

## Bootstrapping

Given data  $((x_i, y_i))_{i=1}^n$  and we want to estimate  $S(z)$

Construct the empirical distribution function  $\hat{F}$ .

Idea: For  $i=1, \dots, n$

- Resample  $z_i^* \sim \hat{F}$
- Calculate  $S_i \equiv S(z_i^*)$

Return distribution  $\{S_i\}_{i=1}^B$

$$\hat{Err}_{\text{boot}} = \frac{1}{N} \sum_{i=1}^N \frac{1}{B} \sum_{b=1}^B L(y_i, \hat{f}_b^*(x_i))$$

## Bagging - Bootstrap Aggregation

Define an estimator

$$\hat{f}_{\text{AVE}}(x) = \sum_{i=1}^B \underbrace{w_i}_{\text{weights}} \underbrace{\hat{f}_i^*(x)}_{\text{Bootstrapped estimate}}$$

General Aggregation models

Suppose we have models  $\{m_i\}_{i=1}^M$  that we wish to

combine to reduce uncertainty.

Assume that  $m_i$  is parameterized by  $s_i$

$$P(S|z) = \frac{P(S, z)}{P(z)} = \sum_{i=1}^m \frac{P(S, z, m_i)}{P(z)} = \sum_{i=1}^m \frac{P(S, z, m_i)}{P(z, m_i)} \frac{P(z, m_i)}{P(z)}$$

$$= \sum_{i=1}^m \underbrace{P(S|z, m_i)}_{\text{model param posterior}} \underbrace{P(m_i|z)}_{\text{Post. Prob of } m}$$

$$\mathbb{E}[S|z] = \sum_{i=1}^m \mathbb{E}[S|z, m_i] P(m_i|z)$$