

Practical Machine Learning in R

Introduction

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²slides available at <http://www.cs.uwo.edu/~larsko/ml-fac>

What is Machine Learning?

- ▷ “gives computers the ability to learn without being explicitly programmed” (Wikipedia)

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- ▷ “A computer program is said to learn from experience E with respect to some class of tasks T and performance measure P if its performance at tasks in T , as measured by P , improves with experience E .” (Tom Mitchell)

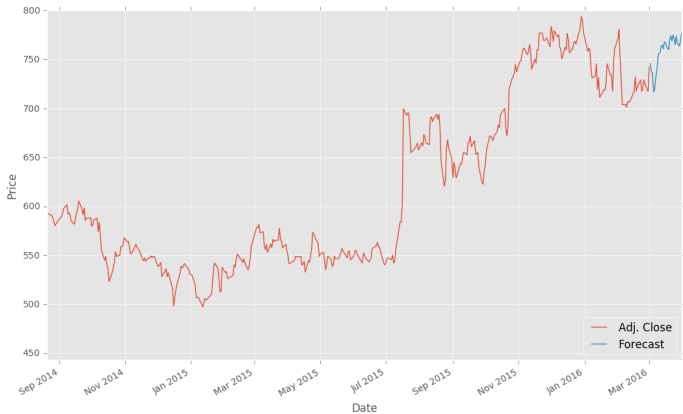
Examples



Examples

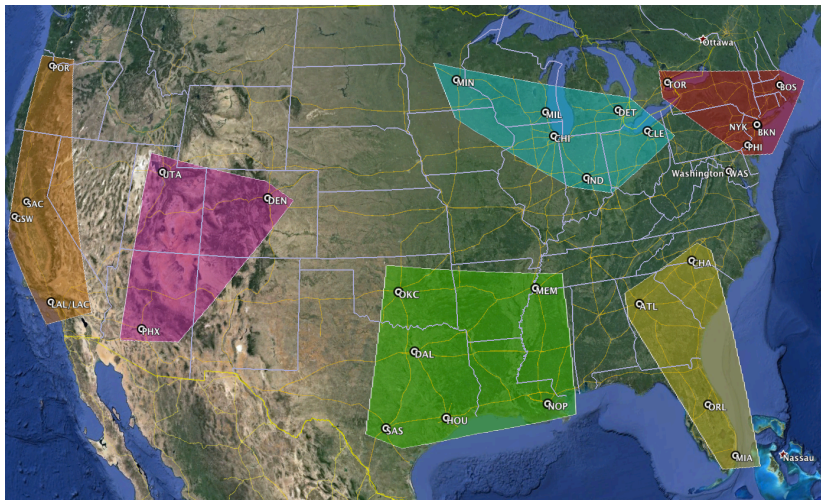
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Examples



<https://pythonprogramming.net/forecasting-predicting-machine-learning-tutorial/>

Examples



<https://squared2020.com/2015/09/09/redefining-nba-divisions-by-clustering/>

Supervised Learning

- ▷ learn the relationship between input x and output y
- ▷ training data with labels available – y known for given x
- ▷ can see this as function approximation – find an f such that

$$y \approx f(x)$$

Supervised Learning

- ▷ x are features or attributes
- ▷ y is the ground truth
- ▷ denote predictions $f(x) = \hat{y}$
- ▷ loss function $L(y, \hat{y})$ measures how good predictions are, e.g.

$$L(y, \hat{y}) = (y - \hat{y})^2$$

- ▷ want to minimize loss given training data $X_{\text{train}} = \{(x_i, y_i)\}^n$:

$$\arg \min \sum_{i=1}^n L(y_i, \hat{y}_i)$$

Supervised Learning

- ▷ want to learn a general function that is predictive on new data
- ▷ second set X_{test} that is **not** used in training to test generalization performance:

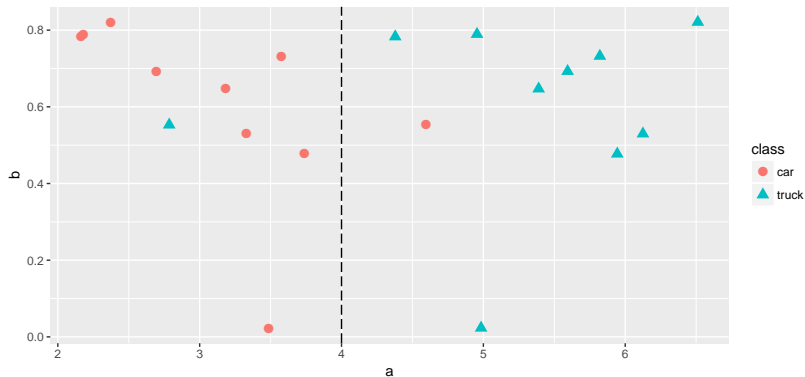
$$\sum_{i=1}^n L(y_i, \hat{y}_i)$$

- ▷ usually full data set X is split into non-overlapping train and test sets:

$$X_{train} \cup X_{test} = X$$

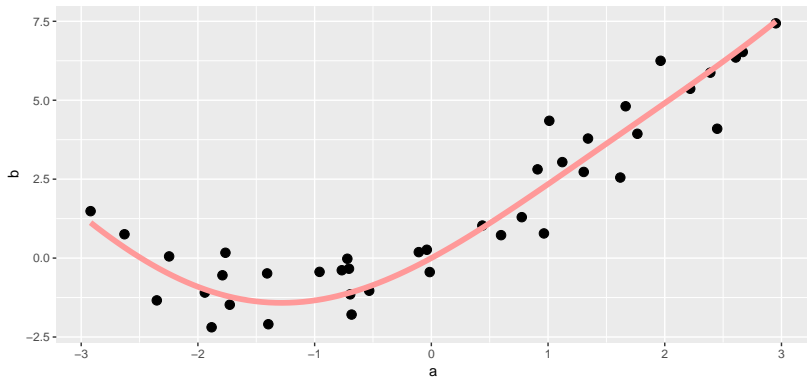
$$X_{train} \cap X_{test} = \emptyset$$

Supervised Classification



Goal: Predict a class (discrete quantity), or membership probabilities

Supervised Regression

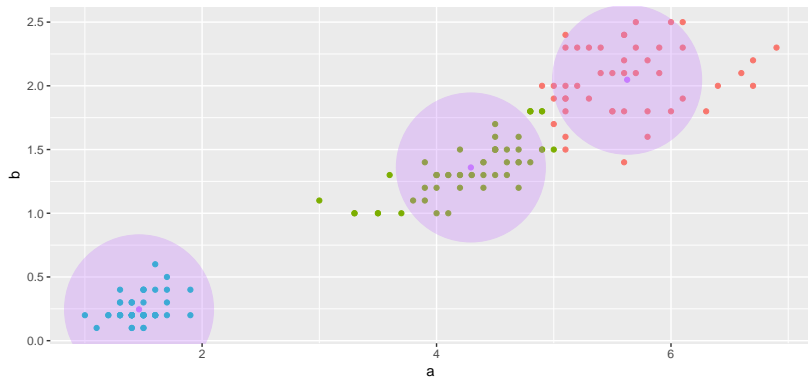


Goal: Predict a continuous quantity

Unsupervised Learning

- ▷ no ground truth y available
- ▷ determine group membership or assign labels
- ▷ loss function measures properties of groups, e.g. homogeneity wrt. features
- ▷ still want to minimize loss given training data and generalize

Unsupervised Clustering



Goal: Group data by similarity, or estimate membership probabilities

In this Course

- ▷ classification
- ▷ regression
- ▷ clustering
- ▷ data preprocessing (missing values, dimensionality reduction)
- ▷ performance evaluation
- ▷ parameter tuning

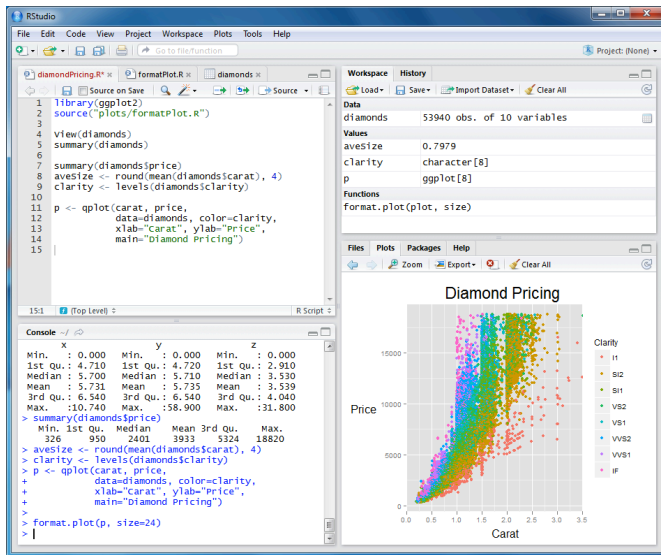
Not in this Course

- ▷ R tutorial
- ▷ details on particular methods
- ▷ deep learning
- ▷ time series
- ▷ Big Data

What you'll need



Install RStudio



<https://www.rstudio.com/products/rstudio/download/>

Install mlr

- ▷ on the R console:

```
install.packages("mlr")
```

- ▷ or see <http://derekogle.com/IFAR/supplements/installations/InstallPackagesRStudio.html>
- ▷ extensive tutorial available: <https://mlr-org.github.io/mlr-tutorial/devel/html/>

Format

- ▷ meetings roughly every week
- ▷ half lecture, half practical exercises
- ▷ happy to discuss specific problems