



# Introduction to Active Learning

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Presented by: Yongchao Zhou

# Outline

- What is active learning?
- What kind of examples are most informative?
- How does AL fit into the ML workflow?

# What is Active Learning?

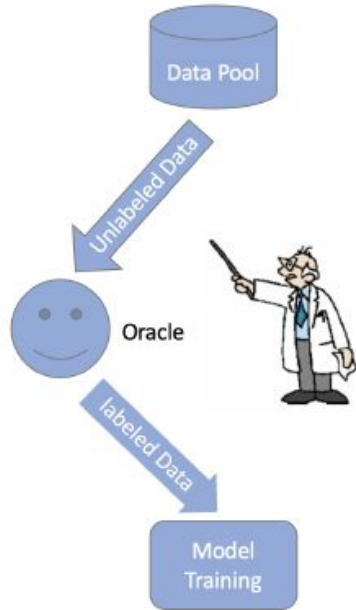


Figure 1: Passive Machine Learning

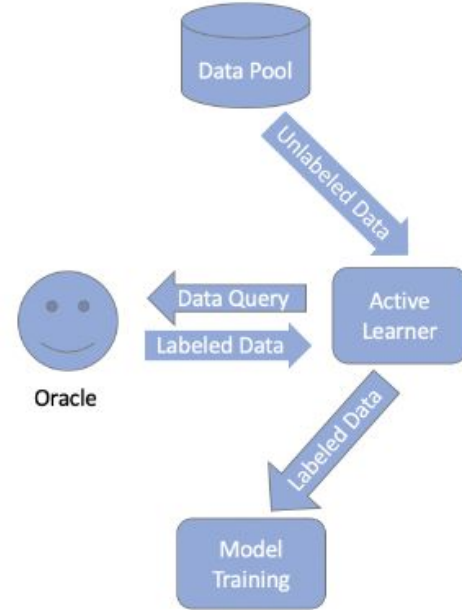


Figure 2: Active Machine Learning

# Active Learner & Label Efficiency

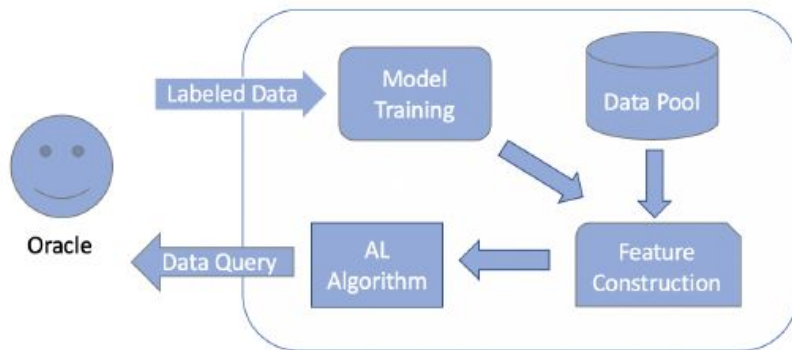


Figure 3: Active Learner

- AL's goal: Build a **high-quality dataset** such that the model trained on it can achieve as high test accuracy as possible.

# Typical Results

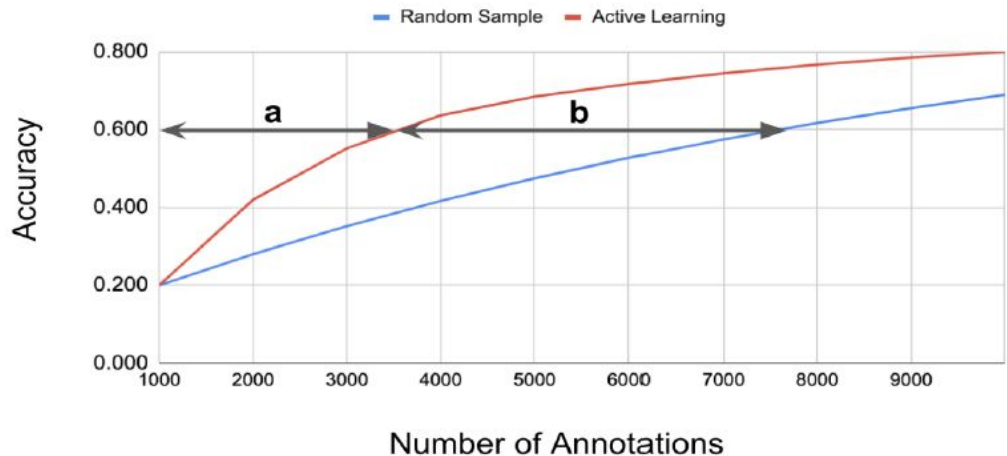


Figure 4: Typical Active Learning Result

- Given a target accuracy, the AL algorithm reduces the labeling effort by x%.
- Given a fixed labeling budget, the AL algorithm improves the performance by x%.

# What kind of examples are most informative?

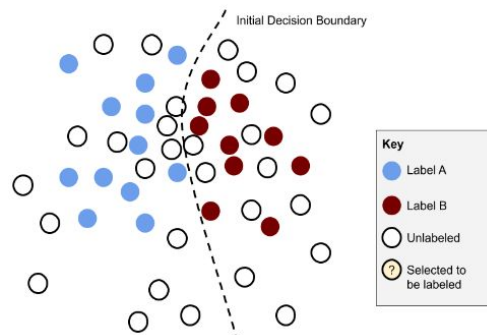


Figure 5: Current Decision Boundary

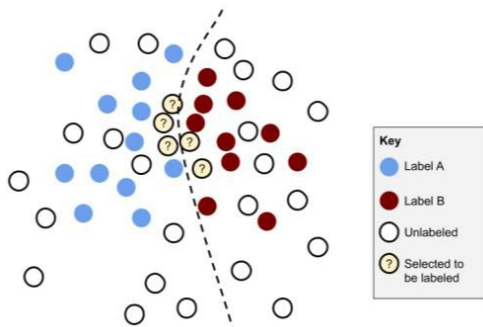


Figure 6: Uncertainty Sampling

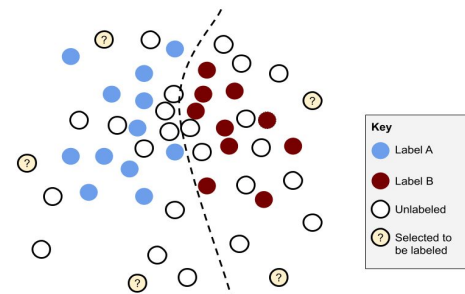


Figure 7: Diversity Sampling

# Uncertainty Sampling

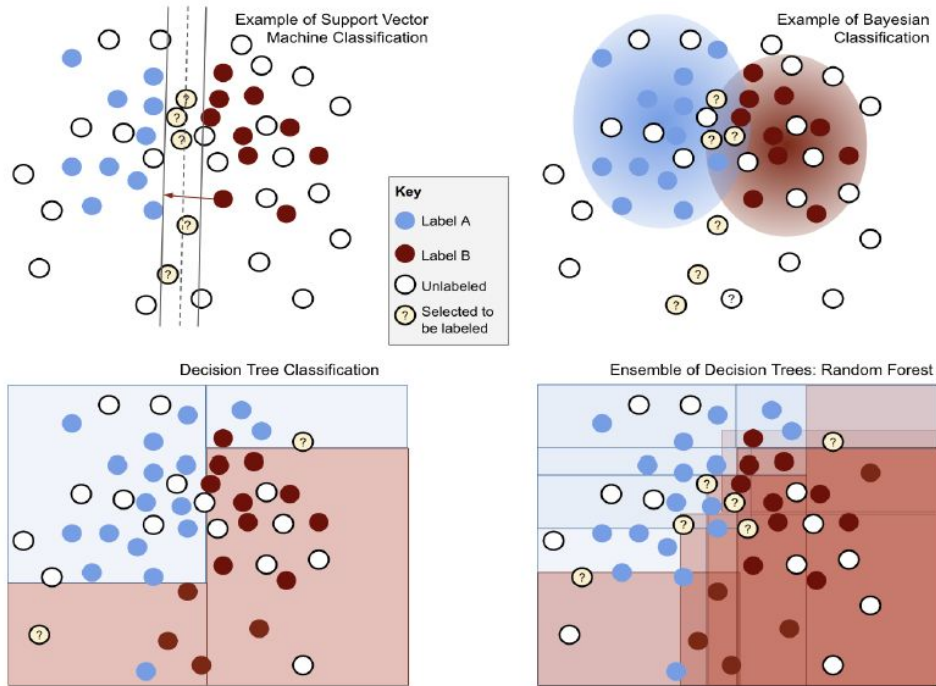


Figure 8: Uncertainty Sampling for different ML Models

# Diversity Sampling

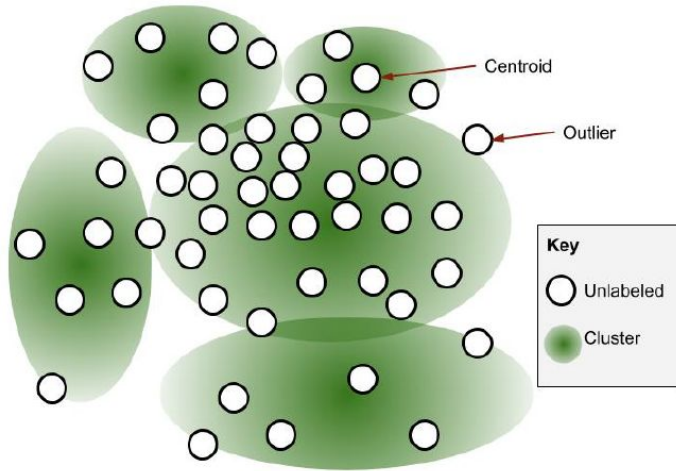


Figure 10: Cluster-based Sampling

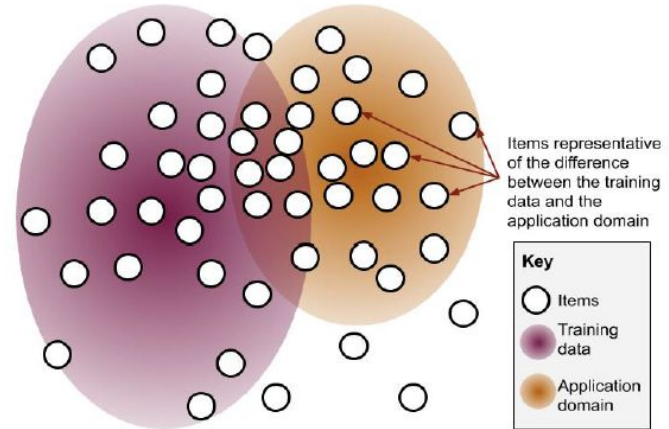


Figure 11: Representative Sampling



# Hybrid Method

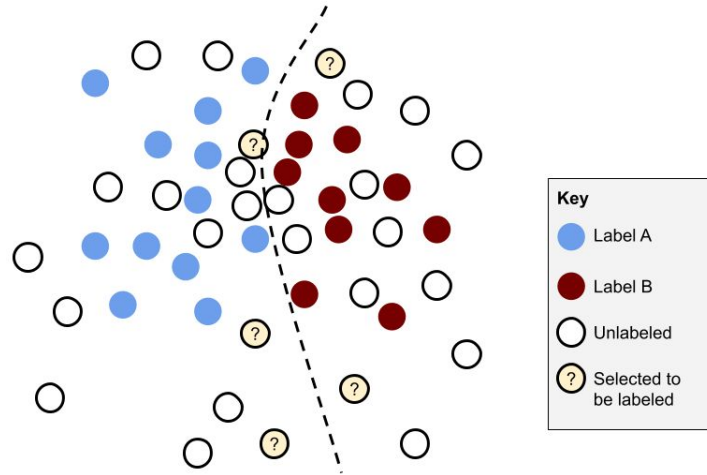


Figure 12: Hybrid Method

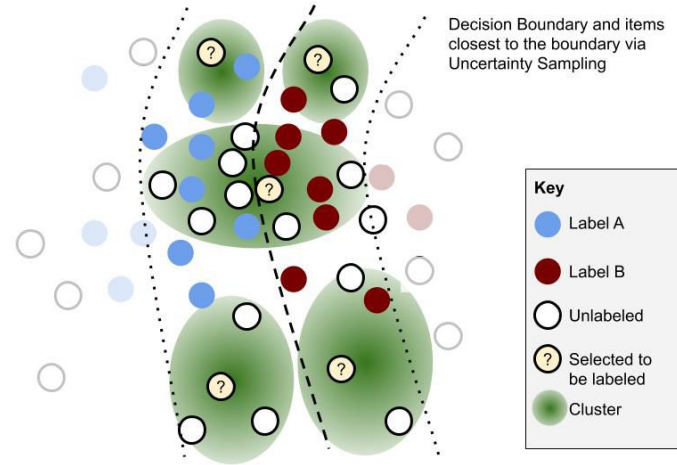


Figure 13: Uncertainty + Clustering

# Active Learning for Neural Networks

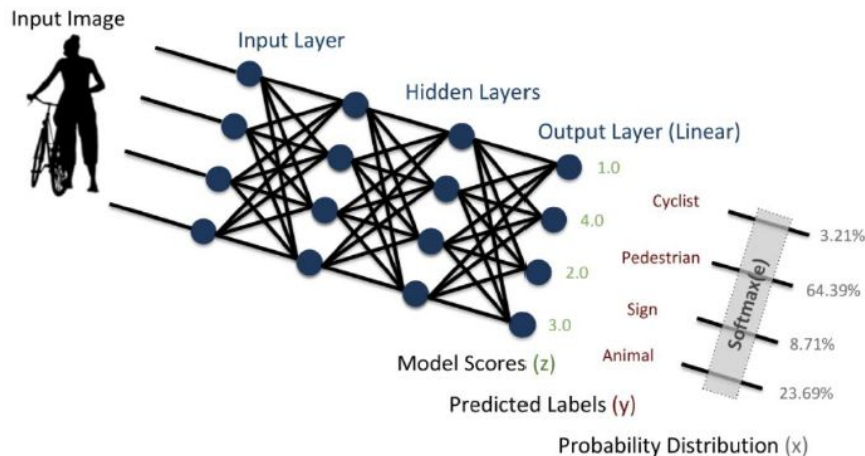


Figure 9: Neural Network Visualization

- Prediction-based
  - Least Confidence
  - Max Entropy
  - Margin Sampling
  - Bayes-Coreset
- Model-based
  - Expected Parameter Change
  - Maximum Variance Reduction
  - Adversarial
  - Coreset
  - BADGE
- Ensemble-based
  - Query by Committee
  - Bayesian Disagreement

# Pros and Cons of different AL algorithm

Table 1: Pros and Cons of different AL algorithms

AL Type	Pros	Cons
Uncertainty Sampling	Simple to implement	Select redundant points
Diversity Sampling	Good sample diversity	Select easy points
Hybrid Method	Consider both uncertainty and diversity	Poor scalability

# Active Learning - An iterative process

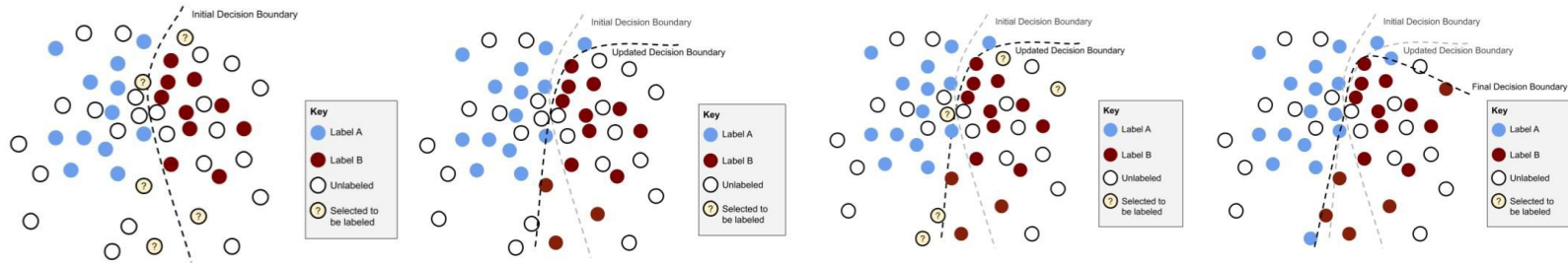


Figure 14: Active Learning Process

- Step 1: Apply Active Learning to sample items that require a human label to create additional training items.
- Step 2: Retrain the model with the new training items, resulting in a new decision boundary.
- Step 3: Apply Active Learning again to select a new set of items that require a human label.
- Step 4: (and beyond): Retrain the model again, and repeat the process to keep getting a more accurate model.

# How does AL fit into the ML workflow?

*Knowledge Quadrant for Machine Learning and strategies to solve different problems*

	Knowns	Unknowns
Known	Confident Predictions from Model (Known Knowns)  <b>Current Model State</b>	Non-Confident Predictions from Model (Known Unknowns)  <b>Uncertainty Sampling</b>
Unknown	Latent Information in Related Models (Unknown Knowns)  <b>Transfer Learning</b>	Gaps in Model Knowledge (Unknown Unknowns)  <b>Diversity Sampling</b>
	<i>Solve with Machine Learning</i>	<i>Solve with Active Learning &amp; Annotation</i>

Figure 15: Knowledge Quadrant for ML

# Human-in-the-Loop Machine Learning

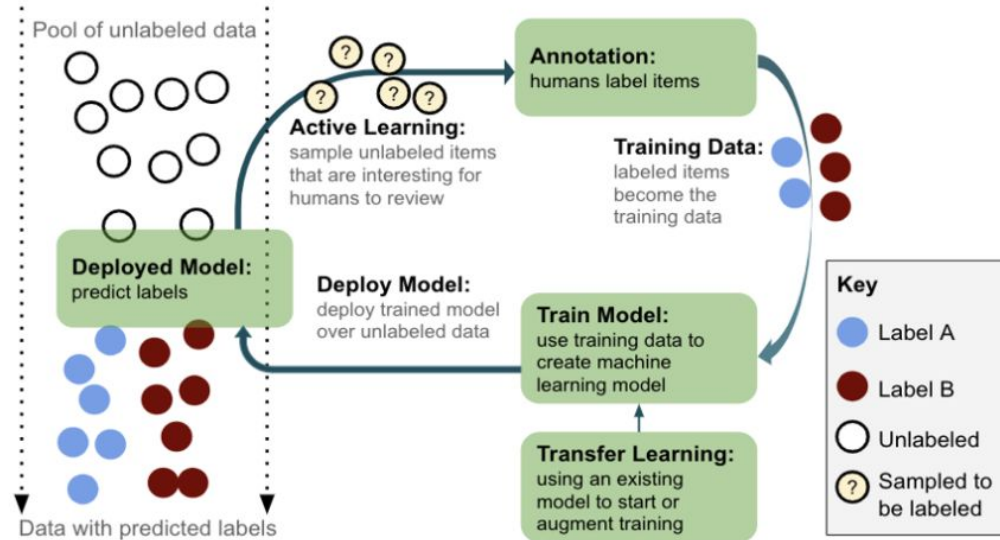


Figure 16: Human-in-the-loop Machine Learning Workflow



# Questions?



# Thank you!