# Practical Machine Learning in R

Regression

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<sup>&</sup>lt;sup>2</sup>slides available at http://www.cs.uwyo.edu/~larsko/ml-fac

#### Regression



Goal: Predict a continuous quantity

#### Linear Models

- assumes linear relationship between features and variable of interest
- prediction is linear combination of feature values with coefficients (similar to logistic regression)
- determine coefficients by minimizing loss (e.g. sum of squared error) wrt training data

#### Linear Models



#### Linear Models



# **Regression Splines**

- Multivariate Adaptive Regression Splines (MARS)
- non-parametric technique no assumptions about the underlying relationship
- ▷ prediction is weighted sum of basis functions
- construction similar to trees repeatedly add basis functions to improve performance (recursive partitioning), then remove some to improve generality (pruning)

# **Regression Splines**



- "boost" a weak learner by training set of models that fix each other's errors (ensemble)
- after adding each model, reweigh training data such that examples with higher error get more weight
- $\,\triangleright\,$  aggregate by combining weighted predictions from each model
- ▷ idea similar to random forests
- ▷ technique not specific to regression















# Support Vector Machines and Random Forests

- ▷ regression versions exist
- ▷ SVMs: minimize error of support vectors
- Random Forests: predict constant quantity (or simple linear model) at leaves



#### http://www.cs.uwyo.edu/~larsko/ml-fac/ 02-regression-exercises.Rmd