

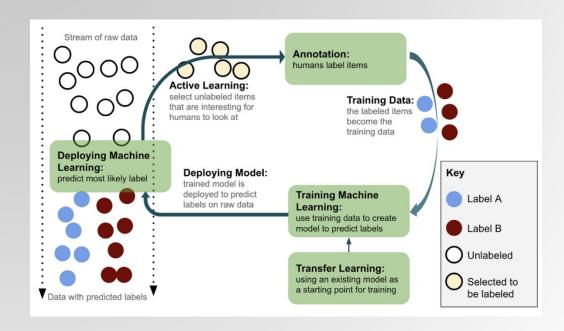
CSC 462: Machine Learning

7.8 Active Learning

Dr. Sultan Alfarhood

7.8 Active Learning

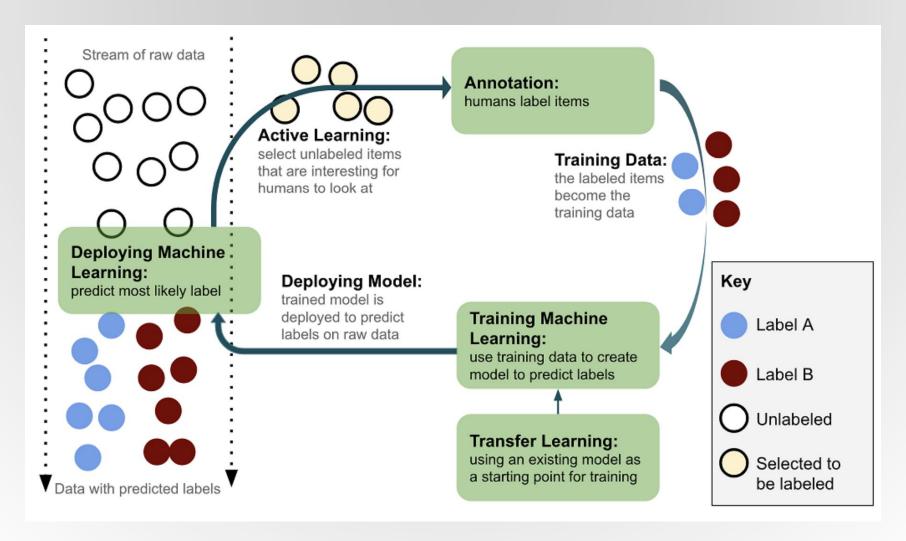
- Usually applied when obtaining labeled examples is costly
- The idea is that we start the learning with relatively few labeled examples, and a large number of unlabeled ones, and then add labels only to those examples that contribute the most to the model quality
- Strategies
 - 1. Data density and uncertainty based
 - 2. Support vector-based



CSC462: MACHINE LEARNING (SPRING 2024)

2

Active Learning



CSC462: MACHINE LEARNING (SPRING 2024)

Data density and uncertainty based active learning

- 1. Apply the ML model (trained using the existing labeled examples) to each of the remaining unlabeled examples
 - Or to some random sample of them to save the computing time
- 2. For each unlabeled example x, the following **importance score** is computed:

 $density(x) \cdot uncertainty(x)$

- Density reflects how many examples surround x in its close neighborhood
- Uncertainty reflects how uncertain the prediction of the model f is for x
- 3. Pick the one with the highest importance score and ask the expert to annotate it
- 4. Add the new annotated example to the training set, **rebuild** the model and continue the process until some stopping criterion is satisfied
 - A stopping criterion can be chosen in advance (the maximum number of requests to the expert based on the available budget)
 - Or depend on how well our model performs according to some metric.

Support vector-based active learning

- 1. Build an SVM model using the labeled data
- 2. Ask our expert to annotate the unlabeled example that lies the closest to the hyperplane that separates the two classes
- The idea is that if the example lies closest to the hyperplane, then it is the least certain and would contribute the most to

