E²M²: Ecological and Epidemiological Modeling in Madagascar

Data and Models

Hôtel Cyperus Andasibe National Park, Madagascar May 2025

Thanks to our sponsors!



- Lecture contributions from:
- Tanjona Ramiadantsoa
- Steve Bellan

• To explain what we're doing here

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- To define "science"

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What is science?

What is science?

the systematic observation of natural events and conditions in order to discover facts about them and to formulate laws and principles based on these facts.

- Academic Press Dictionary of Science & Technology

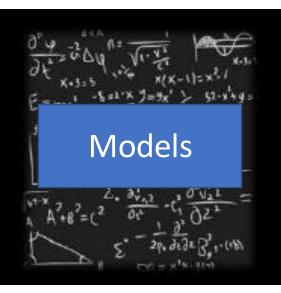
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Observations and Laws and Principles

Data and Models





Data and Models

Data

• What are data?

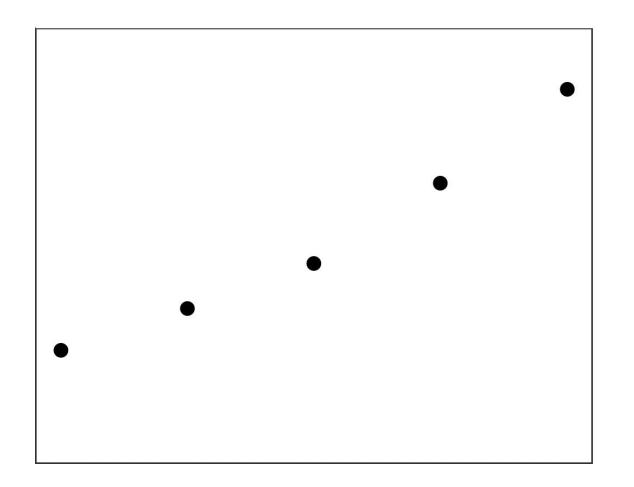
Data and Models

- What are data?
 - Evidence to support a claim

Data

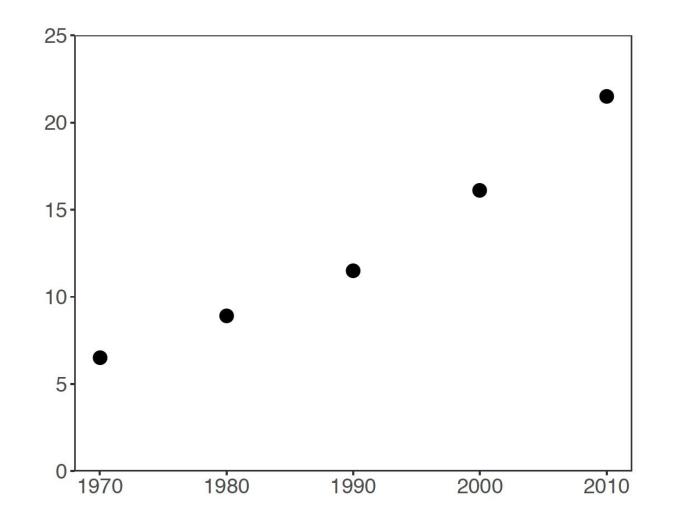




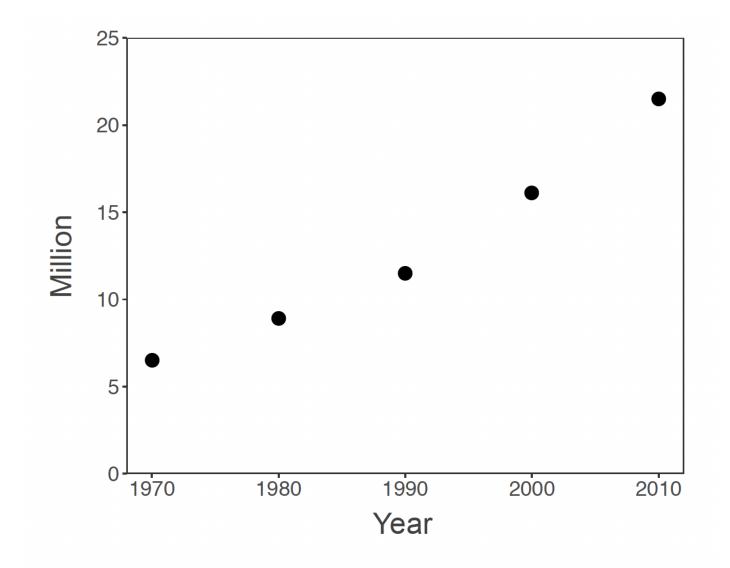


What do we need to make these data?



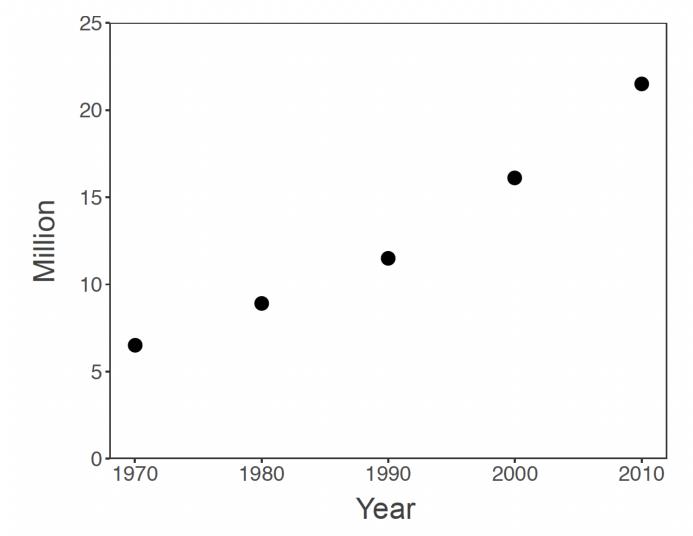


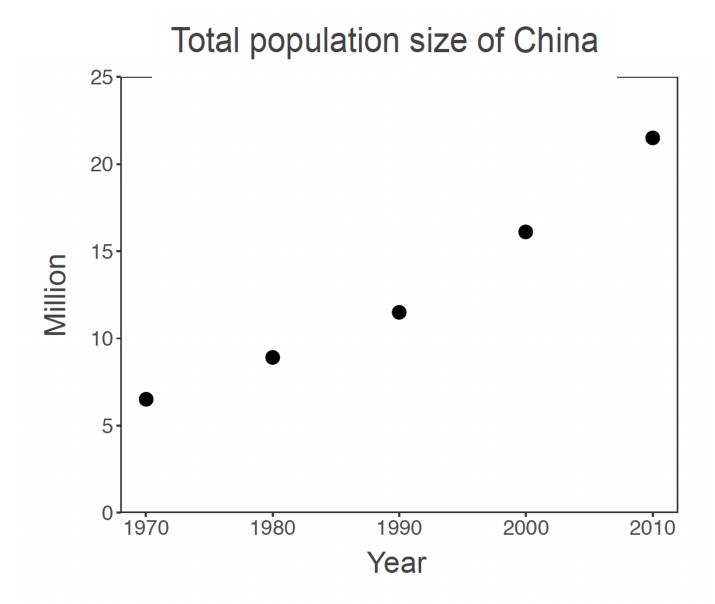


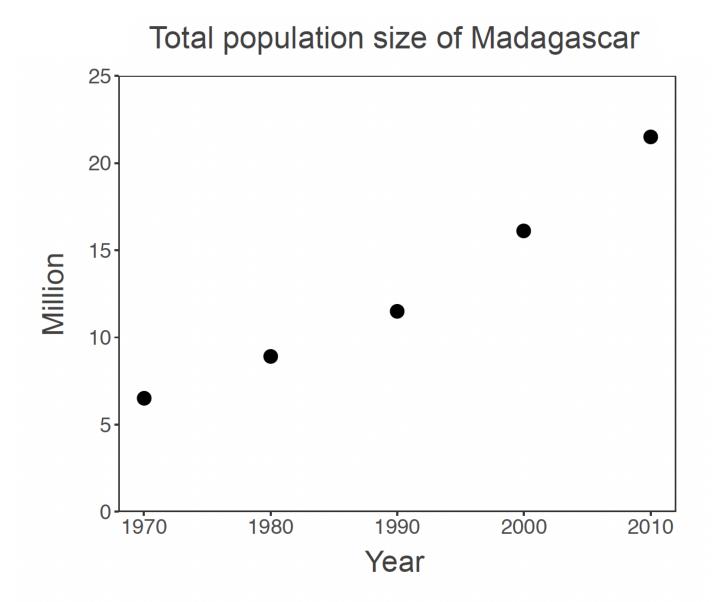




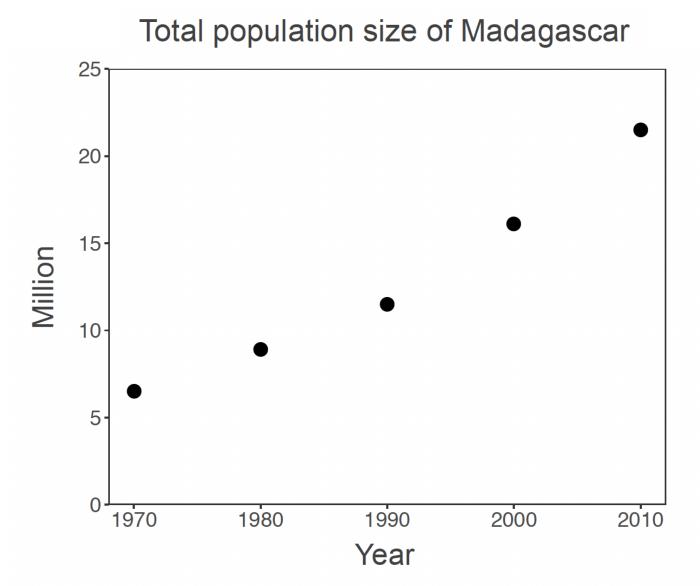
Number of views for Justin Bieber's "Baby" video on Youtube







Data



Source: World Bank (accessed 2017)



What are data?

- A relationship between at least two variables
 - x: explanatory, control, driver, independent variable(s)
 - y: response, dependent variable(s)



What are data?

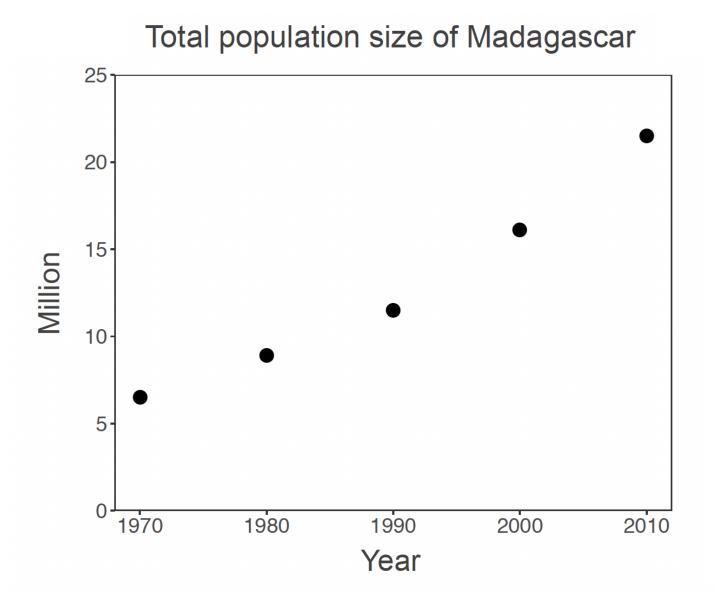
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- x and y should be clearly defined
 - with respect to the **question!**



What are data?

- A relationship between at least two variables
 - x: explanatory, control, driver, independent variable(s)
 - y: response, dependent variable(s)
- x and y should be clearly defined
 - with respect to the **question!**
- Evidence to support a claim

Data provide evidence to support a claim.

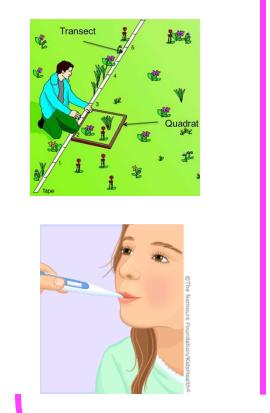


CLAIM: The population size of Madagascar has increased throughout the past 50 years

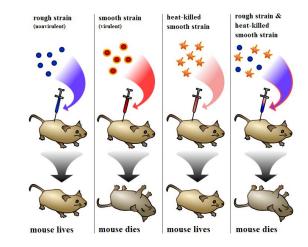
Source: World Bank (accessed 2017)

Data: Sources of x and y

Observational

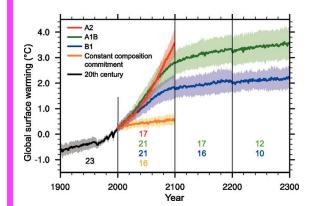


Experimental



Simulated

Data



Empirical data

Data: Types



Numerical

Categorical

Data: Types



Numerical

Categorical

- A variable is numerical when you can transform it with mathematical operation
- Examples?

Data: Types



Categorical

Data

- A variable is numerical when you can transform it with mathematical operation
- Examples:
- Integer, real number, multidimensional number

Data: Types



- A variable is numerical when you can transform it with mathematical operation
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Categorical

• A variable is categorical when it is not numerical but a categorical can be numerical? Data

• Examples?

Data: Types

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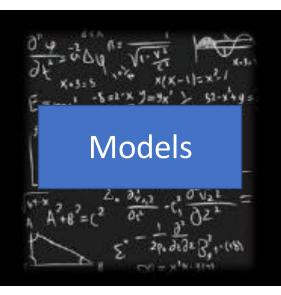
Categorical

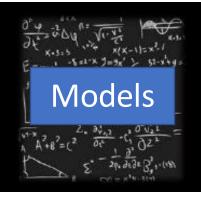
- A variable is categorical when it is not numerical but a categorical can be numerical?
- Examples:
- Colors, (blood) types, species name



Data and Models





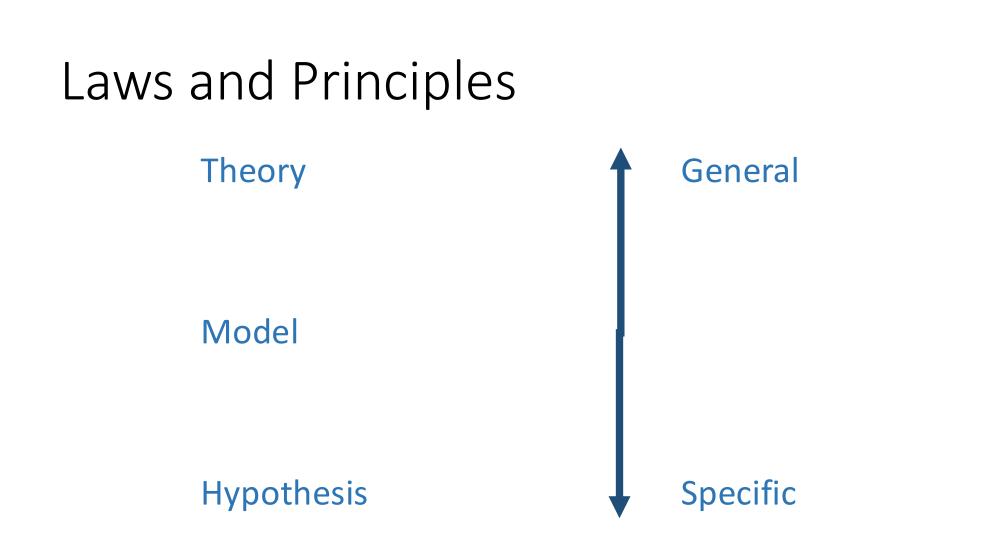


Data and Models

• What is a model?

What is science?

the systematic observation of natural events and conditions in order to discover facts about them and to formulate laws and principles



Models

Laws and Principles

Theory

a declaration to explain a phenomenon

Model

Hypothesis



Specific

 $\int_{a}^{b^{2}} \frac{\varphi}{\partial t_{a}^{k}} = \frac{1}{a^{k} \Delta y} \int_{a}^{b^{2}} \frac{\sqrt{(\sqrt{t_{a}^{k}})^{k}}}{\sqrt{(\sqrt{t_{a}^{k}})^{k}}} \int_{a}^{b^{2}} \frac{\varphi}{\sqrt{(\sqrt{t_{a}^{k}})^{k}}} \int_{a}^{b^{2}} \frac{\varphi}{\sqrt{(\sqrt{t_{a}^{k})^{k}}} \int_{a}^{b^{2}} \frac{\varphi}{\sqrt{(\sqrt{t_{a}^{k})^{k}}}} \int$

Laws and Principles

Theory

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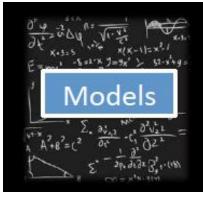
Model

an abstract representation of a phenomenon

Hypothesis

General

Specific



Laws and Principles

Theory

a declaration to explain a phenomenon

Model

an abstract representation of a phenomenon

Hypothesis

a testable declaration that is derived from a theory

General

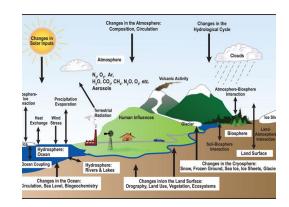


 $\int_{0}^{0} \frac{\psi}{\psi} = \frac{2^{2} \log n^{2}}{\log \sqrt{1 + \frac{1}{\sqrt{2}}}} \int_{0}^{1} \frac{\psi}{\sqrt{1 + \frac{1}{\sqrt{2}}}} \int_{0}^{1} \frac{\psi}$

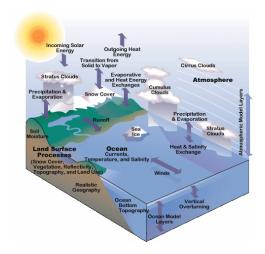
Models

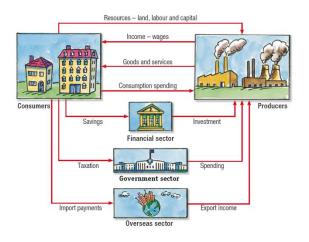


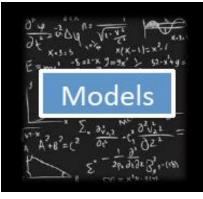






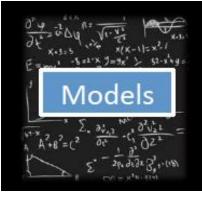






• When you make a model, you include the

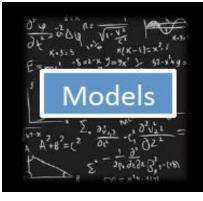
> elements that you feel are most important to explain a phenomenon.



• When you make a model, you include the

elements that you feel are most important to explain a phenomenon.

 Generally, we try to make models that can reproduce real-world data



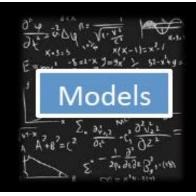
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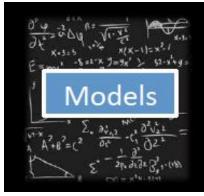
- Generally, we try to make models that can reproduce real-world data
- In E²M², we distinguish between statistical and mechanistic models

Statistical vs. Mathematical Model

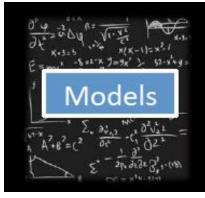
The choice depends on the research question!



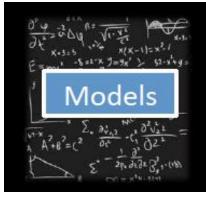
• Goal: To rigorously assess the strength of relationship between x and y



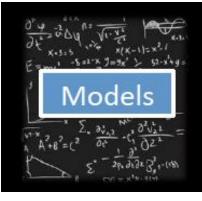
- Goal: To rigorously assess the strength of relationship between x and y
 - Find a significant relationship using a p-value as a measure of relationship strength

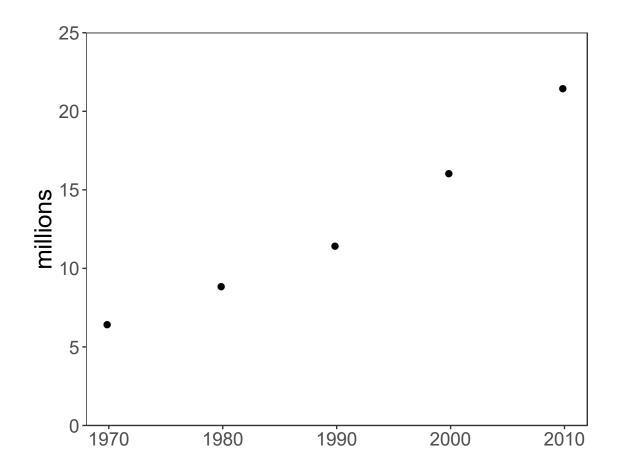


- Goal: To rigorously assess the strength of relationship between x and y
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 - Statistical models can demonstrate correlations.



- Goal: To rigorously assess the strength of relationship between x and y (describe patterns)
 - Find a significant relationship using a p-value as a measure of relationship strength
 - Statistical models can demonstrate correlations.
- Steps:
 - 1. Formulate a research question
 - 2. Formulate a hypothesis
 - 3. Develop a model to demonstrate your hypothesis.
 - 4. Collect data (required!!!)
 - 5. Evaluate hypothesis with appropriate statistical tools
 - t-test, Chi-square, ANOVA
 - Ordination (PCA)
 - Regression (LM, GLM, GLMM, GAM)

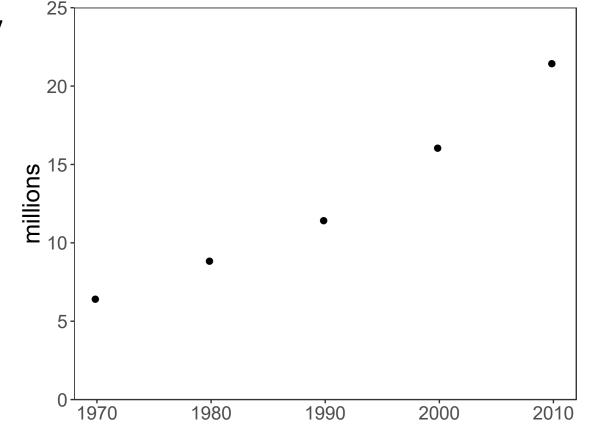


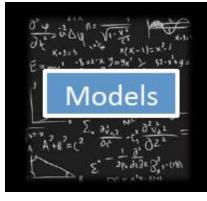




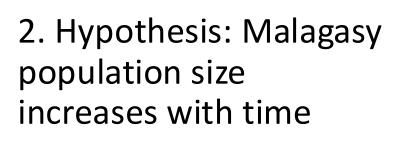
Vlodels

2. Hypothesis: Malagasy population size increases with time

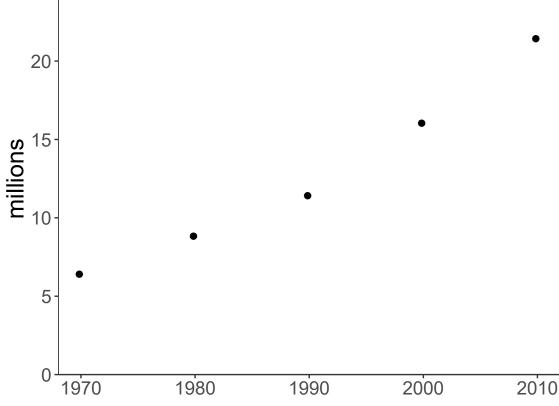


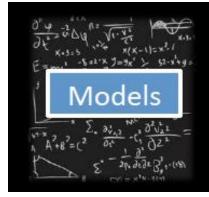


25



3. Statistical Model: y = mx + bLinear Regression

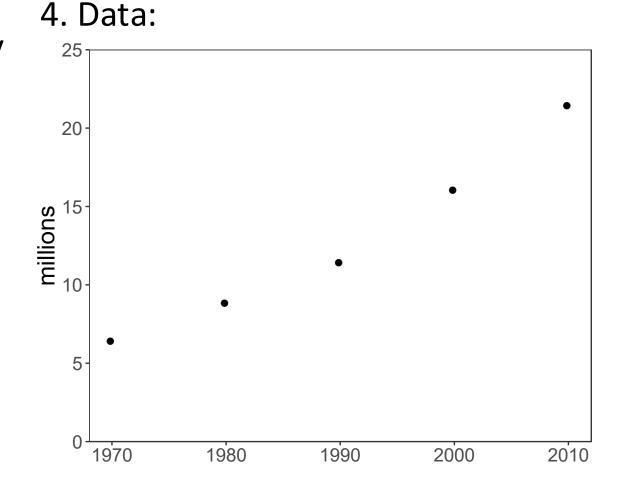




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Linear Regression



Source: World Bank

Viodels

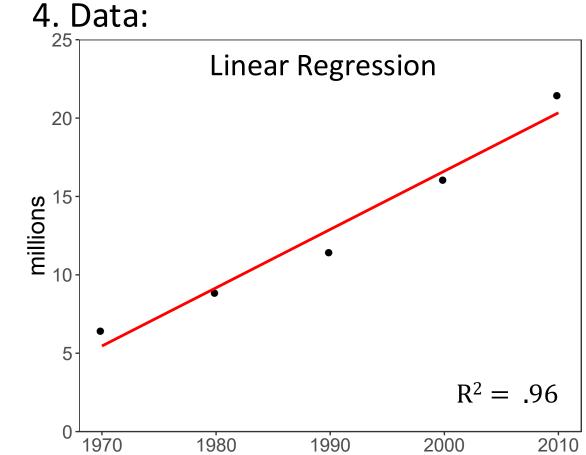
1. Example Question: What is the trajectory of Malagasy population size through time?

2. Hypothesis: Malagasy population size increases with time

3. Statistical Model: y = mx + b

5. Evaluation

m = .372 million p = .003





Vlode

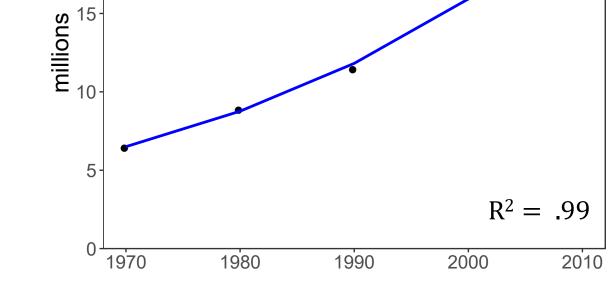
and re-evaluate: $y = e^{mx+b}$ Exponential Regression ω^{15}

25

p < .001

m = 0.029 mil.

6. Adapt your model



Exponential Regression

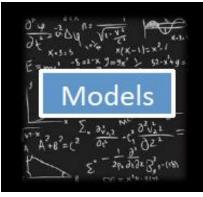
What can we conclude from this fitted model?

Source: World Bank

Vlode

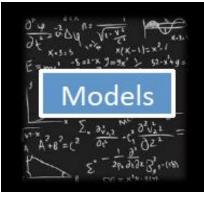
Statistical Models: Beware!

- Statistical models and tests are based on specific assumptions
 - data normally distributed
 - x and y independent



Statistical Models: Beware!

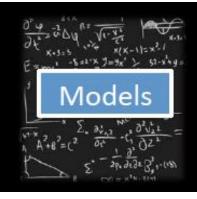
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Statistical Models: Beware!

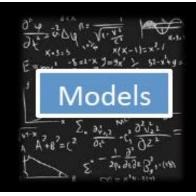
- Statistical models and tests are based on specific assumptions
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- There are so many statistical models...



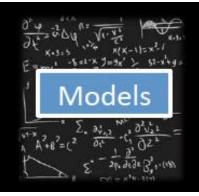


Statistical vs. Mathematical Model

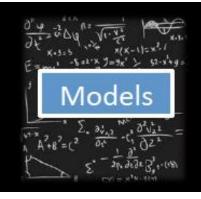
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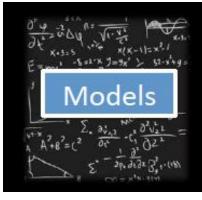
• Goal: To demonstrate the processes that underlie a relationship between x and y



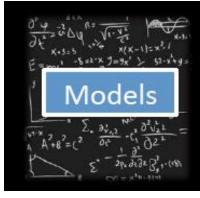
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 - Mechanistic models can demonstrate causation.



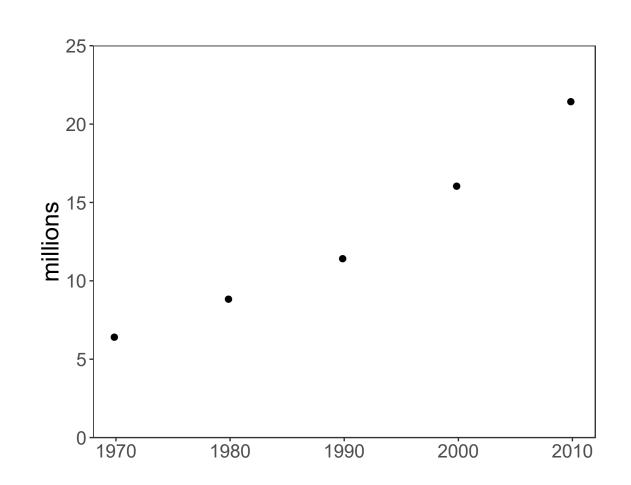
- Goal: To demonstrate the processes that underlie a relationship between x and y
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 - Mechanistic models can demonstrate causation.
- Steps:
 - 1. Formulate a research question
 - 2. Formulate a hypothesis
 - 3. Develop a model to demonstrate your hypothesis.
 - 4. Collect data (for certain questions)
 - 5. Evaluate the extent to which your model-simulated data matches that from the real world.

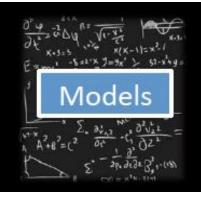


1. Example Question: **How** does Malagasy population size change with time?

2. Hypothesis: Malagasy population size increases because people are having children.

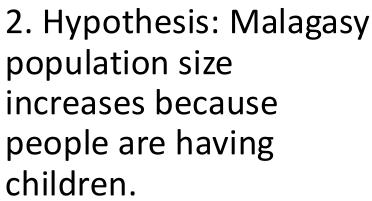
Can you think of an alternative hypothesis?



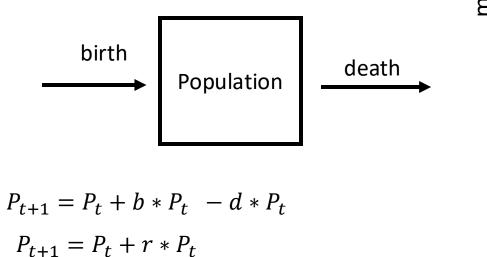


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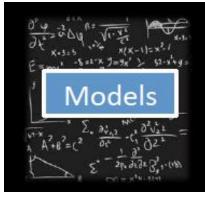
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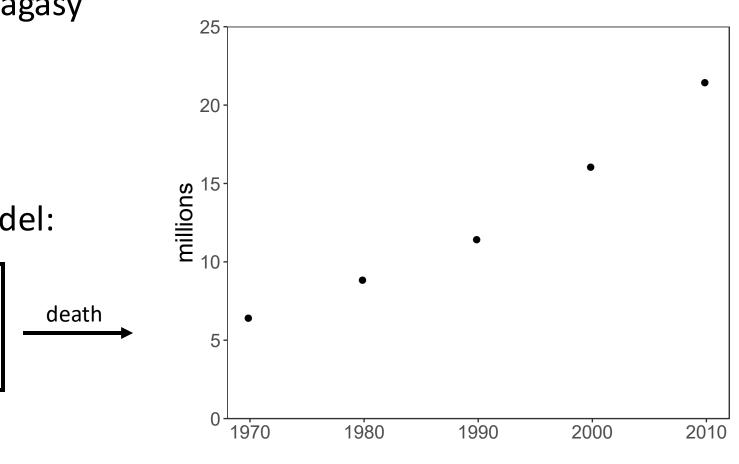


3. Mechanistic Model:





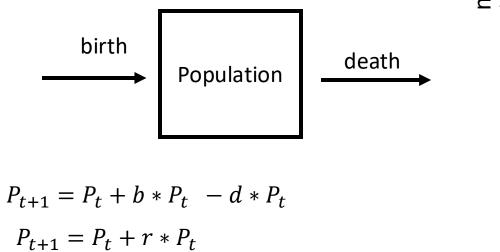


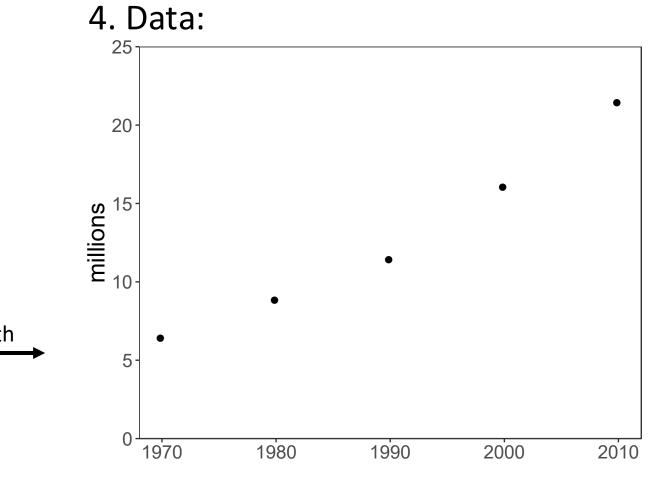


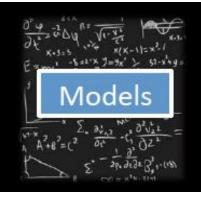
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3. Mechanistic Model:

5. Evaluation:

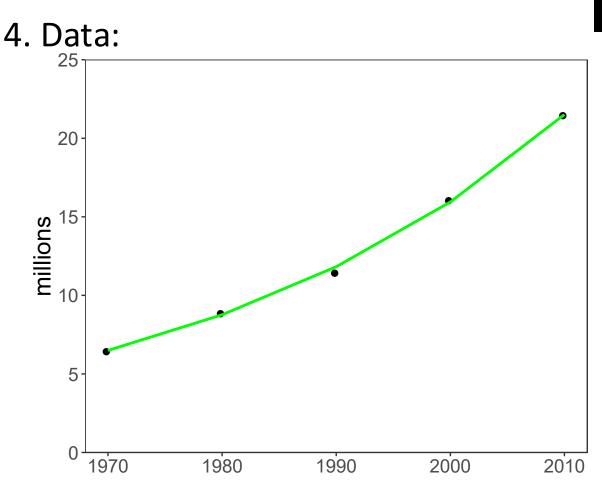
birth

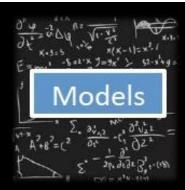
r = .349/person/yr

Population

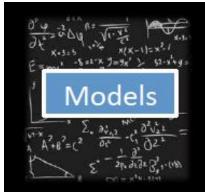
death

What can we conclude from this fitted model?

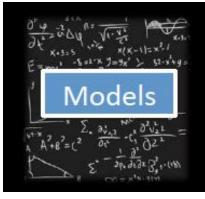




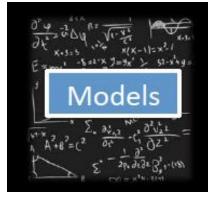
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- Parameters used in the mechanistic models sometimes are not measurable!
- Simulations can be computationally intensive

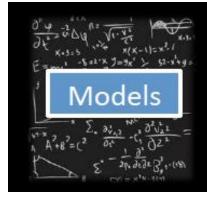


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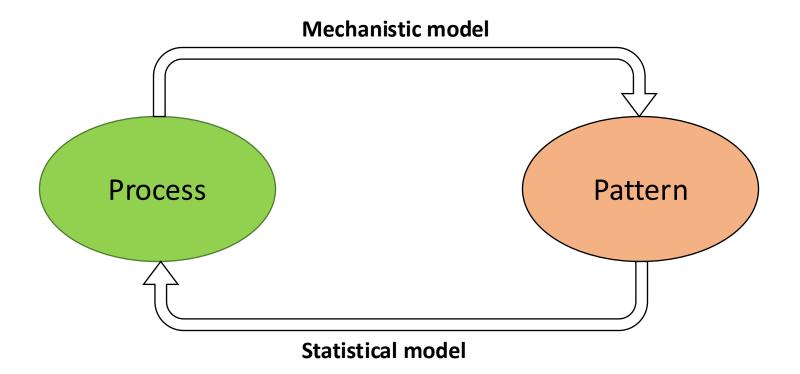
"All models are wrong but some are useful..." -George Box

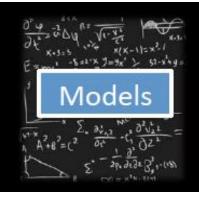


- Parameters used in the mechanistic models sometimes are not measurable!
- Simulations can be computationally intensive
- Advances in computational power often inspire development of more complex models which are not necessarily better

"All models are wrong but some are useful..." -George Box We use models to both **predict** and **explain**.

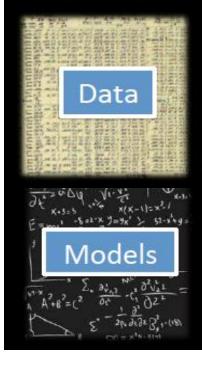
It is ideal when statistical and mechanistic models meet:





A Tool for E^2M^2

- Computer power keeps increasing
- Language/software
 - Fortran, C, C++
 - Julia, Java, Python
 - Matlab, Maple, Mathematica,
 - SAS, SPSS, Stata
- Specific programs
 - Vortex, RAMAS, NetLogo for IBM
 - NicheMapper for physiology, iLand for forest dynamics
 - MaxEnt for species distribution modeling
 - Zonation for reserve selection etc...
- The compromise: R---very powerful for
 - Visualization
 - Data formatting and sorting
 - Statistical analyses
 - Simulation (mechanistic model)





Goals for this lecture

- To explain what we're doing here
- To define "science"
- To define "data"
- To define "models"
- To introduce many different types of models
 - Statistical
 - Mathematical
- To introduce the "E" in E^2M^2
 - Ecology
 - Epidemiology

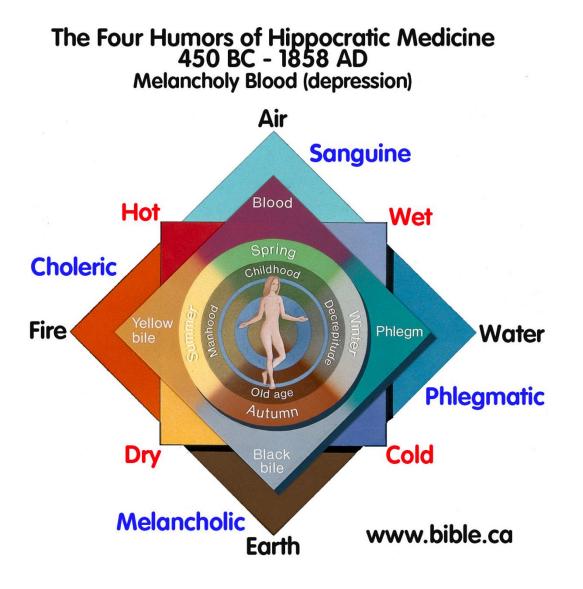
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What is Epidemiology?

- "the study of **what** is on the people"
 - coined in 1802 to describe diseases in the Spanish population
- Emphasis on the study and analysis of the distribution and determinants of health and disease ("risk factors")

1. Sickness caused by an imbalance in the four humors (Hippocrates)

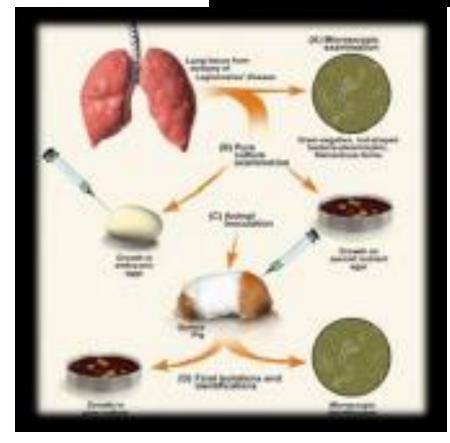


- 1. Sickness caused by an imbalance in the four humors (Hippocrates)
- 2. Miasmatic theory of disease (1500s)
 - Sickness results from emanations of 'bad air'



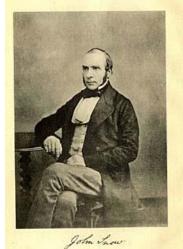
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- 3. Germ theory of disease
 - Leeuwenhoek's microscope (1675)
 - Koch's postulates (1890)



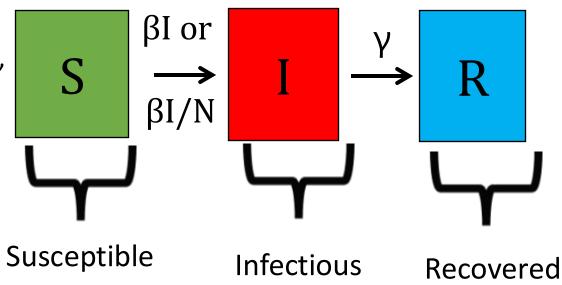


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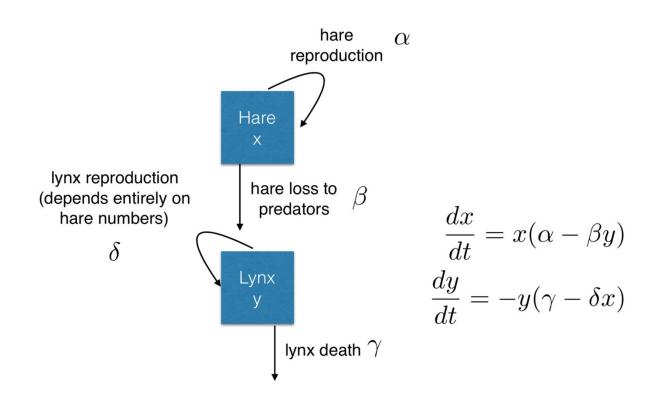
What is Ecology?

- The study of the interactions of organisms and their environment
 - Coined in 1866 by German scientist Ernst Haeckel
 - Nile crocodiles opening mouths for sandpipers (Herodotus)
- Emphasis on explaining dynamical processes in nature

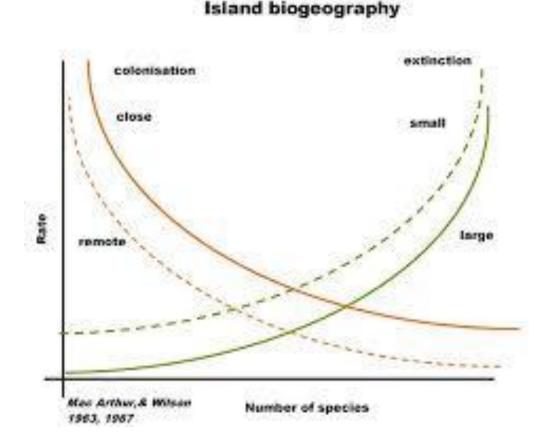
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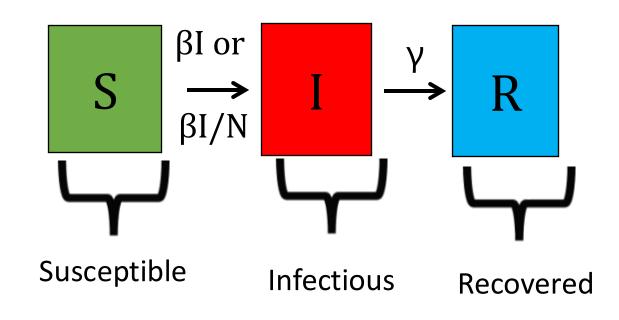
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- 5. Disease Ecology
 - Anderson and May (1980s)
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Misaotra!