#### Constrained vs. Penalized ERM

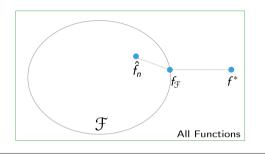
David S. Rosenberg

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October 5, 2017

#### Regularization Paths in Function Space

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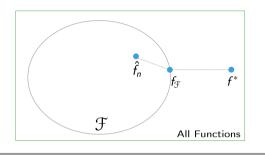


$$f^* = \underset{f}{\arg\min} \mathbb{E}\ell(f(X), Y)$$

$$f_{\mathcal{F}} = \underset{f \in \mathcal{F}}{\arg\min} \mathbb{E}\ell(f(X), Y))$$

$$\hat{f}_n = \underset{f \in \mathcal{F}}{\arg\min} \frac{1}{n} \sum_{i=1}^n \ell(f(x_i), y_i)$$

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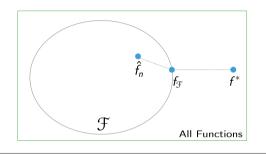


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• Approximation Error (of  $\mathfrak{F}$ ) =  $R(f_{\mathfrak{F}}) - R(f^*)$ 



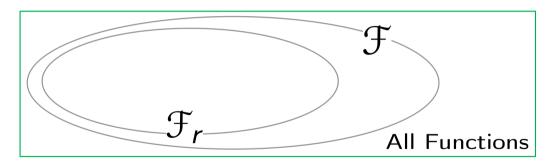
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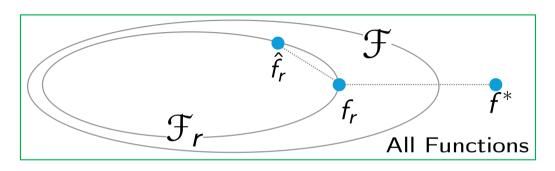
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- Approximation Error (of  $\mathfrak{F}$ ) =  $R(f_{\mathfrak{F}}) R(f^*)$
- Estimation error (of  $\hat{f}_n$  in  $\mathcal{F}$ ) =  $R(\hat{f}_n) R(f_{\mathcal{F}})$

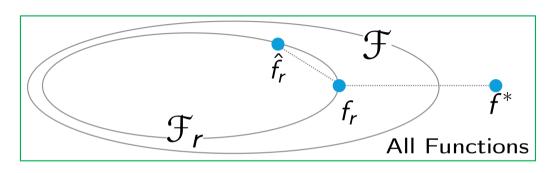
• Introduce complexity-constrained hypothesis space:  $\mathcal{F}_r = \{f \in \mathcal{F} \mid \Omega(f) \leqslant r\}$ 





• Revised notation:

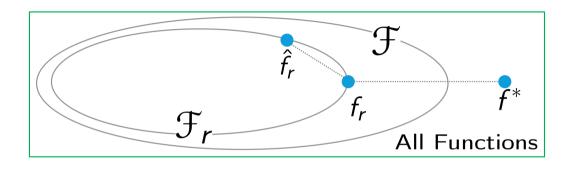
$$\hat{f}_r = \arg\min_{f \in \mathcal{F}_r} \frac{1}{n} \sum_{i=1}^n \ell(f(x_i), y_i) \qquad f_r = \arg\min_{f \in \mathcal{F}_r} \mathbb{E}\ell(f(X), Y)) \qquad f^* = \arg\min_{f} \mathbb{E}\ell(f(X), Y)$$



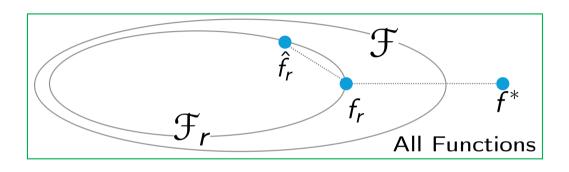
Revised notation:

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- Typically,  $\hat{f}_r$  will have  $\Omega(\hat{f}_r) = r$ , since with more complexity we can usually fit the data better.

• Consider complexity constraints r = .001, .01, 1.0, 10, 1000, corresponding to nested spaces:

$$\mathfrak{F}_{0.001} \subset \mathfrak{F}_{0.1} \subset \mathfrak{F}_{1.0} \subset \mathfrak{F}_{10} \subset \mathfrak{F}_{1000}$$

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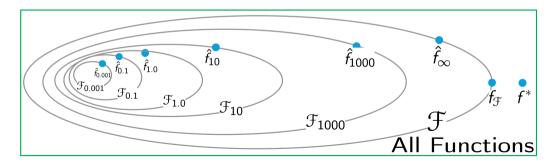
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• We get corresponding sequence of ERMs:  $\hat{f}_{0.001}$ ,  $\hat{f}_{0.1}$ ,  $\hat{f}_{1.0}$ ,  $\hat{f}_{10}$ ,  $\hat{f}_{1000}$ 

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- ullet Define the **regularization path** for constrained optimization in  ${\mathcal F}$  with complexity  $\Omega$  as

$$P_{\mathfrak{F},\Omega}^{\mathsf{constrained}} = \left\{\hat{f}_r \mid r \in [0,\infty]\right\},$$

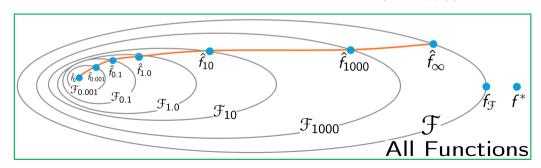
where  $\hat{f}_r$  is the constrained ERM in  $\mathcal{F}$  defined by  $\hat{f}_r = \arg\min_{\{f \in \mathcal{F} | \Omega(f) \leq r\}} \hat{R}(f)$ .

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#### Regularization Path for Penalized ERM

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- For lasso, ridge, and many more,  $P_{\mathcal{F},\Omega}^{\text{constrained}} = P_{\mathcal{F},\Omega}^{\text{penalized}}$ .
  - Precise statement in homework.