

CSE 446

Ensembles

Administrative

- Quiz sections
 - No quiz section this week
 - Will have additional quiz section next week (announced soon)
 - Midterm problems & answers
 - Differentiation (come to this if you found question 2 on the exam or backprop on homework 2 difficult!)
- Midterm grading in progress...

Boosting

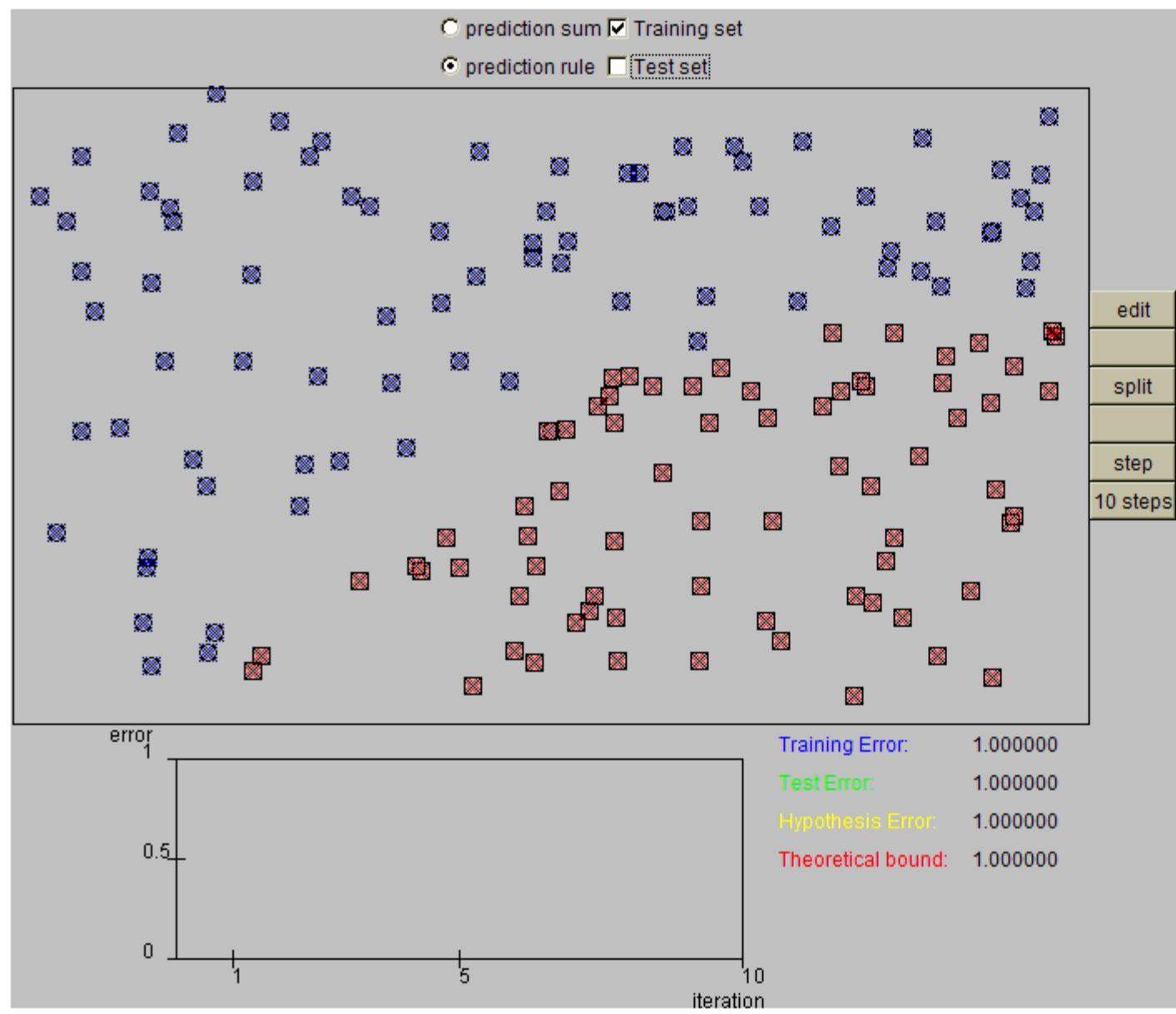
[Schapire, 1989]

- **Idea:** if we give each weak learner a different piece of the dataset, we get a really good complex classifier from letting them vote
- Learners must be different (how was this achieved in Bagging?)
- Learners must be better than random (not too weak)
- Approach: given a weak learner, run it multiple times on (reweighted) training data, then let learned classifiers vote
- On each iteration t :
 - weight each training example by how incorrectly it was classified
 - Learn a hypothesis – h_t
 - A strength for this hypothesis – α_t

- Final classifier:

$$h(x) = \text{sign} \left(\sum_i \alpha_i h_i(x) \right)$$

- **Practically useful**
- **Theoretically interesting**



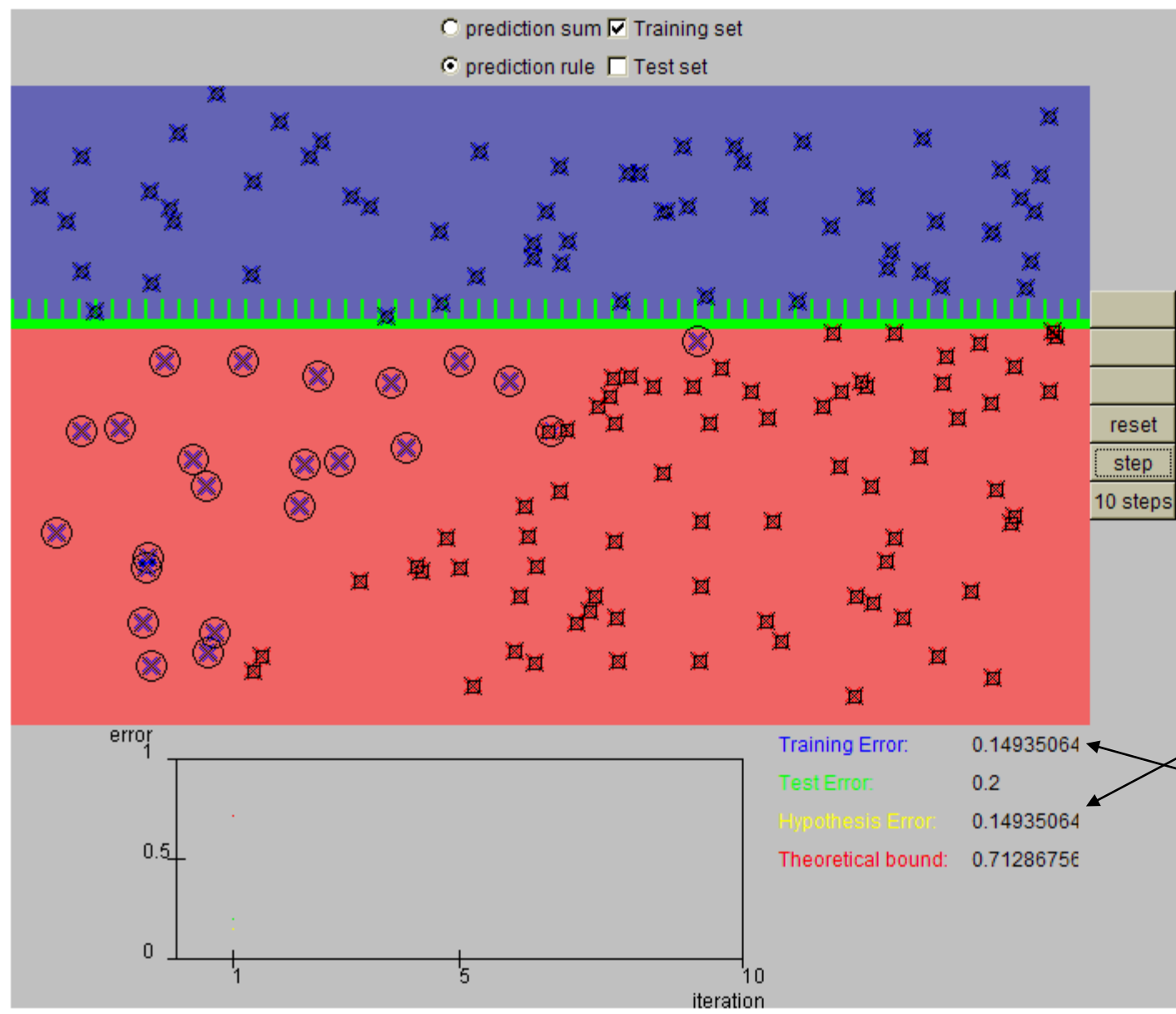
time = 0

blue/red = class

size of dot = weight

weak learner =
Decision stub:
horizontal or vertical

First, generate a data-set by clicking on the left and right buttons in the main window of the applet. Then, press "split" to split the data into training and test sets

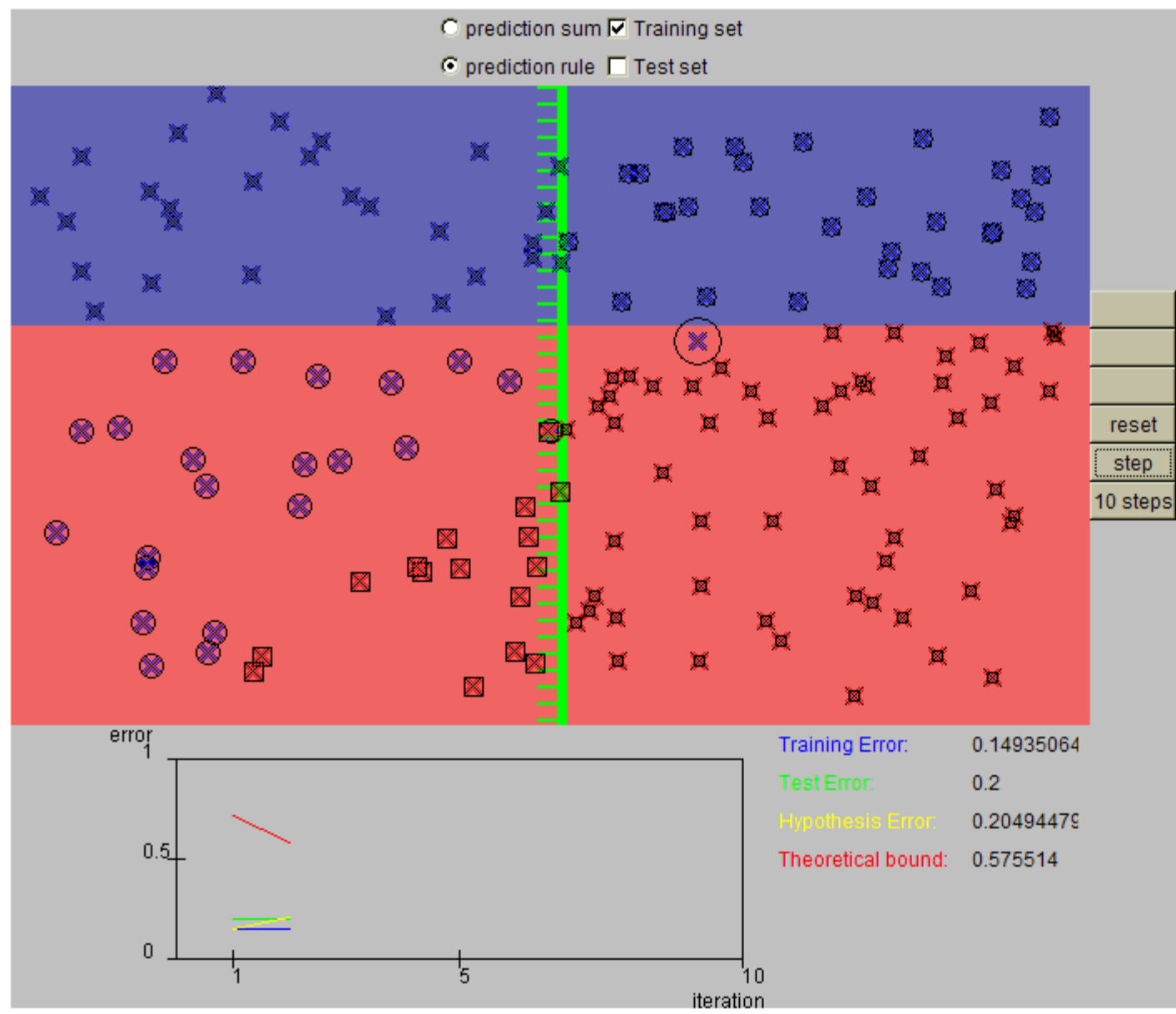


time = 1

this hypothesis has 15% error

and so does this ensemble, since the ensemble contains just this one hypothesis

First, generate a data-set by clicking on the left and right buttons in the main window of the applet. Then, press "split" to split the data into training and test sets



time = 2

First, generate a data-set by clicking on the left and right buttons in the main window of the applet. Then, press "split" to split the data into training and test sets

prediction sum Training set
 prediction rule Test set

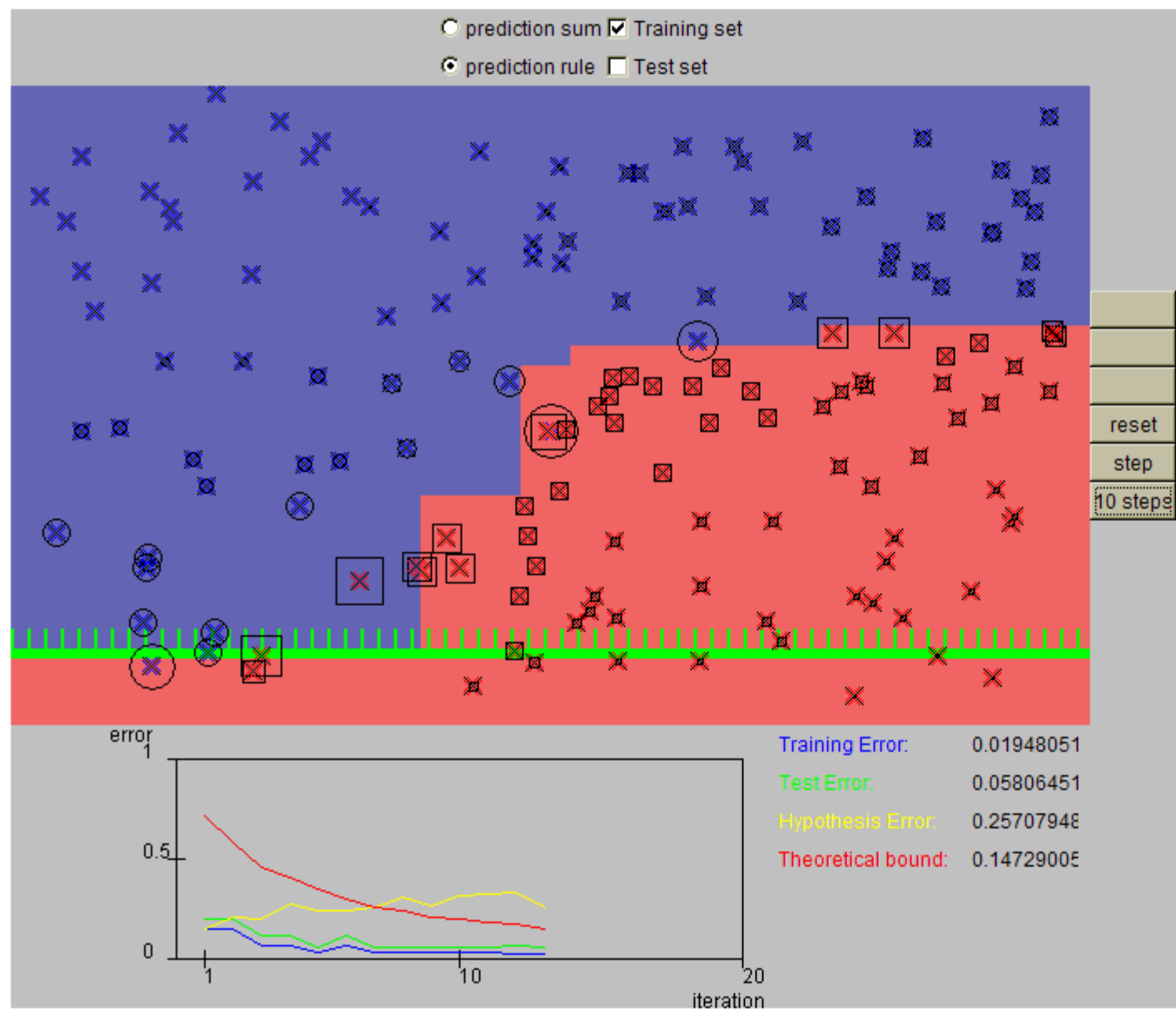
error

Training Error:	0.06493506
Test Error:	0.10967742
Hypothesis Error:	0.1930939
Theoretical bound:	0.45434076

iteration

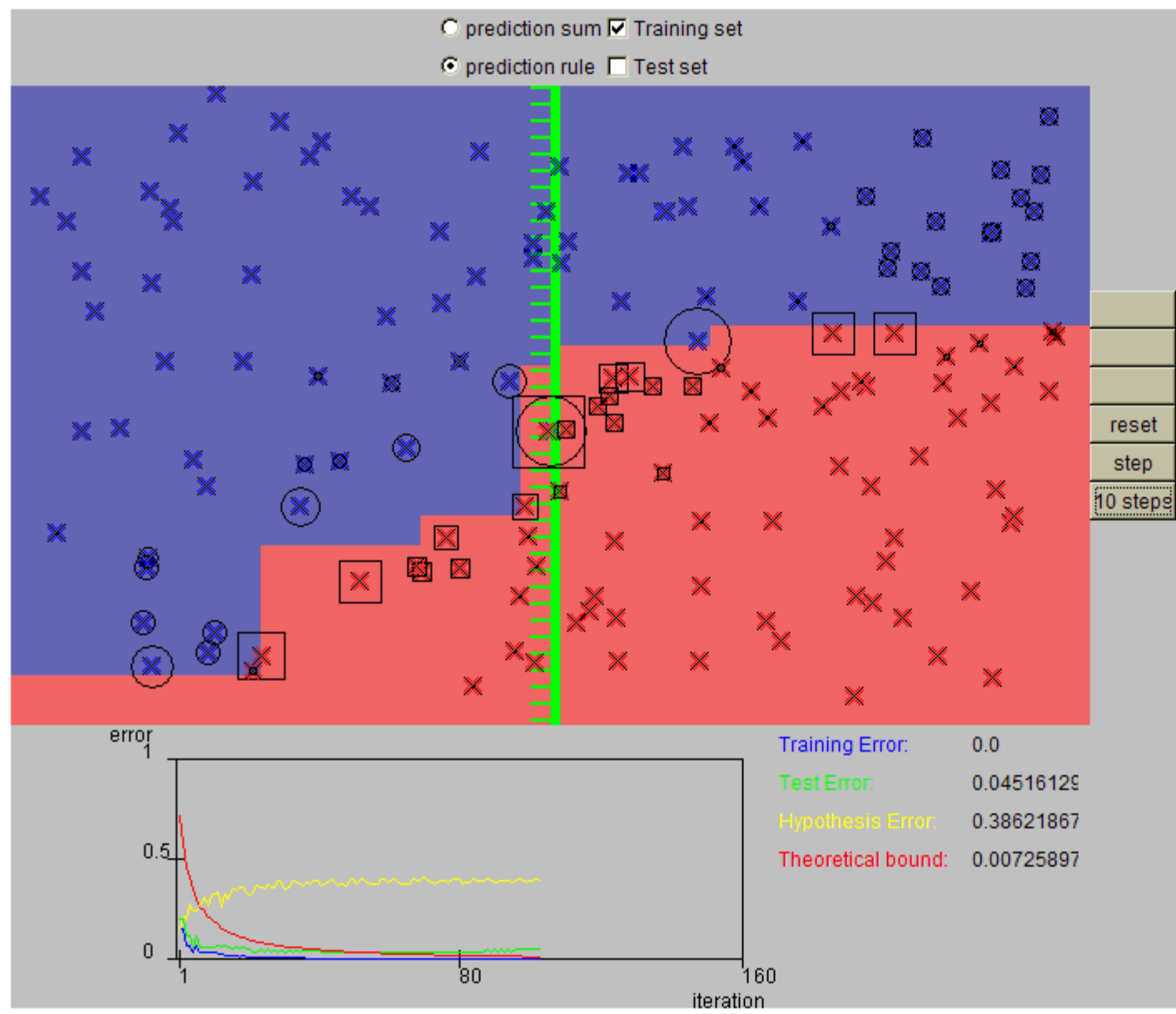
time = 3

First, generate a data-set by clicking on the left and right buttons in the main window of the applet. Then, press "split" to split the data into training and test sets



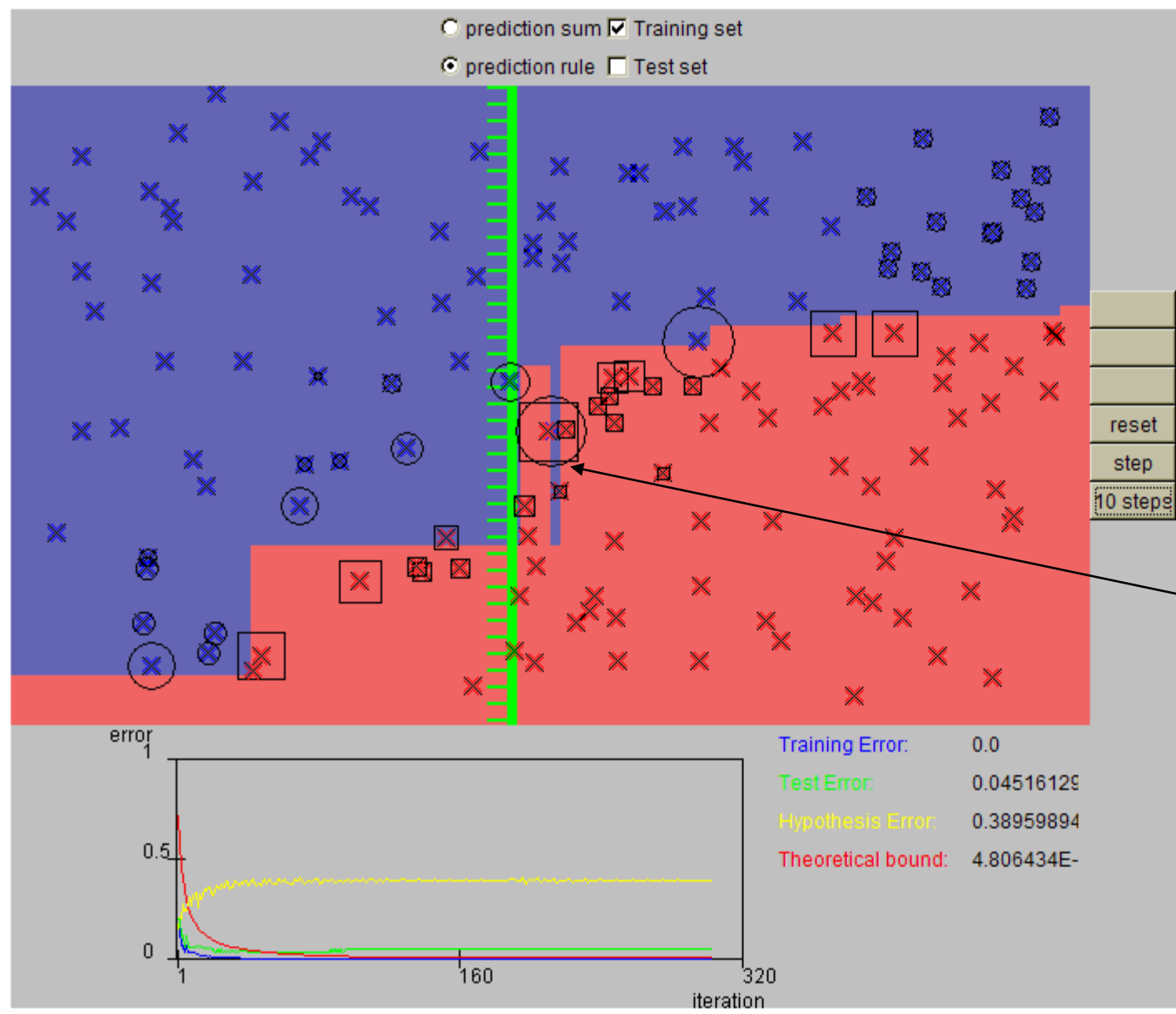
time = 13

First, generate a data-set by clicking on the left and right buttons in the main window of the applet. Then, press "split" to split the data into training and test sets



time = 100

First, generate a data-set by clicking on the left and right buttons in the main window of the applet. Then, press "split" to split the data into training and test sets



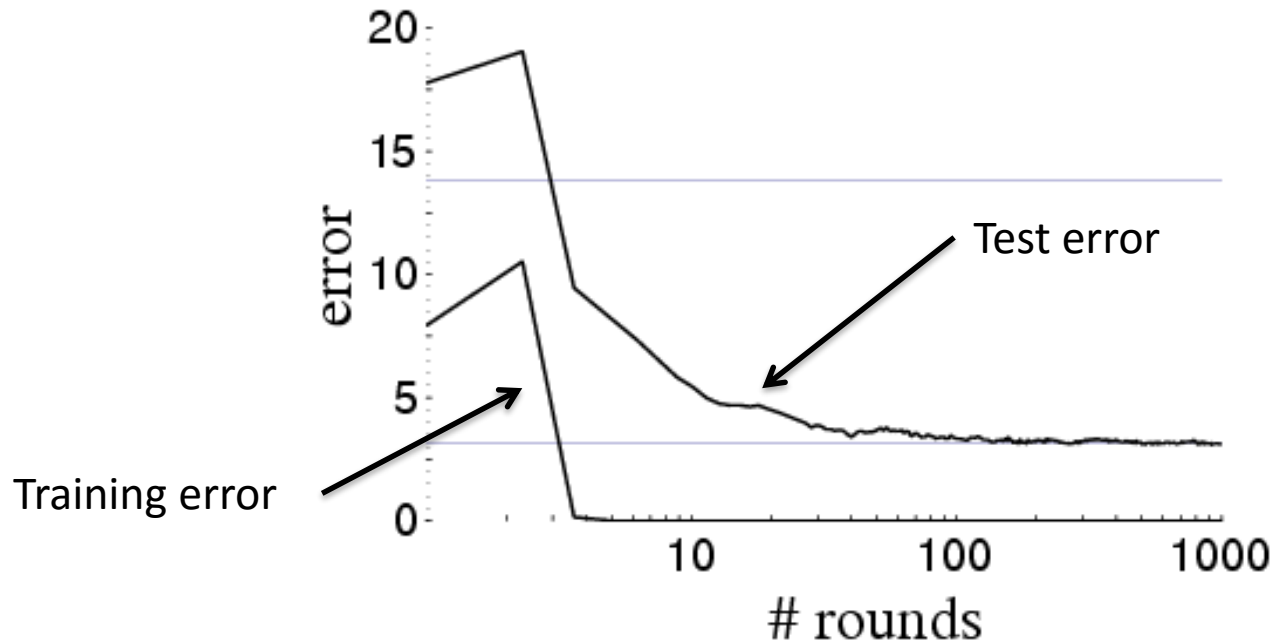
time = 300

overfitting

First, generate a data-set by clicking on the left and right buttons in the main window of the applet. Then, press "split" to split the data into training and test sets

Boosting results – Digit recognition

[Schapire, 1989]



- **Boosting:**
 - Seems to be robust to overfitting
 - Test error can decrease even after training error is zero?

Boosting generalization error bound

[Freund & Schapire, 1996]

$$error_{true}(H) \leq error_{train}(H) + \tilde{O} \left(\sqrt{\frac{Td}{m}} \right)$$

Constants:

- T : number of boosting rounds
 - Higher $T \rightarrow$ Looser bound
- d : measures complexity of classifiers
 - Higher $d \rightarrow$ bigger hypothesis space \rightarrow looser bound
- m : number of training examples
 - more data \rightarrow tighter bound

Boosting generalization error bound

[Freund & Schapire, 1996]

$$error_{true}(H) \leq error_{train}(H) + \tilde{O} \left(\sqrt{\frac{Td}{m}} \right)$$

Constants:

- **Theory does not match practice:**
 - Robust to overfitting
 - Test set error decreases even after training error is zero
 - **Need better analysis tools**
 - we'll come back to this later in the quarter
- more data \rightarrow tighter bound

Boosting: Experimental Results

[Freund & Schapire, 1996]

Comparison of C4.5, Boosting C4.5, Boosting decision stumps (depth 1 trees), 27 benchmark datasets

