

Our algorithm outputs a stochastic classifier,  $h_{\mathcal{A}}$  that is a convex combination of other hypotheses in  $\mathcal{H}$ . We denote the class of functions which are convex combinations of functions in  $\mathcal{H}$  as  $\text{conv}(\mathcal{H})$ .

**Lemma 1.** *The Rademacher complexity defined over a convex combination of distributions is linear with respect to the sample distribution.*

*Proof.*

$$\begin{aligned}
\mathcal{R}_{\mathcal{D}}(\mathcal{F}) &= \mathbb{E}_{S \sim \mathcal{D}} \left[ \hat{\mathcal{R}}_S(\mathcal{F}) \right] \\
&= \sum_S \mathbb{P}_{\mathcal{D}}[S] \left( \hat{\mathcal{R}}_S(\mathcal{F}) \right) \\
&= \sum_S (\alpha \mathbb{P}_{\mathcal{D}_A}[S] + (1 - \alpha) \mathbb{P}_{\mathcal{D}_B}[S]) \left( \hat{\mathcal{R}}_S(\mathcal{F}) \right) \\
&= \sum_S \alpha \mathbb{P}_{\mathcal{D}_A}[S] \left( \hat{\mathcal{R}}_S(\mathcal{F}) \right) + (1 - \alpha) \sum_S \mathbb{P}_{\mathcal{D}_B}[S] \left( \hat{\mathcal{R}}_S(\mathcal{F}) \right) \\
&= \sum_S \alpha \mathbb{P}_{\mathcal{D}_A}[S] \left( \hat{\mathcal{R}}_S(\mathcal{F}) \right) + \sum_S (1 - \alpha) \mathbb{P}_{\mathcal{D}_B}[S] \left( \hat{\mathcal{R}}_S(\mathcal{F}) \right) \\
&= \alpha \sum_S \mathbb{P}_{\mathcal{D}_A}[S] \left( \hat{\mathcal{R}}_S(\mathcal{F}) \right) + (1 - \alpha) \sum_S \mathbb{P}_{\mathcal{D}_B}[S] \left( \hat{\mathcal{R}}_S(\mathcal{F}) \right) \\
&= \alpha \mathcal{R}_{\mathcal{D}_A}(\mathcal{F}) + (1 - \alpha) \mathcal{R}_{\mathcal{D}_B}(\mathcal{F}) \quad \square
\end{aligned}$$

**Lemma 2.** *Let  $\mathcal{F}' = \text{conv}(\mathcal{F})$ . Then  $\mathcal{R}_{\mathcal{D}}(\mathcal{F}') = \mathcal{R}_{\mathcal{D}}(\mathcal{F})$ .*

*Proof.* Find citation. □