

# Introduction to the bartMachine R package

Saint Louis R User Group

John Snyder

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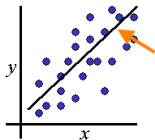
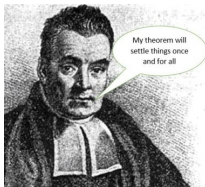
# Outline

1. Brief BART overview
2. Installation and features
3. Demo
4. Further Considerations

# What is BART?

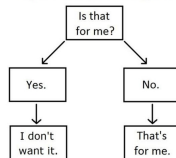


## Bayesian Additive Regression Trees



regression  
line

My Cat's Decision-Making Tree.



# Interpretation

- ▶ Ensemble method combining many shallow trees
- ▶ Bayesian means variation is fully quantified
  - ▶ Yay Statistics

# Powerful Predictive Performance

- ▶ 100 replications of out of sample predictive performance

Function	BART	XGBoost	Random Forest	Linear Reg(lol)
Friedman	1.08	1.21	1.64	2.61
Mirsha's Bird	1.53	2.78	2.90	26.59
Weird Exp	1.04	1.05	1.07	6.08
Linear	1.025	1.032	1.034	1.004

- ▶ bartMachine is relatively unknown
  - ▶ xgboost: ~43k downloads per month
  - ▶ randomForest: ~88k downloads per month
  - ▶ bartMachine: ~2k downloads per month

## Package Features:

- ▶ Functions for Cross Validation
- ▶ Model fitting:
  - ▶ Is done in parallel<sup>1</sup>
  - ▶ Can incorporate missing data
- ▶ Lots of fun statistical things
  - ▶ Credible interval calculation
  - ▶ Diagnostic plots/tests
- ▶ Variable selection
- ▶ Interaction detection
- ▶ Export fit trees

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<sup>1</sup>MCMC

# Installation and loading steps

1. Google “How to install rJava on [your OS]”
2. Do that
3. Run the following

```
install.packages("bartMachine")
```

To load the package with:

- ▶ 10GB of memory
- ▶ All but one core available for compute

```
options(java.parameters = "-Xmx10g")  
library(bartMachine)  
numcores <- parallel::detectCores()  
set_bart_machine_num_cores(numcores - 1)
```

# Code Time

Coding demo



# Computational Considerations

- ▶ Table with memory/time

## John's Final Thought

- ▶ BART is a powerful technique which brings many advantages
  - ▶ At the expense of computational efficiency.
- ▶ Good results with removing expected variation and feeding residuals into BART.