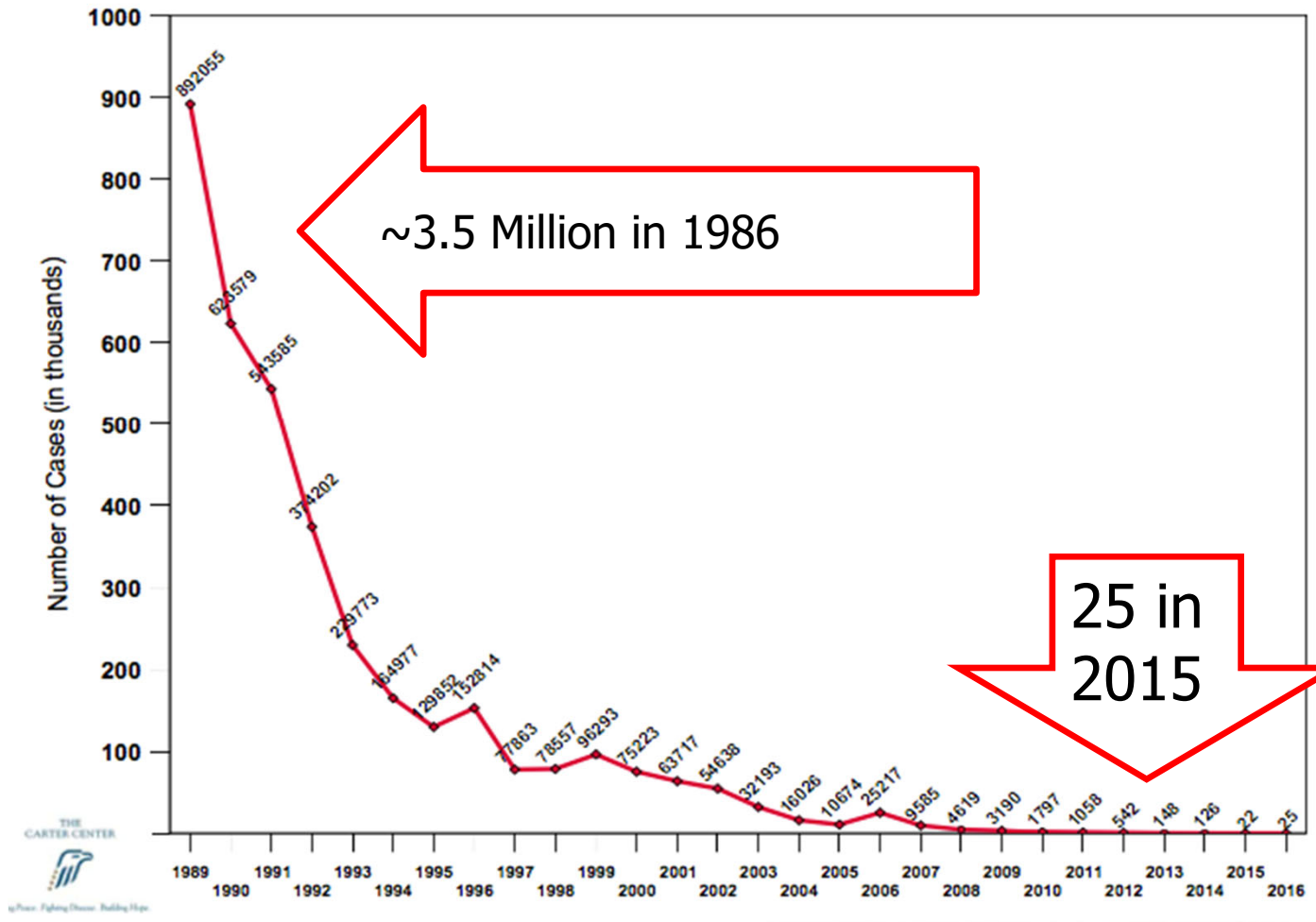


## [Image Description](#)

Source: Centers for Disease Control and Prevention. *Principles of epidemiology*, 2nd ed. Atlanta: U.S. Department of Health and Human Services; 1992.

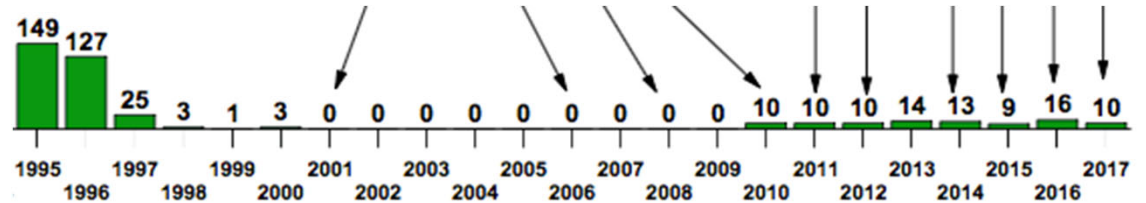
# Guinea Worm Disease

Number of Reported Cases of Guinea Worm Disease by Year: 1989 -2016

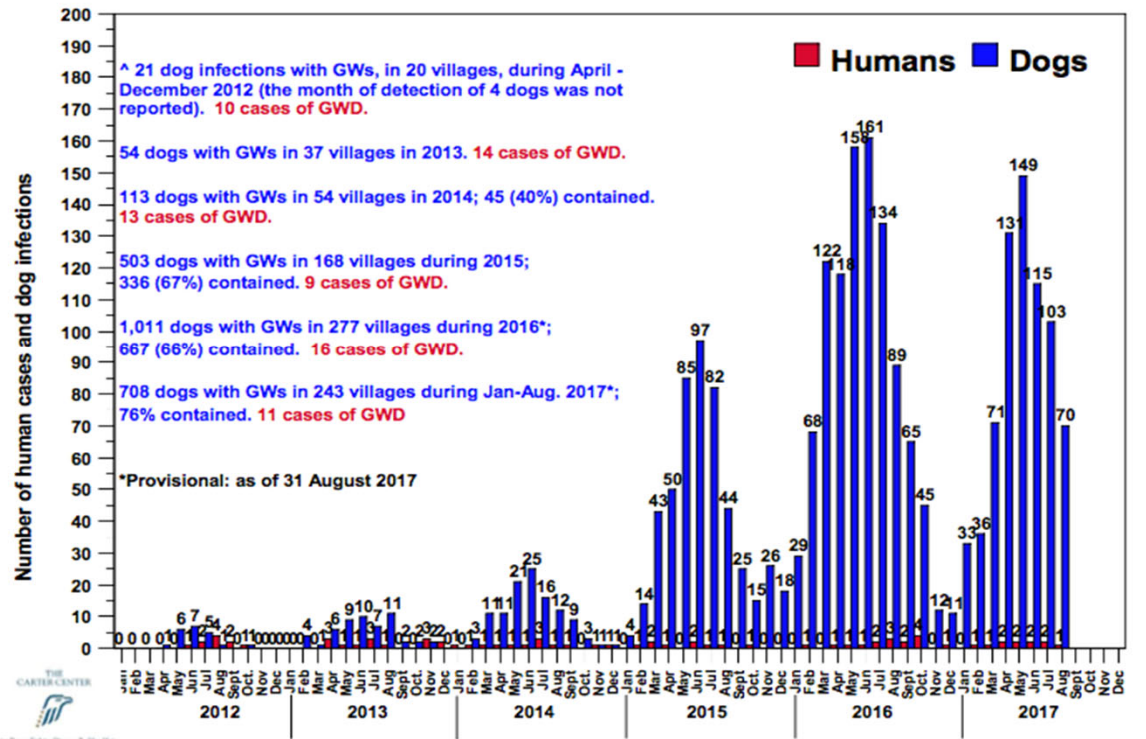


# Guinea Worm Disease in Chad

Zero Human Cases in Chad  
For 9 years

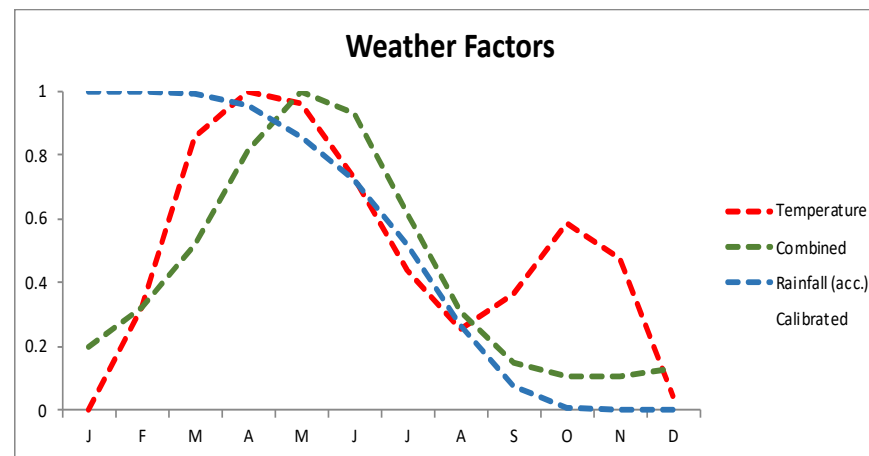


CHAD GUINEA WORM ERADICATION PROGRAM  
FREQUENCY OF DRACUNCULIASIS AMONG HUMANS AND DOGS BY MONTH DURING 2012\*-2017\*



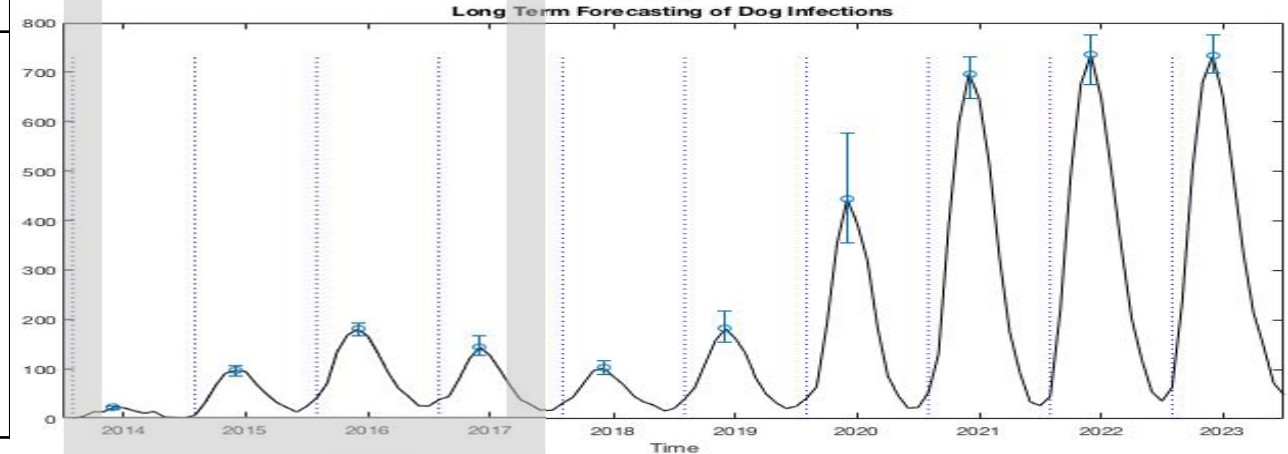
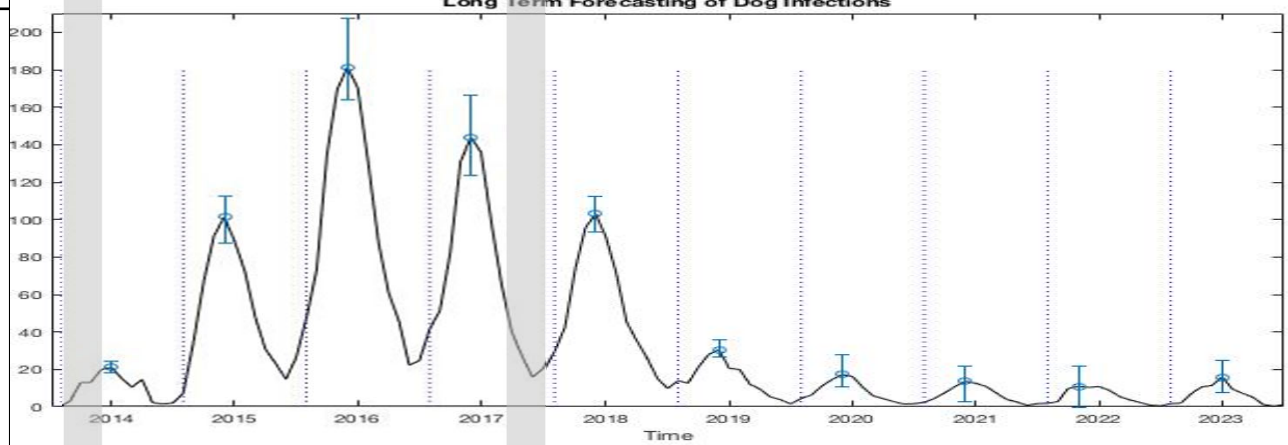
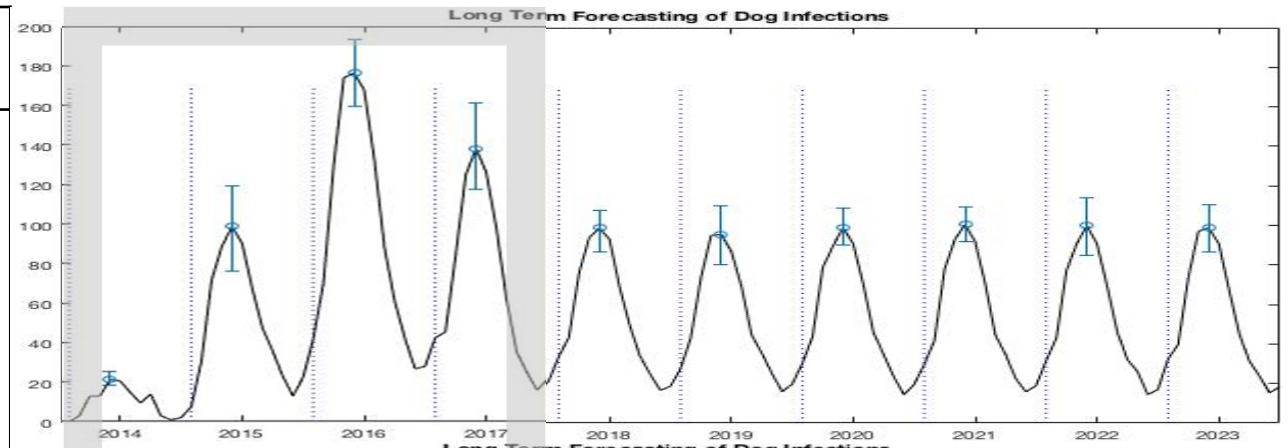
# Guinea Worm Transmission Model

- Agent-based model
- Environmental factors: temperature & rainfall → Worm burden in water → Rate of infection



<https://www.ajtmh.org/view/journals/tpmd/103/5/article-p1942.xml>

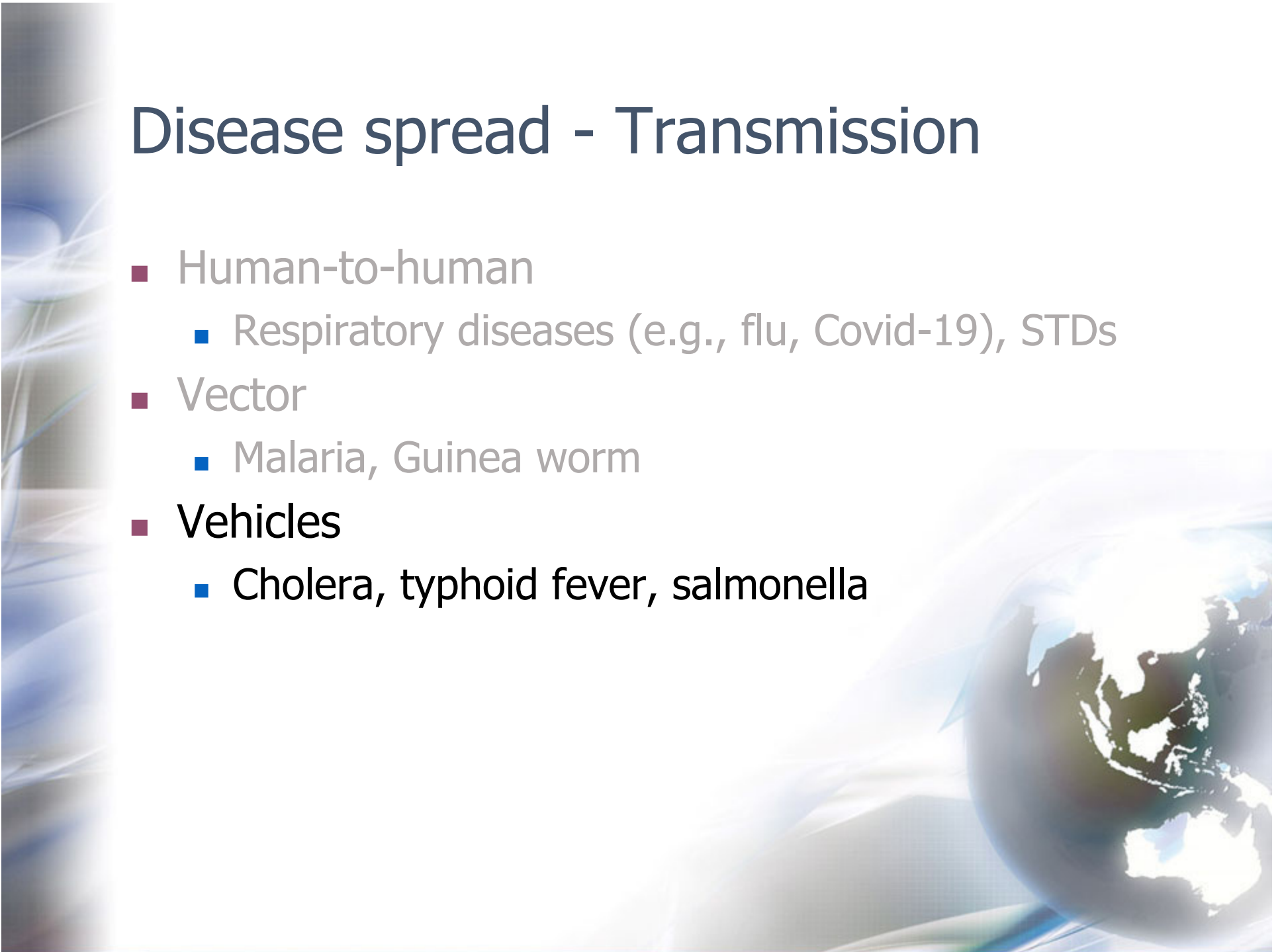
Scenario:	<i>ABATE</i>	<i>Tether</i>	<i>Other</i>
<b>Baseline</b>	40%	76%	17%
<b>Increased</b>	70%	95%	17%
<b>Decreased</b>	20%	50%	17%





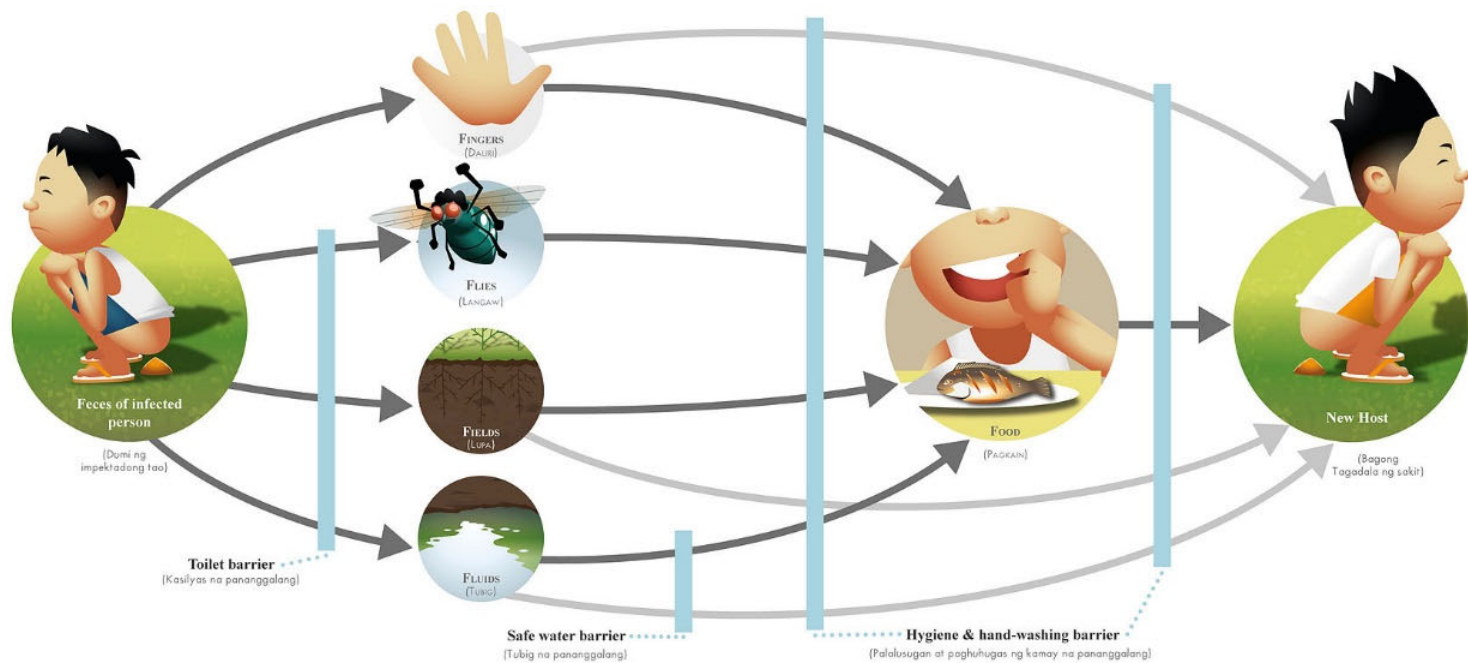
# Disease spread - Transmission

- Human-to-human
  - Respiratory diseases (e.g., flu, Covid-19), STDs
- Vector
  - Malaria, Guinea worm
- Vehicles
  - Cholera, typhoid fever, salmonella



# Cholera Transmission

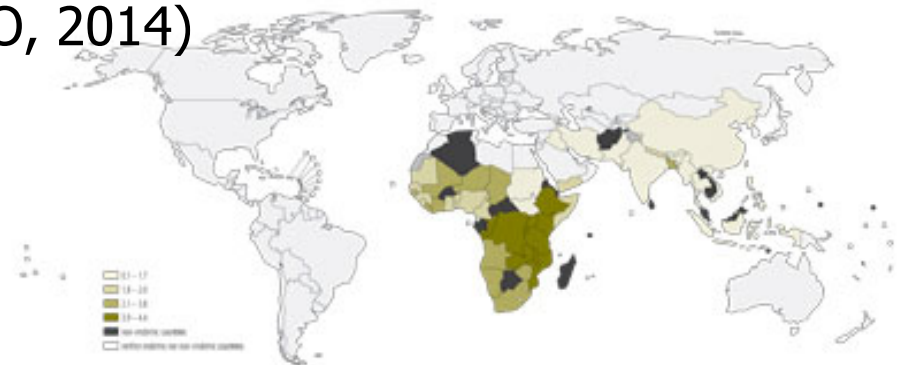
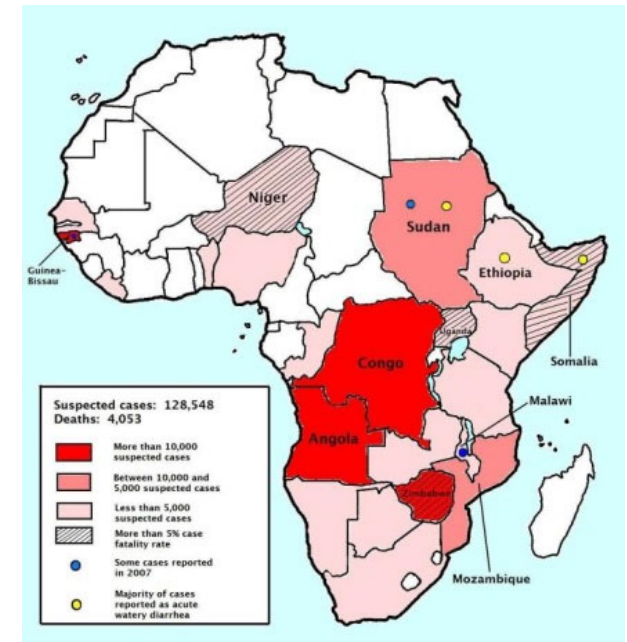
- Oral-Fecal Pathway
  - Five F's: Feces, Fingers, Flies, Fields, Fluids, Food



"F-diagram-01" by UNICEF Philippines and Luis Gatmaitan / 2014 / Gilbert F. Lavidés - <https://www.flickr.com/photos/gtzeccosan/17125224489/in/set-72157648282032913>. Licensed under CC BY 2.0 via Wikimedia Commons - <https://commons.wikimedia.org/wiki/File:F-diagram-01.jpg#/media/File:F-diagram-01.jpg>

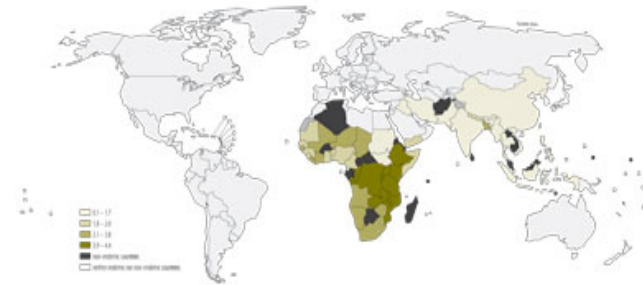
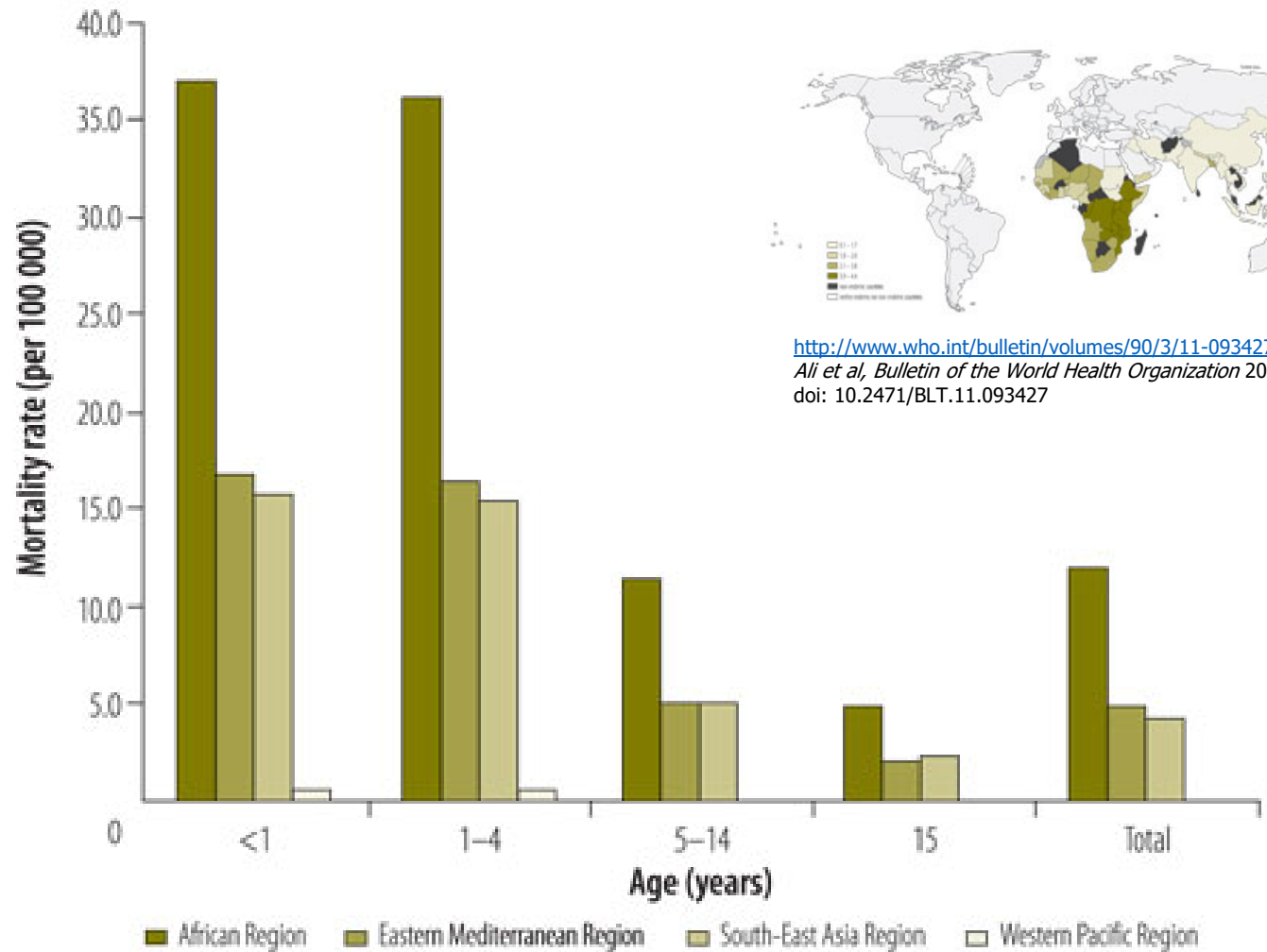
# Cholera Worldwide

- 7 major pandemics in recorded history
- Many notable outbreaks since 1991. Examples:
  - 2008: Zimbabwe
  - 2010: Nigeria, Haiti, Dominican Republic
  - 2014: Ghana
- Cholera endemic in many places
- Overall, 1.4-4.3 million cases of cholera per year, leading to 28,000-142,000 deaths (WHO, 2014)



<http://www.who.int/bulletin/volumes/90/3/11-093427/en/>  
Ali et al, *Bulletin of the World Health Organization* 2012;90:209-218A.  
doi: 10.2471/BLT.11.093427

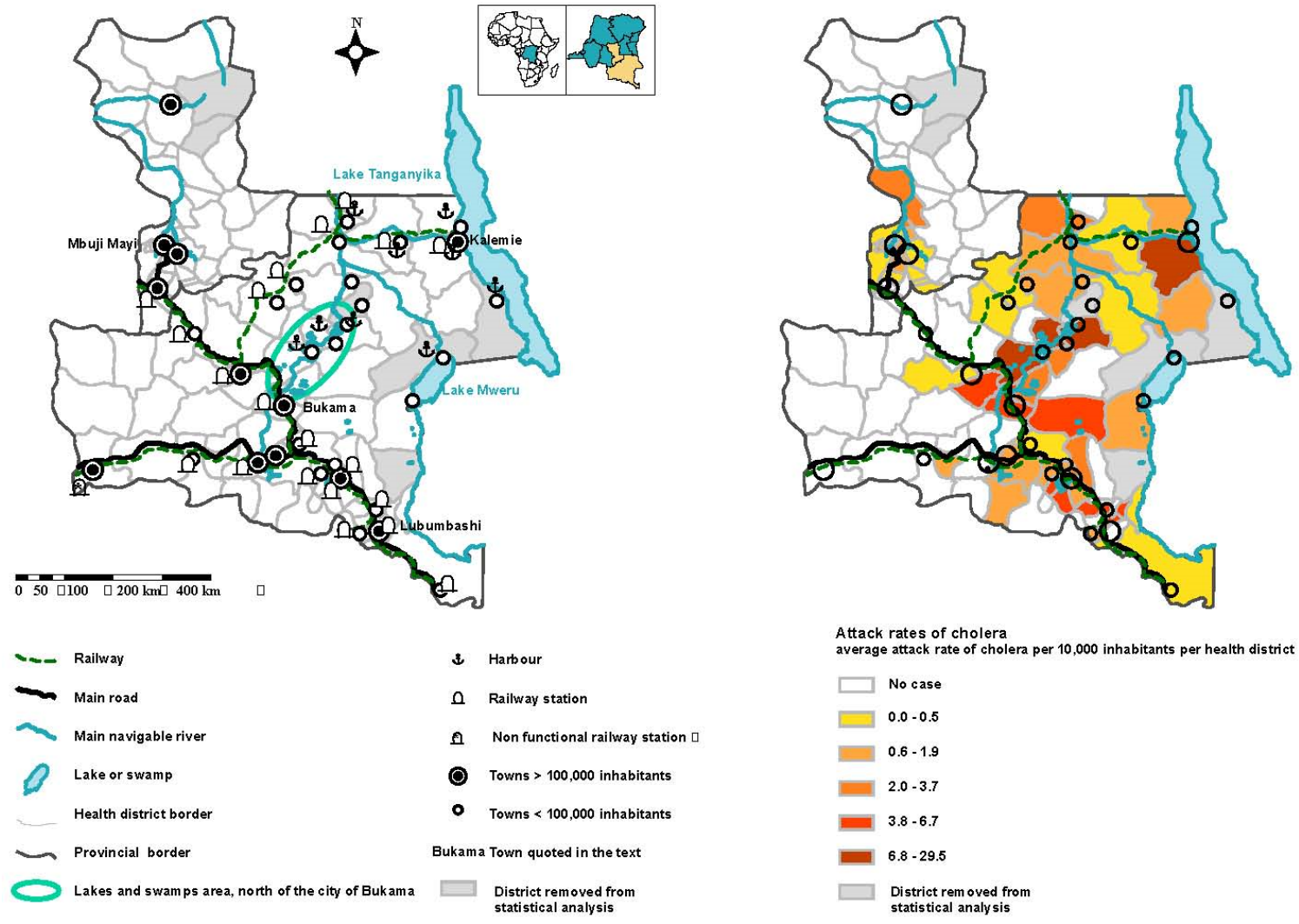
# Cholera Impact - Incidence (new cases) of disease and mortality differ by age



<http://www.who.int/bulletin/volumes/90/3/11-093427/en/>  
Ali et al, *Bulletin of the World Health Organization* 2012;90:209-218A.  
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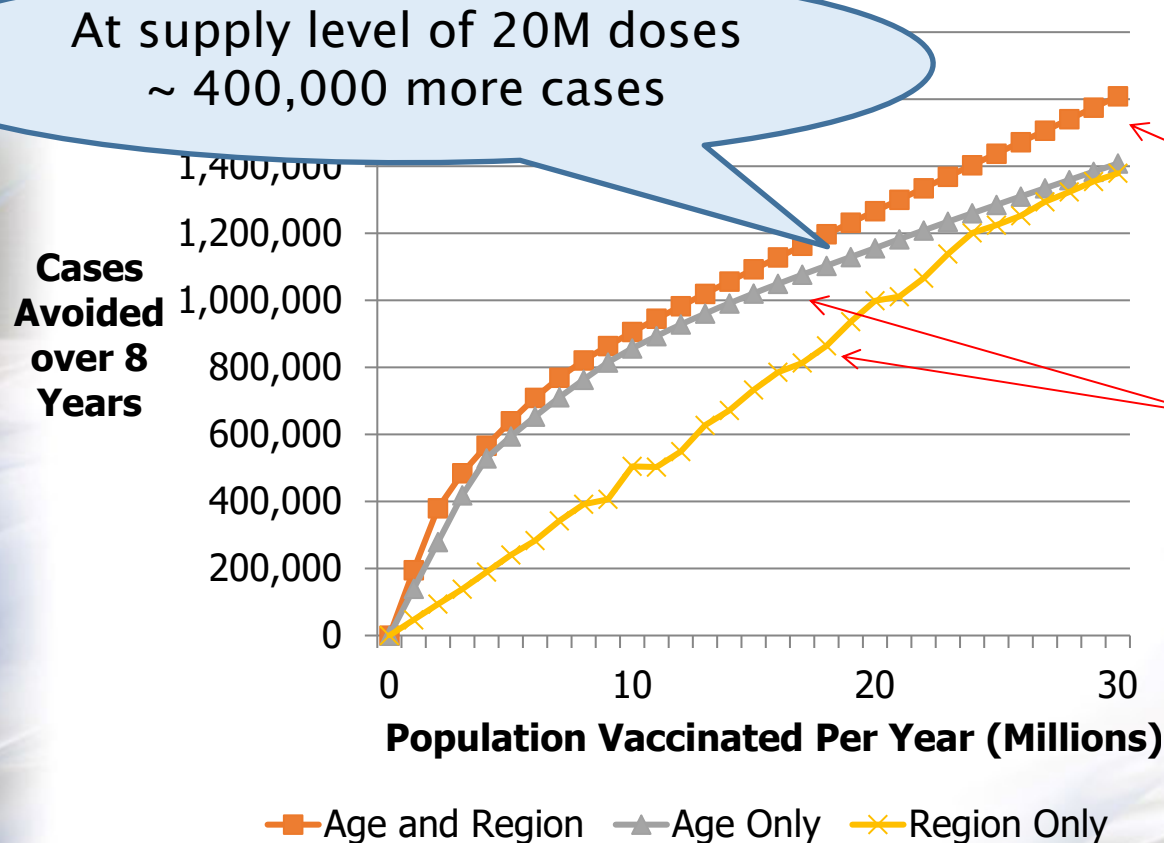
Ali et al (2012), "The global burden of cholera"

# Environmental or other risk factors



Piarroux et al (2009), The journal of field actions. (Democratic Republic of Congo)

# Resource allocation – Example: Oral cholera vaccine



H.K. Smalley, P. Keskinocak, J. Swann, A. Hinman (2015), "Optimized Oral Cholera Vaccine Distribution Strategies to Minimize Disease Incidence: A Mixed Integer Programming Model and Analysis of a Bangladesh Scenario," *Vaccine*, Vol.33, No.46, 6218–6223.

# Our Contribution

- Optimize OCV distribution policies to determine the best **OCV allocation strategy** to **minimize cases or deaths**:
  - Differentiate groups by **age AND region**, with **varying risk levels**
  - Consider fixed and varying **vaccine efficacies based on age and years since vaccination**
- Quantify cost-effectiveness of strategies

- J. Ahmed, P. K. Bardhan, W. Carter, L. Gonzalez, R. Hall, J. Heeger, L. Ivers, A. Khan, P. Keskinocak, H. Matzger, M. Mengel, D. Nazzal, C. Paradiso, F. Qadri, D. Sack, M. Villareal, S.A. Zahan (2013), "Comprehensive Integrated Strategy for Cholera Prevention and Control," Coalition for Cholera Prevention and Control, August.

<http://choleracoalition.org/resources/>



- P. Keskinocak, D. Nazzal, M. Villarreal (2013), "Procurement and Logistics," Meeting of the Coalition for Cholera Prevention and Control, National Institutes of Health, Bethesda, MD, June 3-4.
- H.K. Smalley, P. Keskinocak, J. Swann, A. Hinman (2015), "Optimized Oral Cholera Vaccine Distribution Strategies to Minimize Disease Incidence: A Mixed Integer Programming Model and Analysis of a Bangladesh Scenario," *Vaccine*, Vol.33, No.46, 6218–6223.

# Examples of Interventions

- Pharmaceutical
  - Vaccines, antivirals
- Non-pharmaceutical
  - School closures, Travel restrictions, Physical distancing (e.g., voluntary quarantine), Age-based restrictions, ...
- Combined strategies: Testing, tracing, isolation
- Behaviors/Compliance (geographically, over time, subpopulations)
- How, where, when to allocate limited resources?



# Infectious Disease Modeling

Disease progression  
in an individual –  
Natural history

Environment  
Interventions  
Behaviors

Disease spread

- Metrics/outcomes of interest (by age group, geographically, subpopulation, etc.)
  - **New infections** per time period, e.g., daily
  - Timing and magnitude of the “**peak**”
  - Total number of infections or % of population infected (IAR)
  - Number or % **hospitalized** or **dead**
  - **Resource needs** (e.g., hospital beds, ventilators)

# COVID19 MODELING AND EVALUATING INTERVENTION STRATEGIES

**Collaborators include:** John Asplund, Ph.D.; Emma Baubly; Arden Baxter; Saurabh Doodhwala; Akane Fujimoto; Daniel Kim; Dima Nazzal, Ph.D.; Buse Eylul Oruc; Pelin Pekgun, Ph.D., Lauren Steimle, Ph.D.; Tyler Perini; Josh Rosenblum; Erik Rosenstrom; Nicoleta Serban, Ph.D.; Melody Shellman; Chris Stone; Julie Swann, Ph.D.; Inci Yildirim, MD, Ph.D.; April Yu; Georgia Tech Institute for People and Technology; GA Department of Public Health

**Funding:** “Integrated Systems Model to Inform State and Local Planning for the COVID-19 Pandemic,” Council of State and Territorial Epidemiologists (CSTE); RADx Underserved Populations (RADx-UP) program

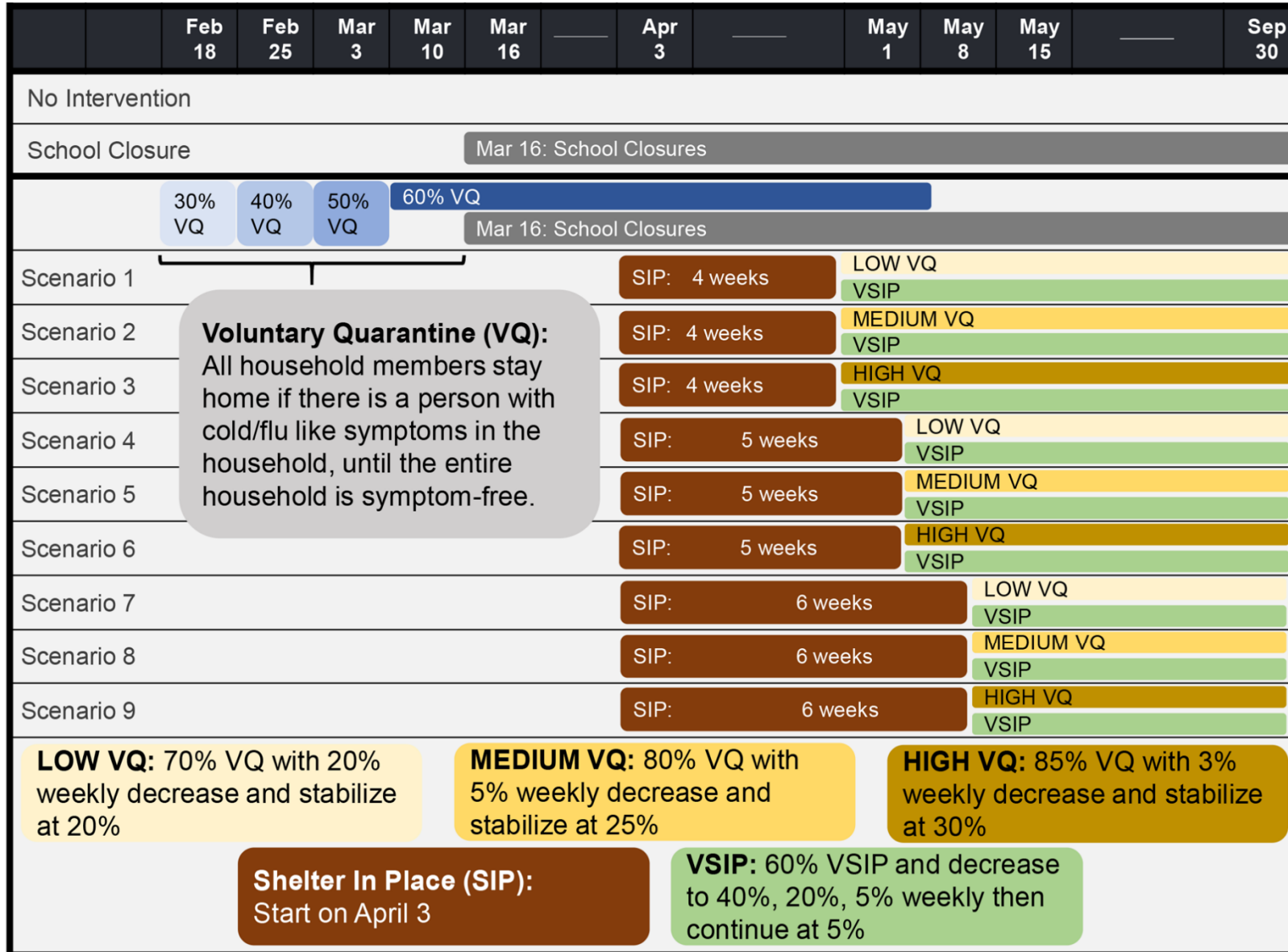
# Research insights

- Modeling the disease spread → Projections
- Evaluating the impact of interventions
  - School closures, Shelter-in-place, voluntary-quarantine
- Estimating resource needs
- Impact of interventions on society: “homebound days” versus reduction in disease spread
- Impact of testing/isolation depending on compliance
- Vaccine allocation – Benefits of serology testing?
- Tradeoff between vaccine efficacy versus reach
- ...

# SCENARIOS FOR PHYSICAL DISTANCING

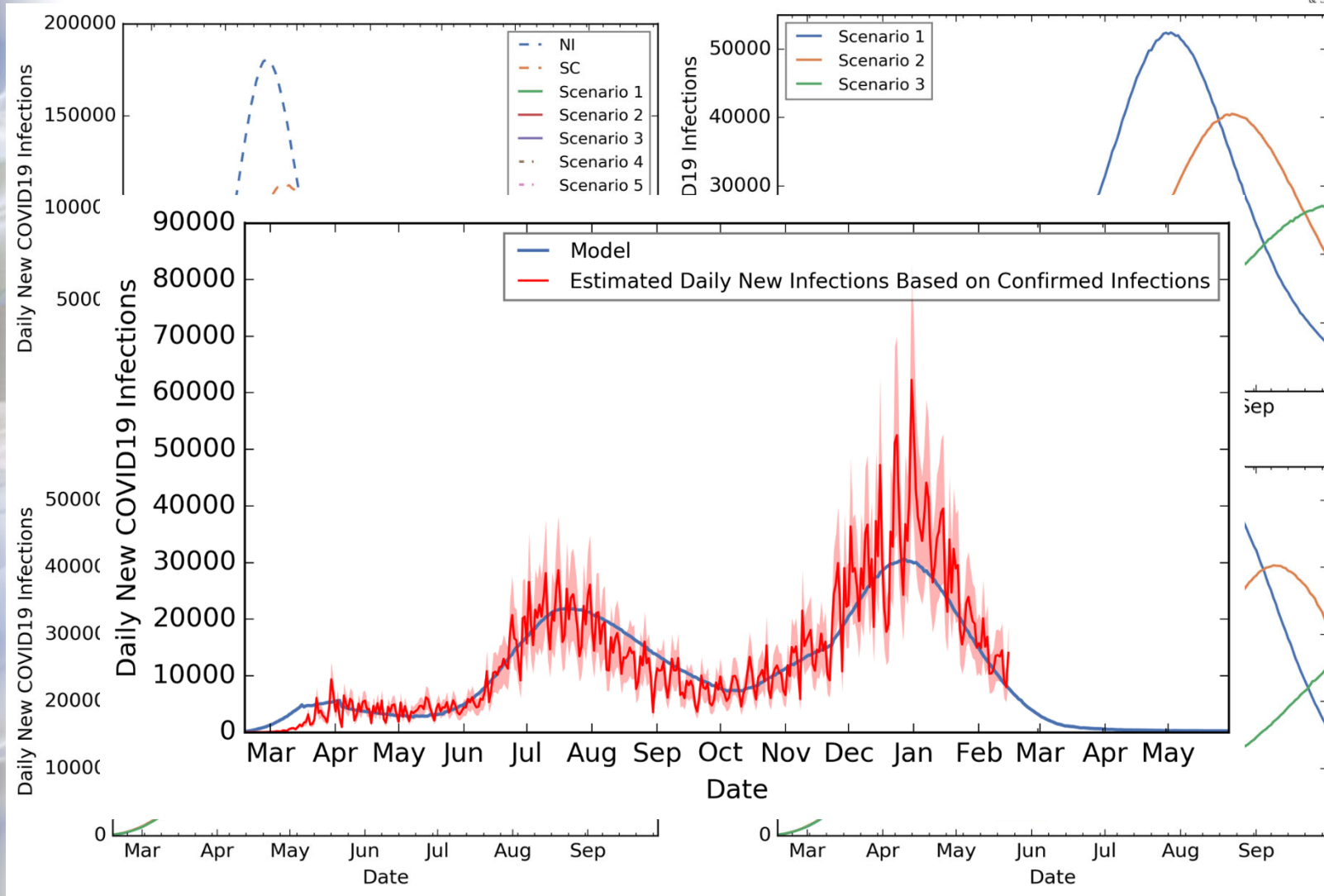
Baselines

Interventions

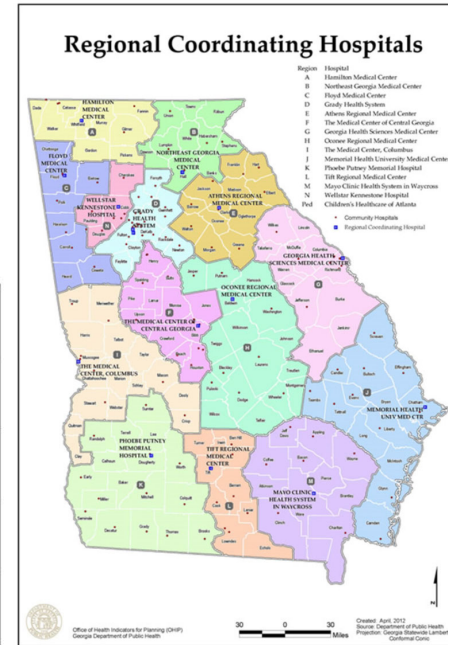
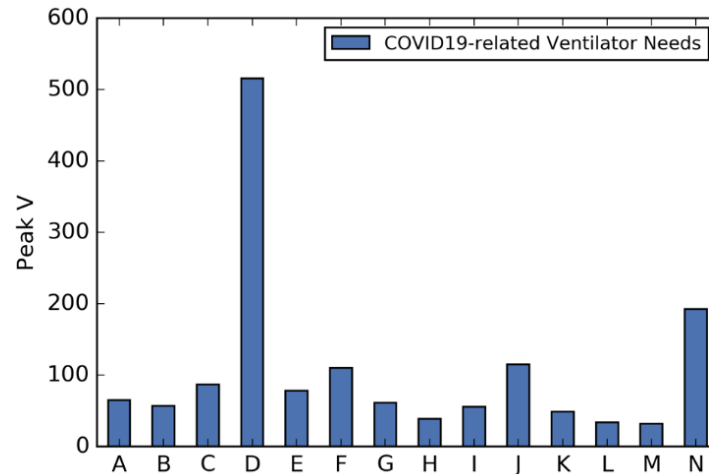
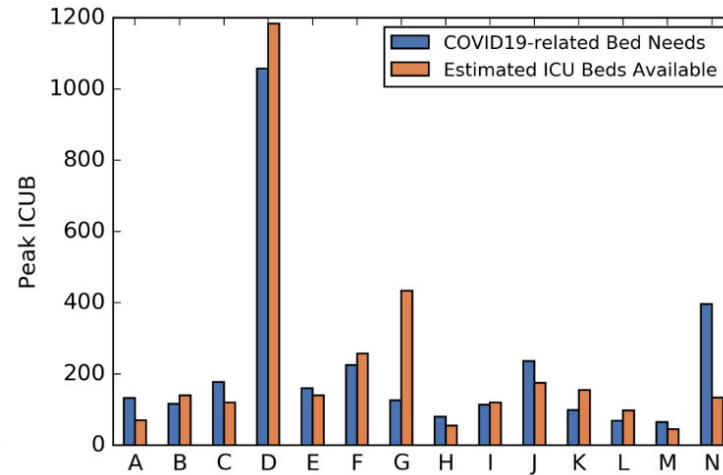
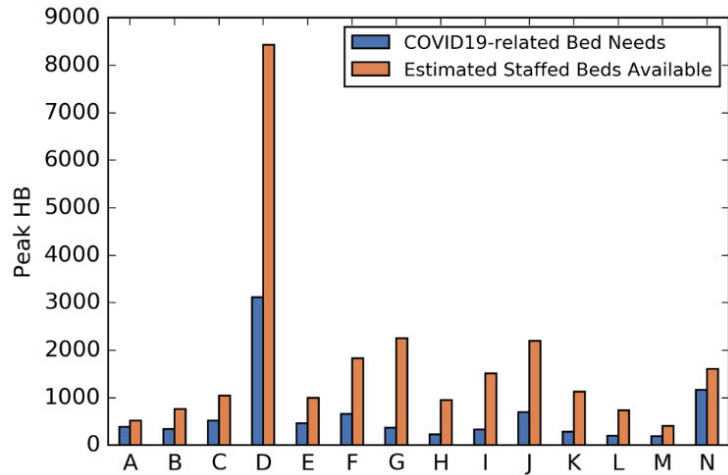


**Voluntary Quarantine (VQ):**  
 All household members stay home if there is a person with cold/flu like symptoms in the household, until the entire household is symptom-free.

# Daily new infections Projections In Georgia



# Healthcare resource estimation



“You may build all the ICUs you want. You may have all the ventilators you need. But you will not have the staff you need,” del Rio said. “There simply are not enough ICU nurses or ICU doctors to take care of the patients.”

# Research questions

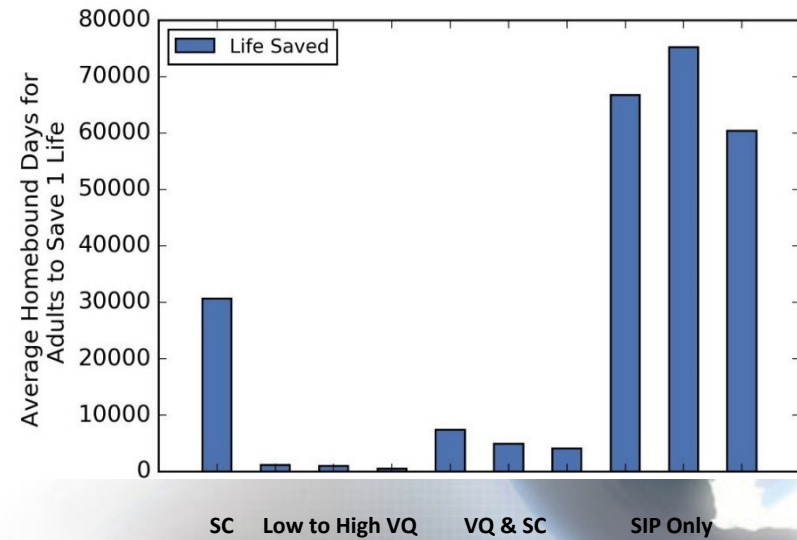
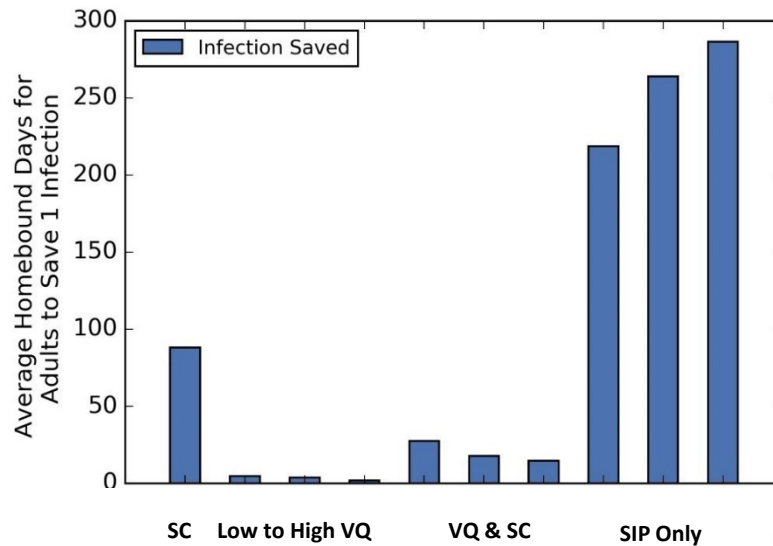
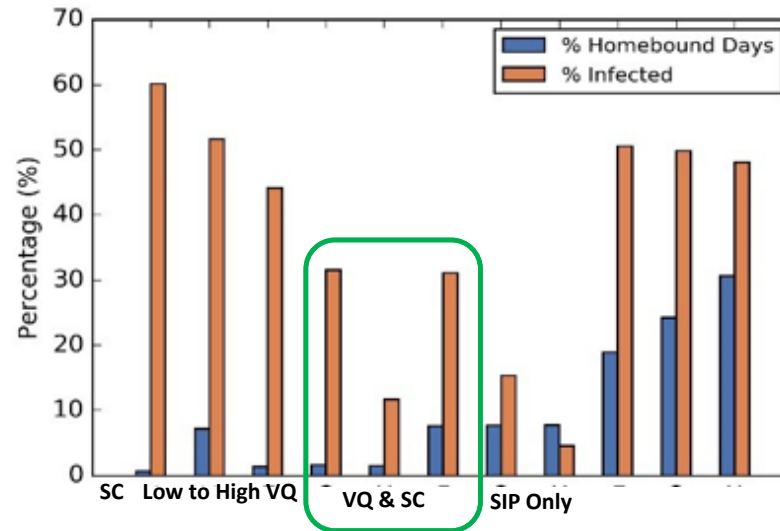
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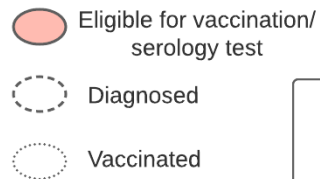


# Health, society, economy, etc. – complex tradeoffs

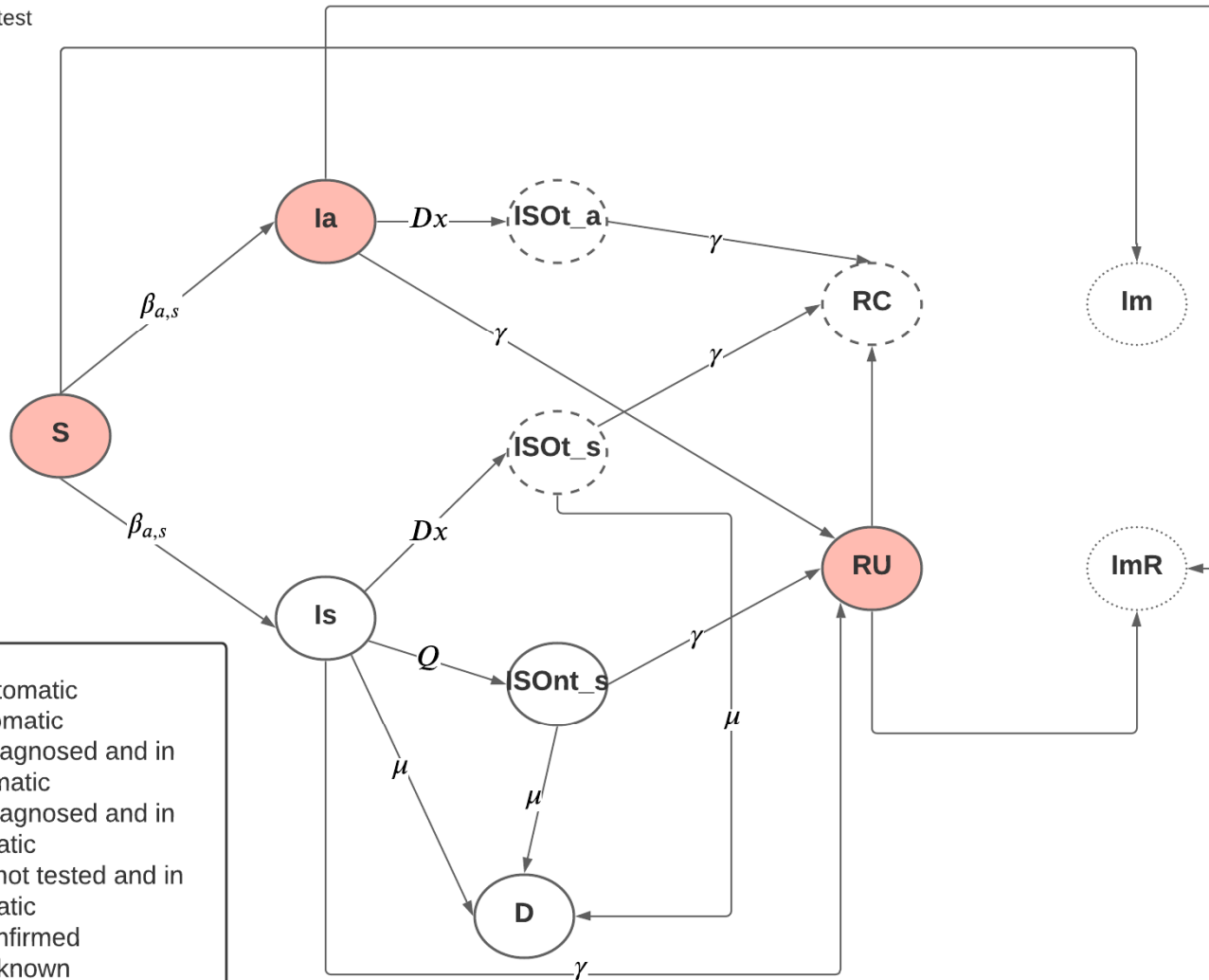


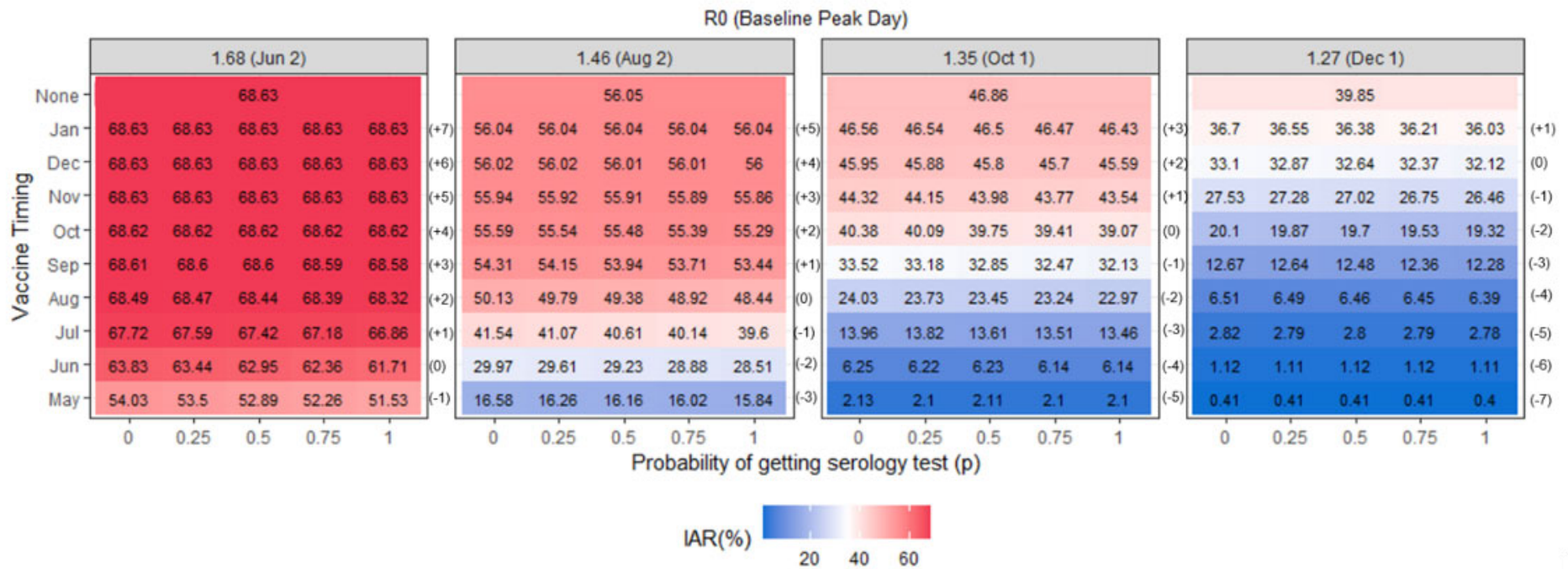
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- **Vaccine allocation – Benefits of serology testing?**



**S:** Susceptible  
**la:** Infected asymptomatic  
**Is:** Infected symptomatic  
**ISOt\_a:** Infected diagnosed and in  
isolation, asymptomatic  
**ISOt\_s:** Infected diagnosed and in  
isolation, symptomatic  
**ISOnt\_s:** Infected not tested and in  
isolation, symptomatic  
**RC:** Recovered confirmed  
**RU:** Recovered unknown  
**Im:** Immunized not previously infected  
**ImR:** Immunized previously infected  
**D:** Dead





**Figure 2:** Infection attack rate for the scenarios evaluated when the vaccine is available for 50% of the population and the vaccine efficacy is 90%.

- Highest IAR: High R0 and vaccination available after the peak.
- Impact of serology test highest when the vaccines are deployed close to the peak time

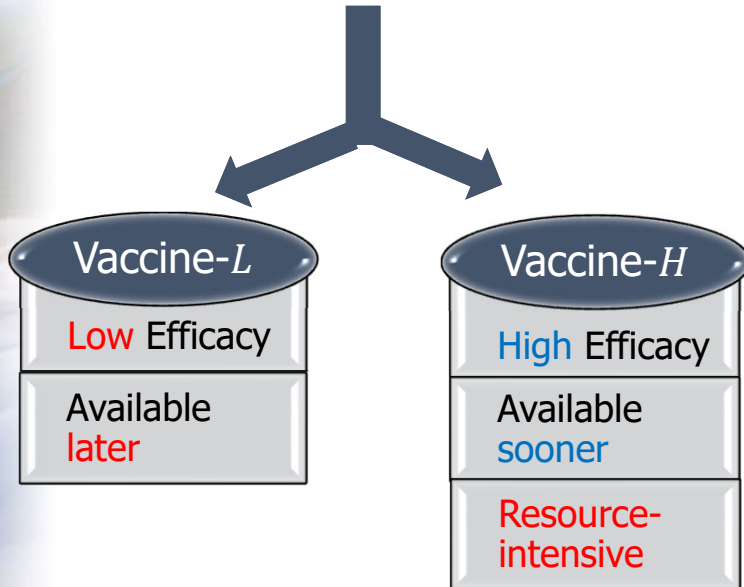
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# Tradeoffs between vaccine efficacy and reach

Limited Resources

Allocation – How?



**Reach** of a vaccine

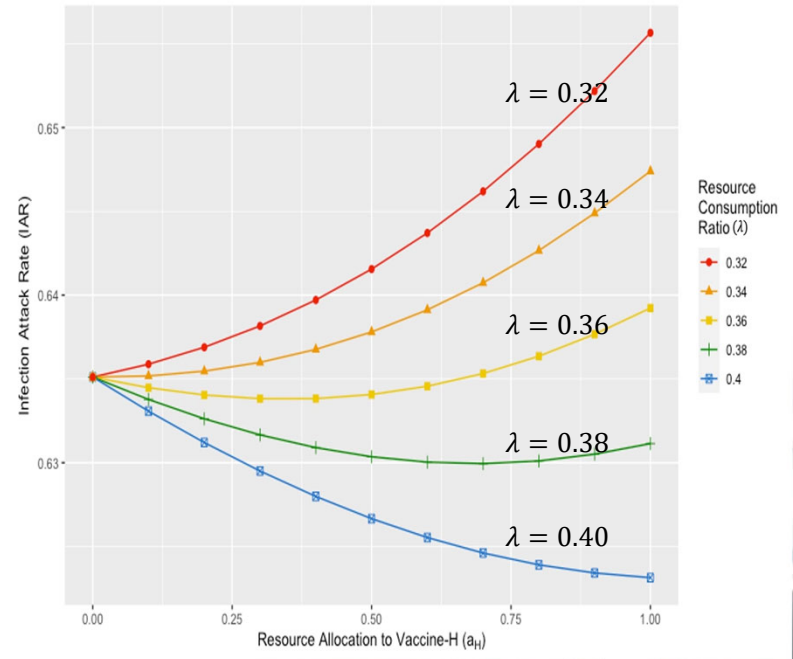


Figure 1 : IAR w.r.t. the percentage of resources allocated to vaccine-H

> Vaccine. 2021 Oct 18;S0264-410X(21)01346-3. doi: 10.1016/j.vaccine.2021.10.025. Online ahead of print.

Resource allocation for different types of vaccines against COVID-19: Tradeoffs and synergies between efficacy and reach

Daniel Kim <sup>1</sup>, Pelin Pekgün <sup>2</sup>, İnci Yıldırım <sup>3</sup>, Pınar Keskinocak <sup>4</sup>

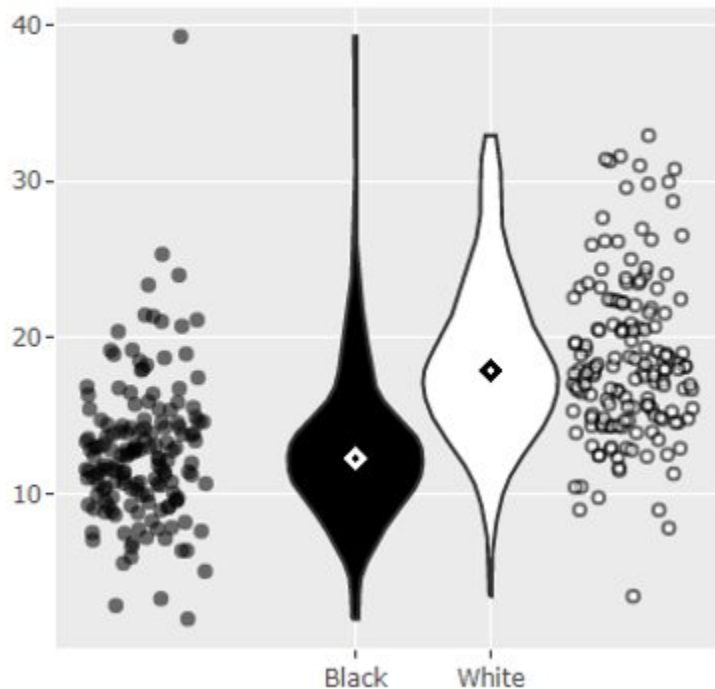
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- **Dashboards**

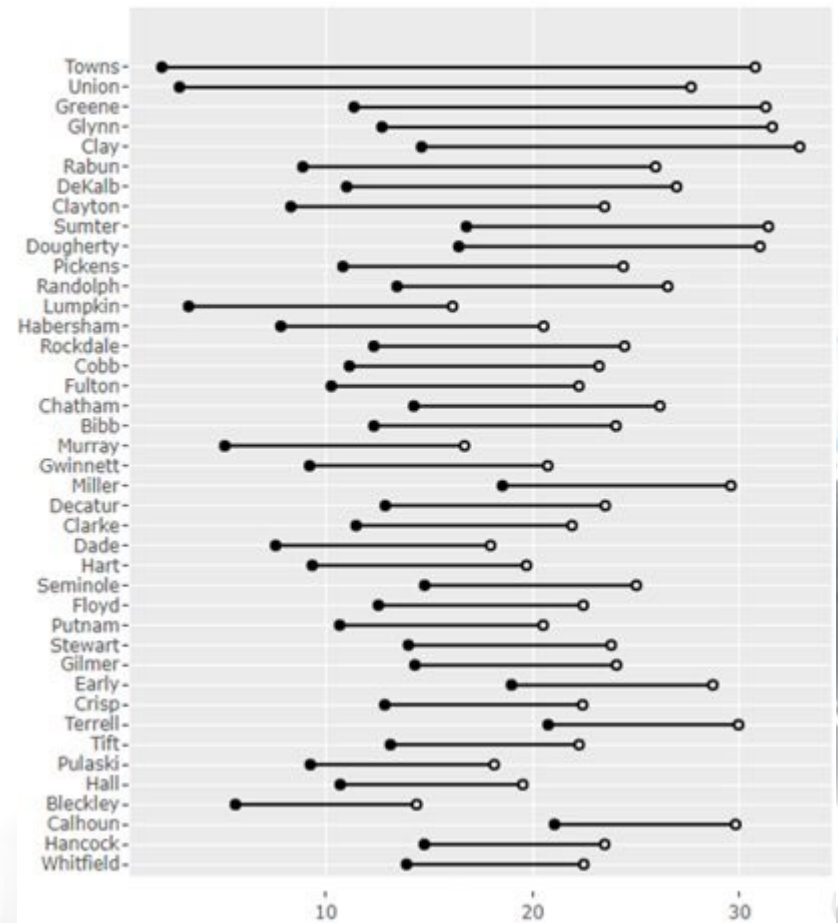
# Dashboards <https://chhs.gatech.edu/covid19-dashboard>

$$\frac{\text{Number of people vaccinated in subpopulation}}{\text{Total number of people in subpopulation}} \times 100$$

Vaccination Rate per Hundred (%)



Rate Differences







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## MISSION

Make a positive “impact” through improved health & humanitarian systems worldwide

