



Bagging



Chenyue Cai, Oscar Xu, Yintang Yang



Ensemble methods:

- If you searched on google, the **definition of ensemble** is as following:
 - A group of musicians, actors, or dancers who perform together.
 - A group of items viewed as a whole rather than individuality.
- If you searched **bagging** on google scholar, the paper you find will say
 - Aggregated prediction based on multiple learners

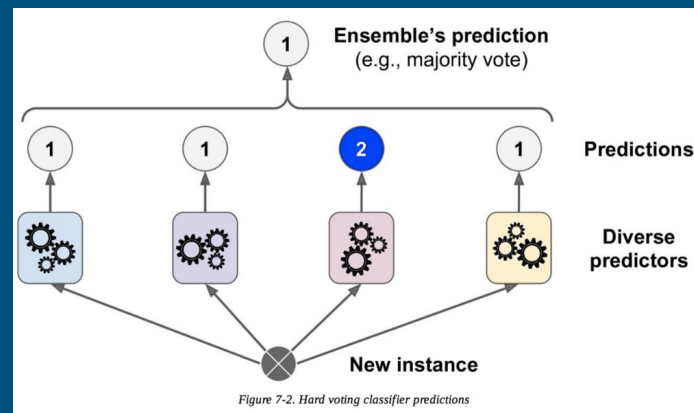
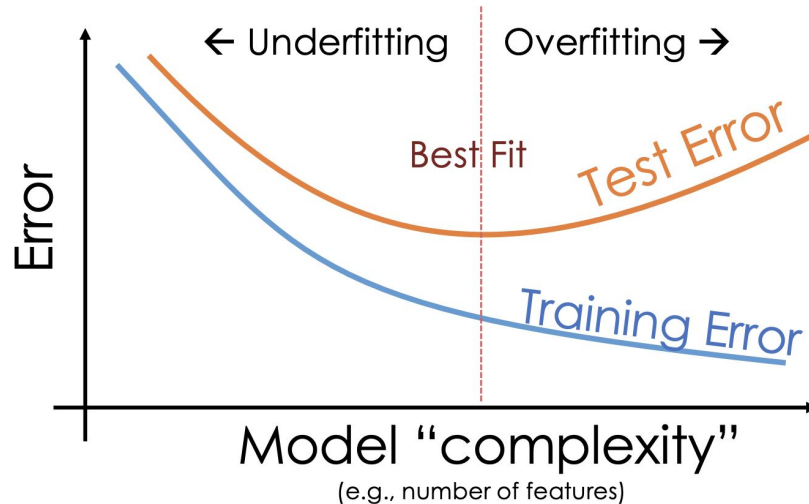


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Before all the math, and
new concept, Let's
identify something
you know !

One issue that ML ppl always encounter :(

Training vs Test Error



Overfitting: A quick recap

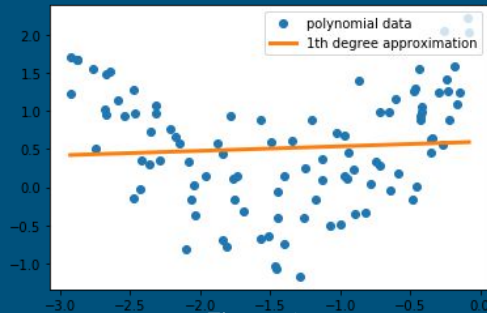


Figure 1

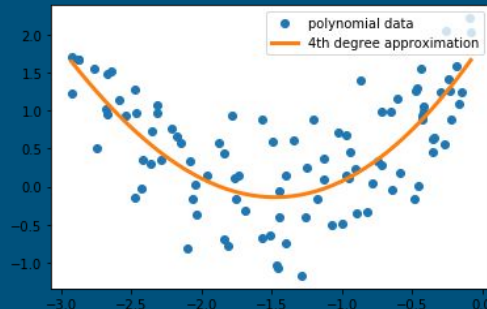


Figure 2

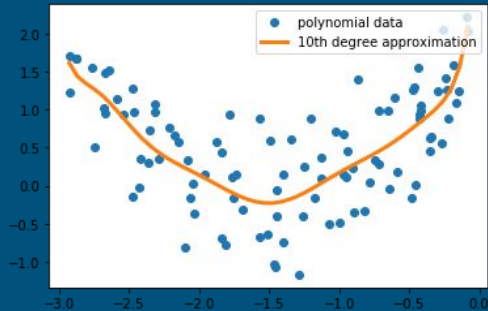


Figure 3

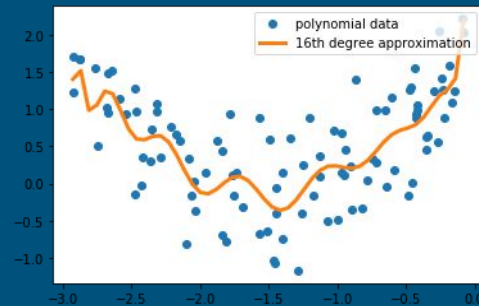
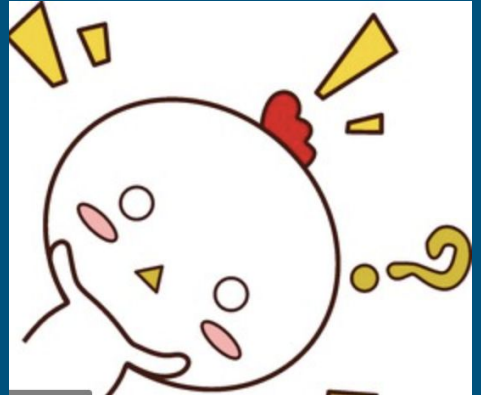


Figure 4

- Questions: given these line of best fit for different degree of polynomials identify the following:
 - Which one has the largest degree?
 - Which one has the smallest degree?
 - Which one is overfitting or underfitting?

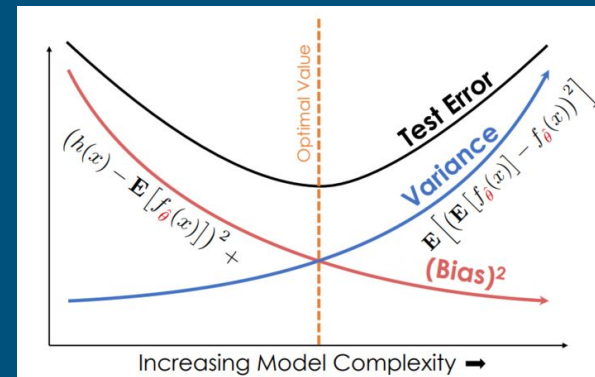
How to solve Overfitting?



Recap: Interpretable metric of error

Bias and Variance

- Bias Variance Definition
 - Bias measures how well the model approximates underlying true function
 - Variance measures how robust the model is towards perturbation
- large bias/small variance means few features, highly regularized, such as highly pruned decision trees, large-k kNN etc;
- small bias/high variance means many features, less regularization, small-k k-NN etc.



It seems that solving
overfitting will always
incur a bias variance
trade off :(

Bagging



“Our secret weapon against the inevitable bias variance trade-off”

- Single model: we are stuck with bias variance trade-off, reducing variance WILL HURT our model accuracy.
- Ensemble model: we can utilize the nature of combining multiple models to achieve our goal: keeping bias low while improving variance.
- In the following slides, we will answer the following questions:
 - How do we implement bagging?
 - Why does bagging work?
 - What does bagging improve?

Bagging: Implementation

- Comparing Bagging with Democracy
 - Dictatorship: Single Model Prediction
 - Democracy: Aggregate multiple learners result by a "Majority vote"
- "Democracy" Elements
 - "Voters": Different models that learns from a subsample of the whole training data
 - "Vote": Simply averages the prediction (in regression case), decide the results based on the most popular vote (in classification case)

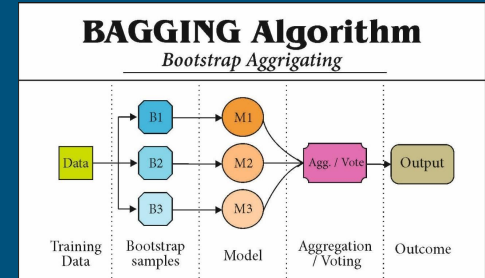


Image Source:

https://analyticsindiamag.com/guide-to-ensemble-methods-bagging-vs-boosting/?utm_source=rss&utm_medium=rss&utm_campaign=guide-to-ensemble-methods-bagging-vs-boosting

Bagging Implementation

- Training:
 - Select a base learner model: M
 - Bootstrap from training data N samples
 - for i in range (N):
 - Train with i th sample and get a trained model M_i
- Predicting:
 - for i in range(N):
 - Get predicted result y_{pred_i}
 - Average out y_preds and attain the final result

Bagging Theory:

- Consider each model as a random variable Y , then we have Y_1, Y_2, \dots, Y_n . Each Y is independent of each other and have the same mean μ and variance σ^2 . We will explore the result of "majority vote".
 - Problem set up: iid. $Y_i \sim (\mu, \sigma^2)$
 - Evaluate: the mean and variance of $1/n * \sum Y_i$
- Since it's iid random variables of identical mean and variance, the mean is just μ , while the variance by linearity of variance is $1/n * \sigma^2$
- Thus, as n increases, the variance drops linearly while mean stays the same.

Bagging improvement:

- Maintain bias, Reduce variance
 - Thus suitable for low bias high variance learners
 - Can you name some of these learners on your own? (High degree polynomial)
- Smoothen out the decision boundary
 - We also see smoothen decision boundary when we do other stuff (e.g. ridge regression, Kernel RBF) It all seems to tackle overfitting!
- Increase test time accuracy

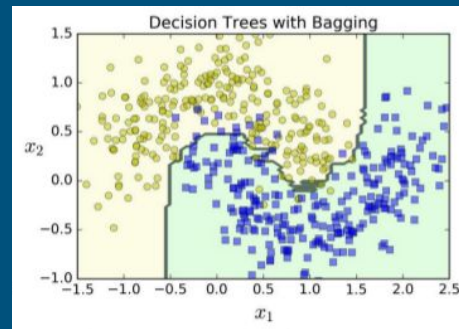
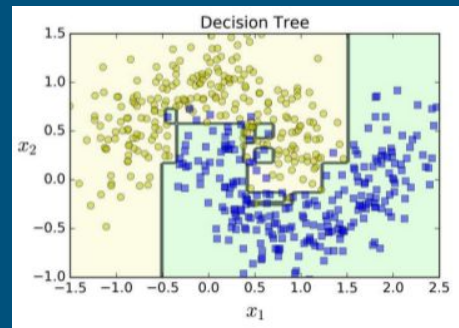


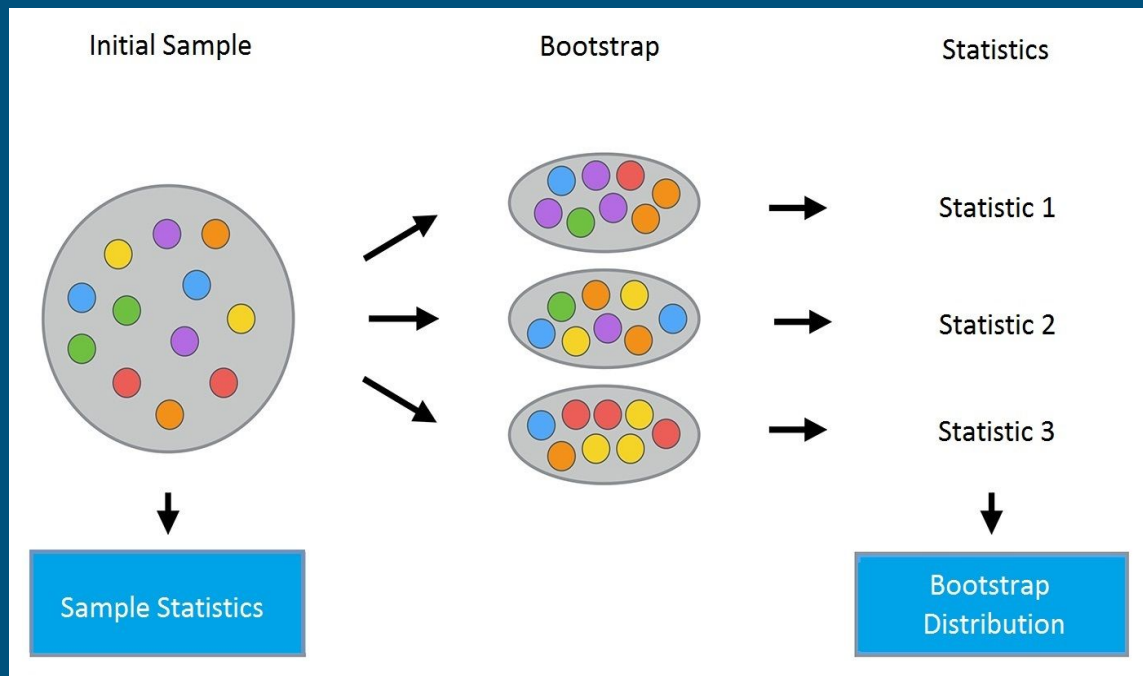
Image Source:

https://www.google.com/search?q=decision+boundary+bagging&tbm=isch&ved=2ahUKEwiZ4e6i2sTtAhWQhZ4KHbolCAYQ2-cCegQIA BAA&oq=decision+boundary+bagging&gs_lcp=CgNpbWcQAzoECAAAQEzoGCAAQCBAAeOgQIABAYUPt9WKqFAWDChgFoAHAAeACAAY2l AZUFkgEBOJgBAKABAaoBC2d3cy13aXotaW1nwAEB&sclient=img&ei=t8HSX9mpBJCL-gS6y6Aw&hl=zh-CN#imgsrc=zU22DM-AQvX3NM

Train - Test Split

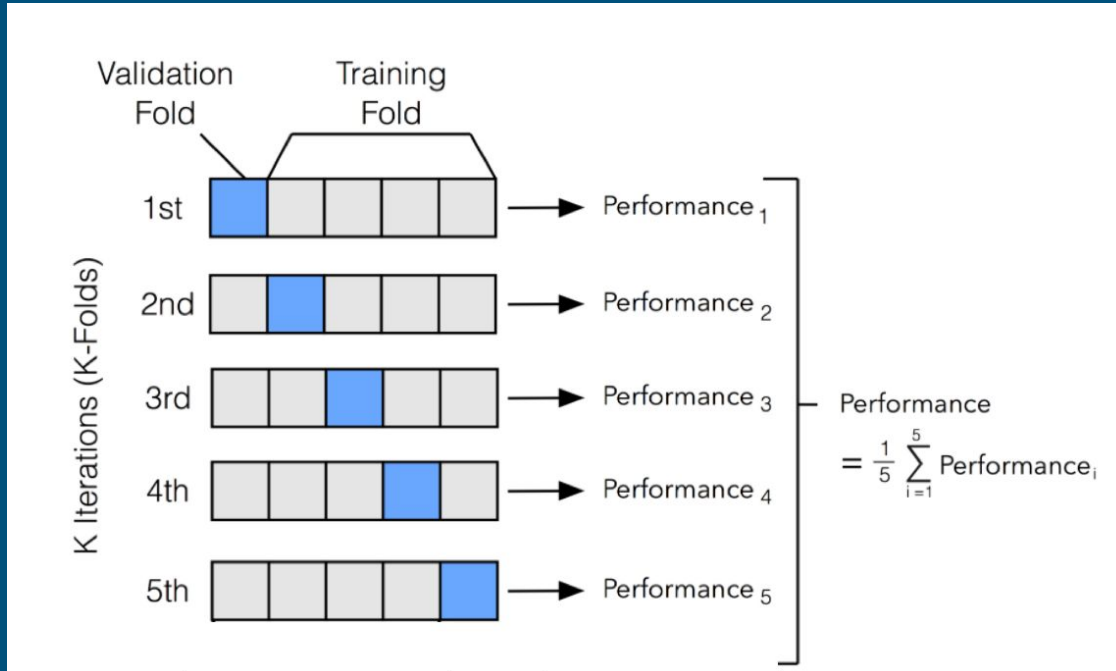
- Training set : used to fit model
- Test set: used to check generalization ability
- Validation set: used to evaluate the model we trained on the training set
- Common ratio: 70% train, 15% val, 15% test, or 80% train, 10% val, 10% test.
- Check `sklearn.model_selection.train_test_split()` function for coding details

Bootstrap



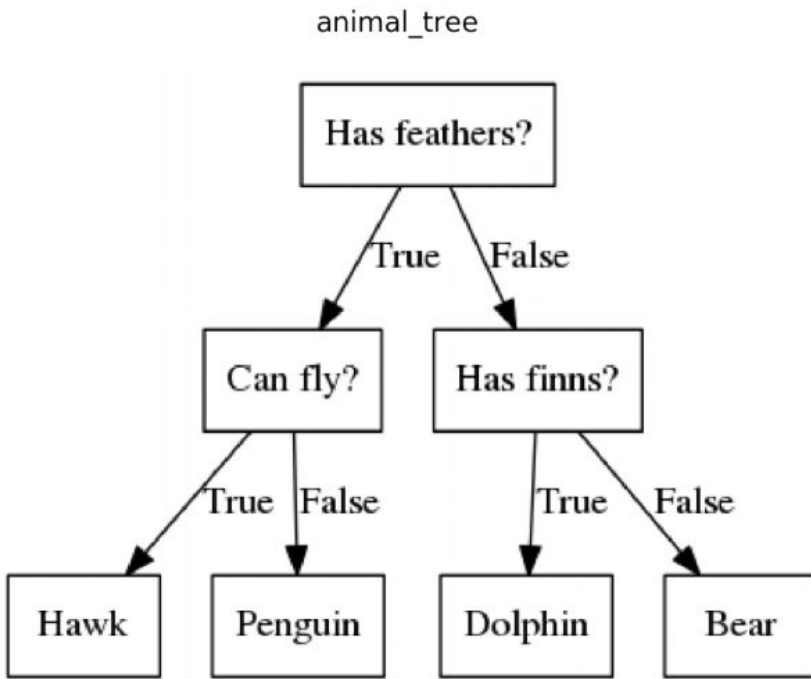
- Treat initial sample as population
- Randomly select
- With replacement
- Average statistics to get bootstrap distribution
- Avoid sample reduction
- Reduce variance
- Avoid overfitting

K-Fold Cross-Validation



- Train the model for Training Fold₁
- Use Validation Fold₁ to find Performance₁
- Repeat for 2...K folds
- Overall Performance is the average of each Performance_i

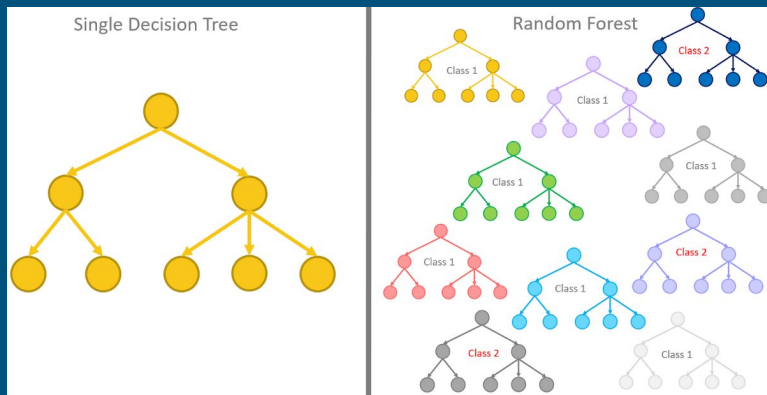
Decision Tree



- Used for classification and regression problems
- Answer sequential questions
- “If A, then B”

Random Forest

- An ensemble of many decision trees
- Each decision tree is used as parallel estimators
- It takes the mean value of the results from decision trees



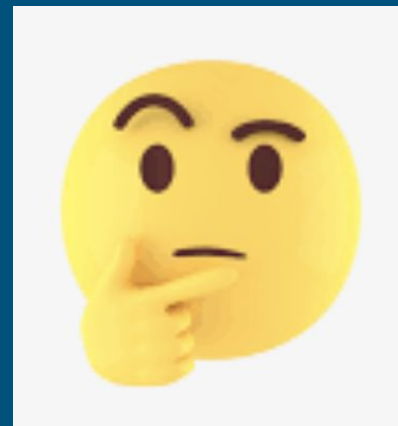
Exercise

True or False: One example of bagging is random forest, it is used for classification. Thus bagging is a method specifically for classification problem but not regression problem.

Solution

False.

Bagging can be used for both classification and regression.



The End.

Thank you!