

# Ensemble Methods

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# Bagging

## Algorithm

- ▶ Input: Data  $D$ , bags  $K$ , base learner  $\lambda$
- ▶ For  $k = 1, \dots, K$ 
  - Sample **with replacement**  $D_k \sim \text{Unif}(D)$  – Obtain predictor  $\pi_k = \lambda(D_k)$ .
- ▶ Return  $\{\pi_k\}$

## The bagged predictor

$$\pi = f\left(\sum_k \pi_k\right)$$

# Bagging classifiers

## Classification setting

- ▶ Weak learner  $\lambda : D \rightarrow \Pi$
- ▶ Base hypotheses  $\pi_k : X \rightarrow \{-1, 1\}$

with

$$\pi_k = \lambda(D_k), \quad D_k \sim D$$

- ▶ Aggregate hypothesis

$$\pi(x) = \operatorname{sgn} \left( \sum_{k=1}^K \pi_k(x) \right)$$

## PAC property

For any  $\delta \in (0, 1)$ , and any  $\pi^* : X \rightarrow \{-1, 1\}$  and a hypothesis class  $\Pi$  with VC dimension  $d$ , for  $T$  data points, and  $K \in [0.02 T, T]$  bootstrap samples, then

$$\mathbb{L} \in O \left( \frac{1}{T} [d + \ln(1/\delta)] \right), \quad \text{w.p. } 1 - \delta.$$

# Sub-sample-and-aggregate

## Algorithm

- ▶ Input: Data  $D$ , number of experts  $K$ , base learner  $\lambda$
- ▶ For  $k = 1, \dots, K$ 
  - Sample **without replacement**  $D_k \sim \text{Unif}(D)$  – Obtain predictor  $\pi_k = \lambda(D_k)$ .
- ▶ Return  $\{\pi_k\}$

## The aggregated predictor

$$\pi = f(\pi_1, \dots, \pi_k)$$