

## Data, models and science

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DAIDD 2020

# Outline

Introduction

Public health

Maternal mortality

Cholera

Yellow fever and malaria

Approaches to epidemiology

Summary

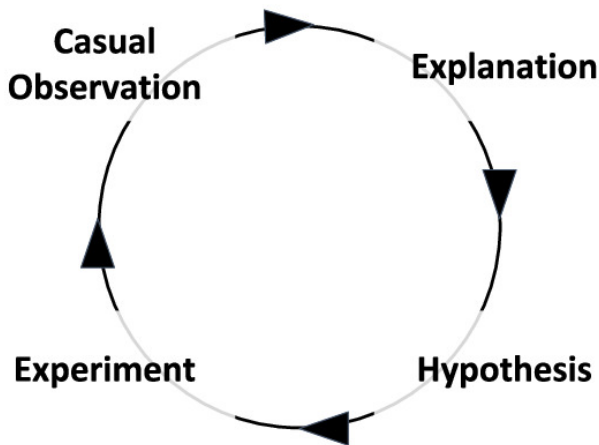
# Goals

- ▶ Process of science
- ▶ How science informs public health
  - ▶ Specific examples
- ▶ Approaches to epidemiology

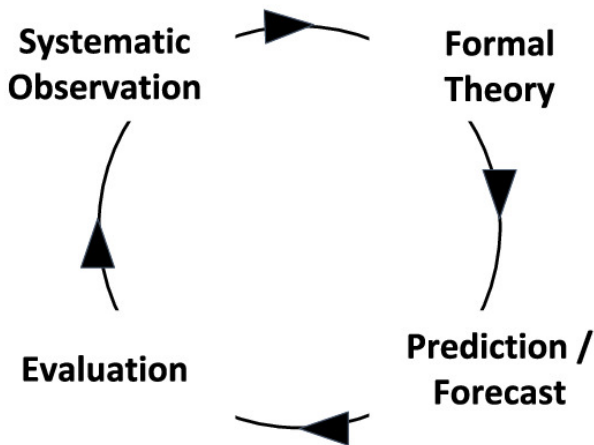
# Science is a *process*

- ▶ Observe and experiment with reality to *discover* and *challenge* ideas about how it works
- ▶ A key to science is that everything is open to question
  - ▶ Science is the belief in the ignorance of experts – *Feynman*

# The process of science



# Science without experiments



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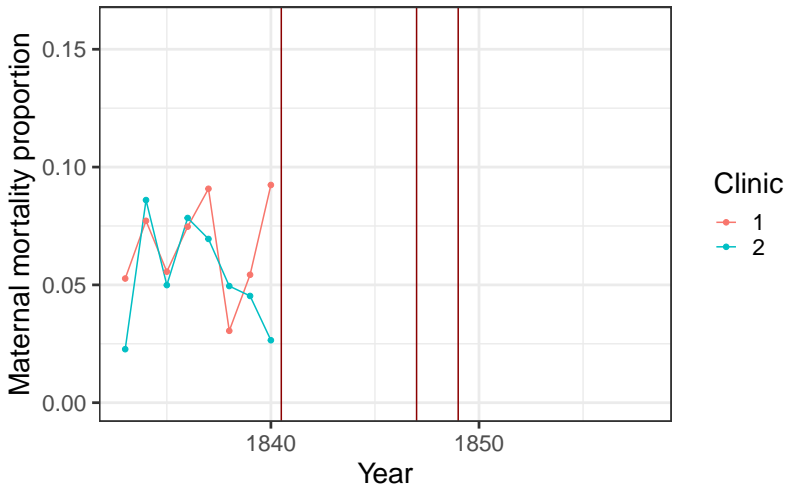
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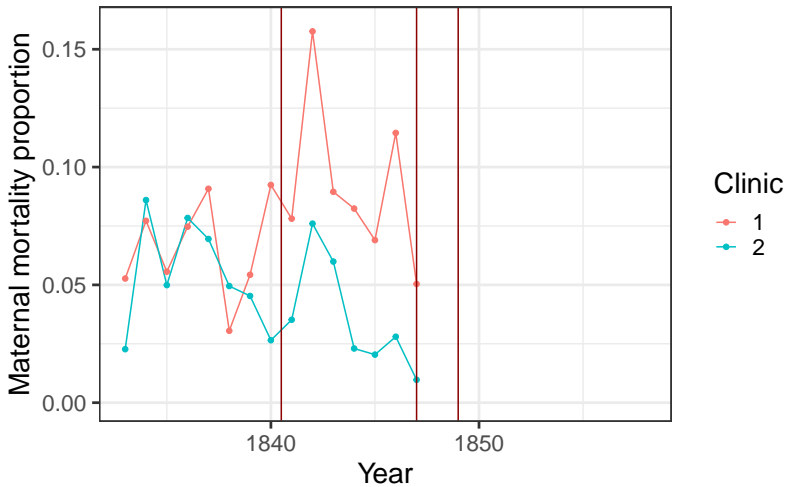


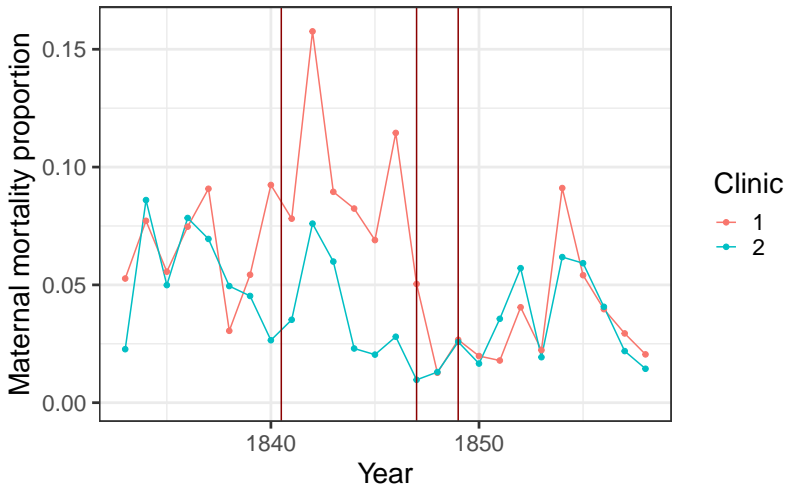
# Maternal mortality

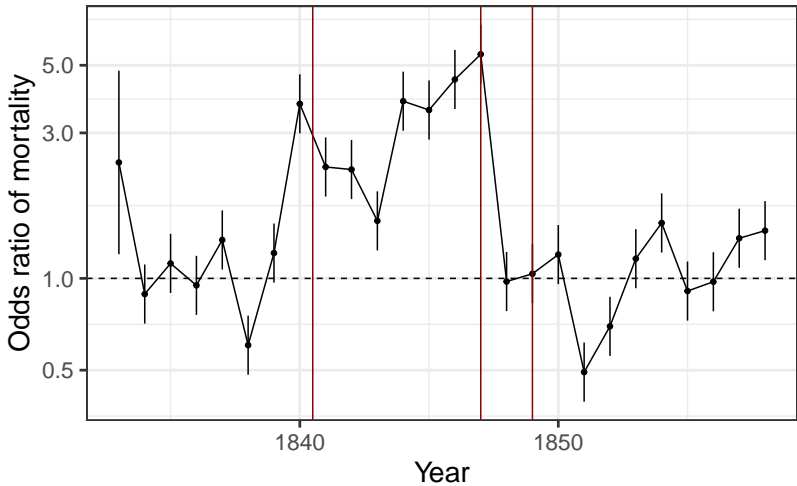


# Observation and action

- ▶ In 1840, medical students stopped visiting Clinic 1
- ▶ In 1847, a surgeon died from infection following a scalpel injury
  - ▶ Igor Semmelweiss made medical students wash their hands







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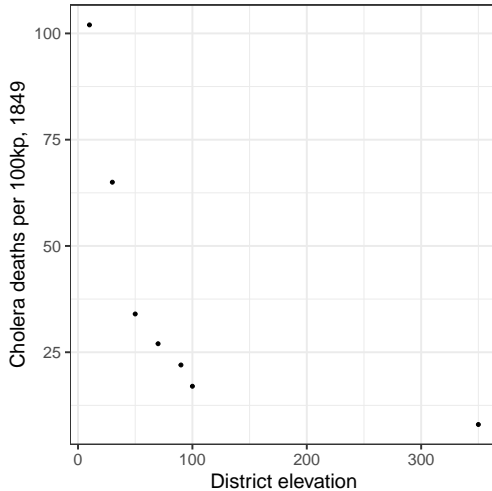
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# Cholera

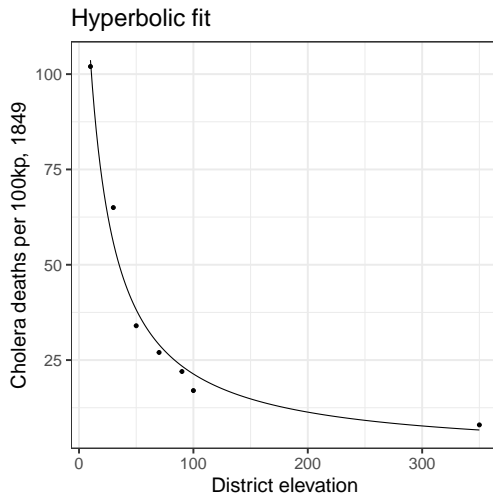
- ▶ Is it caused by bad air, or bad water?
- ▶ What's bad about it?

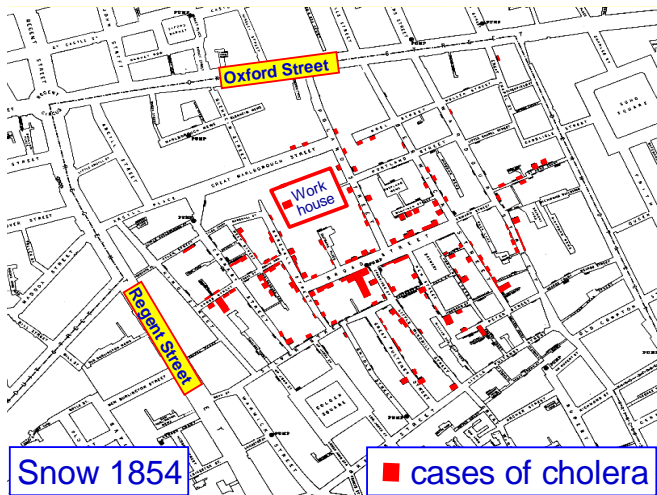


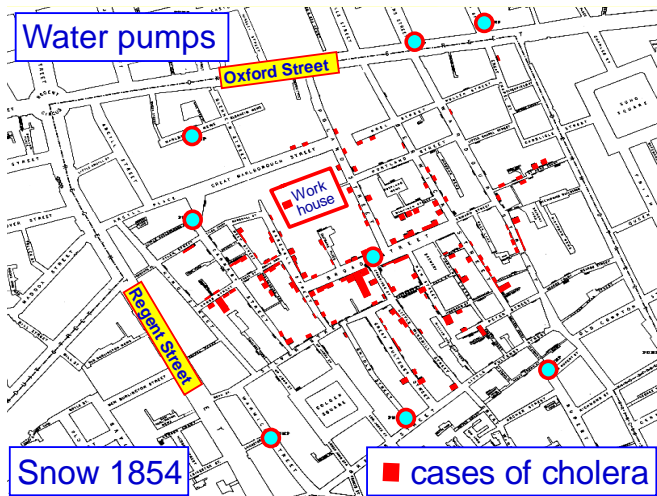
## *Cholera and air (present)*



# Cholera and air







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# Yellow fever and malaria

- ▶ Ross determined the cause of malaria primarily by experiments on mosquitoes
- ▶ Reed determined the cause of yellow fever primarily by experiments on human volunteers

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# Data, models and science

- ▶ We're never finished, we compare models to data over and over again
- ▶ Data is what we use to develop and understand models
- ▶ Models are what we use to interpret data
  - ▶ and they can suggest what data we need to collect
- ▶ Complicated or hard-to-test theories may require *dynamical* models



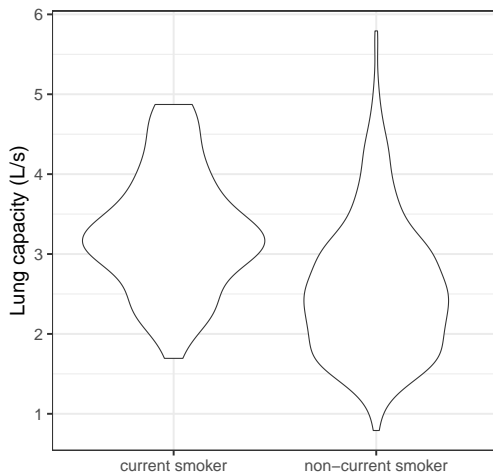
## Classical epidemiology

- ▶ Avoid mechanism
- ▶ Control for non-independence of “units”

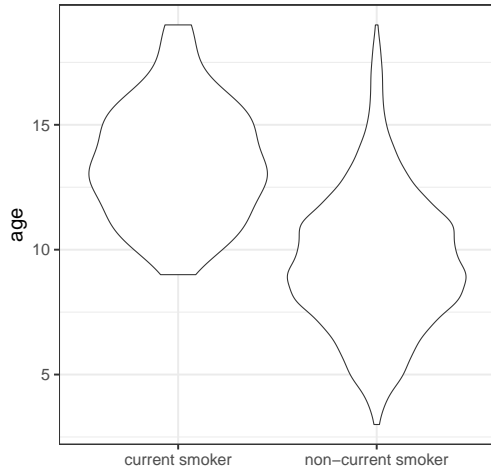
## Dynamical epidemiology

- ▶ Embrace mechanism
- ▶ Explicitly incorporate dependence between units
  - ▶ X is infected because Y infected them

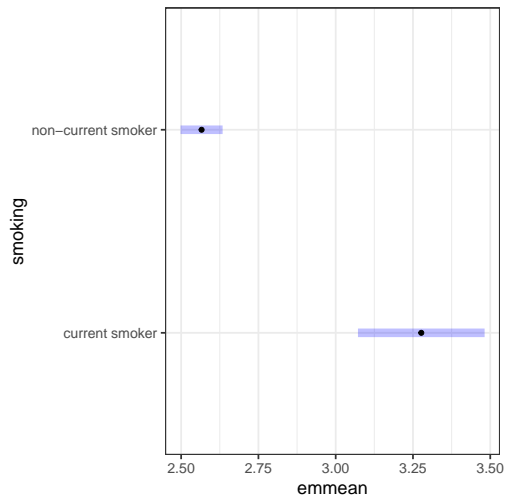
# Classical example



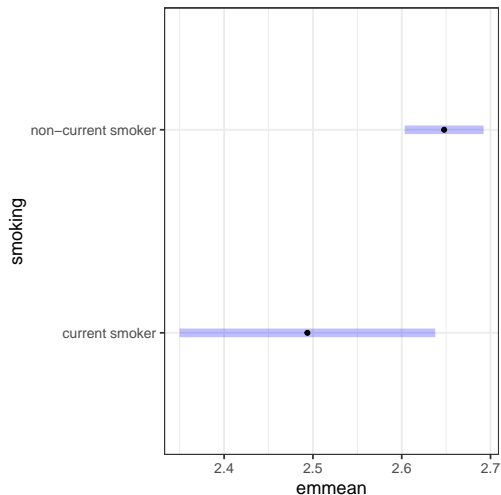
# Classical example



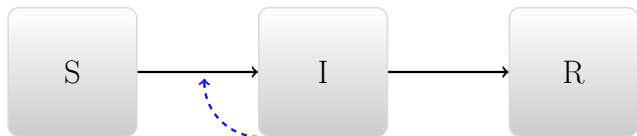
# Univariate means



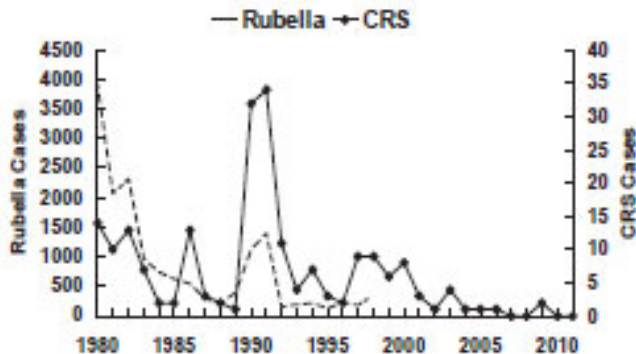
# Multivariate means



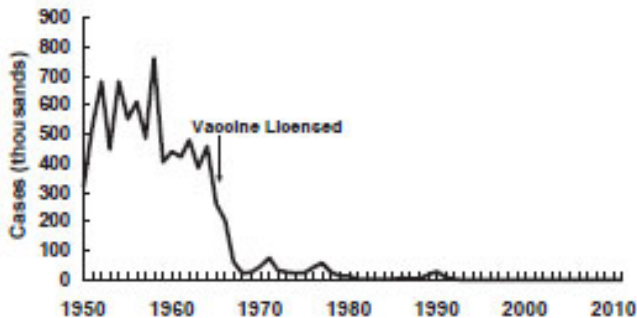
# Dynamical examples



## *Rubella (present)*

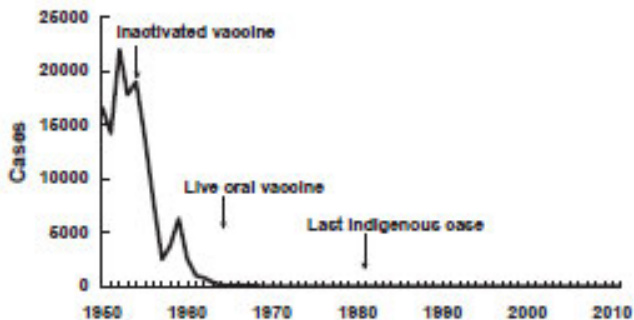


## *Measles (present)*

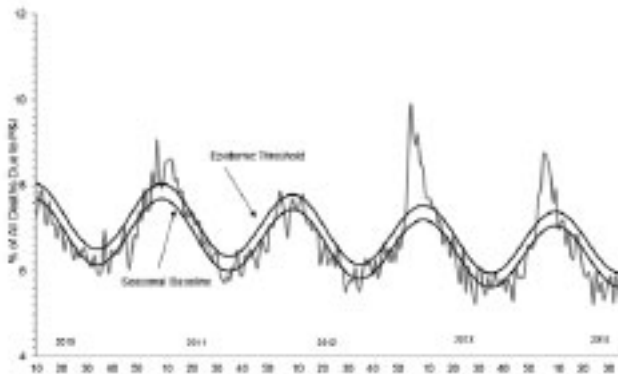




## Polio (present)



## Influenza (present)



# Bridging

- ▶ Classical epidemiology relies on statistics, avoids mechanism
- ▶ Mathematical epidemiology (the traditional approach to dynamical epidemiology) explores mechanism, avoids statistics
- ▶ Much modern dynamical epidemiology seeks ways to put dynamical mechanisms into a statistical framework
  - ▶ This is hard

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- ▶ Science is an ongoing process
- ▶ Models are the way that we bridge between theory and reality
- ▶ Dynamical models have a key role
  - ▶ When we can't do experiments
  - ▶ When mechanisms are complex
- ▶ We should work to combine dynamics with statistical approaches



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