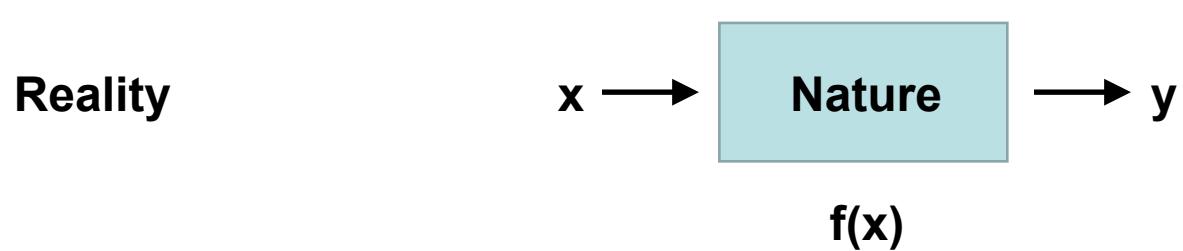


Reminders

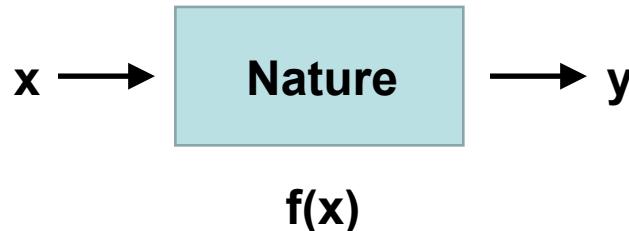
- Take charge of your GitHub repo
- git clone . . .

Trying to learn a function f

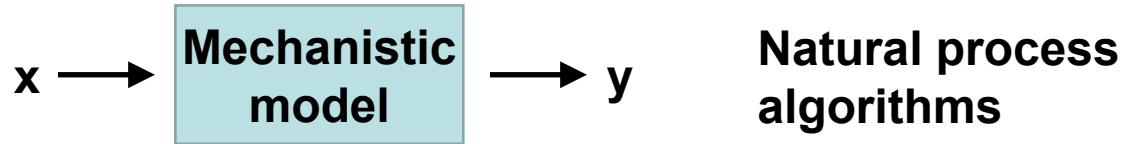


Trying to learn a function f

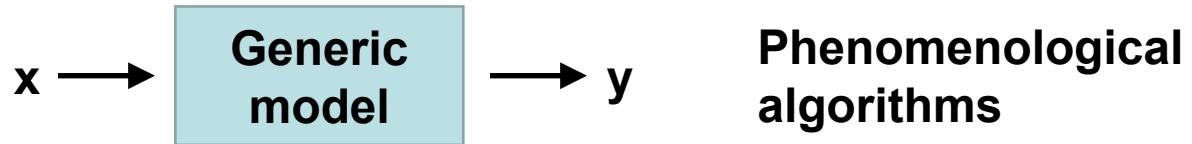
Reality



Natural processes
culture

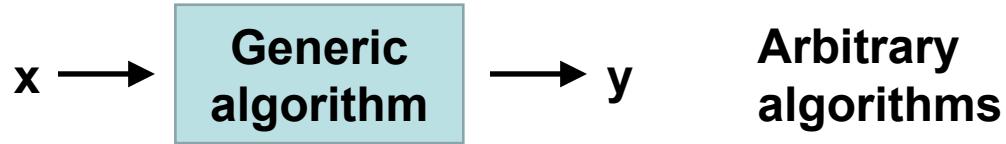


Generative modeling
culture



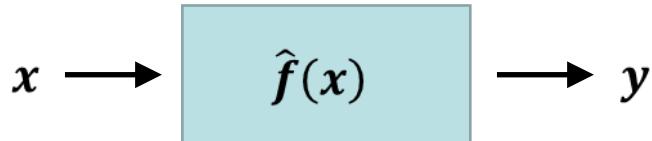
Machine learning

Algorithmic modeling
culture



f can mean different things in different cultures

Prediction is the goal of ML



Goal: find function \hat{f} that has good predictive performance

Accurate on **new observations** of y
(out-of-sample accuracy)

Machine Learning algorithms

- Model algorithm
- Training algorithm
- Inference algorithm

This week

- Full machine learning workflow
- Machine learning with ants data
 - polynomial **model** algorithm (nb pedagogical)
 - least squares **training** algorithm
 - cross validation **inference** algorithm
- R & Python code

Machine learning workflow

Overall algorithm:

1. Create model algorithm(s) for $\hat{f}(x)$
2. Use a training algorithm to find parameter values of $\hat{f}(x)$
3. Use an inference algorithm to measure prediction error and compare predictive skill among models (model families, tuning parameters, x sets, etc).

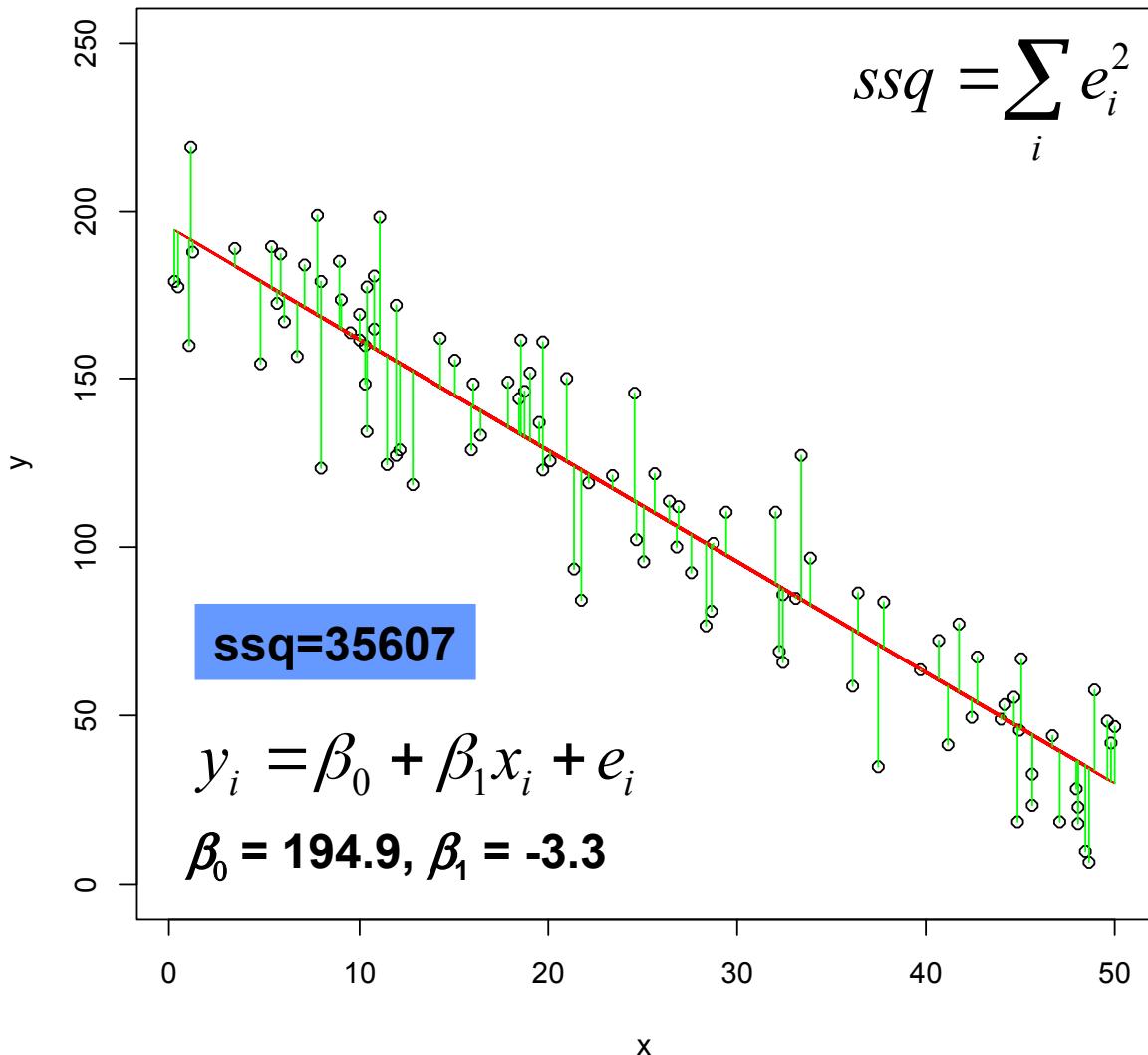
Basic full ML setup

- 3 algorithms:
 - **model:** flexible function $\hat{f}(x)$;
e.g. polynomial linear model
$$y = \beta_0 + \beta_1 x + \beta_2 x^2 + \beta_3 x^3 + \dots + \beta_m x^m \quad m=\text{order}$$
 - **training:** optimize objective function
e.g. least squares
 - minimize $SSQ = \sum_{i=1}^n (y_i - \hat{y}_i)^2$ for training data
 - **inference:** measure error by cross validation;
tuning parameter (order of poly)

Code

- 02_3_ants_cv_polynomial.R
- 02_3_ants_cv_polynomial.py

Least squares optimization



General algorithmic idea:

Vary model parameters until we find the parameter values that minimize the distance of the model from the data

Optimization algorithms

Strategies

1. Systematically try all combinations of parameters - [Grid search algorithms](#)
2. Narrowing in: keep changing parameters in the direction that leads to lower SSQ - [Descent algorithms](#)
3. Try random values for parameter combinations - [Monte Carlo algorithms](#)
4. Solve for parameters using math - [Analytical or numerical algorithms](#)

Linear regression in R uses strategy 4

$\text{lm}(y \sim x)$ solves a system of linear equations using
[linear algebra](#)

Mathematical theory shows what to do ([QR decomposition](#))

Numerical algorithm is needed to do it ([householder algorithm](#))

Fast, [specialized](#), guaranteed to find the minimum SSQ.
Only works for SSQ: **limited to ordinary linear regression.**