Download the FULL Version with Token NOW!

CSCI4230 Computatioal Learning Theory

Lecturer: Siu On Chan

Spring 2019

Based on Rocco Servedio's notes

Notes 5: VC dimension

1. Vapnik-Chervonenkis dimension

Related to mistake lower bounds in Online Learning

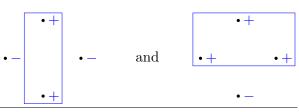
Usually an integer, telling us how expressive a concept class $\mathcal C$ is

Given concept class \mathcal{C} over instance space X, subset $S \subseteq X$ is **shattered by** \mathcal{C} if all "dichotomies" of S can be induced by \mathcal{C} , i.e.:

$$\forall T \subseteq S, \exists c \in C \text{ s.t. } c \cap S = T$$

 $X = \text{the plane} = \mathbb{R}^2$ $\mathcal{C} = \text{axis-aligned rectangles}$

S = ullet has all dichotomies such as



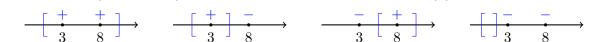
VCDim(C) is the size of the largest subset $S \subseteq X$ shattered by C VCDim(C) = d if and only if

- (1) some subset $S \subseteq X$ with |S| = d is shattered by C; and
- (2) all subsets of size d+1 is not shattered by C

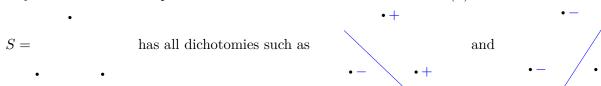
VCDim(C) can be ∞

Example: Closed intervals of the real line

 $X = \mathbb{R}$ $\mathcal{C} = \text{closed intervals} = \{[a, b] \mid a, b \in \mathbb{R}\} \text{ where } [a, b] = \{x \in \mathbb{R} \mid a \leqslant x \text{ and } x \leqslant b\}$ Every two points (e.g. 3 and 8) can be shattered \Longrightarrow VCDim(\mathcal{C}) $\geqslant 2$



Example: Halfspaces in the plane $X = \mathbb{R}^2$ $\mathcal{C} = LTF$ Any three non-collinear points can be shattered \Longrightarrow $VCDim(\mathcal{C}) \geqslant 3$

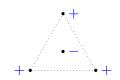


No four points can be shattered \implies VCDim(\mathcal{C}) \leq 3 Case 1: contains three collinear points



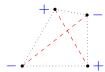
Case 2: No three points collinear

Case 2a: Some point inside the triangle formed by three other points



Case 2b: Four points form a convex quadrilateral $\quad\iff\quad$ the two diagonals cross

Download the FULL Version with Token NOW!



endpoints of two diagonals get different labels

Example: $X = \{0, 1\}^n$ $\mathcal{C} = \{\text{monotone conjunction}\}$ e.g. $c(x) = x_2 \land x_5 \land x_7 \in \mathcal{C}$ VCDim $(\mathcal{C}) \geqslant n$: $S = \{a_j = \text{vector with 0 at position } j \text{ and 1 everywhere else } | 1 \leqslant j \leqslant n \}$

e.g.
$$n = 4$$
 $S = \begin{cases} 0111, \\ 1011, \\ 1101, \\ 1110 \end{cases}$, $T = \begin{cases} 0111, \\ 1101, \\ 1110 \end{cases}$ induced by $c(x) = x_2$

Every subset $T \subseteq S$ is induced by $c \in \mathcal{C}$ containing precisely variables x_j s.t. $a_j \notin T$

 $VCDim(\mathcal{C}) \leq n$: because $|\mathcal{C}| = 2^n$ and

Observation: $VCDim(\mathcal{C}) \geqslant d \text{ implies } |\mathcal{C}| \geqslant 2^d$

2. Online Mistake Lowerbounds from VC dimension

Claim 2.1. Any deterministic algorithm for learning C makes $\geqslant VCDim(C)$ mistakes on some sample sequence

Proof. $S = \{x^1, \dots, x^d\}$ be shattered set of size $d = \text{VCDim}(\mathcal{C})$

Instance sequence is x^1, \ldots, x^d

On sample x^i , algorithm predicts $b_i \in \{0, 1\}$

Can find $c \in \mathcal{C}$ s.t. $c(x^i) = \overline{b}_i$ for all $1 \le i \le n$ (opposite of all predictions)

Claim 2.2. Some fixed sample sequence causes every randomized algorithm for learning C to make $\geqslant VCDim(C)/2$ mistakes in expectation

Previous claim follows from the next claim (via Yao's minimax principle, not covered in this course)

Claim 2.3. Some distribution of random sample sequences causes every deterministic algorithm for learning C to make $\geq VCDim(C)/2$ mistakes in expectation

Proof. $S = \{x^1, \dots, x^d\}$ be shattered set of size $d = \text{VCDim}(\mathcal{C})$

Sample sequence is $(x^1, y^1), \ldots, (x^d, y^d)$, where y^1, \ldots, y^d are uniformly random bits

Any algorithm predicting d uniformly random bits makes d/2 mistakes in expectation

For every choice of random bits y^1, \ldots, y^d , some $c \in \mathcal{C}$ correctly labels all instances x^1, \ldots, x^d

This is the bottom of preview version. Please download the full version with token.