

# ECE595 / STAT598: Machine Learning I

## Lecture 26.2: Growth Function - Examples of $mH(N)$

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# Outline

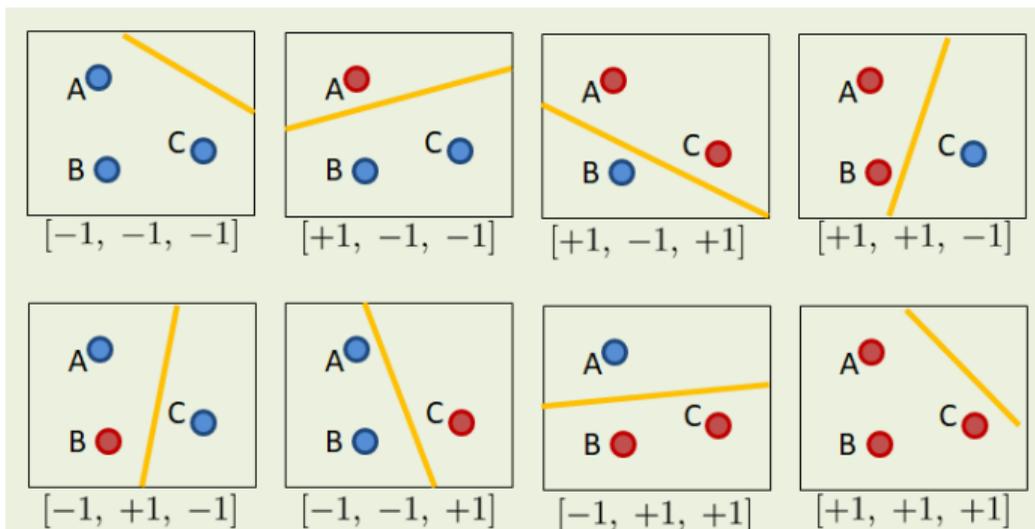
- Lecture 25 Generalization
- **Lecture 26 Growth Function**
- Lecture 27 VC Dimension

## Today's Lecture:

- Overcoming the  $M$  Factor
  - Decisions based on Training Samples
  - Dichotomy
- **Examples of  $m_{\mathcal{H}}(N)$** 
  - **Finite 2D Set**
  - **Positive ray**
  - **Interval**
  - **Convex set**

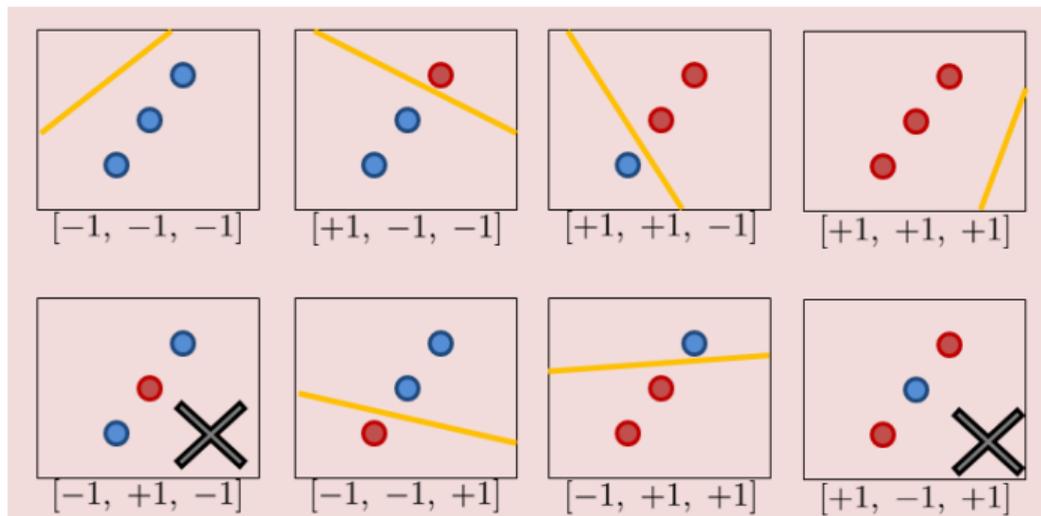
## Examples of $m_{\mathcal{H}}(N)$

- $\mathcal{H}$  = linear models in 2D
- $N = 3$
- How many dichotomies can I generate by moving the three points?
- This gives you 8. Are we the best?

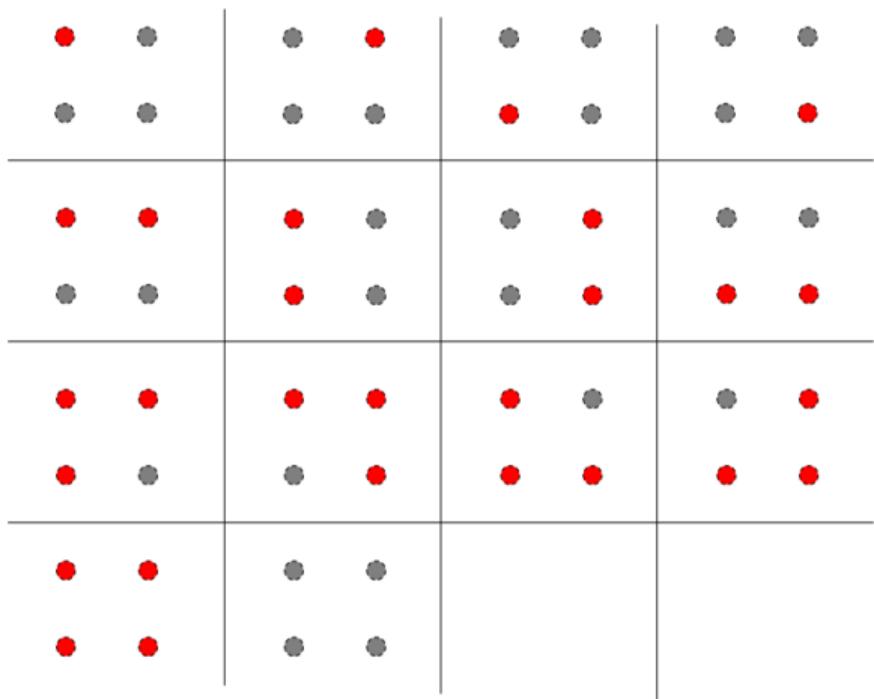


## Examples of $m_{\mathcal{H}}(N)$

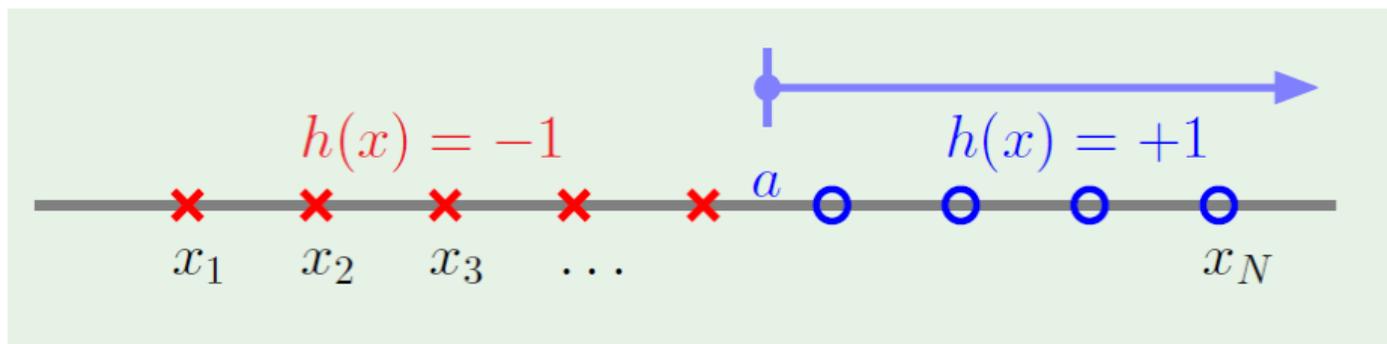
- $\mathcal{H}$  = linear models in 2D
- $N = 3$
- How many dichotomies can I generate by moving the three points?
- This gives you 6. The previous is the best. So  $m_{\mathcal{H}}(3) = 8$ .



What about  $m_{\mathcal{H}}(4)$ ? Ans: 14.

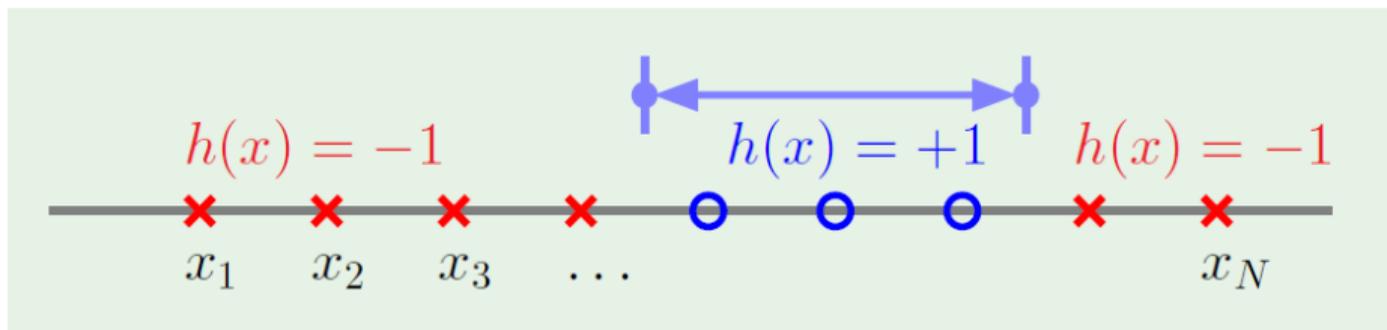


## Another Example



- $\mathcal{H}$  = set of  $h: \mathbb{R} \rightarrow \{+1, -1\}$
- $h(x) = \text{sign}(x - a)$
- Cut the line into two halves
- You can only move along the line
- $m_{\mathcal{H}}(N) = N + 1$
- The  $N$  comes from the  $N$  points
- The  $+1$  comes from the two ends

## Another Example



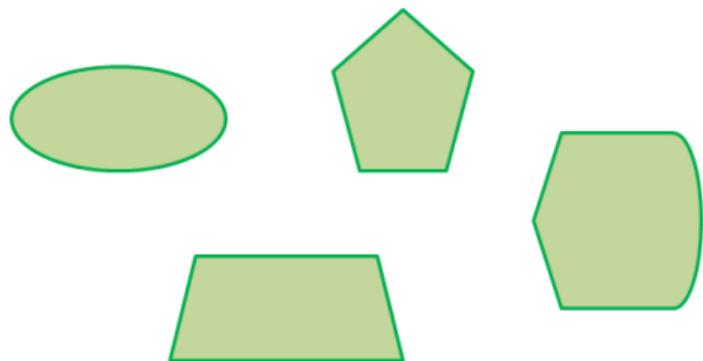
- $\mathcal{H} = \text{set of } h: \mathbb{R} \rightarrow \{+1, -1\}$
- Put an interval
- Length of the interval is  $N$  points

$$m_{\mathcal{H}}(N) = \binom{N+1}{2} + 1 = \frac{N^2}{2} + \frac{N}{2} + 1$$

- Think of  $N + 1$  balls, pick 2.

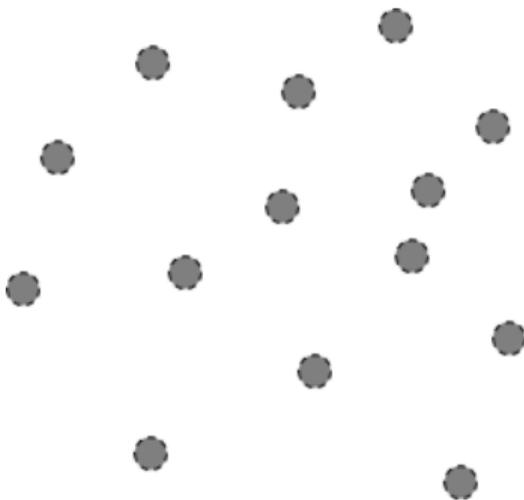
## Another Example

- $\mathcal{H} =$  set of  $h: \mathbb{R}^2 \rightarrow \{+1, -1\}$
- $h(\mathbf{x}) = +1$  is convex
- Here are some examples

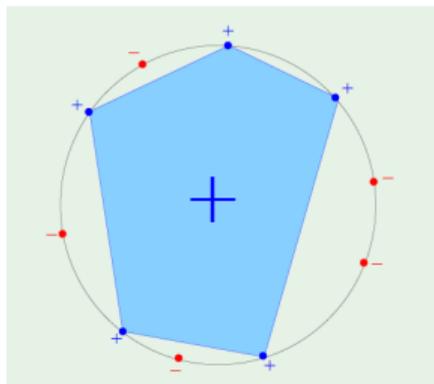


## Another Example

- How about this collection of data points?
- Can you find an  $h$  such that you get a convex set?
- Yes. Do convex hull.
- Does it give you the maximum number of dichotomies?
- No. All interior points do not contribute.



## Another Example



- The best you can do is this.
- Put all the points on a circle.
- Then you can get at most  $2^N$  different dichotomies
- So

$$m_{\mathcal{H}}(N) = 2^N$$

- That is the best you can ever get with  $N$  points

## Summary of the Examples

- $\mathcal{H}$  is positive ray:

$$m_{\mathcal{H}}(N) = N + 1$$

- $\mathcal{H}$  is positive interval:

$$m_{\mathcal{H}}(N) = \binom{N+1}{2} + 1 = \frac{N^2}{2} + \frac{N}{2} + 1$$

- $\mathcal{H}$  is convex set:

$$m_{\mathcal{H}}(N) = 2^N$$

- So if we can replace  $M$  by  $m_{\mathcal{H}}(N)$
- And if  $m_{\mathcal{H}}(N)$  is a polynomial
- Then we are good.

## Reading List

- Yasar Abu-Mostafa, Learning from Data, chapter 2.1
- Mehrya Mohri, Foundations of Machine Learning, Chapter 3.2
- Stanford Note <http://cs229.stanford.edu/notes/cs229-notes4.pdf>