Boosting

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Boosting Motivation

- Question of Kearns: Can you turn a "weak" learning algorithm (that is barely better than random guessing) into a "strong" learning algorithm (whose error rate is arbitrarily close to 0)?
- We could ask the algorithm to create a lot of classifiers and figure out how to combine them... how to do that?

Boosting Motivation

Schapire and Freund's answer:

- Reweight the data in many ways
- Use the weak learning algorithm to create a weak classifier for each (reweighted) dataset
- Compute a weighted average of the weak classifiers.

Weak classifiers used by Viola and Jones

· Subtract the white areas from the black ones









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Subtract the white areas from the black ones



Weak classifiers used by Viola and Jones

· Subtract the white areas from the black ones

Doesn't detect anything



Black and white areas are very similar

Subtract the white areas from the black ones

Doesn't detect anything



Black and white areas are very similar

Weak classifiers used by Viola and Jones

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Weak classifiers used by Viola and Jones

• Subtract the white areas from the black ones

Now it detects!



• Subtract the white areas from the black ones

Weak classifiers





Detects eyes!

Weak classifiers used by Viola and Jones

• Subtract the white areas from the black ones

Weak classifiers









Detects eyes!

 Used hundreds of thousands of these weak classifiers at all different scales



Weak classifiers used by Viola and Jones

 Used hundreds of thousands of these weak classifiers at all different scales



AdaBoost Pseudocode

Assign observation *i* the weight of $d_{1i}=1/n$ (equal weights).

For t=1:T

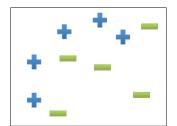
Train weak learning algorithm using data weighted by d_{ti} . This produces weak classifier h_t .

Choose coefficient α_t . $y_i h_t(x_i)$ is 1 if correct, -1 if incorrect Update weights: $d_{t+1,i} = \frac{d_{t,i} \exp(-\alpha(y_i h_t(x_i)))}{Z_t}$ Z_t is a normalization factor.

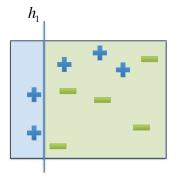
End

Output the final classifier: $H(x) = \text{sign}\left(\sum_{i=1}^{T} \alpha_i h_i(x_i)\right)$

Boosting Example



All points start with equal weights.

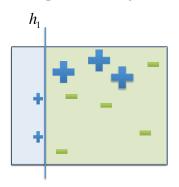


Run the weak learning algorithm to get a weak classifier.

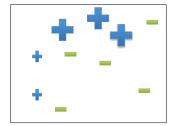
Choose coefficient $\alpha_1 = .41$

(Credit: Example adapted from Freund and Schapire)

Boosting Example

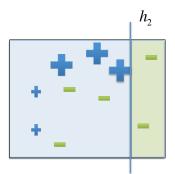


Increase the weights on the misclassified points, decrease the weights on the correctly classified points.



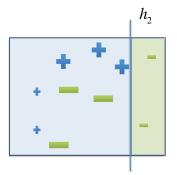
(Credit: Example adapted from Freund and Schapire)

Boosting Example



Run the weak learning algorithm to get a weak classifier for the weighted data.

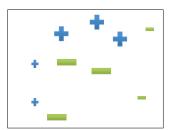
Choose coefficient $\alpha_2 = .66$



Increase the weights on the misclassified points, decrease the weights on the correctly classified points.

(Credit: Example adapted from Freund and Schapire)

Boosting Example



Increase the weights on the misclassified points, decrease the weights on the correctly classified points.



Increase the weights on the misclassified points, decrease the weights on the correctly classified points.

Choose coefficient $\alpha_3 = .93$

(Credit: Example adapted from Freund and Schapire)

Boosting Example