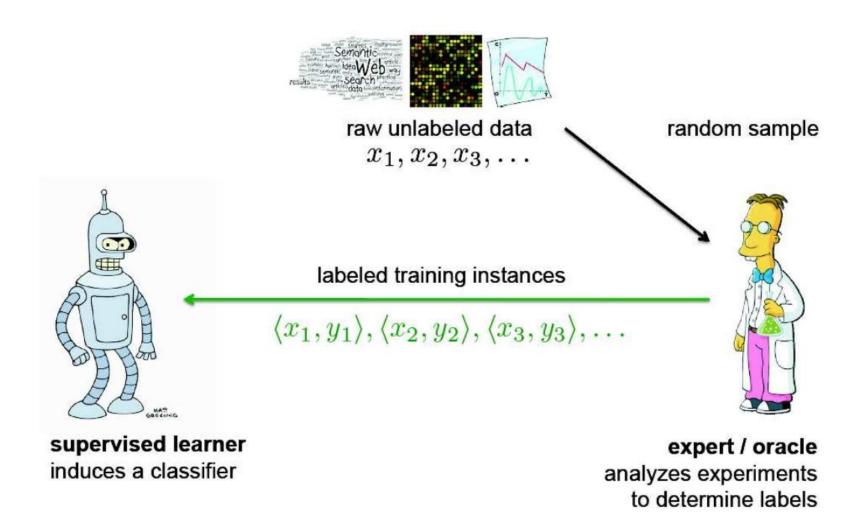
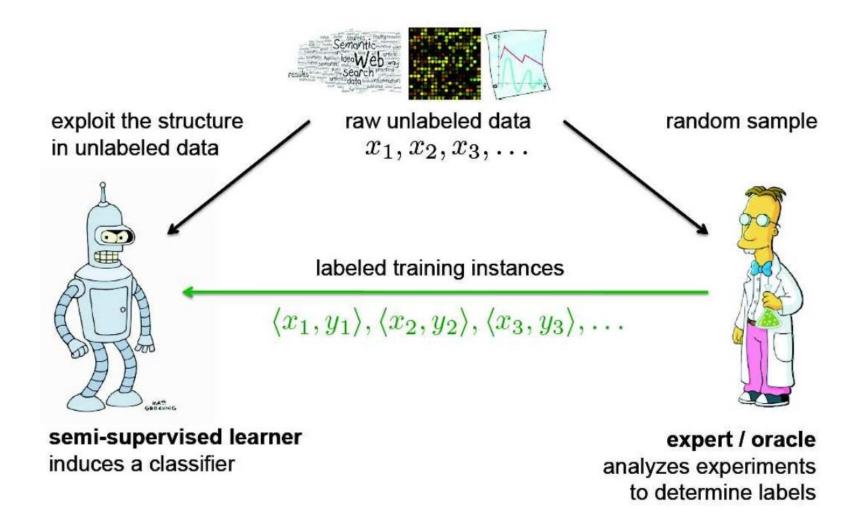
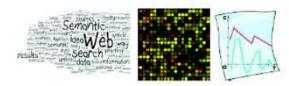
(Passive) Supervised Learning

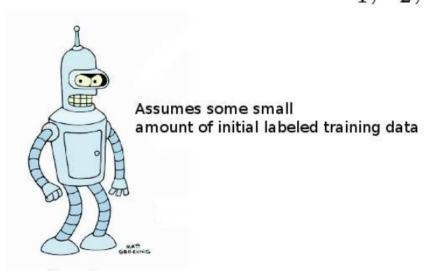


Semi-Supervised Learning





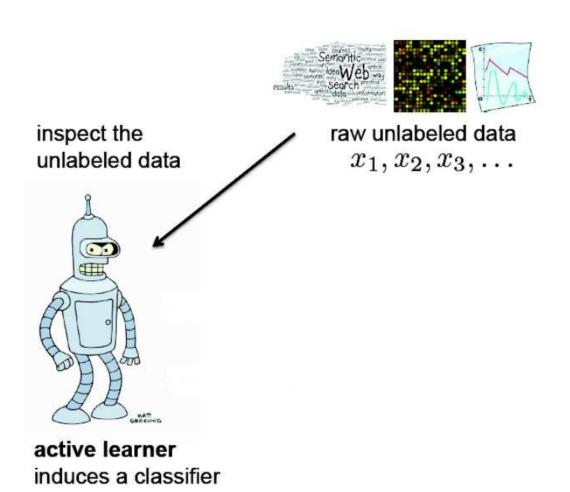
raw unlabeled data x_1, x_2, x_3, \dots



active learner induces a classifier

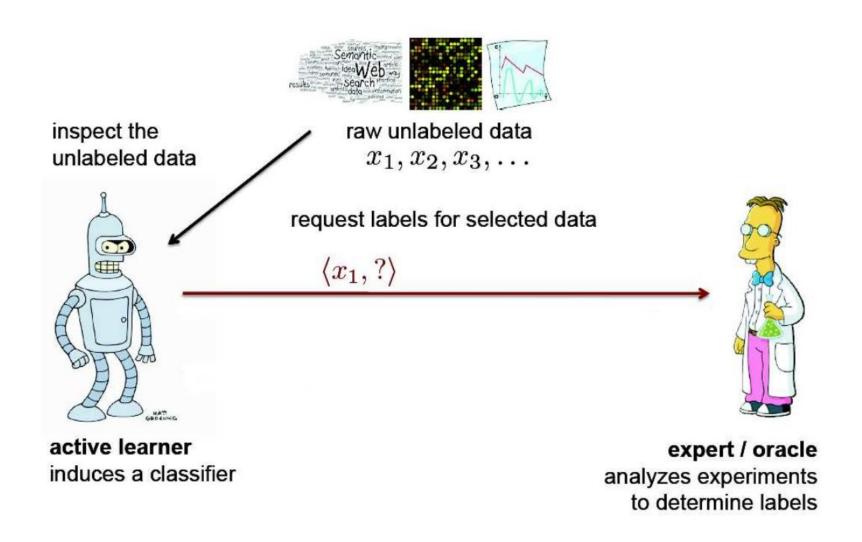


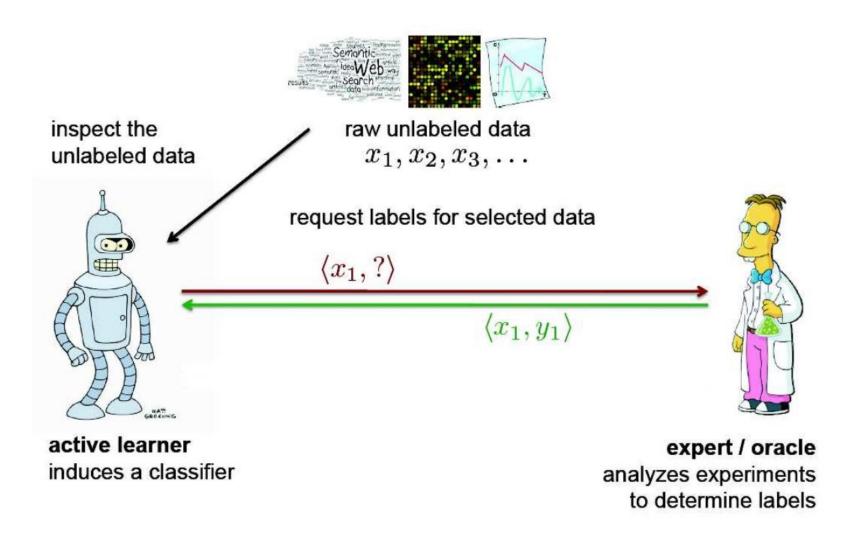
expert / oracle analyzes experiments to determine labels

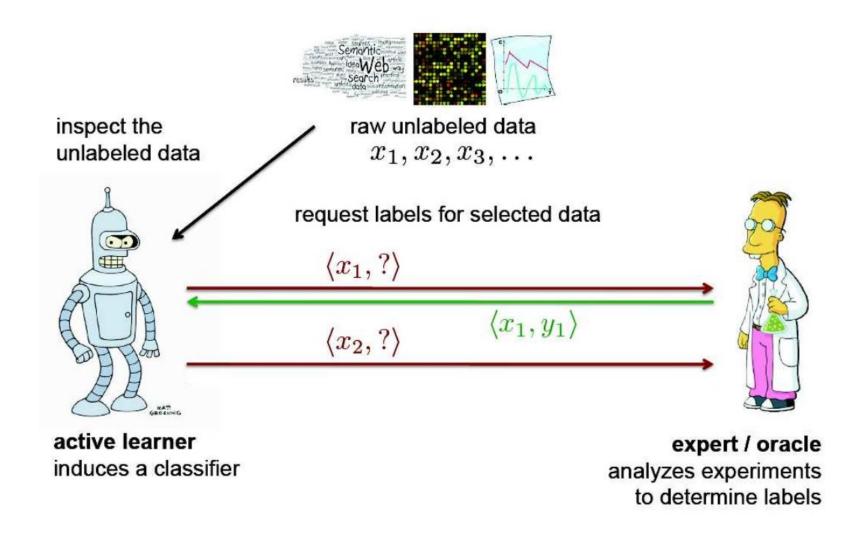


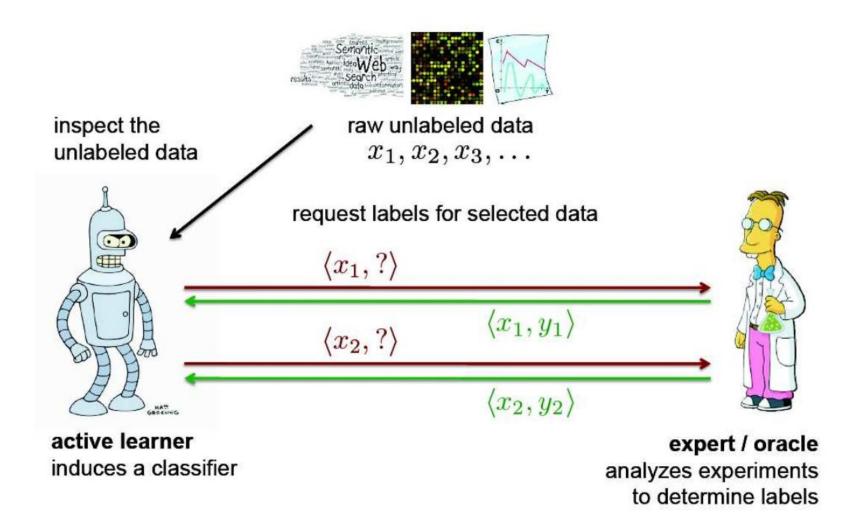


expert / oracle analyzes experiments to determine labels









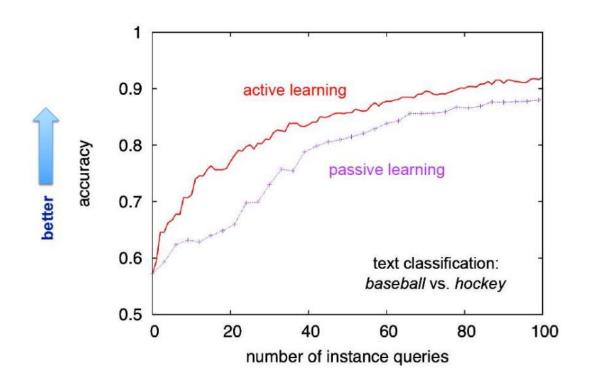
Active Learning vs Random Sampling

Passive learning curve

Randomly selects examples to get labels for

Active learning curve

Active learning selects examples to get labels for



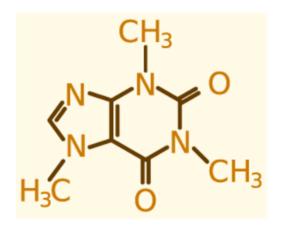
Motivation

• Why do we need active learning?

- Supervised learning can solve all our problems, right?
- Yes, if we have enough labeled data (input-output pairs)
- But Labeling is expensive
- We want to learn a highly-accurate function with few labeled examples
- ▲ Intelligently select the examples for which we want to get labels for (unlabeled data is plentiful and cheap)

Active Learning Example: Drug Design

Goal: find compounds which bind to a particular target



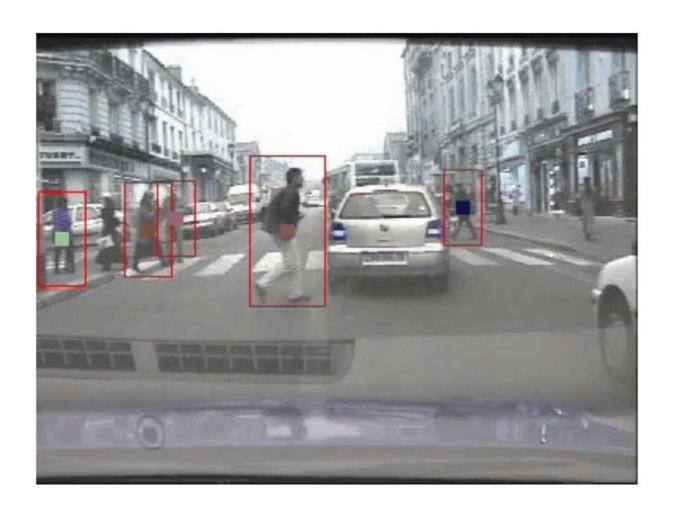
Large collection of compounds, from:

- vendor catalogs
- corporate collections
- combinatorial chemistry

```
unlabeled point \equiv description of chemical compound label \equiv active (binds to target) vs. inactive
```

getting a label ≡ chemistry experiment

Active Learning Example: Pedestrian detection



Who uses Active Learning?



Sentiment analysis for blogs; Noisy relabeling

- Prem Melville



Biomedical NLP & IR; Computer-aided diagnosis

Balaji Krishnapuram



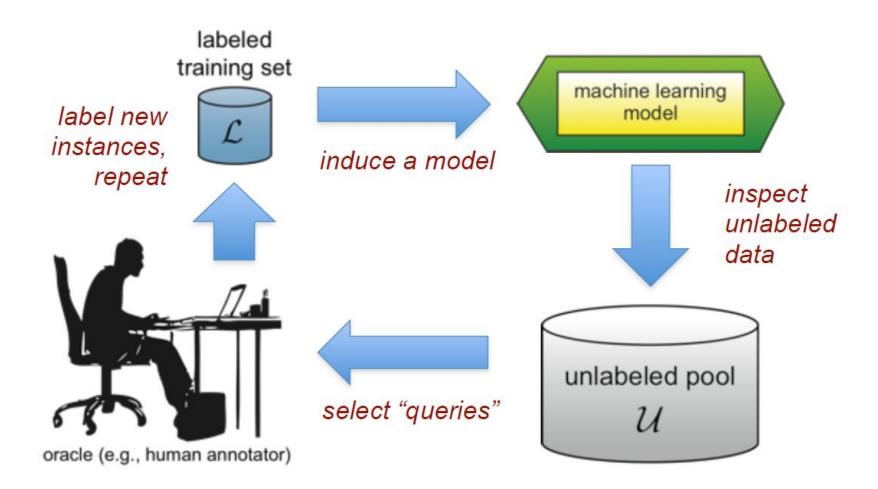
MS Outlook voicemail plug-in [Kapoor et al., IJCAI'07]; "A variety of prototypes that are in use throughout the company." – *Eric Horvitz*



"While I can confirm that we're using active learning in earnest on many problem areas... I really can't provide any more details than that. Sorry to be so opaque!"

David Cohn

Pool based Active Learning



Credit: Burr Settles

Query Selection Strategies

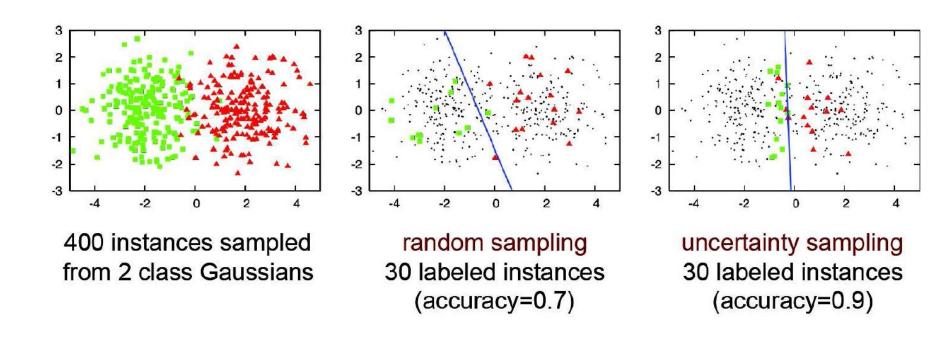
- Any active learning algorithm requires a query selection strategy.
- Some examples
 - Uncertainty Sampling
 - Query By Committee (QBC)
 - Expected Model Change
 - Expected Error Reduction
 - ◆ Variance Reduction
 - Density Weighted Methods

Uncertainty Sampling

- Select examples which the current model is most uncertain about
- Many ways to measure uncertainty
 - Based on distance from the hyperplane
 - Using the probability distribution over labels P(y|x) for probabilistic models
- Some examples based on label probabilities
 - ▲ Least Confident, where confidence is defined as 1 probability of highest scoring label
 - ▲ Smallest Margin, where margin is defined as the difference between the probabilities of the first and second best labels

Uncertainty Sampling: Illustration

 Example: uncertainty sampling based on the distance from hyperplane (i.e., margin based)



Query By Committee (QBC)

- QBC uses a committee of models (say $h_1, h_2, ..., h_k$)
- ullet All models are trained using the currently available labeled data L
- How is the committee constructed?
 - Ensemble methods (e.g., bagging, boosting)
- All models vote their predictions on the unlabeled pool
- The example(s) with maximum disagreement is(are) chosen flabeling
 - simple disagreement rate
 - Entropy over the vote distribution over all labels
- Each model in the committee is re-trained with new example

Summary and Outlook

- Active learning is a label-efficient learning strategy
- Intelligently selects the examples based on their informativeness
- Other variants
 - Different examples having different labeling costs
 - Access to multiple labeling oracles varying in labeling cost and accuracy
 - Active learning on features instead of labels
- Further Reading
 - Active learning survey from Burr Settles
 - http://burrsettles.com/pub/settles.activelearning.pdf