

Intro to Ensemble Methods & Bagging

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Intro to Ensemble Methods & Bagging

LEARNING OBJECTIVES

- Explain the power of using ensemble classifiers
- Know the difference between a base classifier and an ensemble classifier
- Describe how bagging works
- Use the bagging classifier in scikit-learn

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PRE-WORK

- Perform a classification
- Use label encoder
- Use cross validation to evaluate model performance

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OPENING

What is an Ensemble Method?

What is an Ensemble Method?

What is Bagging?





What classifiers have we learned about thusfar? Which is your favorite?

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How do we assess the goodness of a model?

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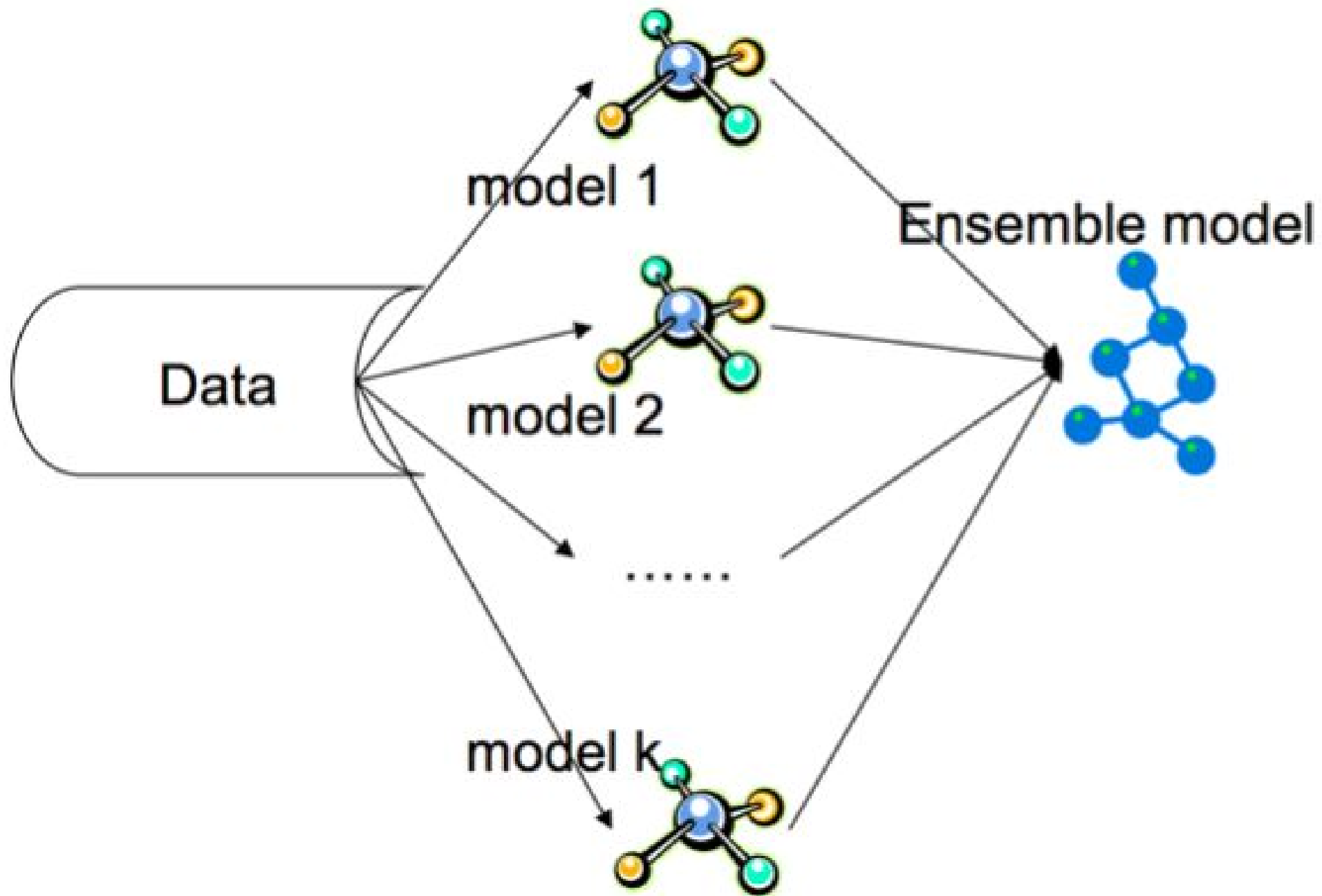
How could we improve the performance of a model?

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Introduction: Ensemble Techniques

What is an Ensemble Method?

Ensemble methods are learning algorithms that construct a set of classifiers and then classify new data points by taking a weighted vote of their predictions.



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When might this be useful?









There are two families of Ensemble Methods

1. Averaging Methods

- Driving principle is to build several estimators independently and then to average their predictions.
- On average, the combined estimator is usually better than any of the single base estimator because its variance is reduced.

EX: Random Forest and Bagging

There are two families of Ensemble Methods

2. Boosting Methods

- Base estimators are built sequentially and one tries to reduce the bias of the combined estimator.
- The motivation is to combine several weak models to produce a powerful ensemble. We will discuss these in a future lecture.

EX: AdaBoost and Gradient Tree Boosting

The Hypothesis Space

In any supervised learning task, our goal is to make predictions of the true classification function f by learning the classifier h . In other words we are searching in a certain hypothesis space for the most appropriate function to describe the relationship between our features and the target.

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Can you give an example of how this search is performed using one of the classifiers you know?

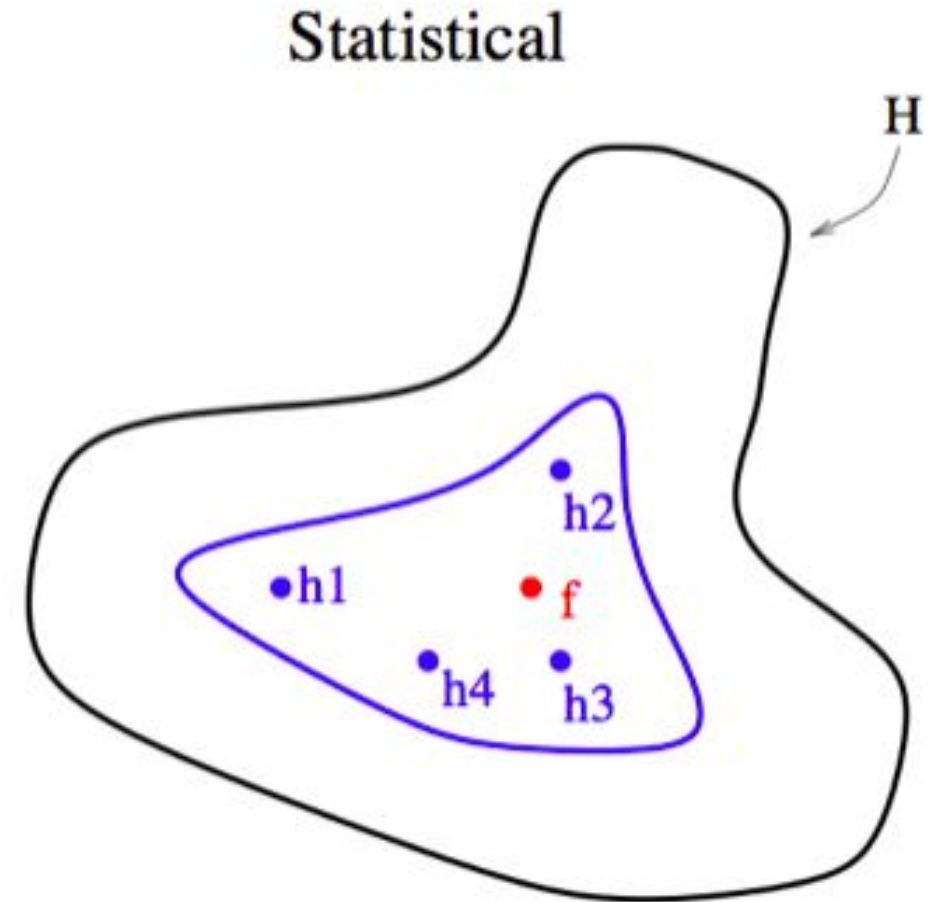
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**What reasons could be preventing our hypothesis
to reach perfect score?**

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The Statistical Problem:

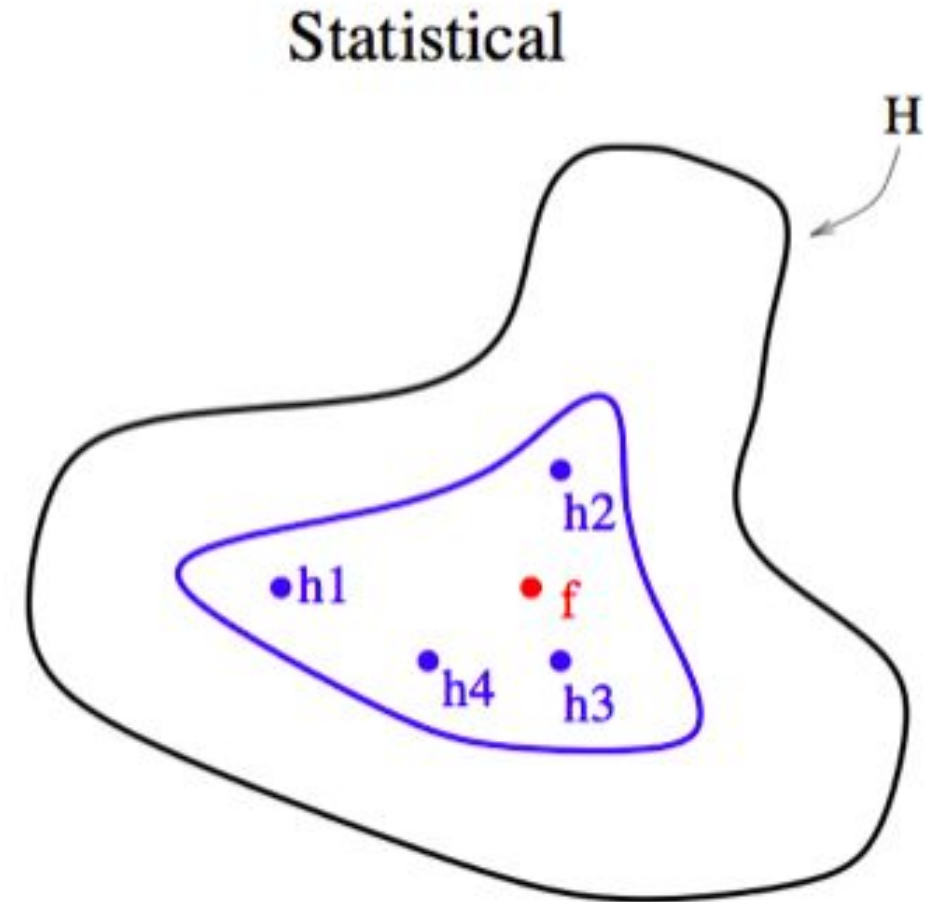
If the amount of training data available is small, the base classifier will have difficulty converging to \mathbf{h} .



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The Statistical Problem:

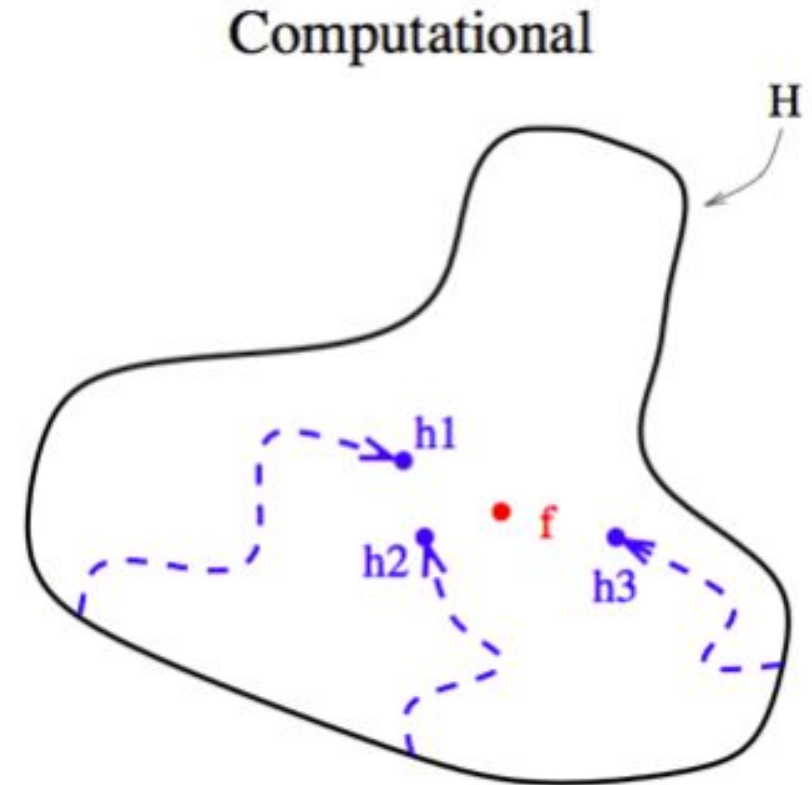
An ensemble classifier can mitigate this problem by "averaging out" base classifier predictions to improve convergence.



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The Computational Problem

Even with sufficient training data, it may still be computationally difficult to find the best classifier h .

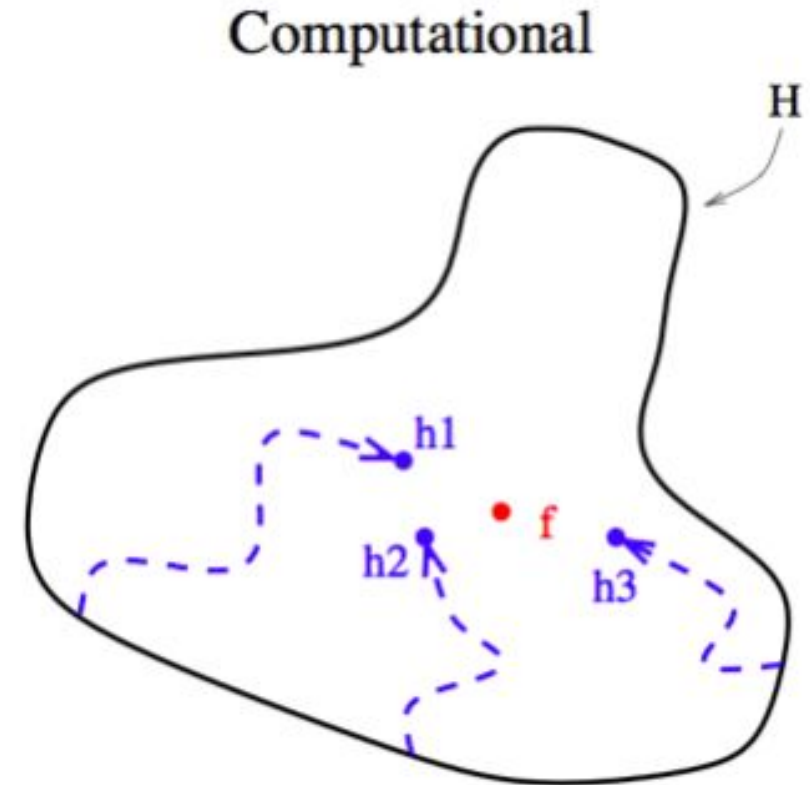


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The Computational Problem

Even with sufficient training data, it may still be computationally difficult to find the best classifier h .

The true function f is often best approximated by using several starting points to explore the hypothesis space



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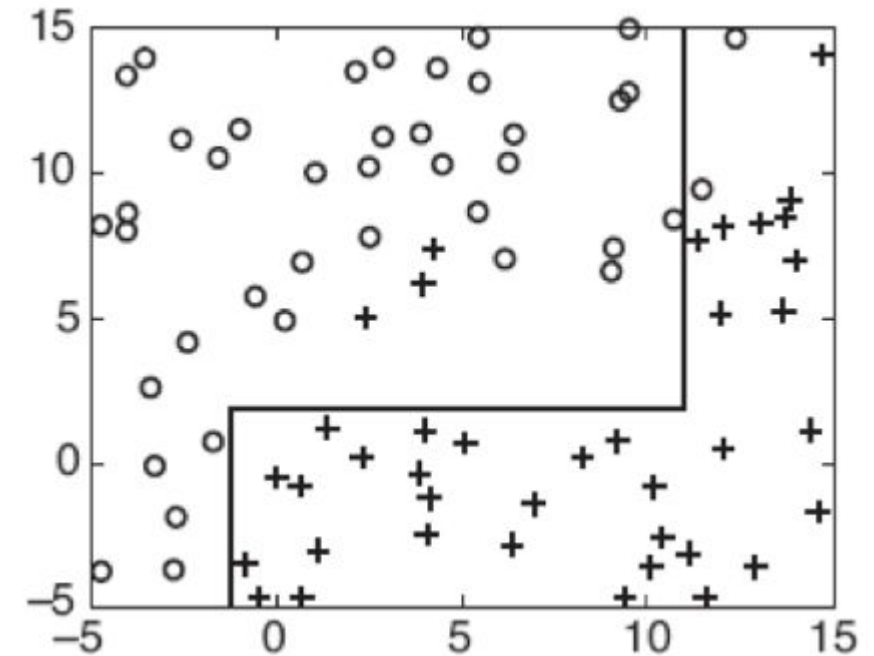
The Representational Problem

Sometimes f cannot be expressed in terms of our hypothesis at all. To illustrate this, suppose we use a decision tree as our base classifier. A decision tree works by forming a rectilinear partition of the feature space, i.e it always cuts at a fixed value along a feature.

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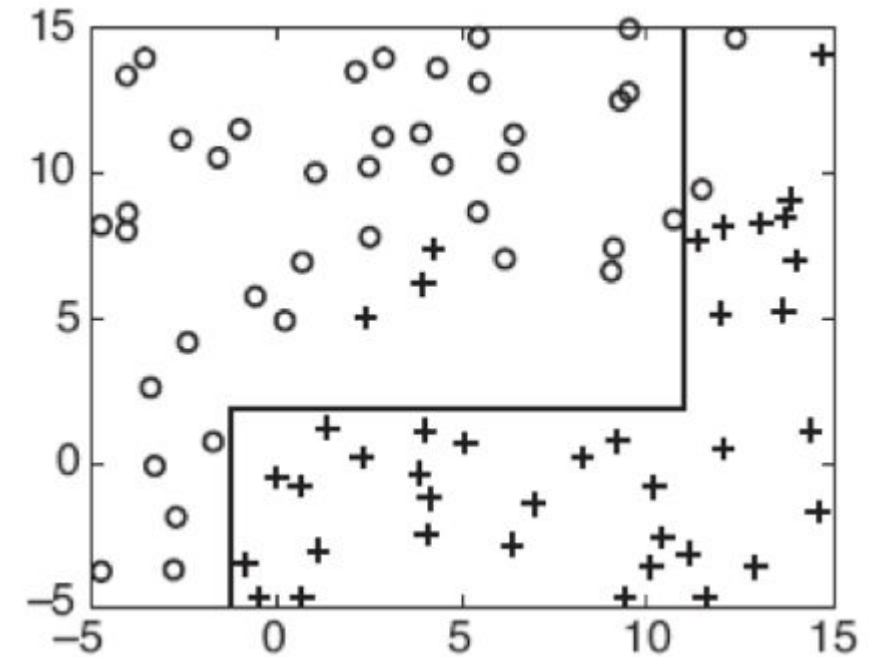


What is f is a diagonal line?

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The Representational Problem

Then it cannot be represented by finitely many rectilinear segments, and therefore the true decision boundary cannot be obtained by a decision tree classifier.

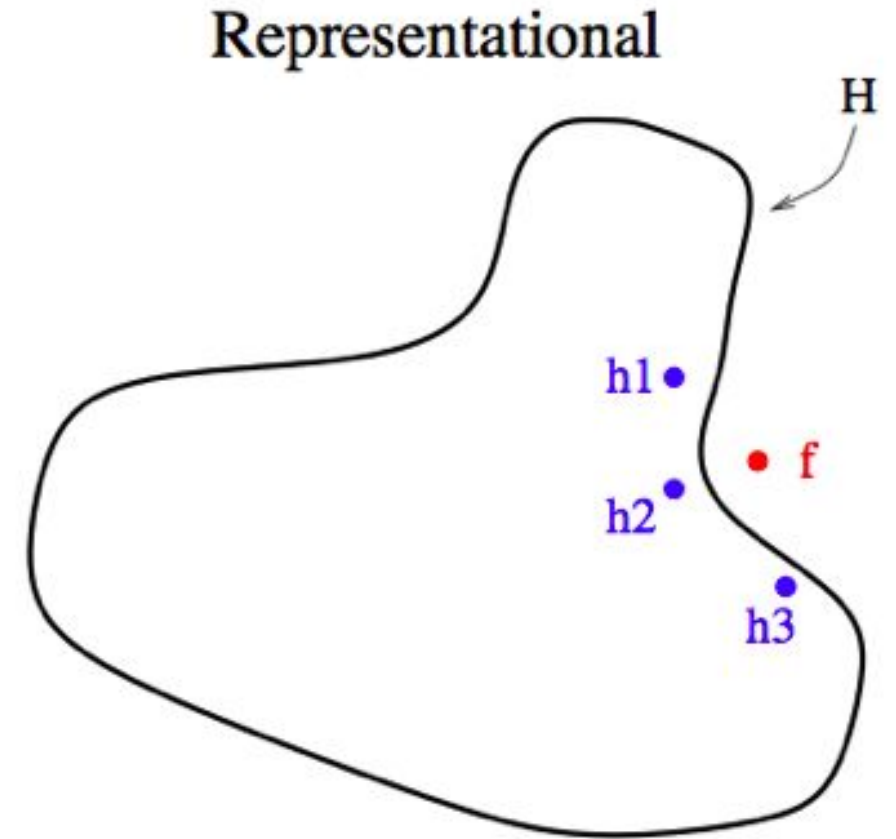


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The Representational Problem

However, it may be still be possible to approximate f or even to expand the space of representable functions using ensemble methods.



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Characteristics of Ensemble Methods

In order for an ensemble classifier to outperform a single base classifier, the following conditions must be met:

- **accuracy:** base classifiers outperform random guessing
- **diversity:** misclassifications must occur on different training examples

Practice: Real World Applications of Ensemble Methods

Introduction: Bagging

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Bagging (Bootstrap Aggregating)

Bagging or bootstrap aggregating is a method that involves manipulating the training set by resampling

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Unlike Boosting, Bagging uses uniform weights (e.g. a uniform sampling distribution)

Demo: Bagging Classifier in Scikit Learn

Independent Practice

Conclusion

Feature Scaling

Q & A

Feature Scaling

Review

Feature Scaling

EXIT TICKET

DON'T FORGET TO FILL OUT YOUR EXIT TICKET