

Data, models and science

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MMED 2021

Introduction

Public health

Maternal mortality
Cholera
Yellow fever and malaria

Approaches to epidemiology

Dynamical epidemiology

Building knowledge from data

Summary

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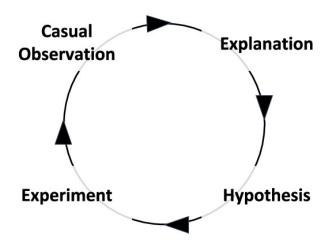
Goals

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

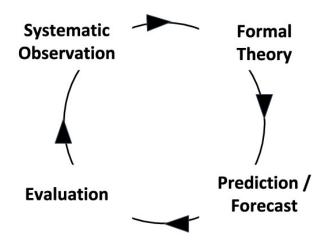
Science is a process

- 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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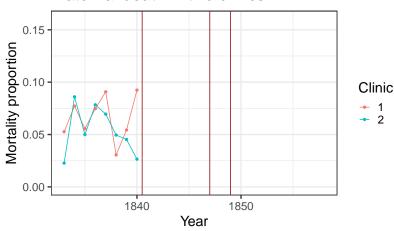
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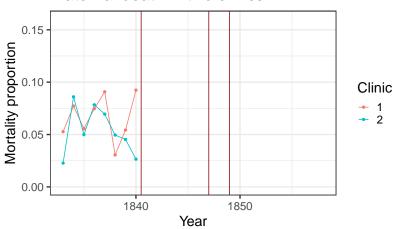




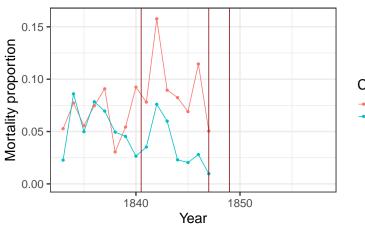
Observation and action

- 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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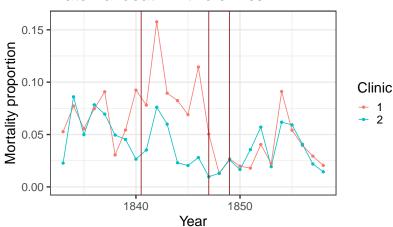






Clinic





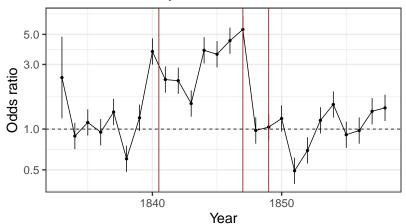


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Looking at the data

- * Why was Clinic 1 so dangerous in the 40s?
- ▶ * ... and so safe in 1851 and 1852?
- * What can we learn from modern statistics?
- * And what can't we learn without more data?

Relative mortality risk in Clinic 1





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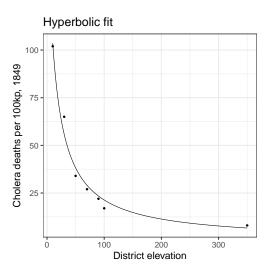
Summarv

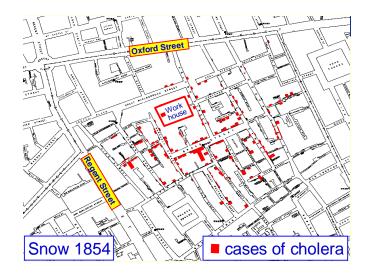


Cholera

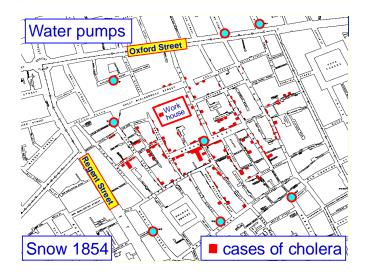
- Is it caused by bad air, or bad water?
- What's bad about it?

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

Dynamical epidemiology

- Avoid mechanism
- Control for non-independence of "units"

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

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Public health

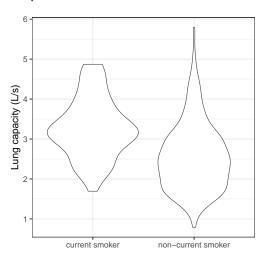
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Approaches to epidemiology Classical epidemiology

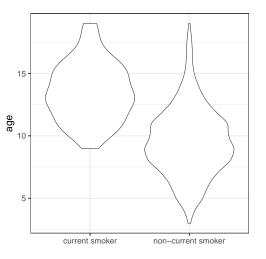
Dynamical epidemiology Building knowledge from data

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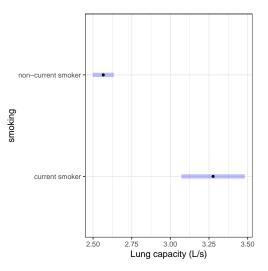
Classical example



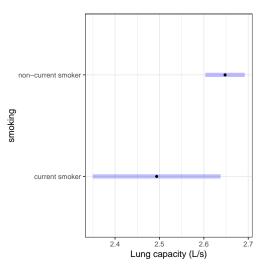
Classical example



Univariate means



Multivariate means



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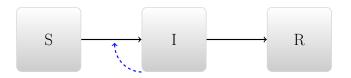
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Dynamical epidemiology



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

Maternal mortality Yellow fever and malaria

Approaches to epidemiology

Building knowledge from data



Building knowledge from data

- We must have communication up and down the analysis pipeline
 - Data are collected in the field
 - Organized and documented
 - Protected (for confidentiality, and often for the rights of collectors)
 - Summarized
 - Modeled

Example: COVID incidence

- Positive test results but no negative test results
 - Can't correct for testing intensity
- Positive and negative test results, but no individual identifiers
 - Can't correct for multiple testing of the same people
- Test results, but not reason for testing
 - Can't correct for testing focus

Example: COVID variants

- Mutational screens not linked to invididuals
 - We can estimate mutations, but not variants
- Reasons for screening or sequencing not provided
 - Can't correct for selection bias
- Personal information not provided
 - Can't look for geographical patterns, vaccine effectiveness,

. .

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Example: West Africa Ebola Outbreak

- Medical vs. public-health priorities
 - Am I responsible if my data request increases the pressure on a front-line responder?
- Individual-level vs. population considerations
 - When it is OK to randomize people to receive a placebo vaccine or treatment?

Example: Microbiome studies

- In some fields, the amount of apparently high-quality data is far outstripping the ability to understand it
- Lots of reasonably intelligent experiments (or observational designs)
- Huge lists of taxonomic (or metagenomic) communities

scientific reports

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Article | Open Access | Published: 21 June 2021
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SARS-CoV-2 viral load in nasopharyngeal swabs is not an independent predictor of unfavorable outcome

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Sonsoles Salto-Alejandre, Judith Berastegui-Cabrera, Pedro Camacho-Martínez, Carmen Infante-
Domínguez, Marta Carretero-Ledesma, Juan Carlos Crespo-Rivas, Eduardo Márquez, José Manuel
Lomas, Claudio Bueno, Rosario Amaya, José Antonio Lepe, José Miguel Cisneros, Jerónimo Pachón

☐, Elisa Cordero, Javier Sánchez-Céspedes & The Virgen del Rocío Hospital COVID-19 Working
Team
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Scientific Reports 11, Article number: 12931 (2021) | Cite this article 256 Accesses | 1 Altmetric | Metrics
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https://www.nature.com/articles/s41598-021-92400-y



Data vs. models

- Models can teach us a lot, but good data with a simplistic model is usually better than poor (or poorly contextualized) data with a good model
- Sometimes, the most valuable thing about the model is that it helps us figure out what data we need
 - * Value-of-information models

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Summary

- Science is an ongoing process
- Models are the way that we bridge between theory and reality
- We can only bridge to reality if we can measure reality
 - Collect and curate good data
- 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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