

Machine Learning from Data

Lecture 5: Spring 2021

Today's Lecture

- Training Vs Testing
 - The Two Questions of Learning
 - Theory of Generalization ($E_{in} \approx E_{out}$)
 - An Effective Number of Hypotheses

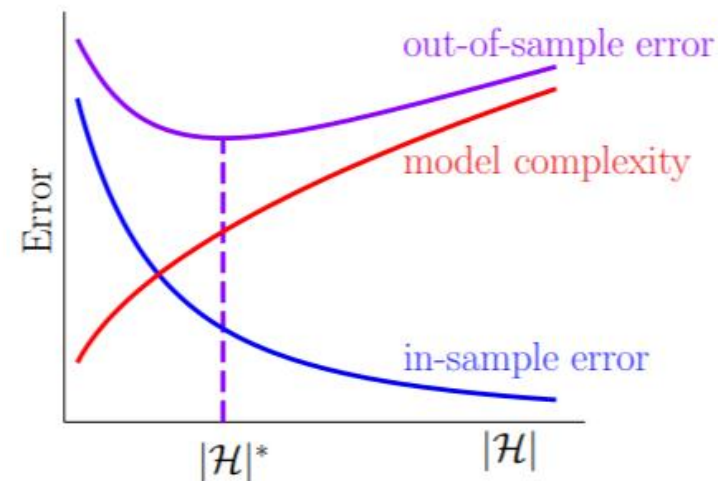
1. Can we make sure that $E_{\text{out}}(g)$ is close enough to $E_{\text{in}}(g)$?
2. Can we make $E_{\text{in}}(g)$ small enough?

The Hoeffding *generalization bound*:

$$E_{\text{out}}(g) \leq E_{\text{in}}(g) + \underbrace{\sqrt{\frac{1}{2N} \ln \frac{2|\mathcal{H}|}{\delta}}}_{\text{generalization error bar}}$$

E_{in} : training (eg. the practice exam)

E_{out} : testing (eg. the real exam)

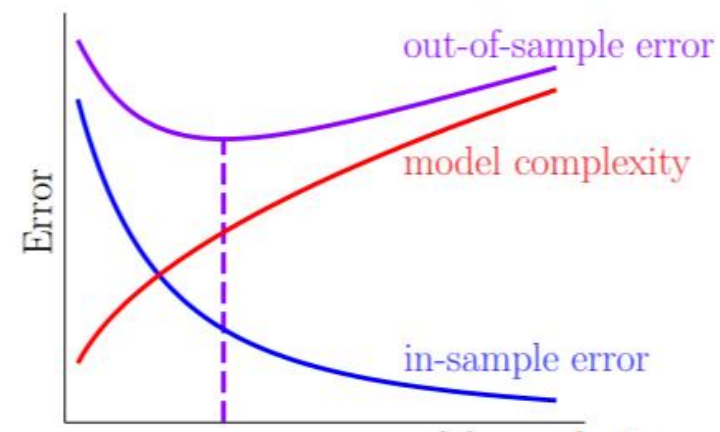
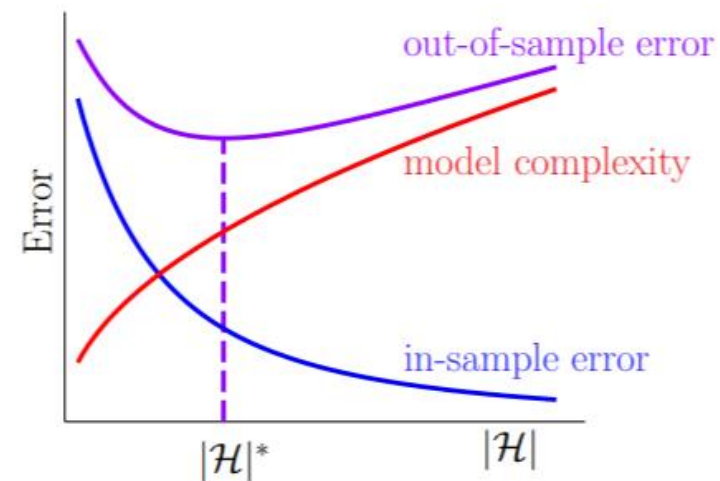


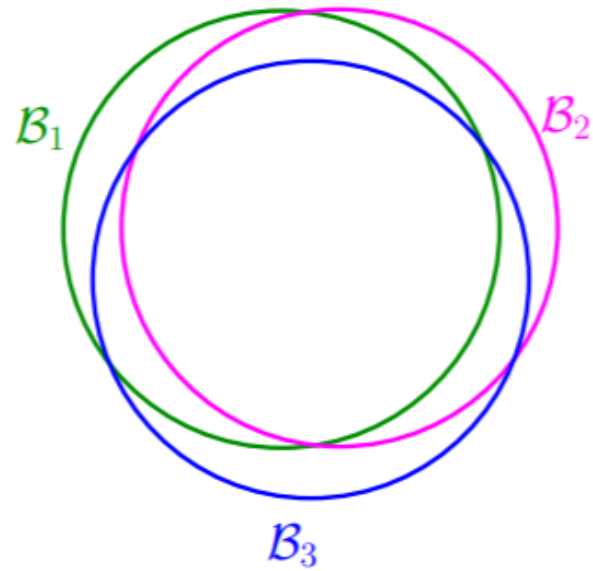
There is a tradeoff when picking $|\mathcal{H}|$.

$$E_{\text{out}}(g) \leq E_{\text{in}}(g) + \sqrt{\frac{1}{2N} \ln \frac{2|\mathcal{H}|}{\delta}}$$



$$E_{\text{out}}(g) \leq E_{\text{in}}(g) + \sqrt{\frac{8}{N} \ln \frac{4m_{\mathcal{H}}}{\delta}}$$





- \mathcal{B}_m are events (sets of outcomes); they can overlap.
- If the \mathcal{B}_m overlap, the union bound is loose.
- If many h_m are similar, the \mathcal{B}_m overlap.
- There are “effectively” fewer than $|\mathcal{H}|$ hypotheses,.
- We can replace $|\mathcal{H}|$ by something smaller.

Meaning of cardinality of \mathcal{H} ?

How did \mathcal{H} come in?

Effective Number of Hypothesis

- We need a way to measure the diversity of H .
- A simple idea:
 - Fix any set of N data points.
 - If H is diverse it should be able to implement all functions . . . on these N points.

The Growth Function

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Another Example

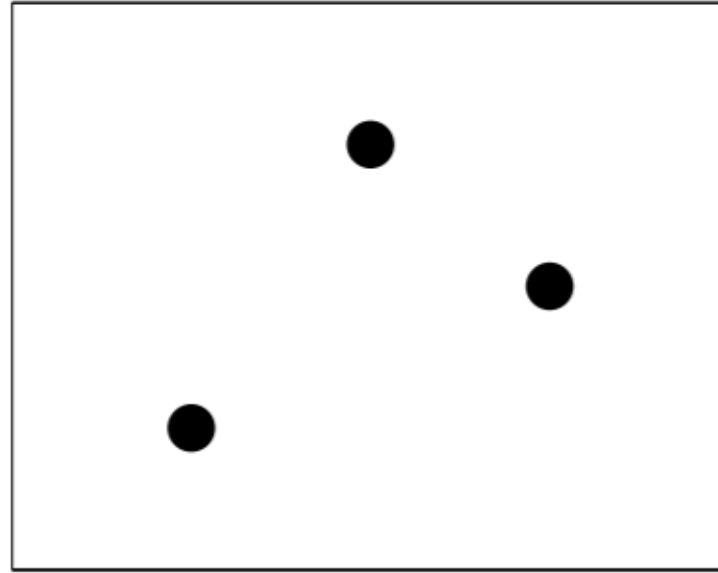
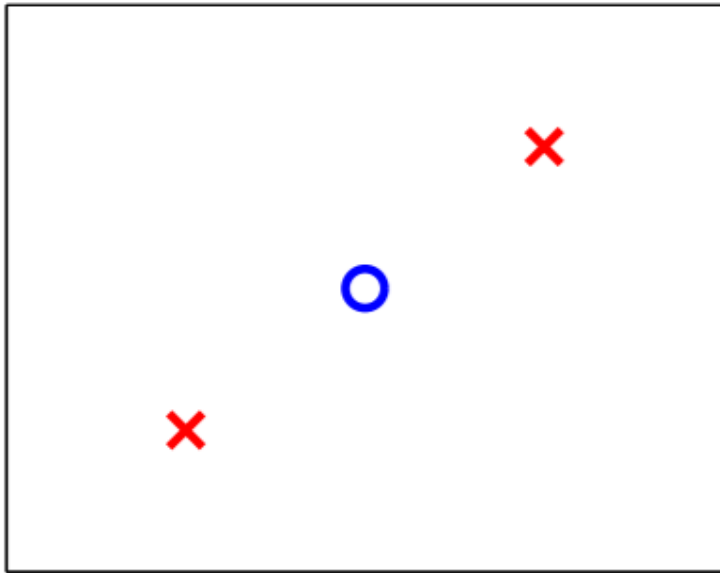
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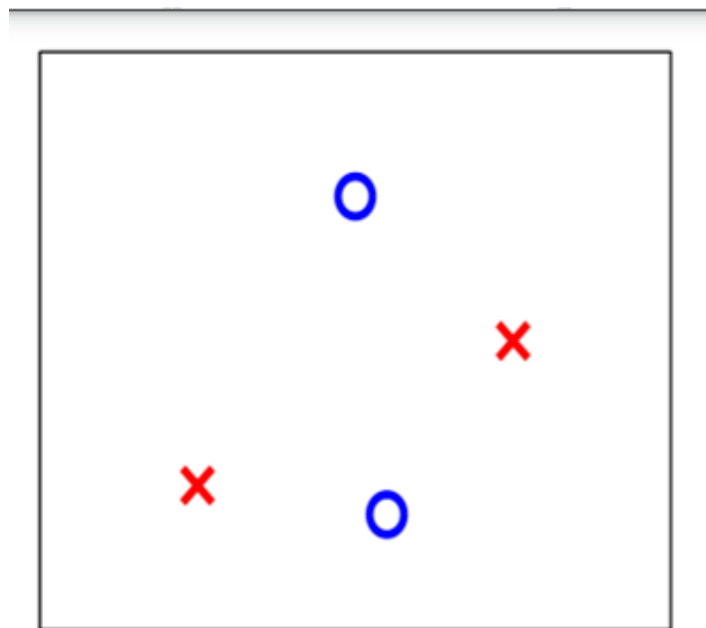
Dichotomies in 2-D Perceptron

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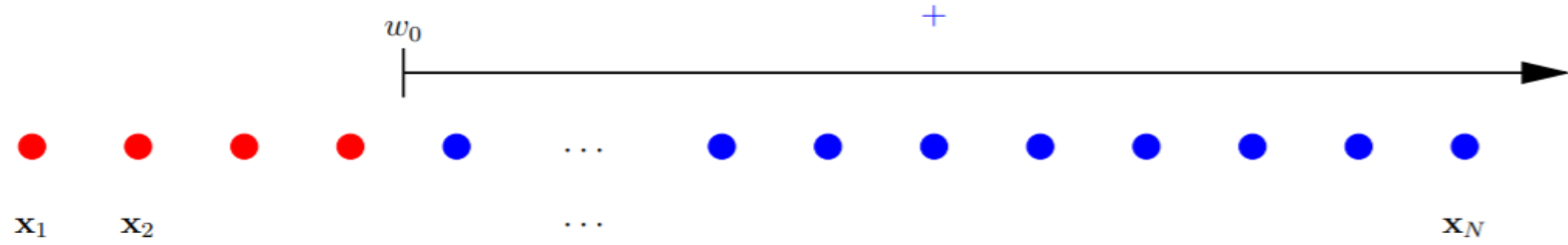
Example - Growth Functions

2-D Perceptron Model





1-D Positive Ray



Positive Rectangles in 2-D

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Summarize













	1	2	3	N 4	5	...
2-D perceptron	2	4	8	14	...	
1-D pos. ray	2	3	4	5	...	
2-D pos. rectangles	2	4	8	16	$< 2^5$...

- $m_{\mathcal{H}}(N)$ drops below 2^N – there is hope for the generalization bound.
- A **break point** is any n for which $m_{\mathcal{H}}(n) < 2^n$.













Definition: Shatter a Data Set

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Combinatorial Puzzle

\mathbf{X}_1	\mathbf{X}_2	\mathbf{X}_3
		
		
		
		

2 points are
shattered

\mathbf{x}_1	\mathbf{x}_2	\mathbf{x}_3
		
		
		
		

No Pair of Points
is Shattered

\mathbf{x}_1	\mathbf{x}_2	\mathbf{x}_3
		
		
		
		

If $N = 4$ how many possible dichotomys with no 2 points shattered?

\mathbf{X}_1	\mathbf{X}_2	\mathbf{X}_3
○	○	○
○	○	●
○	●	○
●	○	○

\mathbf{X}_1	\mathbf{X}_2	\mathbf{X}_3	\mathbf{X}_4
○	○	○	○
○	○	○	●
⋮			

Thanks!