## PROBABILISTIC CLASSIFIERS /2

- Both scoring and probabilistic classifiers can output classes by thresholding (binary case) / selecting the class with the maximum score (multiclass)
- Thresholding:  $h(\mathbf{x}) := [\pi(\mathbf{x}) \ge c]$  or  $h(\mathbf{x}) = [f(\mathbf{x}) \ge c]$  for some threshold c.
- Usually c = 0.5 for probabilistic, c = 0 for scoring classifiers.
- There are also versions of thresholding for the multiclass case





## DISCRIMINANT APPROACH

The **discriminant approach** tries to optimize the discriminant functions directly, usually via empirical risk minimization.

$$\hat{f} = \arg\min_{f \in \mathcal{H}} \mathcal{R}_{emp}(f) = \arg\min_{f \in \mathcal{H}} \sum_{i=1}^{n} L\left(y^{(i)}, f\left(\mathbf{x}^{(i)}\right)\right).$$



## Examples:

- Logistic regression (discriminant, linear)
- Neural networks
- Support vector machines