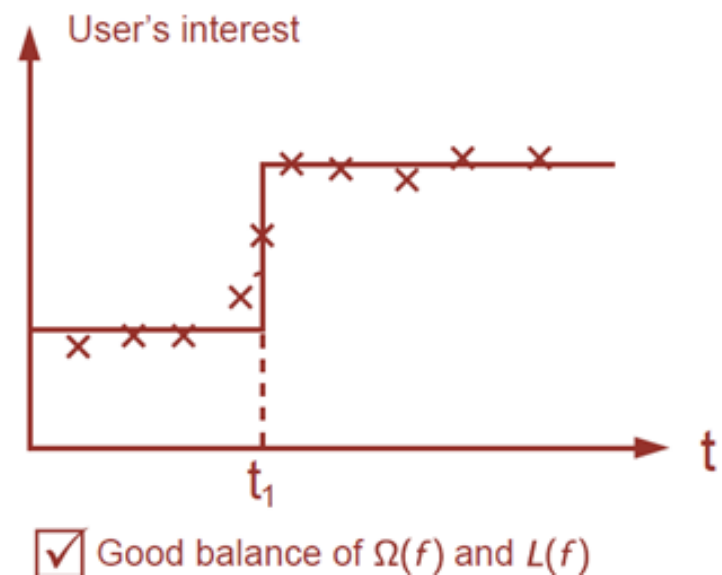
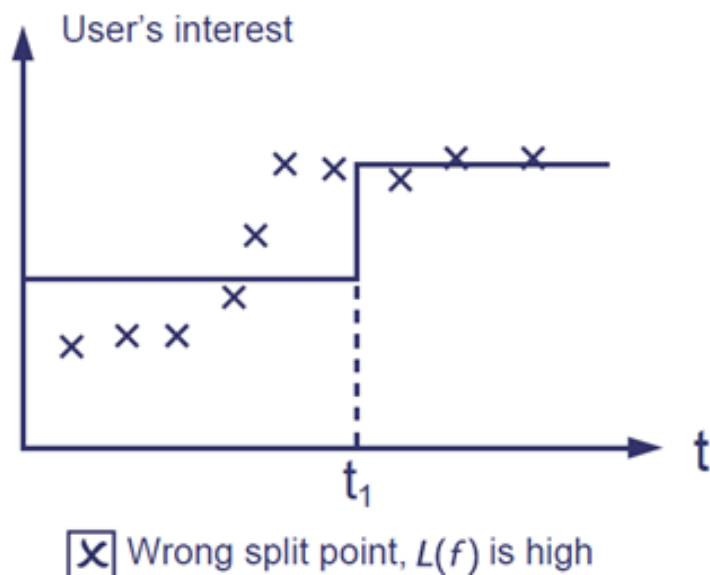
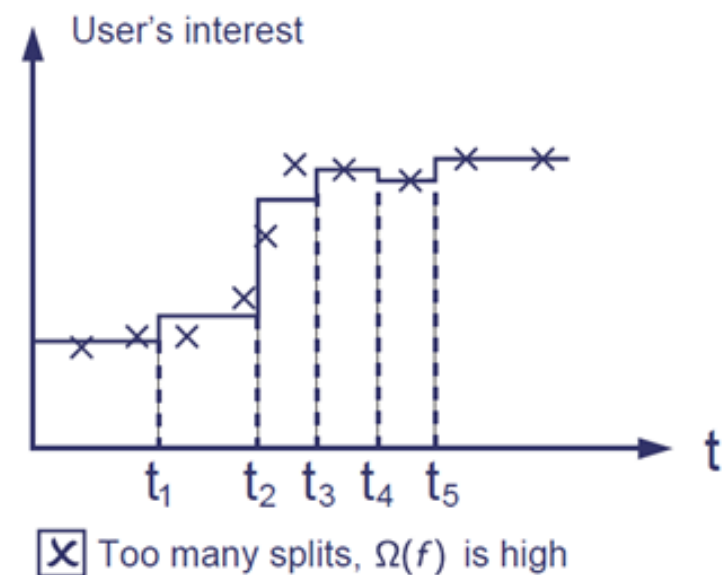
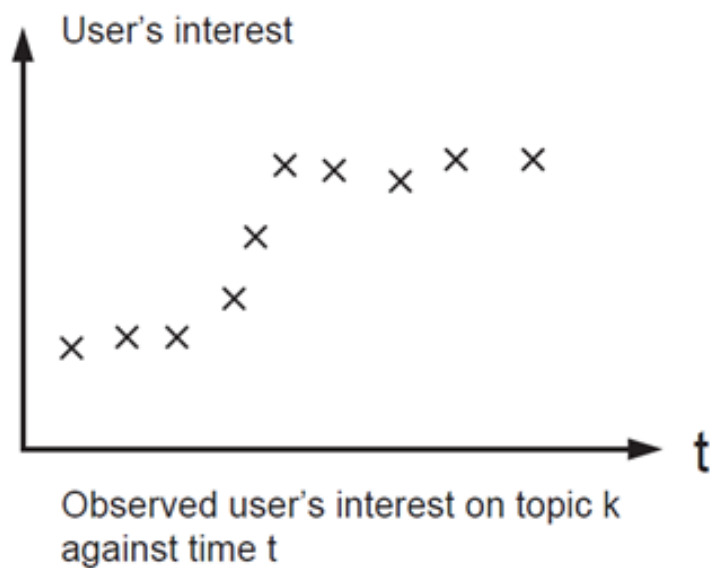


# XGBOOST

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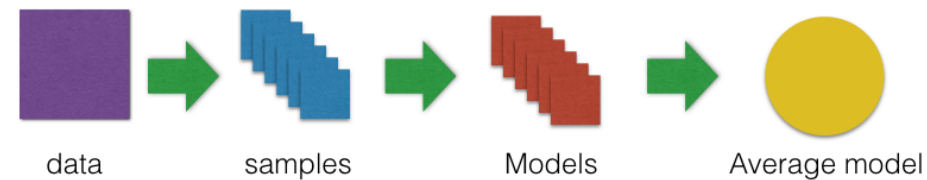
# EXAMPLES



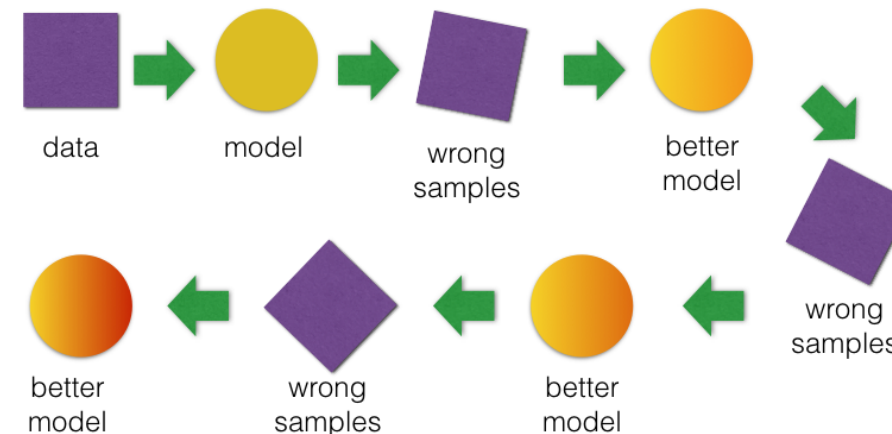
# OPENING

- Recall that **boosting** is a sequential additive modeling technique that fits a model, reweights the observations based on whether they were classified properly or improperly, then fits an additional model, and so on until the algorithm stops.

## Bagging



## Boosting

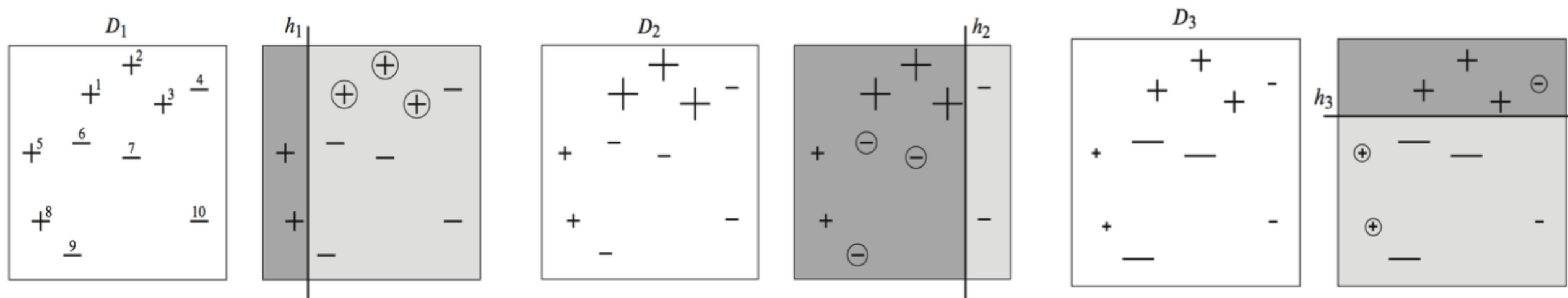


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# OPENING

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## Adaboost



**Figure:** AdaBoost. Source: Figure 1.1 of [Schapire and Freund, 2012]

# OPENING

## Adaboost

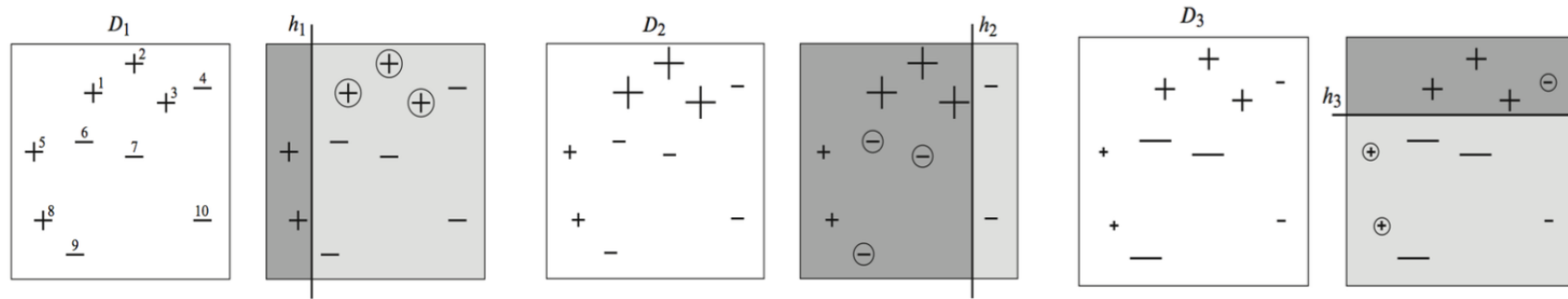


Figure: AdaBoost. Source: Figure 1.1 of [Schapire and Freund, 2012]

- ▶ Fit an additive model (ensemble)  $\sum_t \rho_t h_t(x)$  in a forward stage-wise manner.
- ▶ In each stage, introduce a weak learner to compensate the shortcomings of existing weak learners.
- ▶ In Adaboost, “shortcomings” are identified by high-weight data points.

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# OPENING

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## Adaboost

$$H(x) = \sum_t \rho_t h_t(x)$$

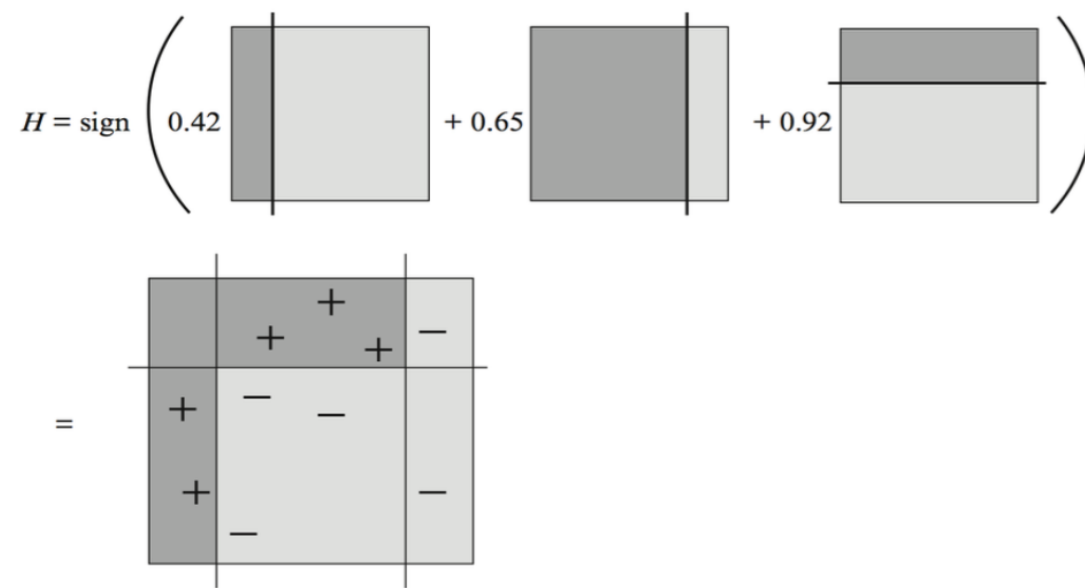


Figure: AdaBoost. Source: Figure 1.2 of [Schapire and Freund, 2012]

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# OPENING

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- This method, AdaBoost, was generalized to include the gradient descent method.  
The highlights:
  - AdaBoost is an iterative, additive model.
  - AdaBoost iterates by reweighting observations based on misclassification.
  - These misclassifications are “residuals” and, treating these residuals as gradients, we can formulate this as a gradient descent problem with a particular loss function.
  - As a gradient descent problem, we can replace the loss function associated with AdaBoost with other loss functions to make this more general.
  - This generalization is known as “gradient boosting.”

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# XGBOOST

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- XGBoost, or “eXtreme Gradient Boosting,” is similar to gradient boosting.
  - Random Forests are our boosted models.
  - $\text{XGBoost Model} = \text{RF1} + \text{RF2} + \dots$
- XGBoost has some very good computational benefits.
- XGBoost is capable of handling missing values.