

E²M²: Ecological and Epidemiological Modeling in Madagascar

Data and Models

Centre ValBio

Ranomafana National Park, Madagascar

December 2022

Thanks to our sponsors!



Mahaliana
IT ALWAYS STARTS WITH A QUESTION

Lecture contributions from:

Tanjona Ramiadantsoa

Steve Bellan

Goals for this lecture

- To explain what we're doing here

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 - Ecology
 - Epidemiology

E2M2.org

Friday: R Bootcamp

- Intro to R Studio
- Exploring and Visualizing Data in R
- For-loops, Functions, and If-Else Statements

Thursday: “Bells and Whistle”

- Introduction to Phylogenetic Modeling • Tutorial
- Introduction to Spatial Visualization and Plotting • Tutorial
- Modeling Insights from the Metapopulation Game • Tutorial
- Introduction to Network Modeling • Tutorial
- Research Snapshots

- Programming
- Data
- Models
- Research Development

Saturday: Travel

Sunday: “Getting Started with the Basics”

- Data and Models
- Student introductions & presentations
- Linear regression & simple statistics
- Linear regression tutorial

Monday: “Using Models with Data”

- Model-Guided Study Design
- Study design tutorial
- Intro to Compartmental Models & Differential Equations
- Building mechanistic models in R
- Refining research questions for modeling
- Defining a model world

Wednesday: “Refining Your Work”

- Model Fitting in Practice – the Basic Concept
- Epidemic Cards
- Model Fitting with Epidemic Cards
- Model Selection and Comparison
- Model Selection Tutorial
- Model Telephone

Tuesday: “Applying Simple Models”

- Dynamical Fever
- Introduction to Mixed Modeling
- Reading a Research Paper

Friday: “Putting it All in Perspective”

- The Life Cycle of a Modeling Project
- C4C Student Presentations

Saturday: Travel

January: “Sharing Your Work”

- Final student presentations

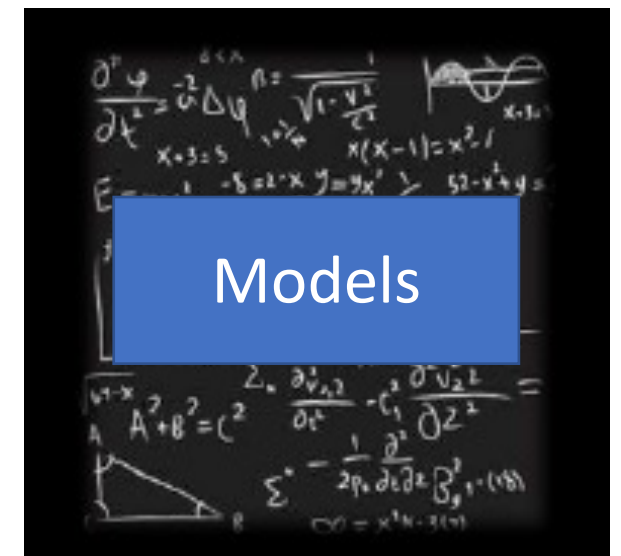
What is science?

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*

Observations and Laws and Principles

Data and Models





Data and Models

- What is data?



Data and Models

- What is data?
 - Backbone of science

What is science?

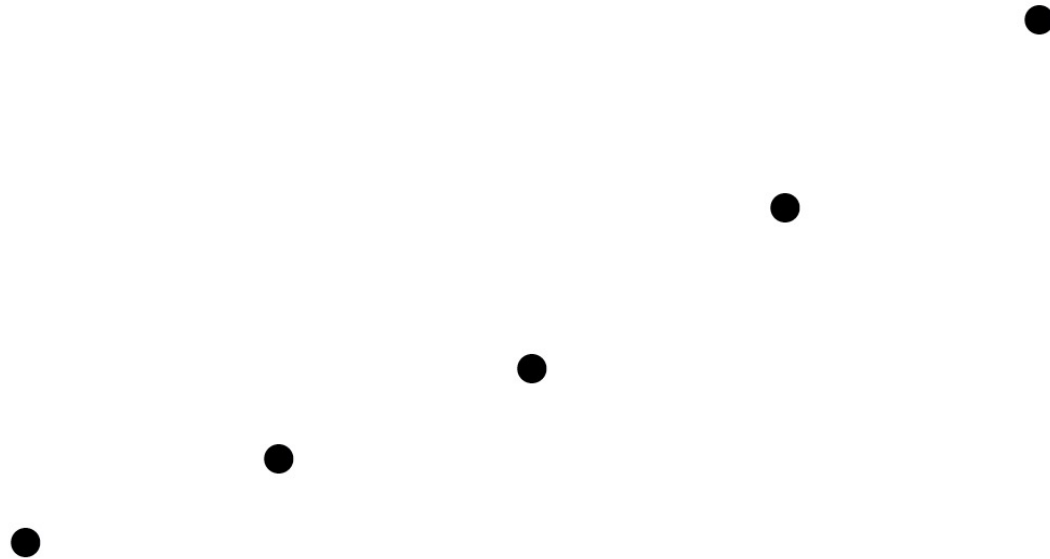
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Data and Models

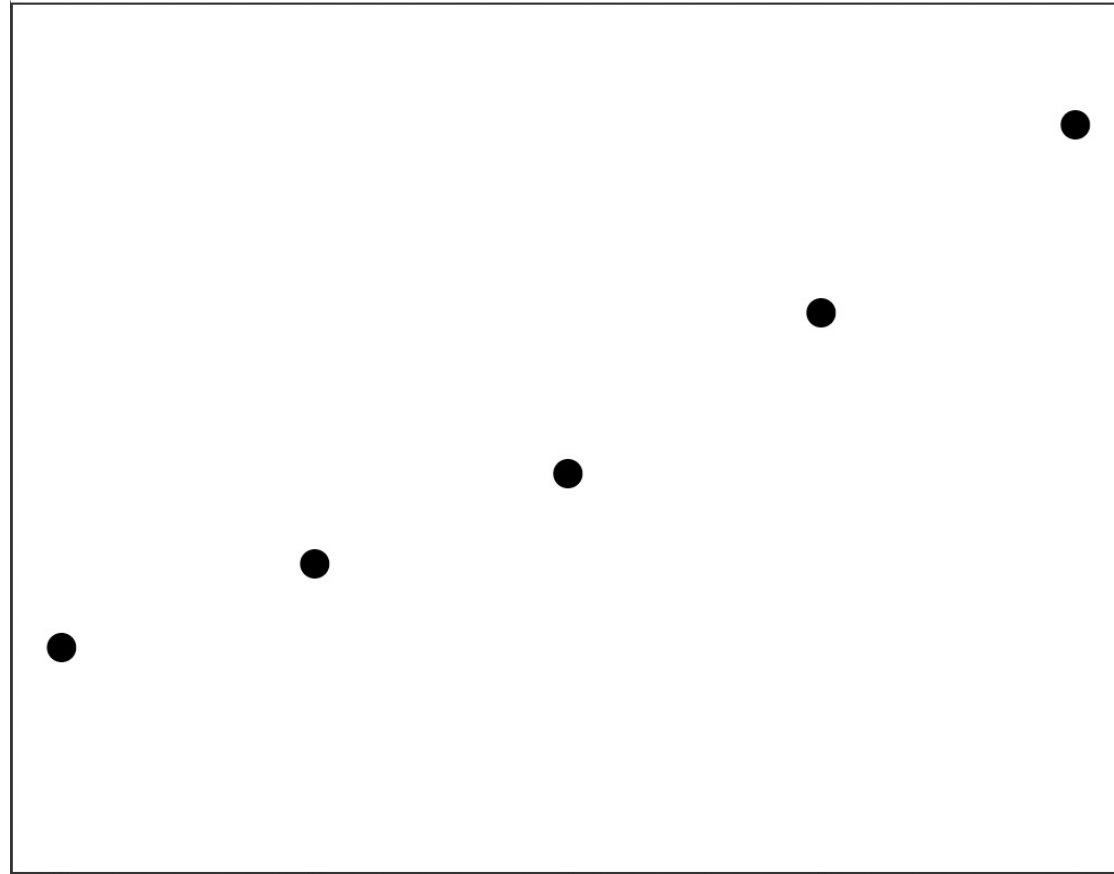


- What are data?
 - Evidence to support a claim

Are these data?

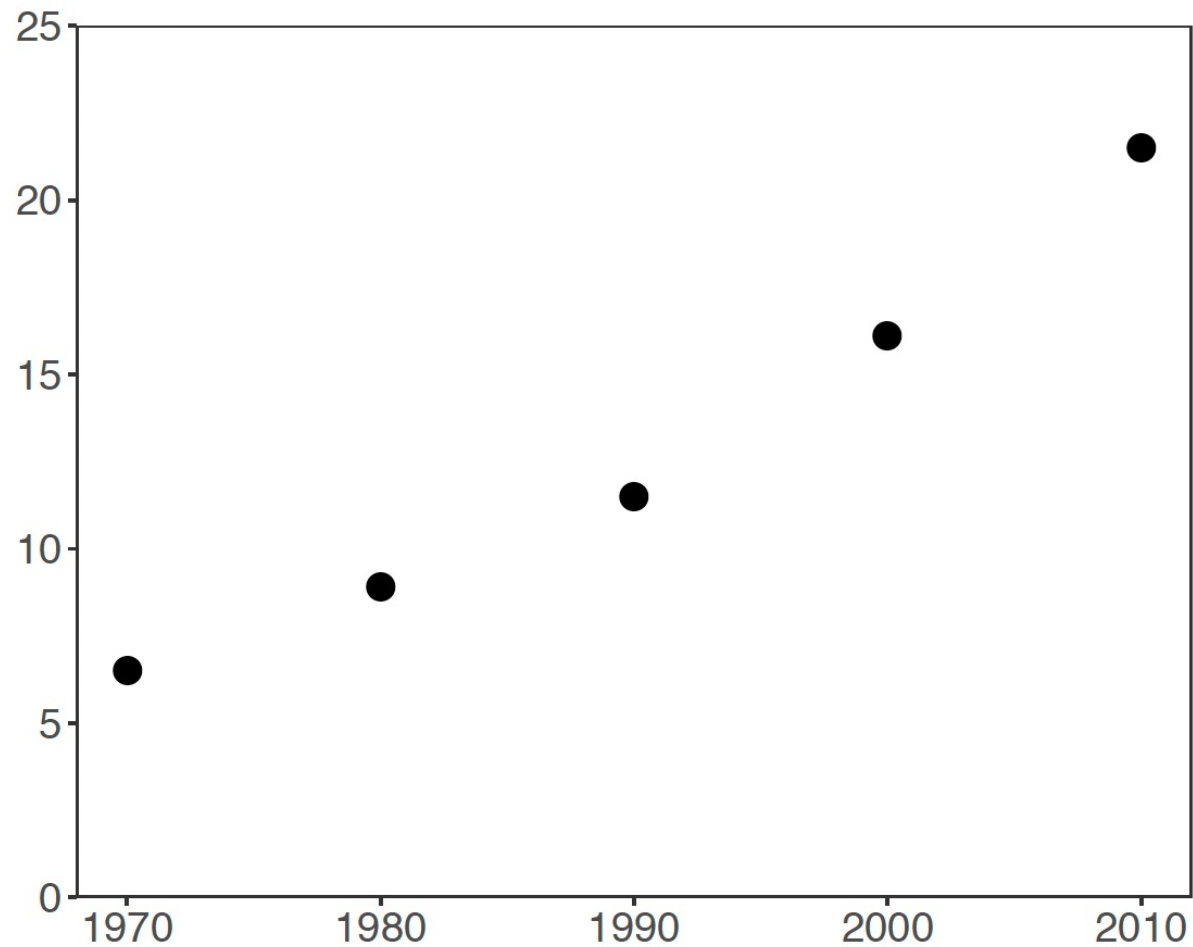


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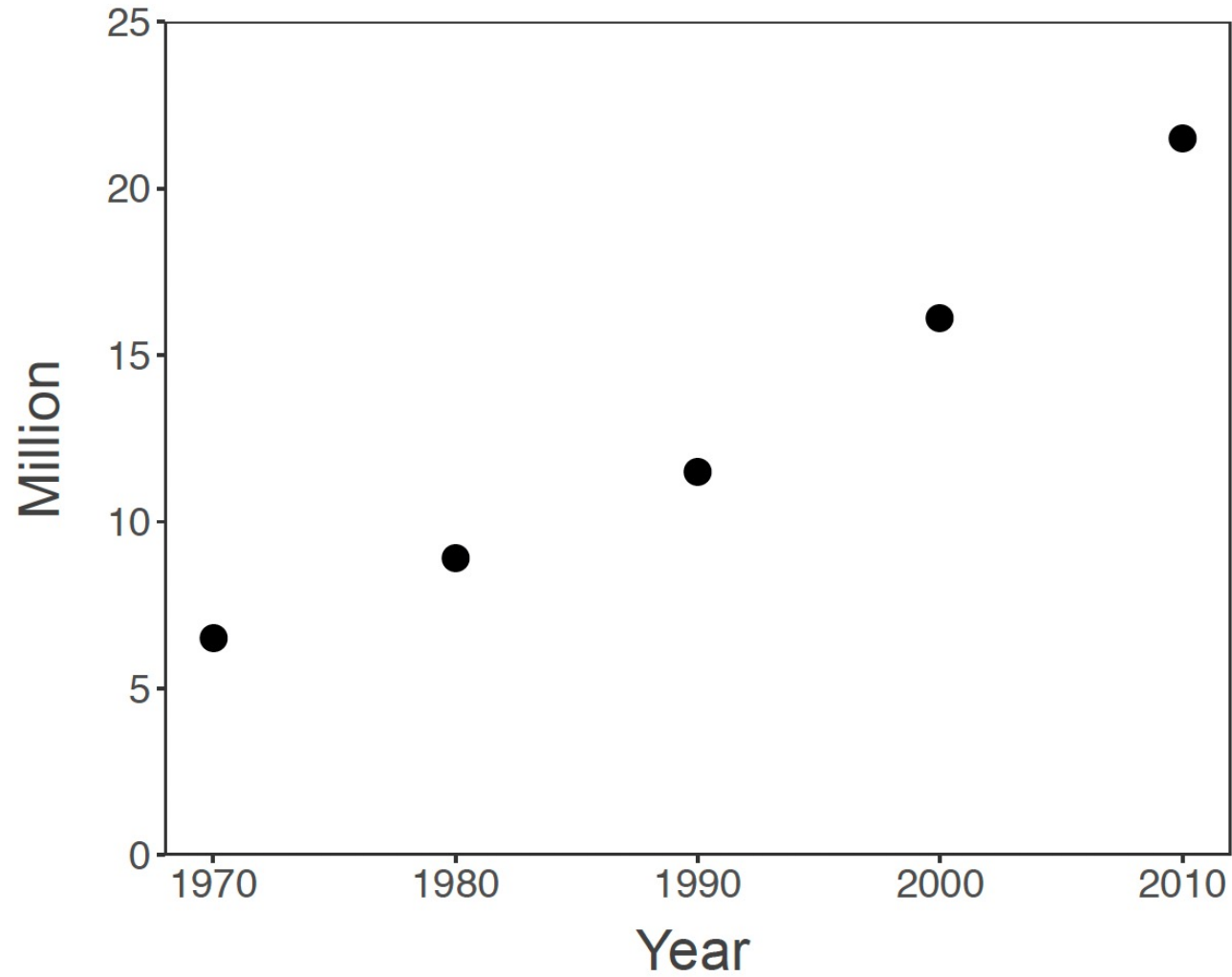


What do we need to make these data?

Are these data?

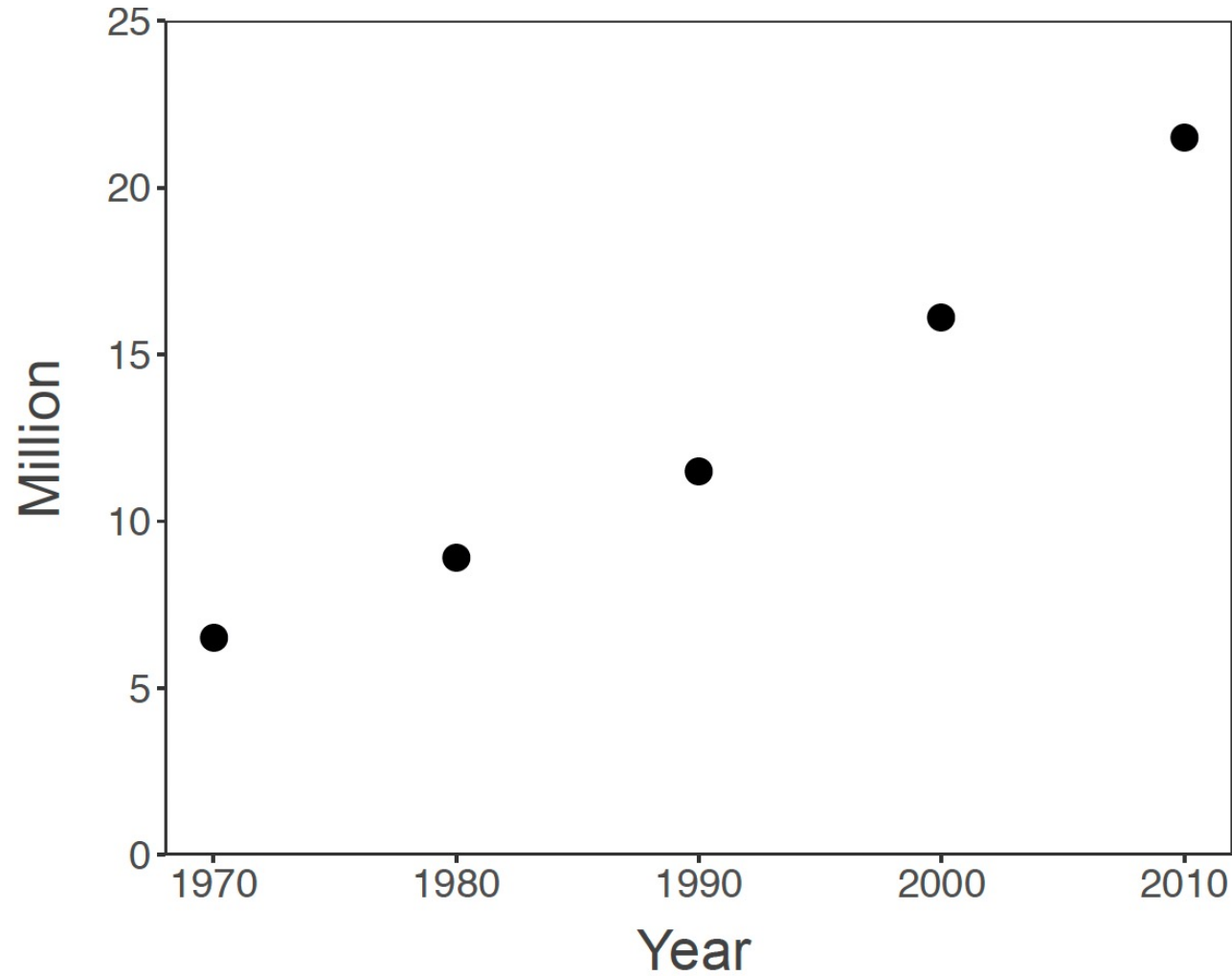


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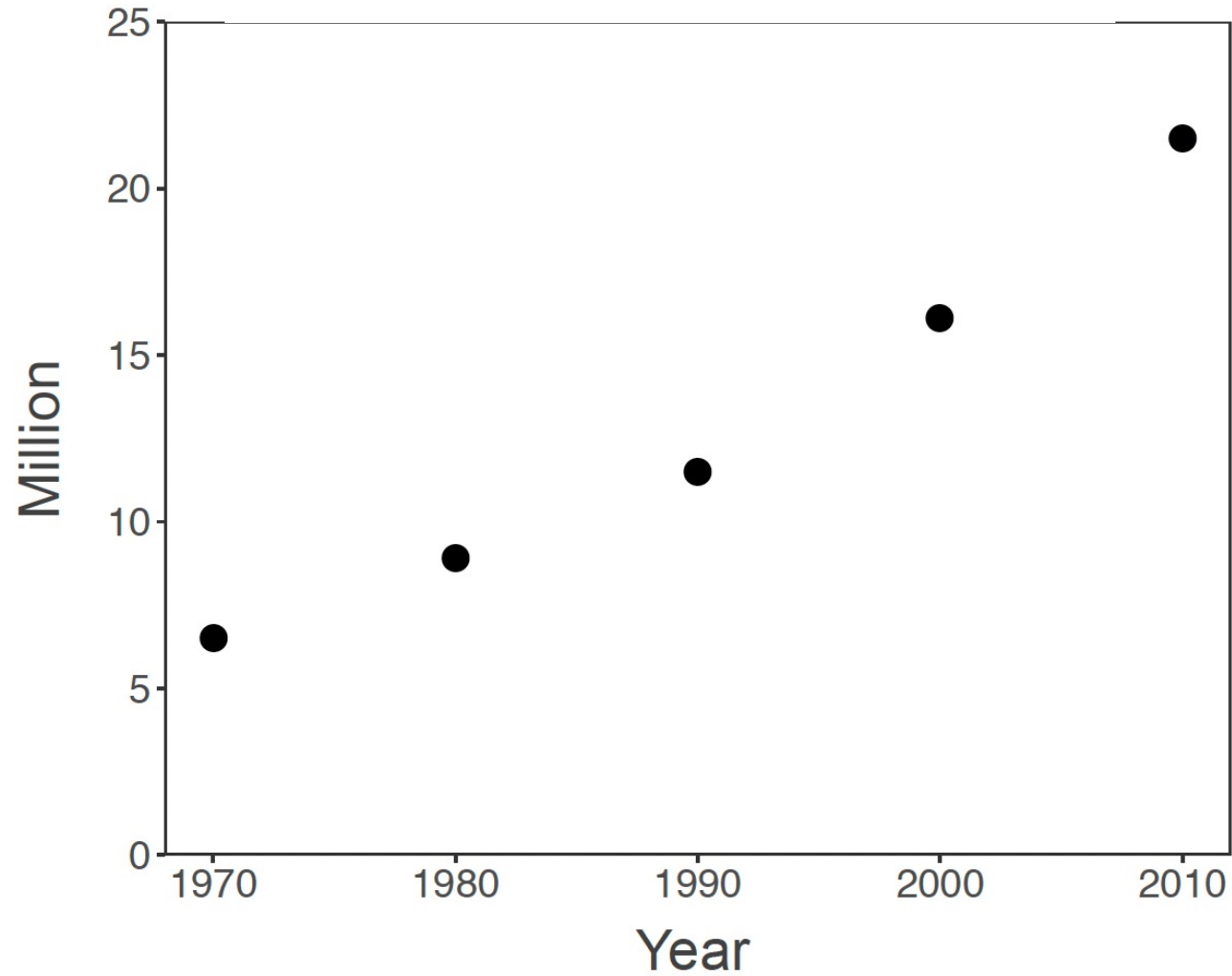
Number of views for Justin Bieber's "Baby" video on Youtube



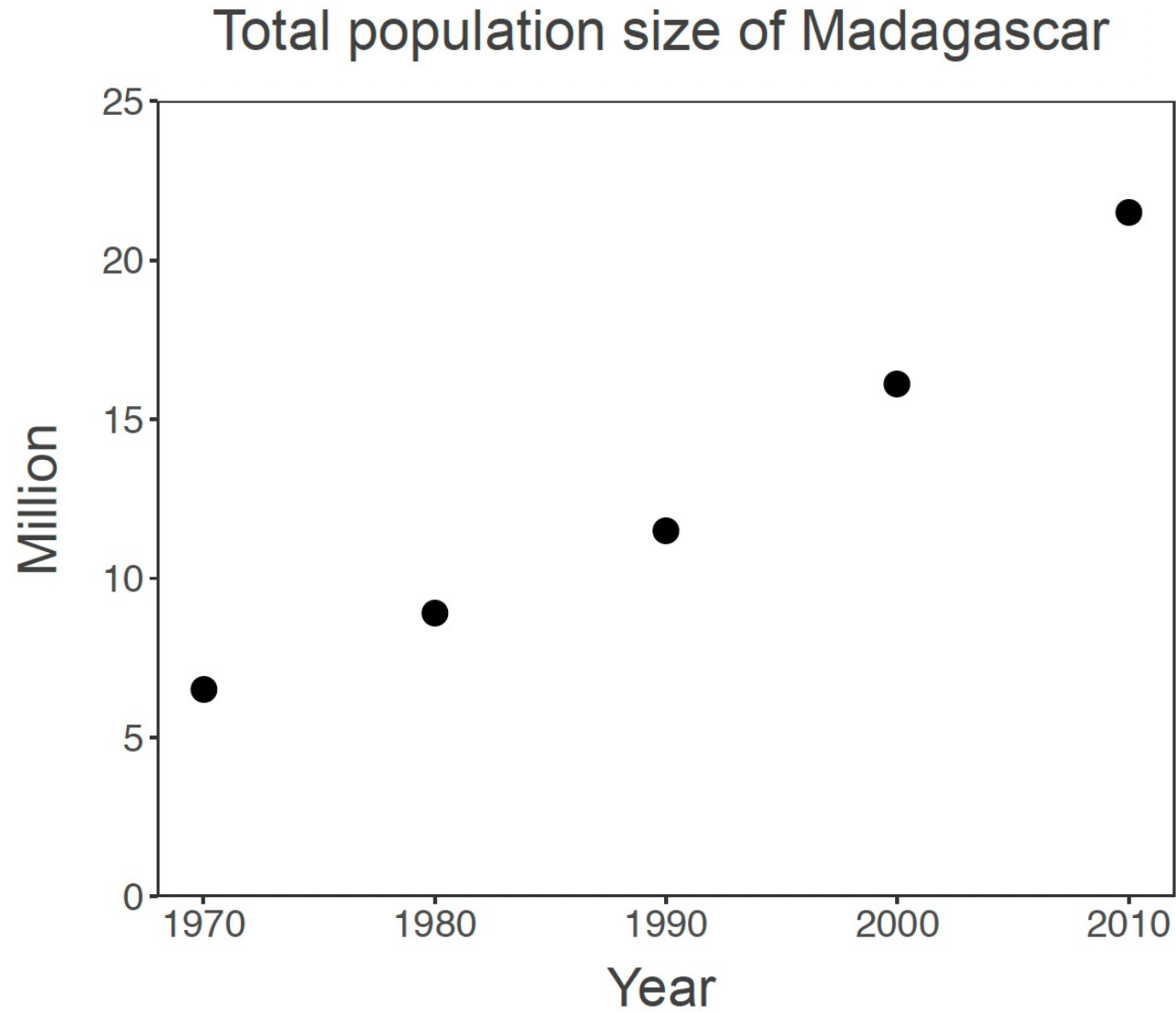
Are these data?



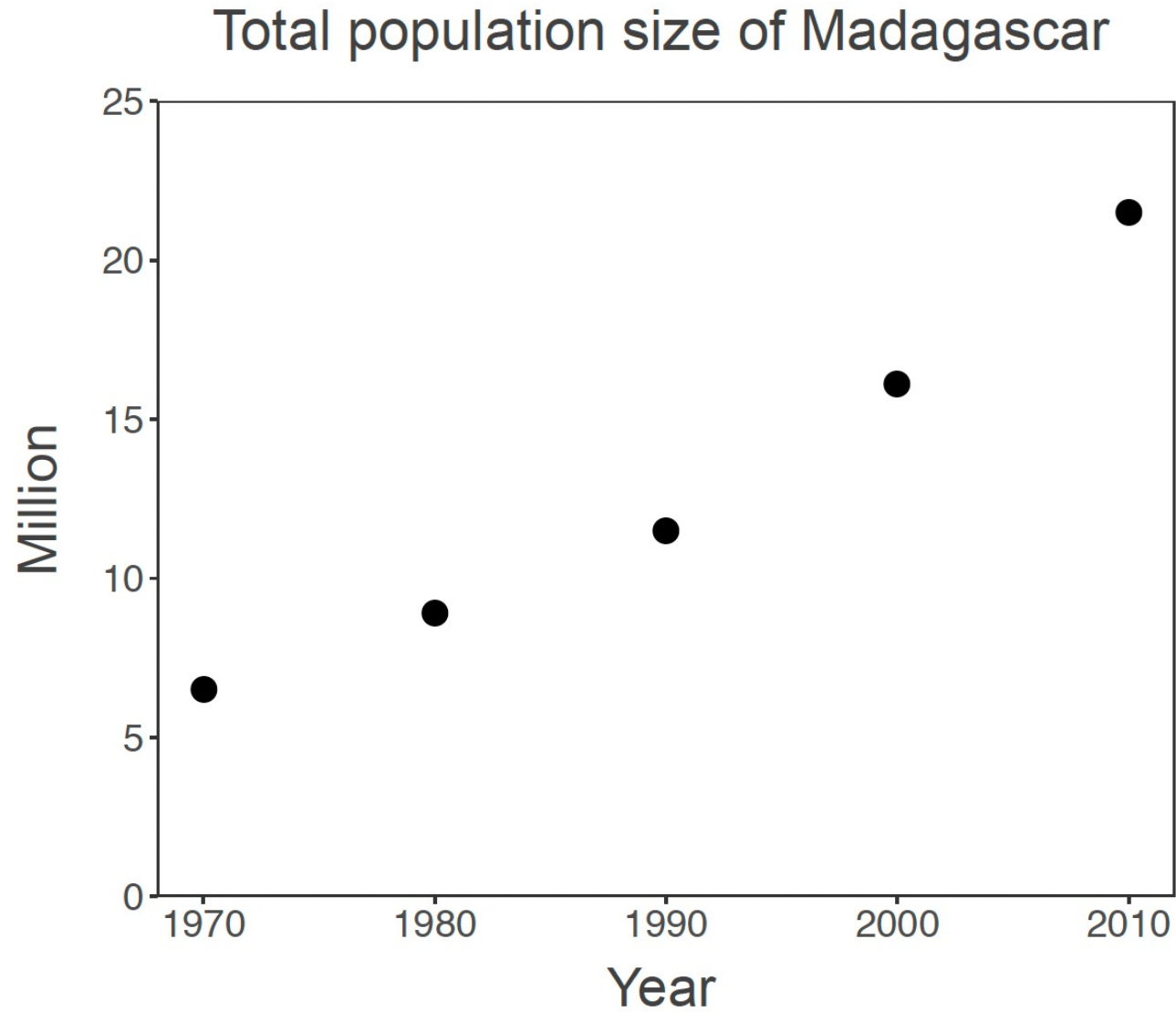
Total population size of China



Are these data?



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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)

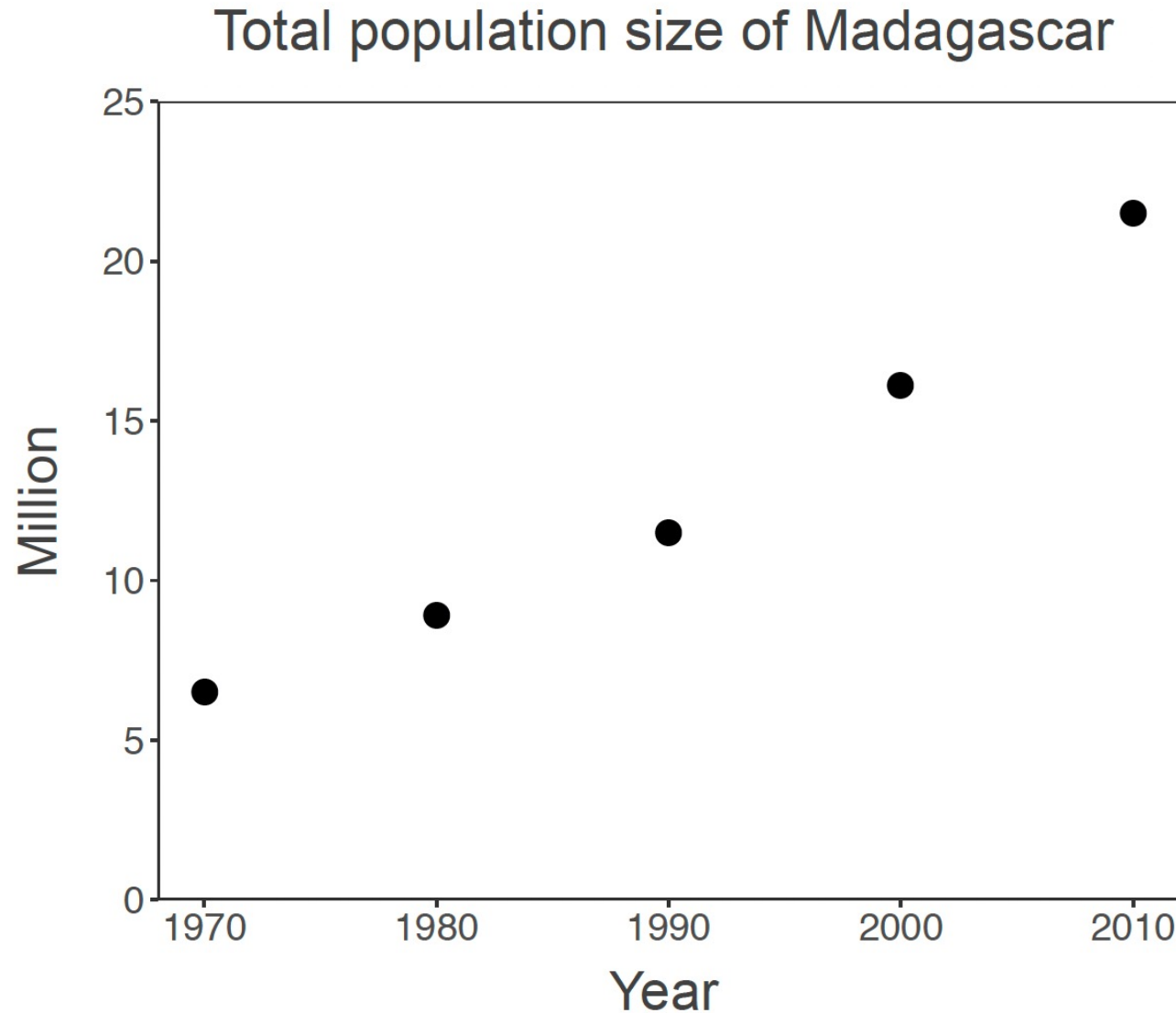
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 - with respect to the **question!**

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- Backbone of science
- **Evidence** to support a **claim**

Data provide **evidence** to support a **claim**.

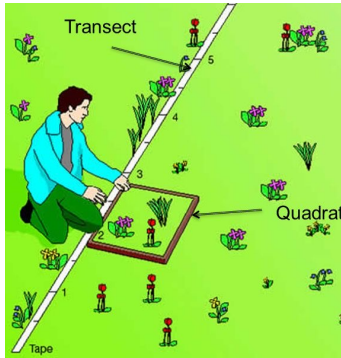


CLAIM: *The population size of Madagascar has increased throughout the past 50 years*

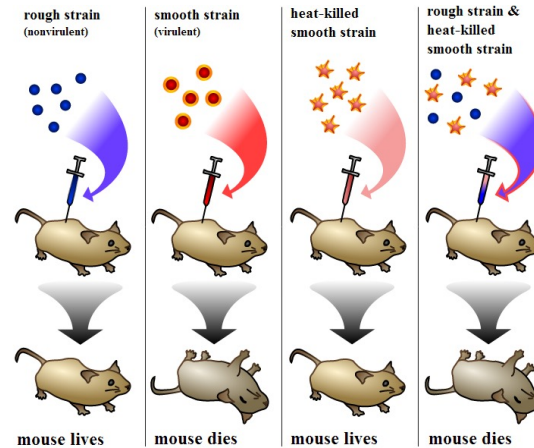
Data: Sources of x and y

Data

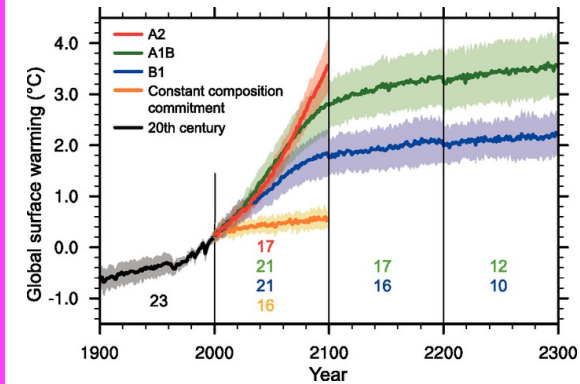
Observational



Experimental



Simulated



Empirical data

Data: Types



Numerical

Categorical

Data: Types



Numerical

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

Categorical

Data: Types



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- Integer, real number, multi-dimensional number

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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: Things to consider





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- Data acquisition
 - Impossible, example?



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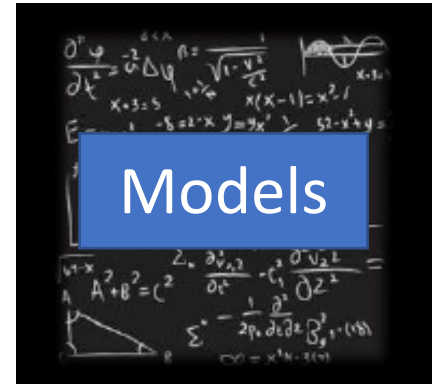
Data and Models

Data

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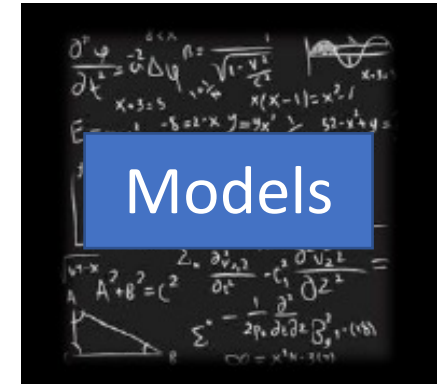
- What is a model?



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

Laws and Principles



- A theory = a declaration to explain a phenomenon
 - Logical and falsifiable
- A **model** = an abstract representation of a phenomenon
- A hypothesis = a testable declaration that is derived from a theory

Theory, Models, Hypotheses

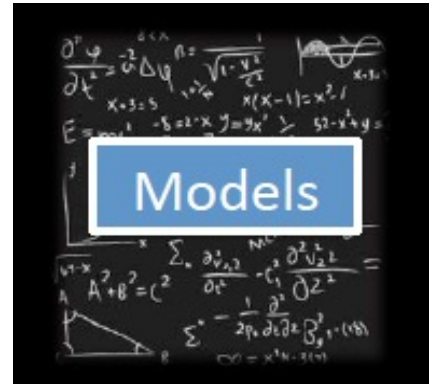
Theory

Model

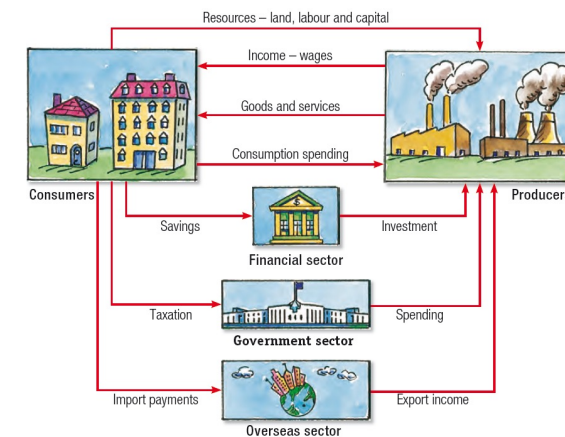
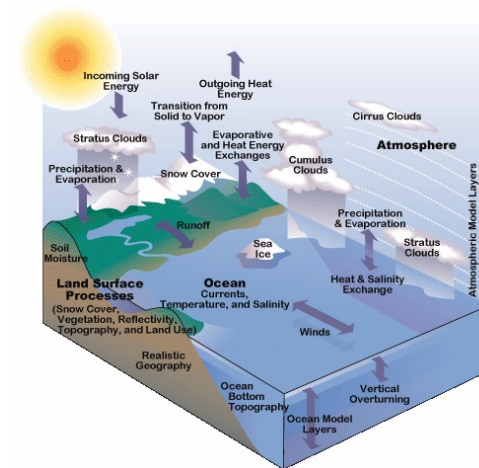
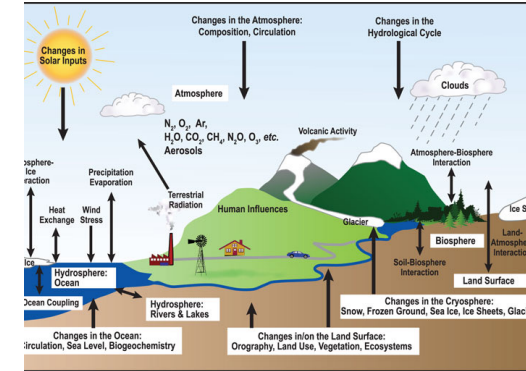
Hypothesis

General

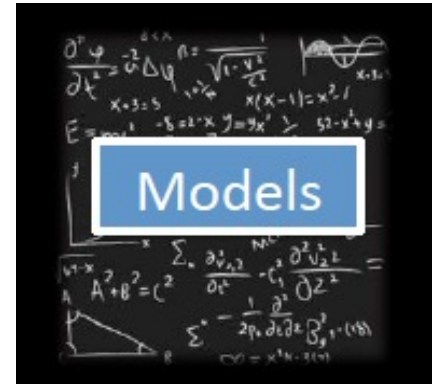
Specific

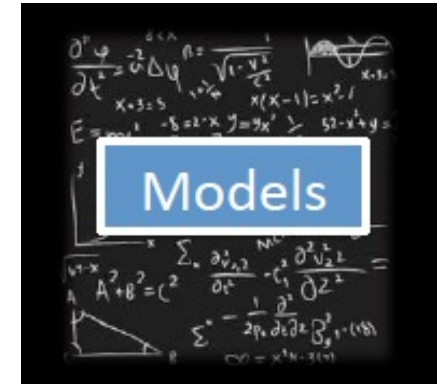


Models



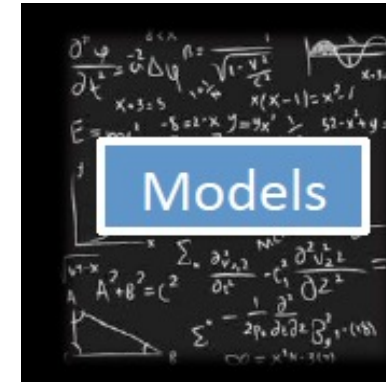
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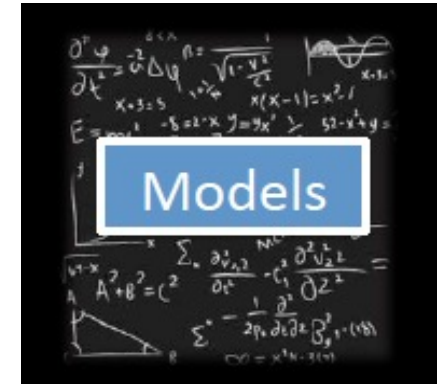


- When you make a **model**, you include the **elements that you feel are most important** to explain a phenomenon.
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- In E^2M^2 , we distinguish between **statistical** and **mechanistic** models



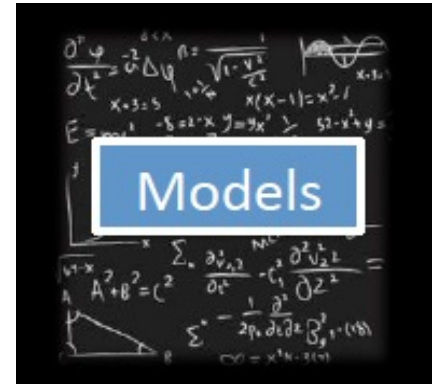
Statistical vs. Mathematical Model

The choice depends on the research question!

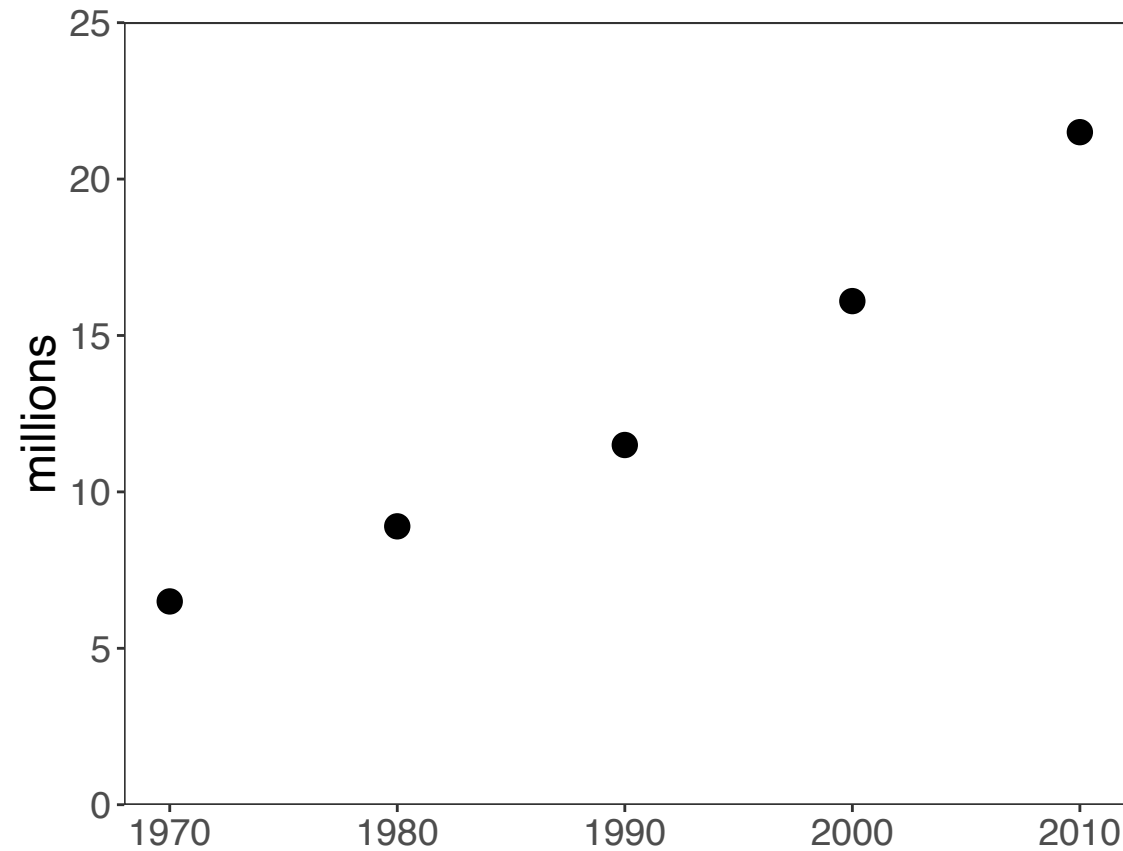
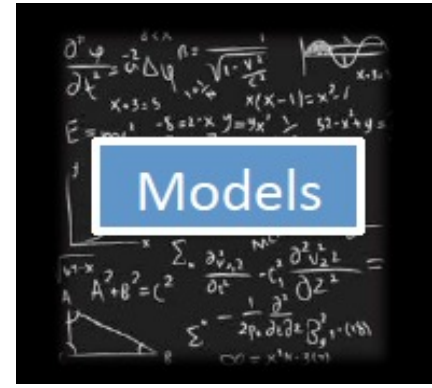


Statistical Models

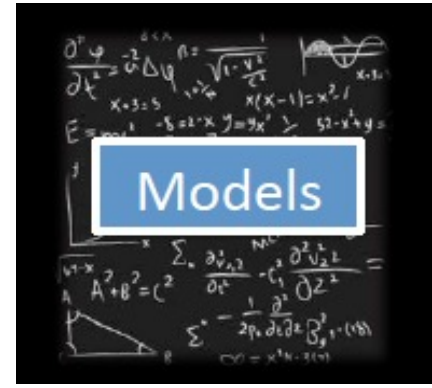
- Goal: To rigorously assess the strength of relationship between x and y
 - Find a significant relationship using a p-value as a measure of relationship strength
 - **Statistical models can demonstrate correlations.**



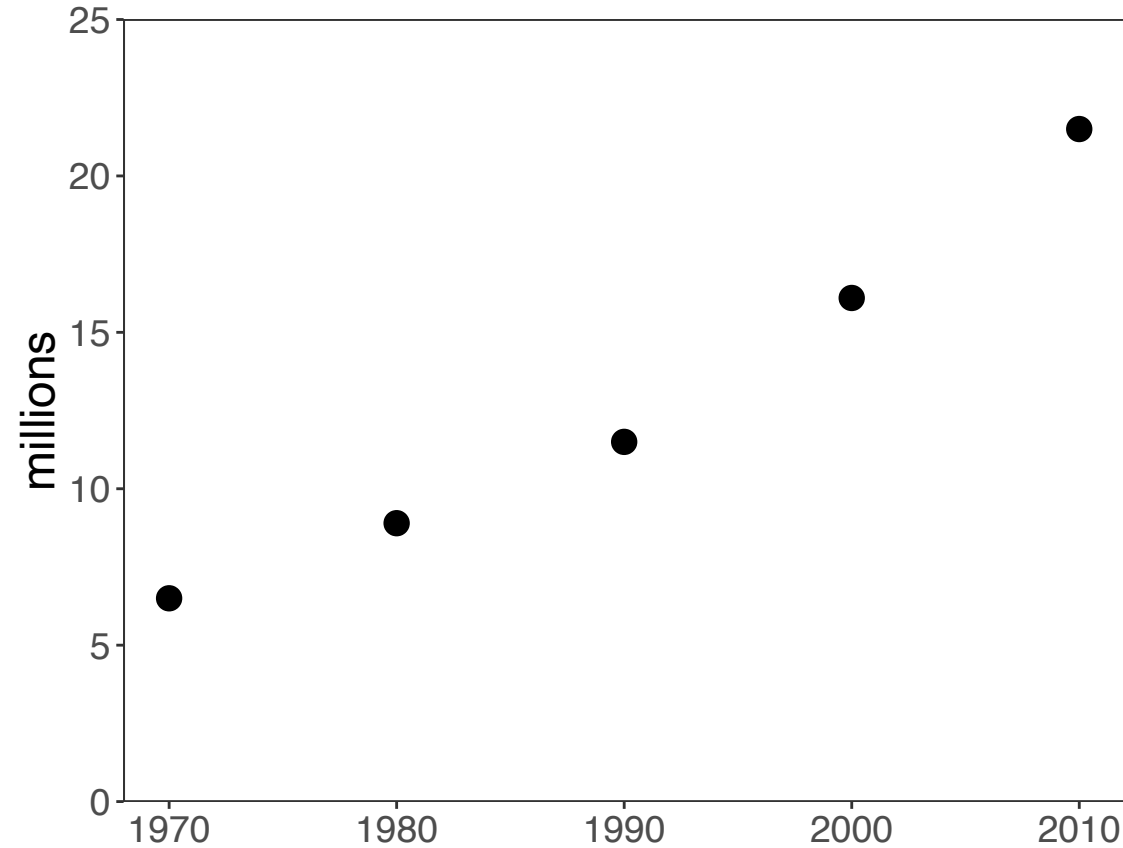
1. Example Question: **What** is the trajectory Malagasy population size through time?



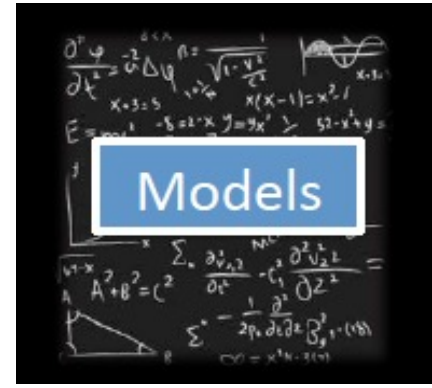
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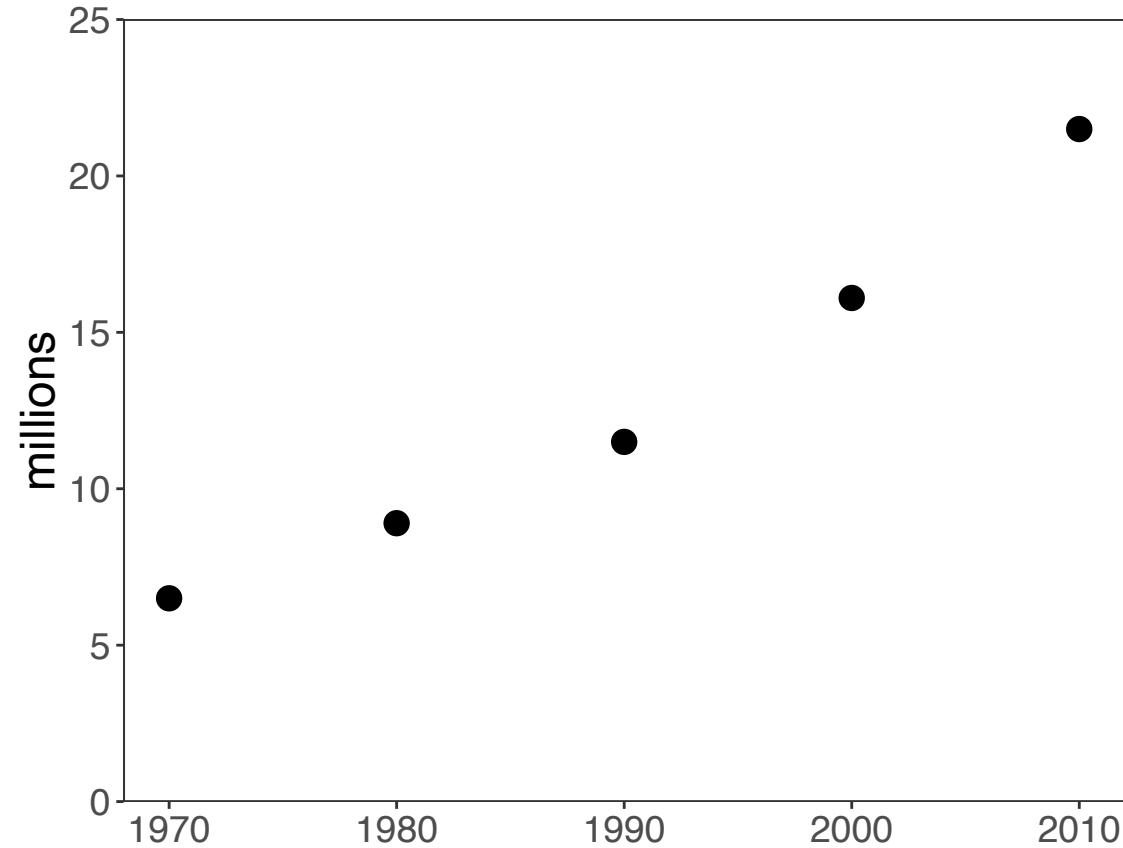


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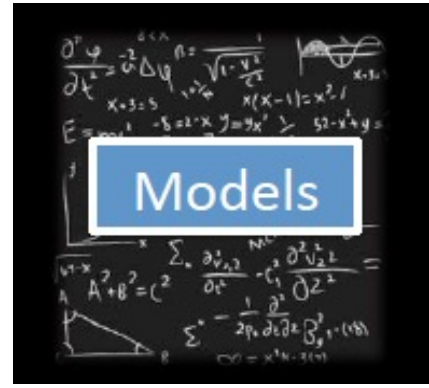
3. Statistical Model:

$$y = mx + b$$

Linear Regression



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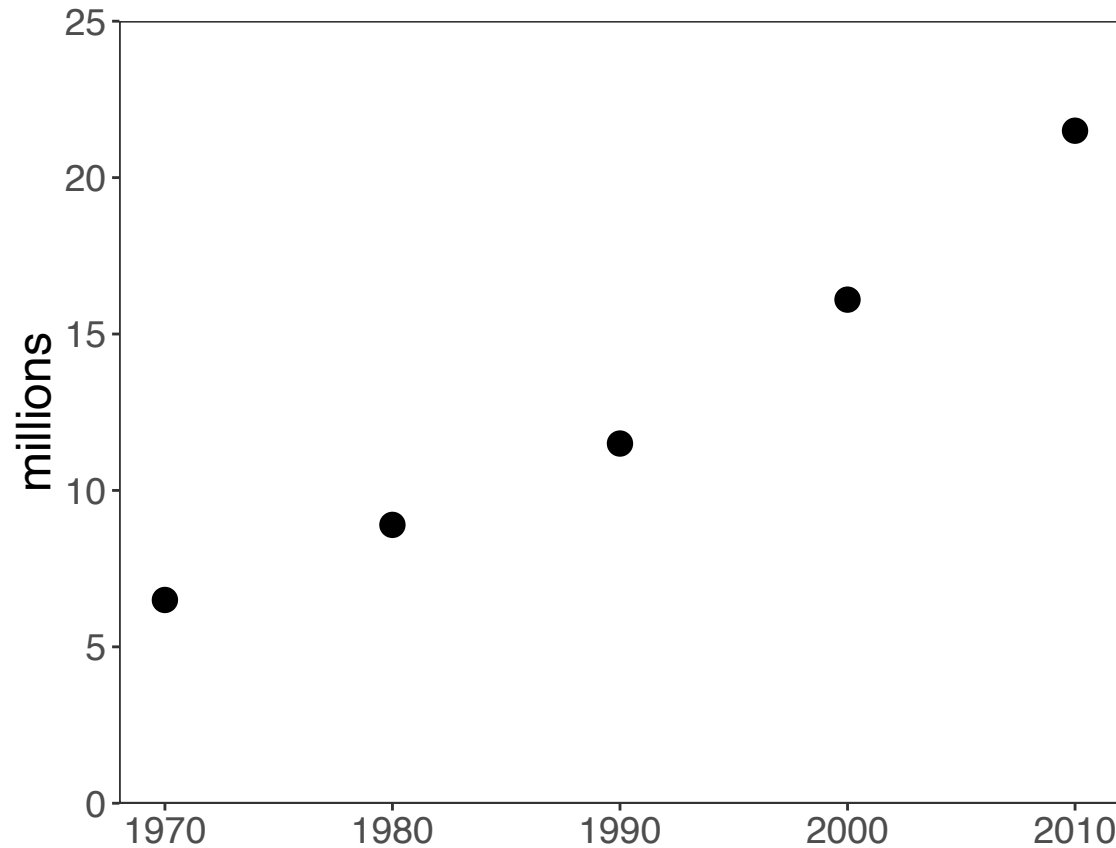
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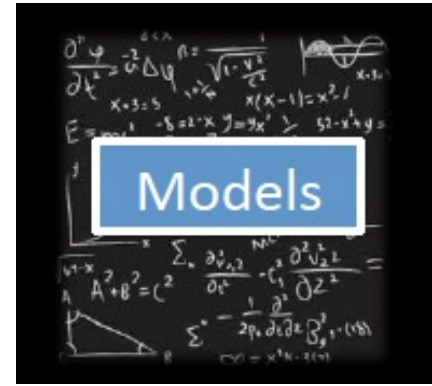
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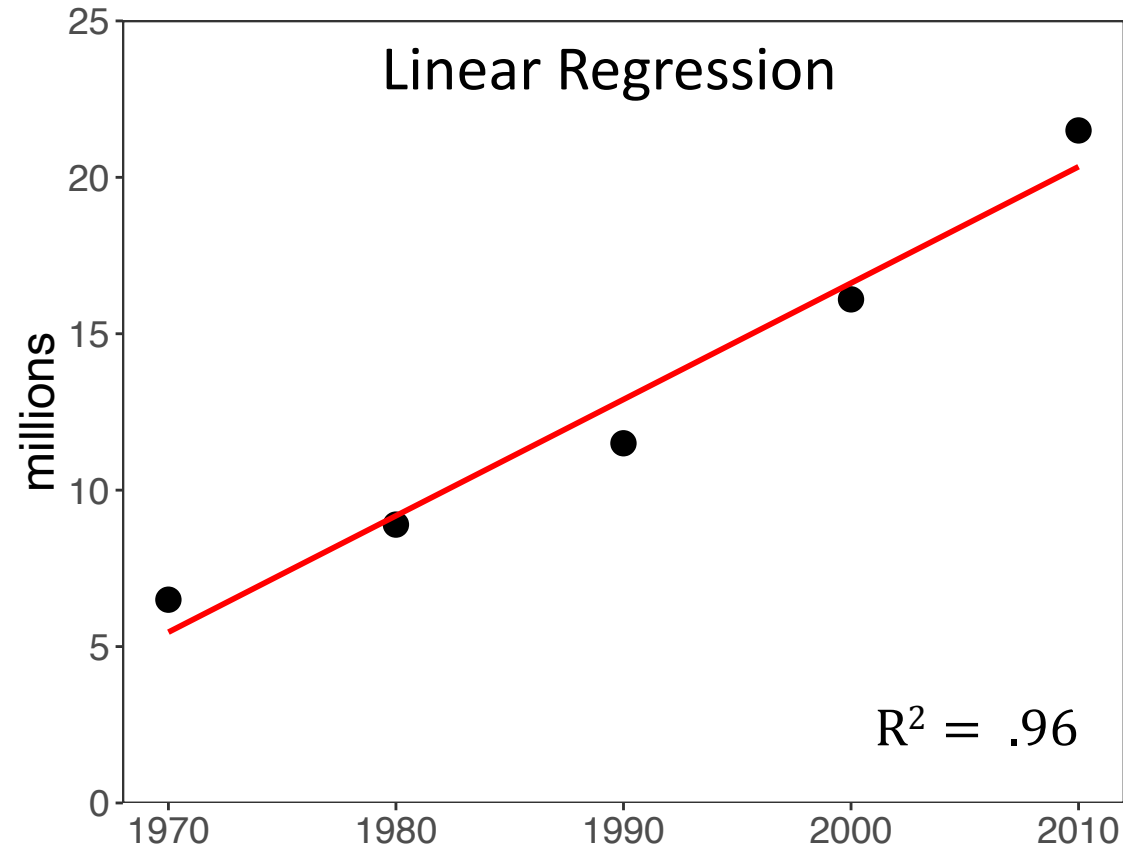
$$y = mx + b$$

5. Evaluation

$$m = .372 \text{ million}$$

$$p = .003$$

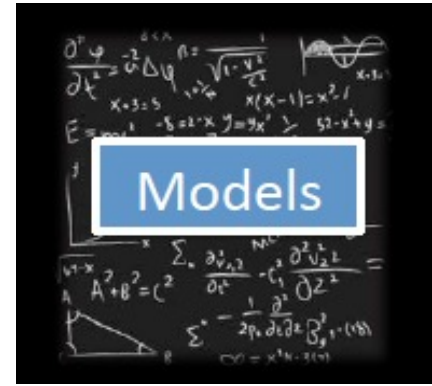
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What can we conclude from this fitted model?

Source: World Bank

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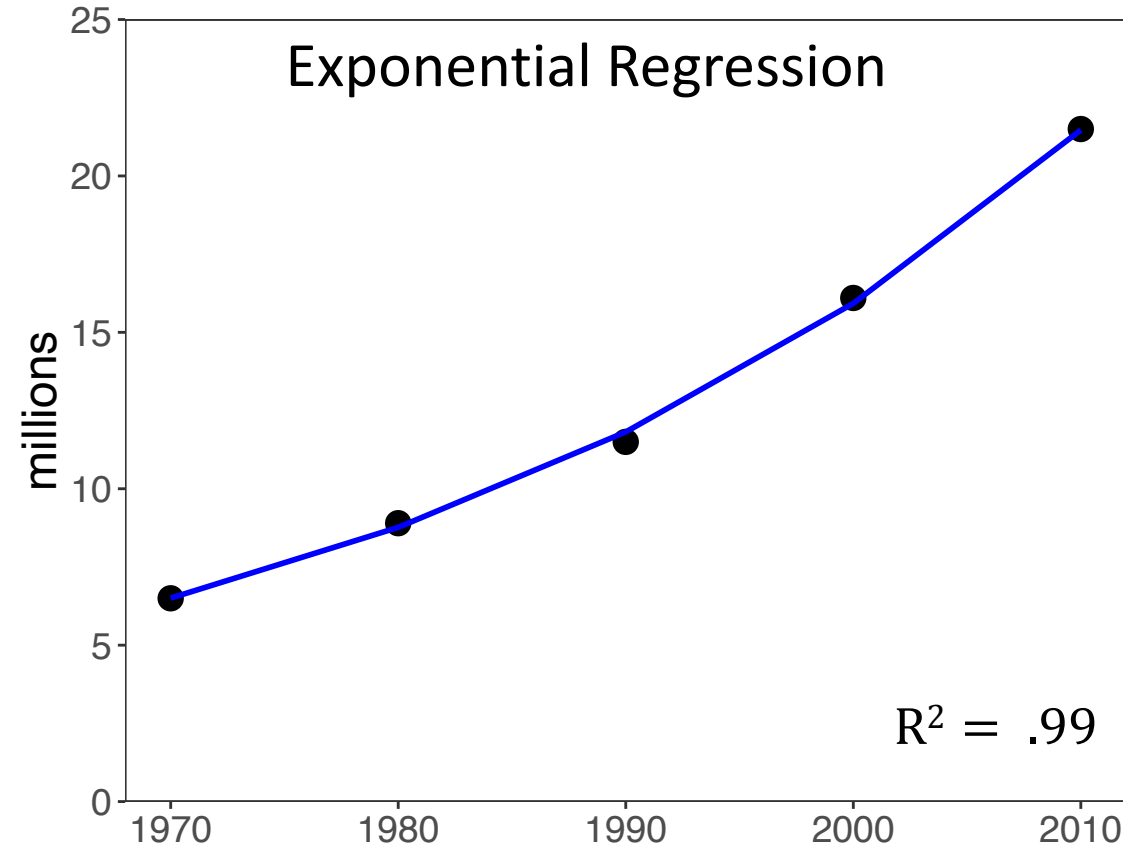
7. Adapt your model and re-evaluate:

$$y = e^{mx+b}$$

Exponential Regression

$$m = 0.029 \text{ mil.}$$

$$p < .001$$

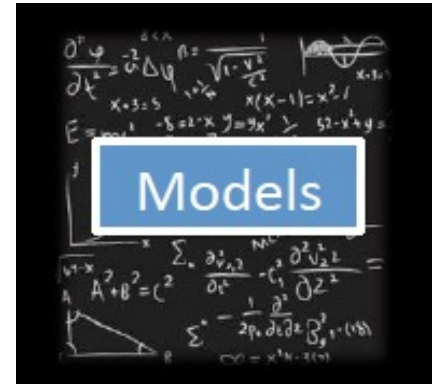


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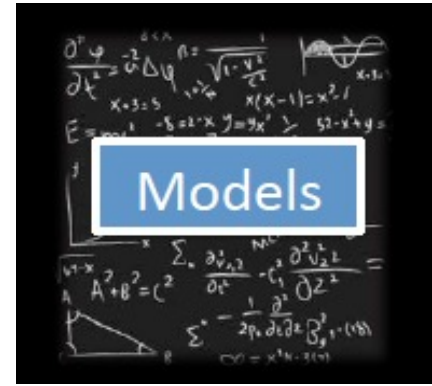
Statistical Models: Beware!

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



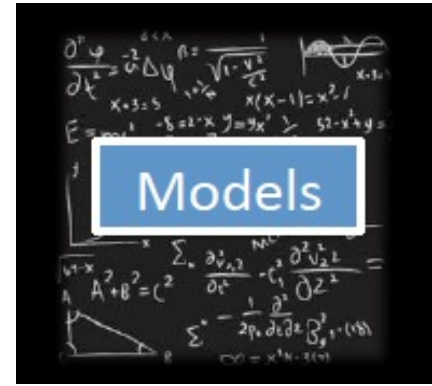
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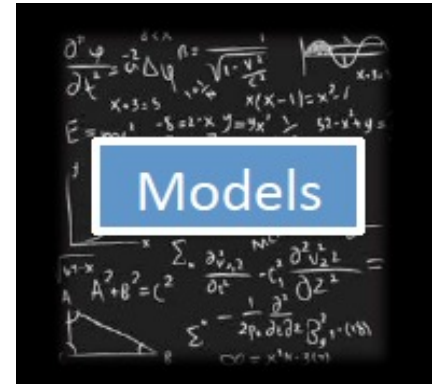
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- There are so many statistical models...



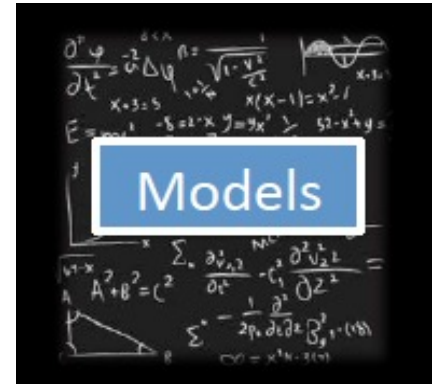
Statistical vs. Mathematical Model

The choice depends on the research question!



Mechanistic Models

- Goal: To **demonstrate the processes** that underlie a relationship between x and y
 - Find a significant relationship using a p-value as a measure of relationship strength
 - **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.

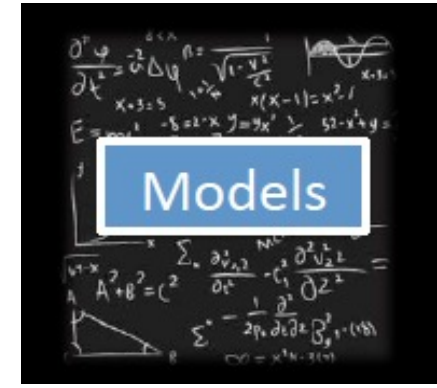
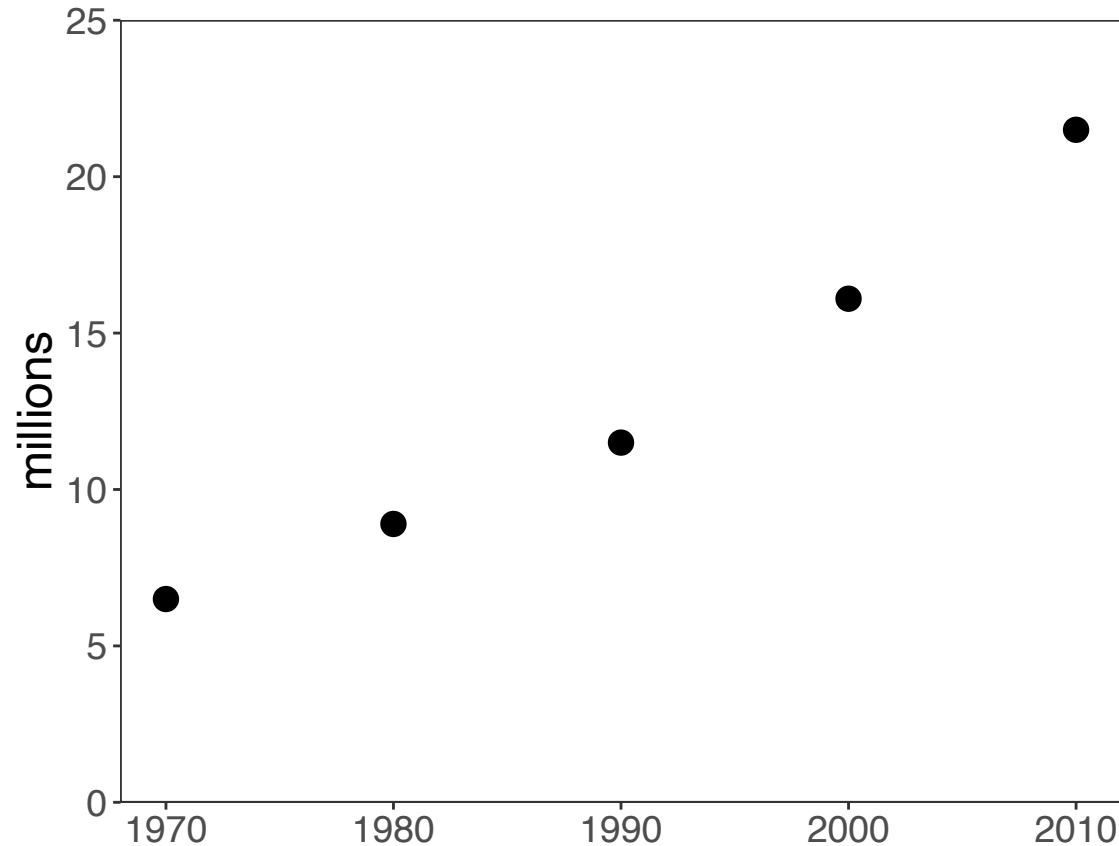


1. Example Question: **How** does Malagasy population size change with time?

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

Can you think of an alternative hypothesis?

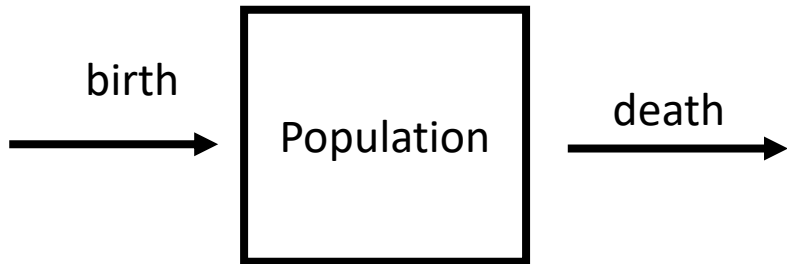
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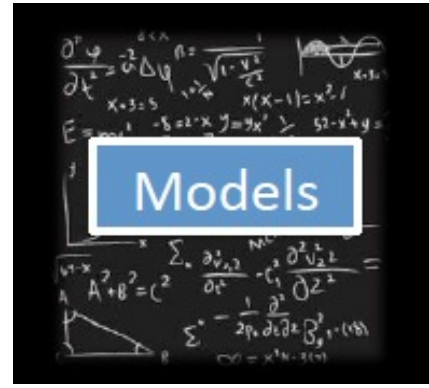
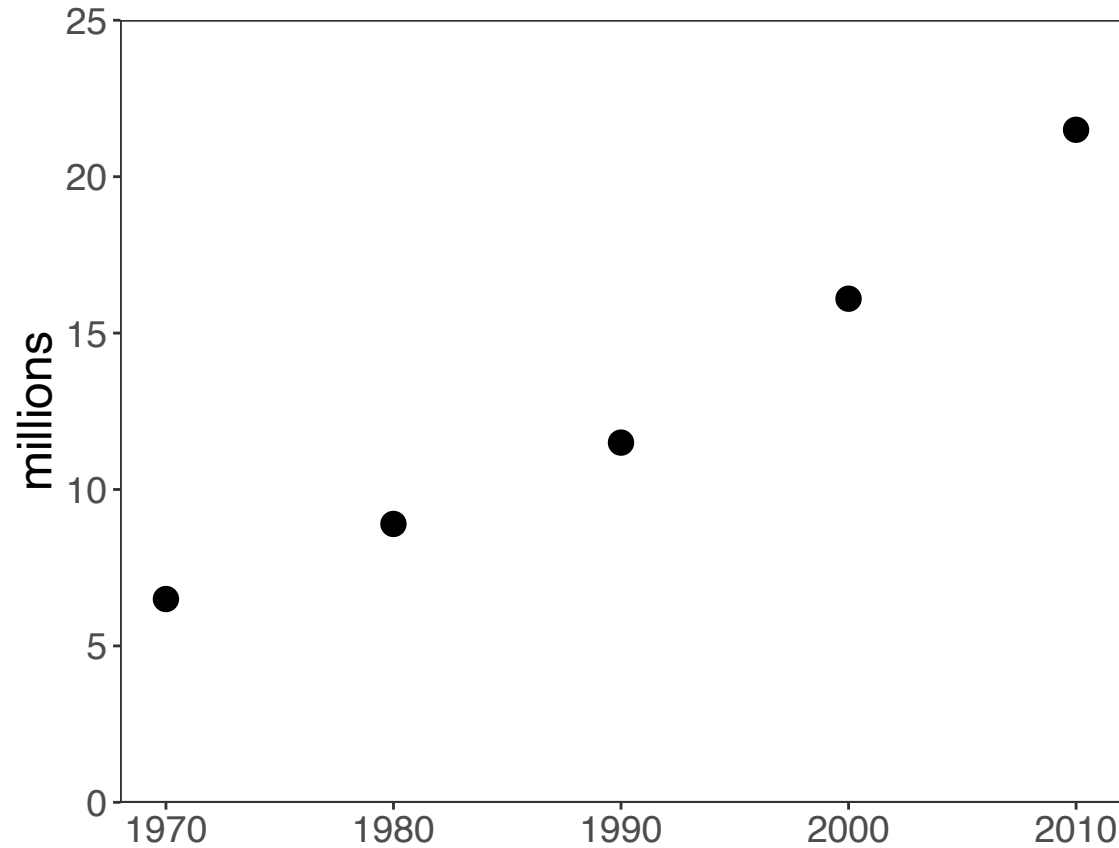
3. Mechanistic Model:



$$P_{t+1} = P_t + b * P_t - d * P_t$$

$$P_{t+1} = P_t + r * P_t$$

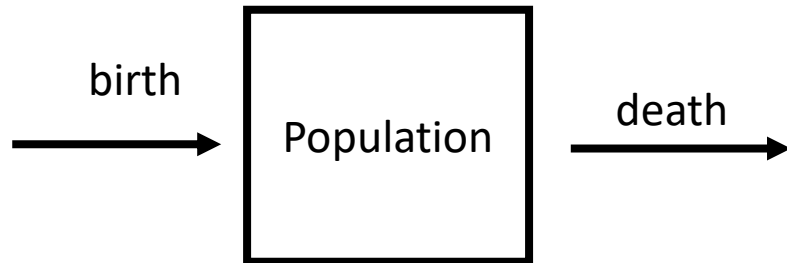
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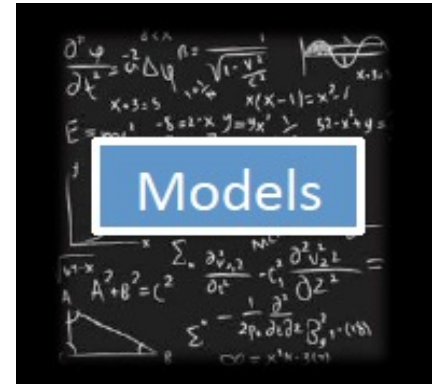
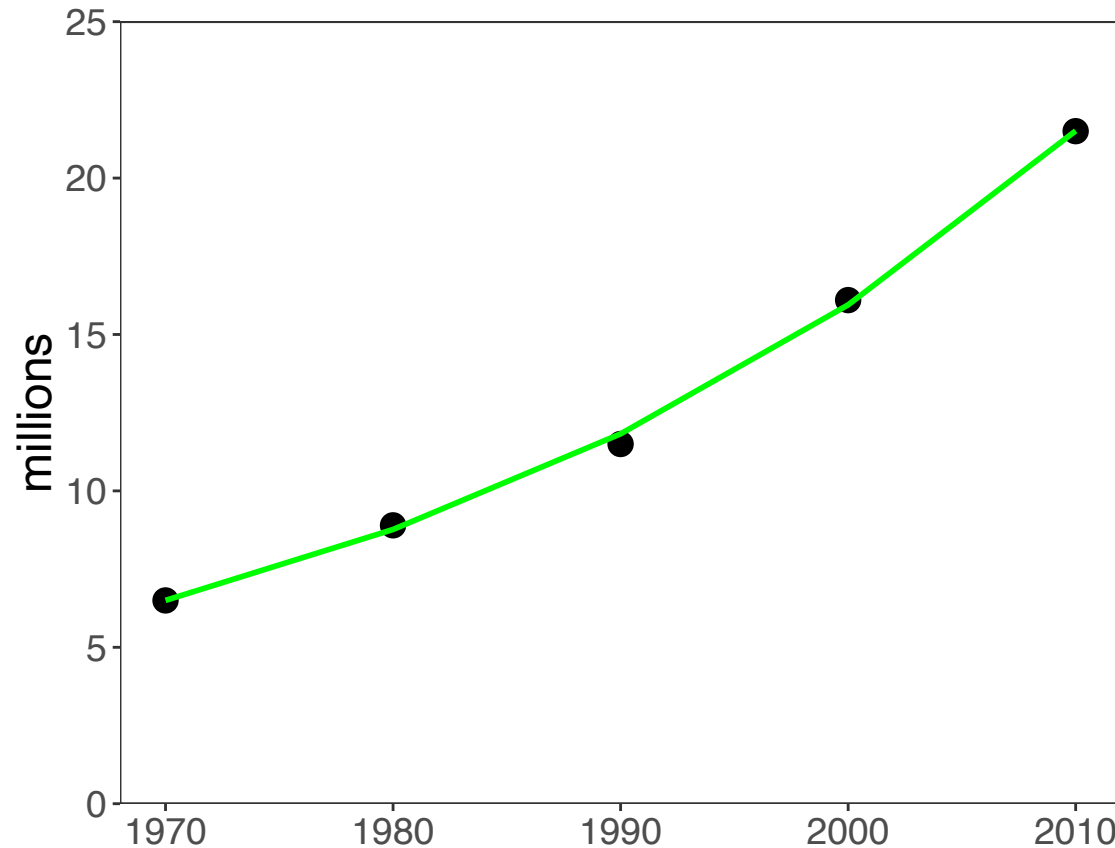
3. Mechanistic Model:



5. Evaluation:

$$r = .349/\text{person}/\text{yr}$$

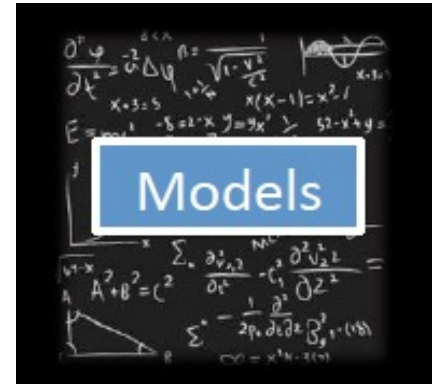
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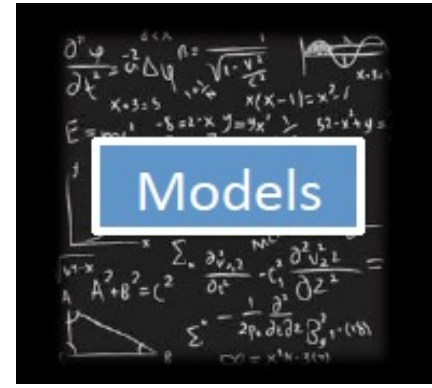
Mechanistic Models: Beware!

- Parameters used in the mechanistic models sometimes are not measurable!



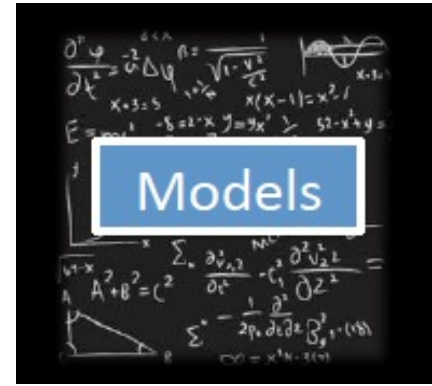
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- Advances in computational power often inspire development of more complex models which are not necessarily better

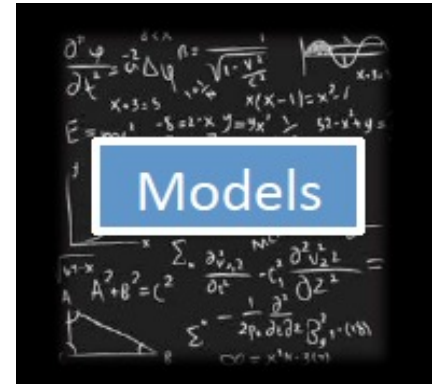


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“All models are wrong but some are useful...”

-George Box



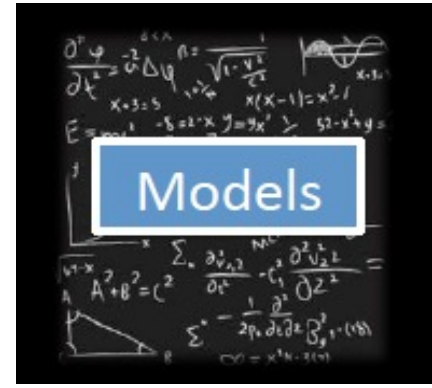
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- Simulations can be computationally intensive
- Advances in computational power often inspire development of more complex models which are not necessarily better

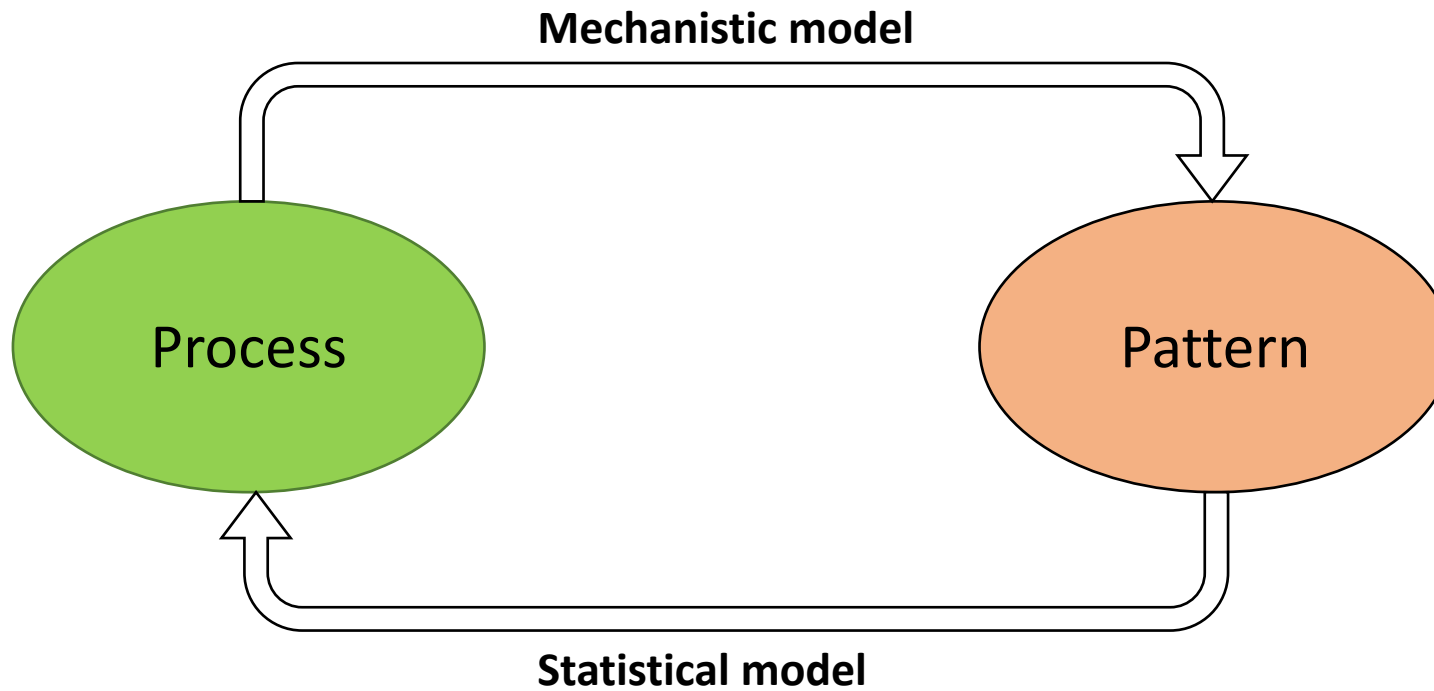
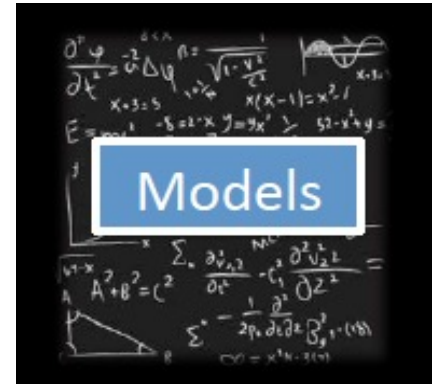
“All models are wrong but some are useful...”

-George Box

We use models to both **predict** and **explain**.

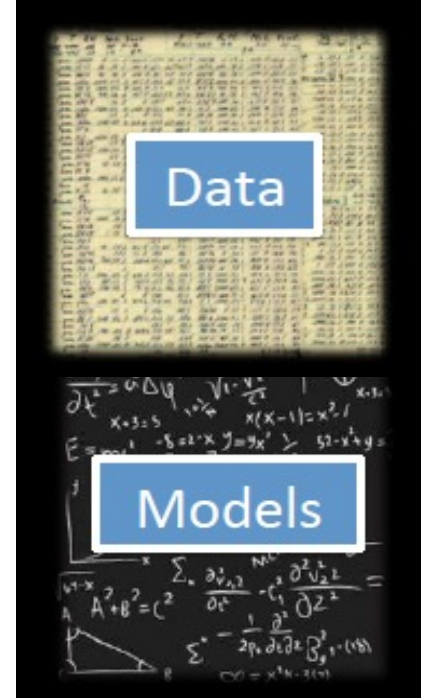


It is ideal when statistical and mechanistic models meet:



A Tool for E²M²

- 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

Goals for this lecture

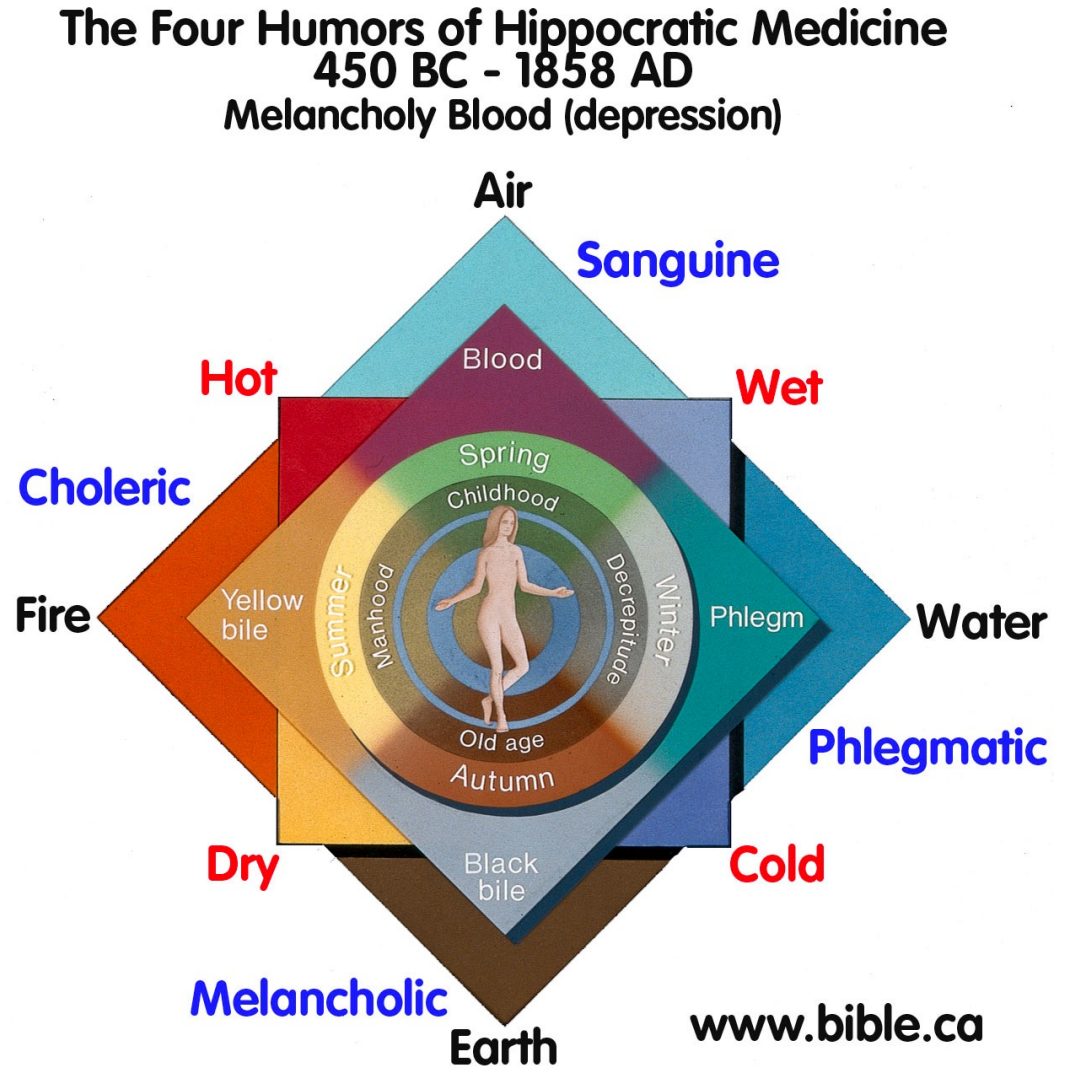
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 - Ecology
 - Epidemiology

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”)

Models in Epidemiology

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



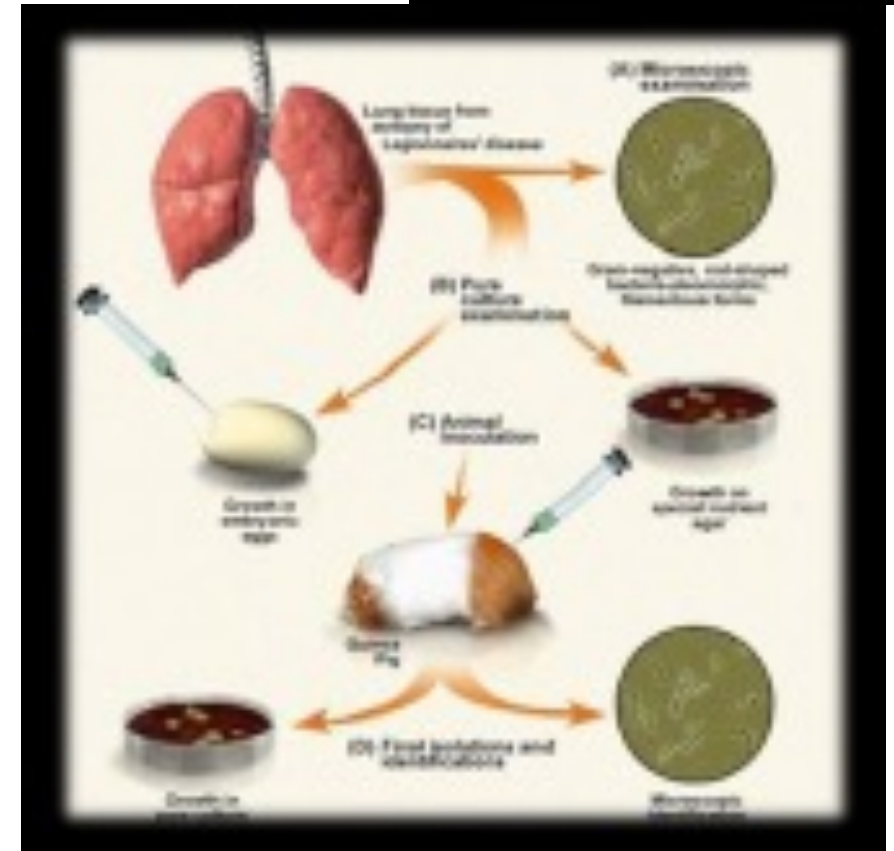
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2. Miasmatic theory of disease (1500s)
 - Sickness results from emanations of 'bad air'



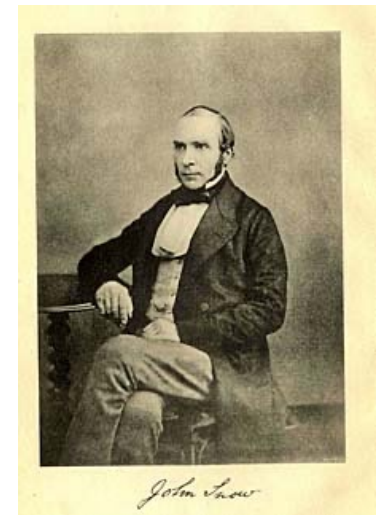
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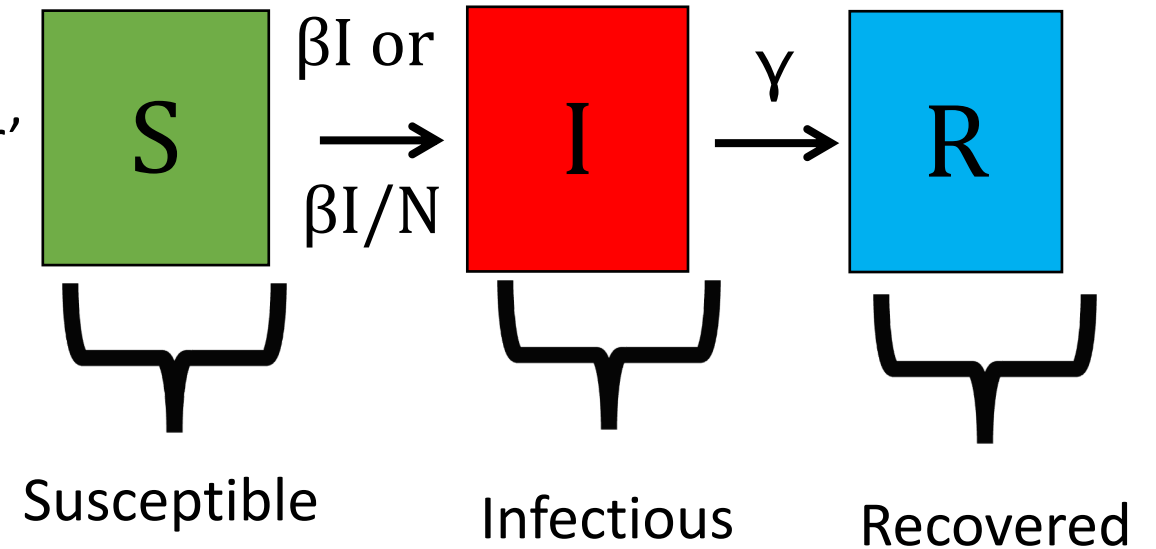
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 - John Snow and London cholera (1854)
5. Mathematical epidemiology
 - Kermack and McKendrik (1927)



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

Models in Ecology

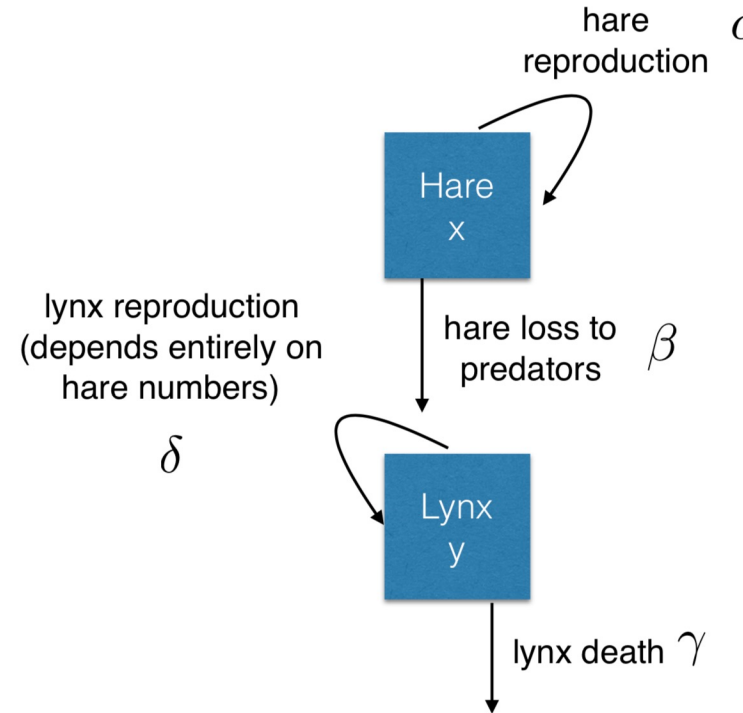
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Models in Ecology

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Models in Ecology

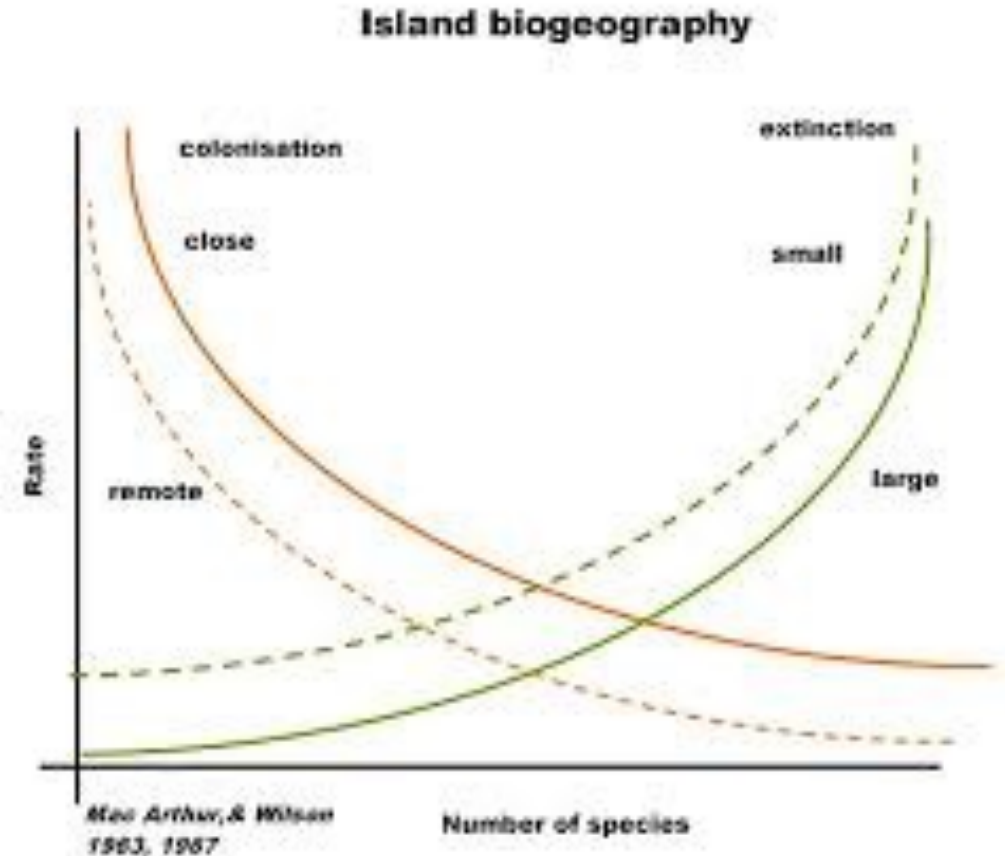
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$$\frac{dx}{dt} = x(\alpha - \beta y)$$
$$\frac{dy}{dt} = -y(\gamma - \delta x)$$

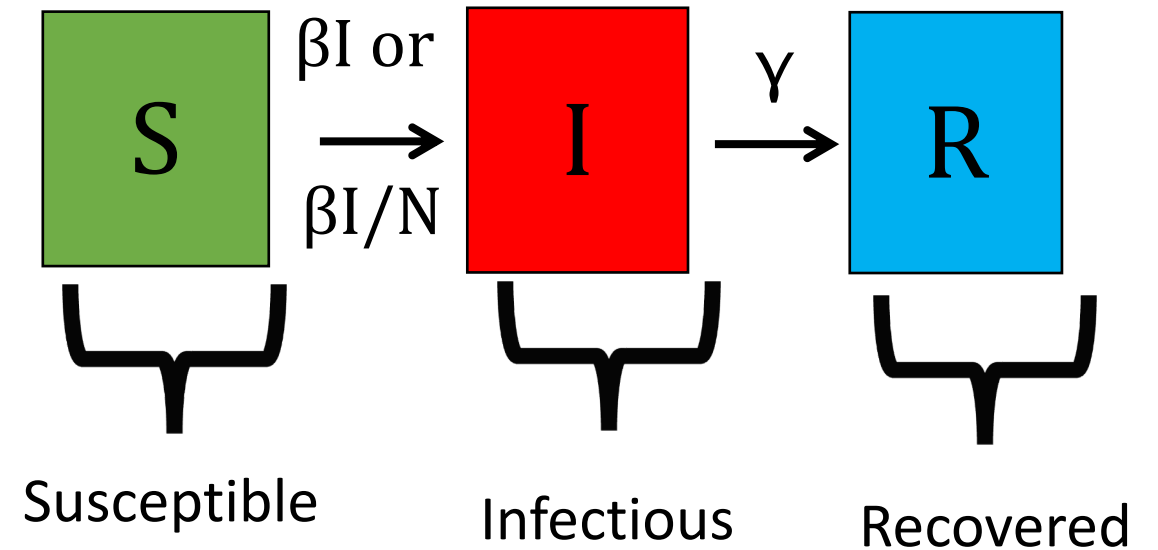
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 - Robert MacArthur (1950s)
 - Island biogeography



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4. Mathematical Ecology
 - Robert MacArthur (1950s)
 - Island biogeography
5. Disease Ecology
 - Anderson and May (1980s)
 - Island biogeography



Friday: R Bootcamp

- Intro to R Studio
- Exploring and Visualizing Data in R
- For-loops, Functions, and If-Else Statements

Thursday: “Bells and Whistle”

- Introduction to Phylogenetic Modeling • Tutorial
- Introduction to Spatial Visualization and Plotting • Tutorial
- Modeling Insights from the Metapopulation Game • Tutorial
- Introduction to Network Modeling • Tutorial
- Research Snapshots

- Programming
- Data
- Models
- Research Development

Saturday: Travel

Sunday: “Getting Started with the Basics”

- Data and Models
- Student introductions & presentations
- Linear regression & simple statistics
- Linear regression tutorial

Monday: “Using Models with Data”

- Model-Guided Study Design
- Study design tutorial
- Intro to Compartmental Models & Differential Equations
- Building mechanistic models in R
- Refining research questions for modeling
- Defining a model world

Wednesday: “Refining Your Work”

- Model Fitting in Practice – the Basic Concept
- Epidemic Cards
- Model Fitting with Epidemic Cards
- Model Selection and Comparison
- Model Selection Tutorial
- Model Telephone

Tuesday: “Applying Simple Models”

- Dynamical Fever
- Introduction to Mixed Modeling
- Reading a Research Paper

Friday: “Putting it All in Perspective”

- The Life Cycle of a Modeling Project
- C4C Student Presentations

Saturday: Travel

January: “Sharing Your Work”

- Final student presentations

Misaotra!



International Clinics on Infectious Disease, Dynamics, & Data

MMED: Clinic on the Meaningful Modeling of Epidemiological Data

May-June 2023, Cape Town, South Africa



DAIDD: Clinic on Dynamical Approaches to Infectious Disease Data

December 2023, Virtual



**South African Center for Epidemiological Modeling and Analysis (SACEMA),
Director**

Dr. Juliet Pulliam
University of Stellenbosch

www.ici3d.org