Ensemble Methods

Overview

Merge predictions

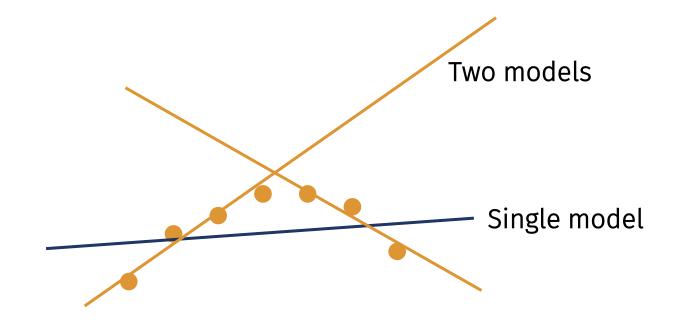
- Individual predictions might be off
- Perhaps bias too high

Core idea

- Build (at least 2, typically more) models
- Models may be unstable
- Either
 - Learn when to trust which one
 - Reduce variance by aggregation

Meta Algorithm

- Concrete base model is arbitrary
- Model mixing possible



Bootstrap Aggregating ("Bagging")

Merge predictions

- Take random subset of training data with replacement
- Random subset of features
- Fit decision trees

Classification

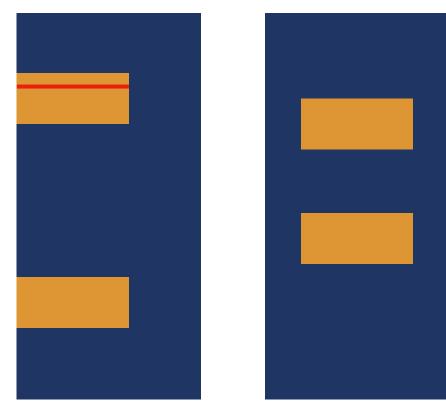
Majority vote

Regression

Average prediction

Random forests

Base learner is decision tree.



Boosting

Refine predictions

- Build base model
- Focus on mispredictions, build new or stronger weighted model for those
- Many flavours

Example: AdaBoost

Build a new weak learner, assign residual errors Until end:

Train new weak learner f with training instances weighted by current error Find global weight w for prediction s.t. total error is minimized: $P_{+} = P + wf$

Stacking

Many learners

- Build diverse base models m_i
 - Could even be different methods

Combiner model

- Take all predictions mi(x) as input
 - Optionally augmented by model features
- Combine prediction
 - Either learn reliability depending on features
 - Weights depending on regions

