## Using $\mathcal{D}_{\mathrm{val}}$ more than once

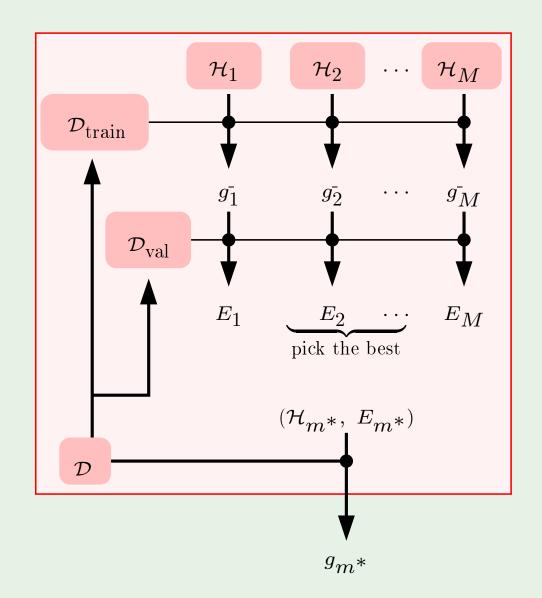
M models  $\mathcal{H}_1,\ldots,\mathcal{H}_M$ 

Use  $\mathcal{D}_{ ext{train}}$  to learn  $g_m^-$  for each model

Evaluate  $g_m^-$  using  $\mathcal{D}_{ ext{val}}$ :

$$E_m = E_{\mathrm{val}}(\underline{g}_m^-); \quad m = 1, \dots, M$$

Pick model  $m=m^*$  with smallest  $E_m$ 



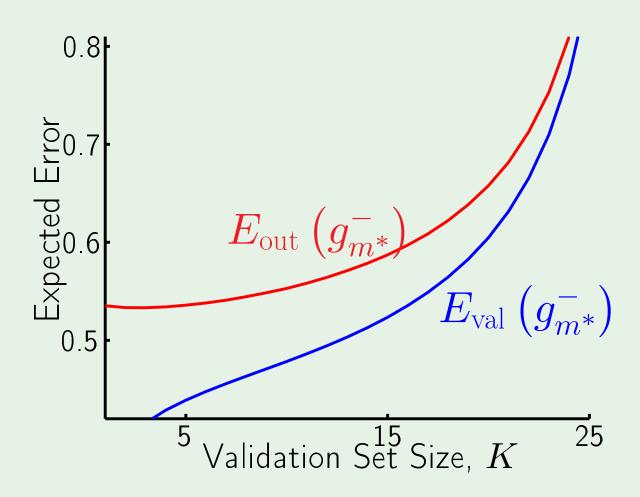
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## The bias

We selected the model  $\mathcal{H}_{m^*}$  using  $\mathcal{D}_{ ext{val}}$ 

 $E_{
m val}(g_{m^*}^-)$  is a biased estimate of  $E_{
m out}(g_{m^*}^-)$ 

Illustration: selecting between 2 models



## How much bias

For M models:  $\mathcal{H}_1,\ldots,\mathcal{H}_M$ 

 $\mathcal{D}_{\mathrm{val}}$  is used for "training" on the **finalists model**:

$$\mathcal{H}_{ ext{val}} = \; \{g_1^-, g_2^-, \dots, g_{ ext{M}}^- \}$$

Back to Hoeffding and VC!

$$E_{\mathrm{out}}(g_{m^*}^-) \leq E_{\mathrm{val}}(g_{m^*}^-) + O\left(\sqrt{\frac{\ln M}{K}}\right)$$

regularization  $\lambda$  early-stopping T

## Data contamination

Error estimates:  $E_{
m in},\,E_{
m test},\,E_{
m val}$ 

Contamination: Optimistic (deceptive) bias in estimating  $E_{
m out}$ 

Training set: totally contaminated

Validation set: slightly contaminated

Test set: totally 'clean'