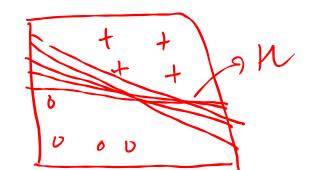
Machine Learning from Data

Lecture 6: Spring 2021

Today's Lecture

- Bounding the Growth Function $\gamma \gamma_{\mu}(\gamma)$
- Models are either Good or Bad
- The VC Bound

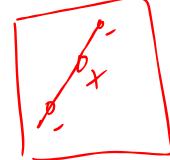
Putting Everything Together



• The growth function:

The growth function $m_{\mathcal{H}}(N)$ considers the worst possible $\mathbf{x}_1, \dots, \mathbf{x}_N$.

 $m_{\mathcal{H}}(N) = \max_{\mathbf{x}_1, \dots, \mathbf{x}_N} |\mathcal{H}(\mathbf{x}_1, \dots, \mathbf{x}_N)|.$



1) Can we bound $m_h(N)$ by a polynomial in N? 2) Can we replace |h| by $m_h(N)$ in the generalization bound? Pump han: N=4 (16) -> 14 1-D positive vay: N=2(4) -> 3 2-D rectangle: N=5 mn(5) <2 -> my(N) drops below 2^N
-> A break-point is any k for which mylk) 42^k

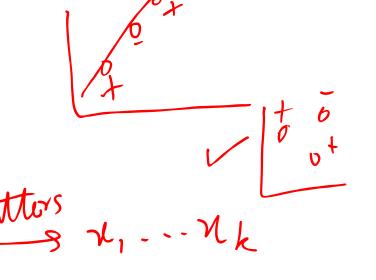
I give you a set of k^* points $\mathbf{x}_1, \dots, \mathbf{x}_{k^*}$ on which \mathcal{H} implements $< 2^{k^*}$ dichotomys.

- (a) k^* is a break point.
- (b) k^* is not a break point.
- (c) all break points are $> k^*$.
- (d) all break points are $\leq k^*$.
- (e) we don't know anything about break points.

For every set of k^* points $\mathbf{x}_1, \dots, \mathbf{x}_{k^*}$, \mathcal{H} implements $< 2^{k^*}$ dichotomys.

- $\sqrt{(a)} k^*$ is a break point. \longrightarrow definition
 - (b) k^* is not a break point.
- \checkmark (c) all $k \ge k^*$ are break points.
 - (d) all $k < k^*$ are break points.
 - (e) we don't know anything about break points.

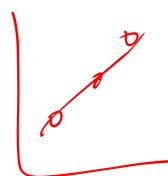
To show that k is *not* a break point for \mathcal{H} :



- \checkmark (a) Show a set of k points $\mathbf{x}_1, \dots \mathbf{x}_k$ which \mathcal{H} can shatter.
- \rightarrow (b) Show \mathcal{H} can shatter any set of k points.
 - (c) Show a set of k points $\mathbf{x}_1, \dots \mathbf{x}_k$ which \mathcal{H} cannot shatter.
 - (d) Show \mathcal{H} cannot shatter any set of k points.

$$\sqrt{\text{(e)}}$$
 Show $m_{\mathcal{H}}(k) = 2^k$. (similar to (9))

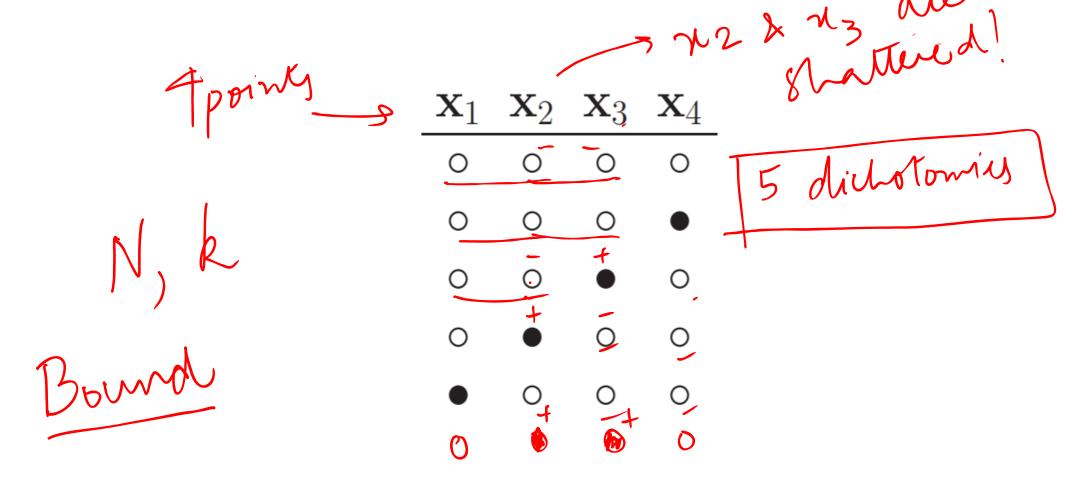
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- (d) Show \mathcal{H} cannot shatter any set of k points.
- (e) Show $m_{\mathcal{H}}(k) > 2^k$.

Back to the puzzle

How many dichotomies can you list on 4 points so that no 2 is shattered.



The combinatorial relationship

B(N, K): The manimum no of data points subset of size k is shattend.

B(N,K)

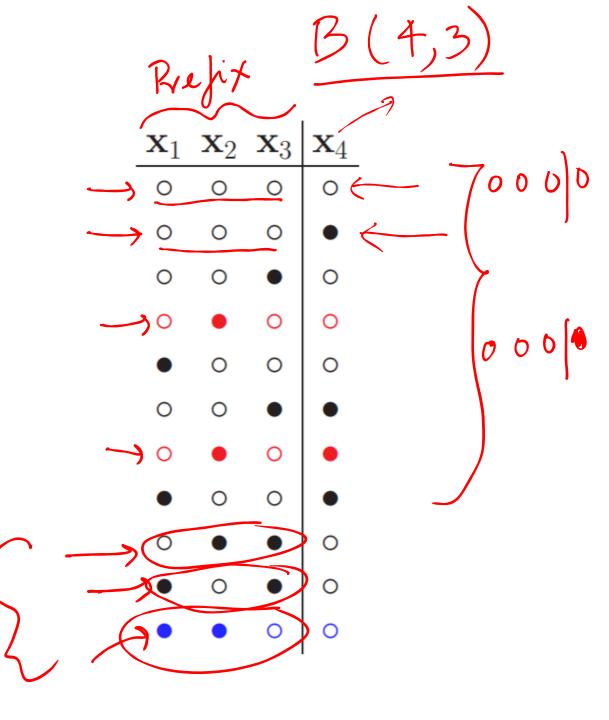
How many dichotomies can you list on 4 points so that no subset of 3 is shattered.

\mathbf{x}_1	\mathbf{x}_2	\mathbf{x}_3	\mathbf{x}_4	_	
0	0	0	0	7	
0	0	0	•		
0	0	•	0		
0	•	0	0		
•	0	0	0		11
0	0	•	•		1 1
0	•	0	•		
•	0	0	•		
0	•	•	0		
•	0	•	0		
•	•	0	0		

$$B(N,K)$$

 $B(4,3)=[]$

Based on prefix's we will segregate the dichotomics.



 α : prefix appears once \mathbf{V}

 β : prefix appears twice

$$B(4,3) = \alpha + 2\beta = 1$$

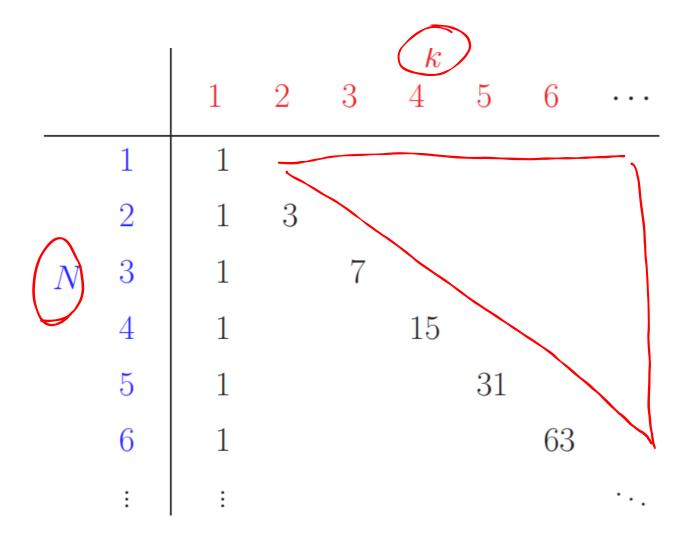
$$B(4,3) = \alpha + 2\beta = 1$$

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 $\frac{|B(4,3)|}{|A+B| \leq B(3,3)}$

Suppose a pair is Shattere d. $\mathbf{x}_1 \ \mathbf{x}_2 \ \mathbf{x}_3 \ \mathbf{x}_4$

Fill the table values



$$B(N,1) = 1$$

$$B(N, N) = 2^N - 1$$

$$B(N, k) \le B(N - 1, k) + B(N - 1, k - 1)$$

$$B(N, k) \le B(N - 1, k) + B(N - 1, k - 1)$$

Analytic Bound

2 pax cod Proof: (Induction on N.) 1. Verify for N = 1: B(1,1)2. Suppose $B(N,k) \leq \sum_{i=1}^{N-1} {N \choose i}$. Lemma. 70 p

i) $\frac{1}{k-1}$ \frac B(N, K)

< Polynomial in N 2-D perception 1-D Ray (N+N 2-D Restergle dichotomies that can be mplenented, max. no. of implemented, the break points on Npoints by a M.

Subset of size k

K -> break point # Any subset of
k points is not
shattered by these
shattered by these
my(N) dichotomics $B(N,k) \rightarrow max.$ $M_{H}(N) \leq B(N,k)$

Theorem: Let H be any hypotheris set:

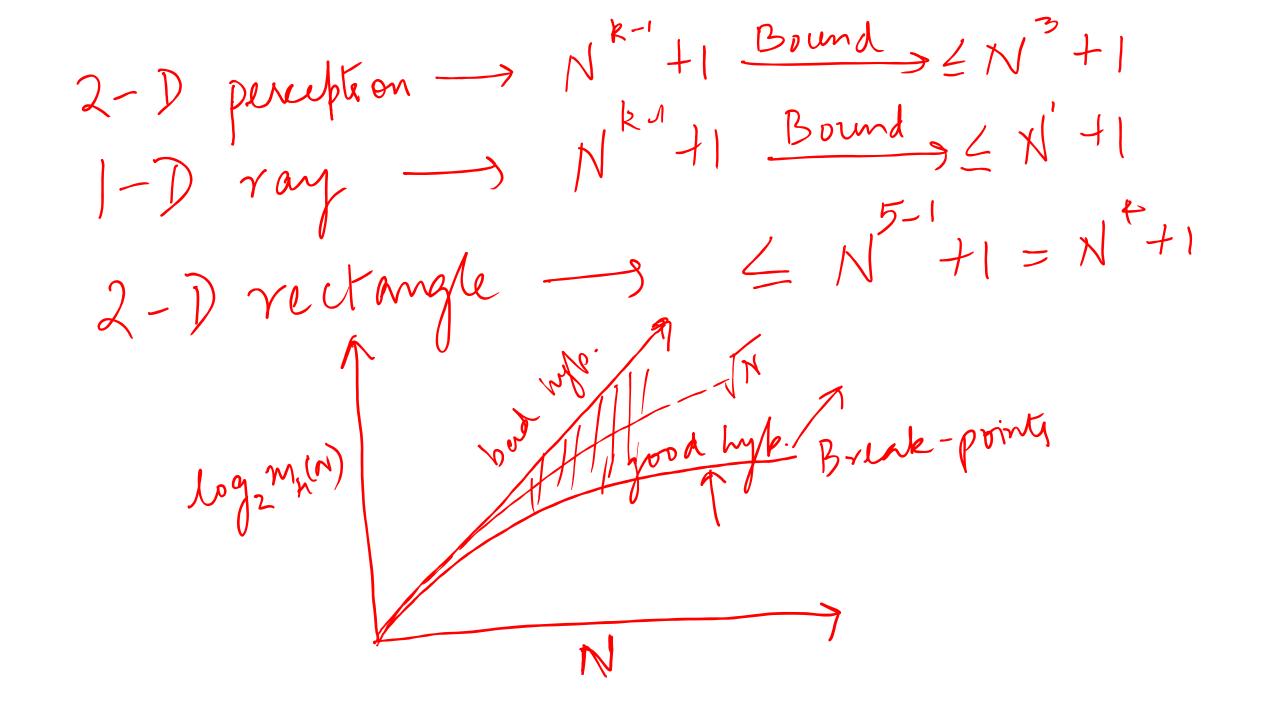
i) M is entremely complen (Never breaks)

i) M in my (N) = 2 N + N (Not very useful) ii) H does get broken. In this case it has

a breakpoint (k). Then:

my(N) & B(N,k) & N +1

my(N)



Error bar In Hill Market Eaut LEin + 0/ In (N) —> O E out = E in (i)

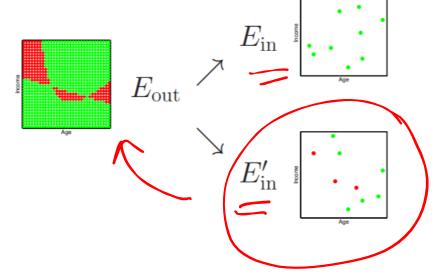
 \checkmark Can we get a polynomial bound on $m_{\mathcal{H}}(N)$ even for infinite \mathcal{H} ?

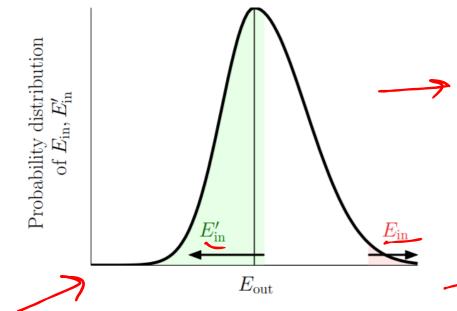
Can we replace $|\mathcal{H}|$ with $m_{\mathcal{H}}(N)$ in the generalization bound?

Let \mathcal{L} but \mathcal{L} the form \mathcal{L} but \mathcal{L}

Test

The ghost data set: a 'fictitious' data set \mathcal{D}' :





 $E'_{\rm in}$ is like a test error on N new points.

 $E_{\rm in}$ deviates from $E_{\rm out}$ implies $E_{\rm in}$ deviates from $E'_{\rm in}$.

 $E_{\rm in}$ and $E'_{\rm in}$ have the same distribution.

$$\mathbb{P}[(E'_{\text{in}}(g), E_{\text{in}}(g)) \text{ "deviate"}] \ge \frac{1}{2} \mathbb{P}[(E_{\text{out}}(g), E_{\text{in}}(g)) \text{ "deviate"}]$$



	\mathbf{x}_1	\mathbf{x}_2	\mathbf{x}_3	 \mathbf{x}_N	$ \mathbf{x}_{N+1} $	\mathbf{x}_{N+2}	\mathbf{x}_{N+3}	 \mathbf{x}_{2N}	
	0	0	•	 0	•	•	0	 0	
•	_				_				

Number of dichotomys is at most $m_{\mathcal{H}}(2N)$.

Abus not affect the polynomial matrix of the bound)

Up to technical details, analyze a "hypothesis set" of size at most $m_{\mathcal{H}}(2N)$.

The Vapnik-Chervonenkis Bound (VC Bound)

$$\mathbb{P}\left[|E_{\mathrm{in}}(oldsymbol{g}) - E_{\mathrm{out}}(oldsymbol{g})| > \epsilon
ight] \leq 4m_{\mathcal{H}}(2N)e^{-\epsilon^2N/8},$$

for any $\epsilon > 0$.

$$\mathbb{P}\left[|oldsymbol{E}_{ ext{in}}(oldsymbol{g}) - oldsymbol{E}_{ ext{out}}(oldsymbol{g})| \leq \epsilon
ight] \geq 1 - 4m_{\mathcal{H}}(2oldsymbol{N})e^{-\epsilon^2N/8},$$

for any $\epsilon > 0$.

$$E_{ ext{out}}(g) \leq E_{ ext{in}}(g) + \sqrt{rac{8}{N}} \log rac{4m_{\mathcal{H}}(2N)}{\delta},$$

w.p. at least $1 - \delta$.

Thanks!