## $\varepsilon$ -Nets and VC Dimension

- Sampling is a powerful idea applied widely in many disciplines, including CS.
- There are at least two important uses of sampling: estimation and detection.
- CNN, Nielsen, NYT etc use polling to estimate the size of a particular group in the larger population.
- By sampling a small segment of the population, one can predict the winner of a presidential election (with high confidence). How many prefer Bush to Gore; how many will use a new service etc.
- In detection, the goal is to sample so that any group with large probability measure will be caught with high confidence.
- Random traffic checks, for example. Frequent speeders (drinkers) are likely to get caught.

# Sampling

- A network monitoring application.
- Want to detect flows that are suspiciously big, in terms of fraction of total packets.
- Set a threshold of  $\theta\%$ . Any flow that accounts for more than  $\theta\%$  of traffic at a router should be flagged.
- Keeping track of all flows is infeasible; millions of flows and billions of packets per second.
- By taking a number of samples that depends only on  $\theta$ , we can detect offending flows with high probability.
- Track only sampled flows.

# **Basic Sampling Theorem**



- U is a ground set (points, events, database objects, people etc.)
- Let  $R \subset U$  be a subset such that  $|R| \ge \varepsilon |U|$ , for some  $0 < \varepsilon < 1$ .
- Theorem: A random sample of  $(\frac{1}{\varepsilon} \ln \frac{1}{\delta})$ points from U intersects R with probability at least  $1 - \delta$ .
- Proof: A particular sample point is in R with prob ε, and not in R with prob. 1 - ε.
  Prob. that none of the sampled points is in R is

$$\leq (1-\varepsilon)^{\frac{1}{\varepsilon}\ln\frac{1}{\delta}} \leq e^{-\ln\frac{1}{\delta}} = \delta.$$

## **Universal Samples**

- Sample size is independent of |U|.
- Basic sampling theorem guarantees that for a given set *R*, a random sample set works.
- If we want to hit each of the sets  $R_1$ ,  $R_2$ , ...,  $R_m$ , then this idea is too limiting. It requires a separate sample for each  $R_i$ .
- Can we get a single universal sample set, which hit all the  $R_i$ 's?



•  $\varepsilon$ -Nets and VC dimension characterize when this is possible.

#### $\varepsilon$ -Nets

- Let  $(\mathcal{U}, \mathcal{R})$  be a finite set system, and let  $\varepsilon \in [0, 1]$  be a real number.
- A set  $N \subseteq \mathcal{U}$  is called an  $\varepsilon$ -net for  $(\mathcal{U}, \mathcal{R})$  if  $N \cap R \neq \emptyset$  for all  $R \in \mathcal{R}$  whenever  $|R| \ge \varepsilon |\mathcal{U}|$ .



 A more general form of ε-net can be defined using probability measures. Think of this as endowing points of U with weights.

#### **Shatter Function**

- A set system  $(\mathcal{U}, \mathcal{R})$ , where  $\mathcal{U}$  is the ground set and  $\mathcal{R}$  is a family of subsets.
- $\mathcal{R} = \{R_1, \ldots, R_m\}$ , with  $R_i \subset \mathcal{U}$ , are ranges that we want to hit.
- A subset  $X \subset \mathcal{U}$  is shattered by  $\mathcal{R}$  if all subsets of X can be obtaind by intersecting X with members of  $\mathcal{R}$ .
- That is, for any  $Y \subseteq X$ , there is some  $A \in \mathcal{R}$  such that  $Y = X \cap A$ .
- Examples:  $U = \text{points in the plane. } \mathcal{R} = \text{half-spaces.}$



# **VC** Dimension



- The shatter function measures the complexity of the set system.
- If instead of half-spaces, we used ellipses, then (ii) and (iii) can be shattered as well.
- So, the set system of ellipses has higher complexity than half-spaces.

VC Dimension: The VC dimension of a set system  $(\mathcal{U}, \mathcal{R})$  is the maximum size of any set  $X \subset \mathcal{U}$  shattered by  $\mathcal{R}$ .

• Thus, the half-spaces system has VC dimension 3.

## **Other Examples**

- Set system where  $\mathcal{U} = \text{points in } d\text{-space}$ , and  $\mathcal{R} = \text{half-spaces}$ , has VC-dimension d+1.
- A simplex is shattered, but no (d+2)-point set is shattered (by Radon's Lemma).
- Set system where  $\mathcal{U} = \text{points in the plane}$ , and  $\mathcal{R} = \text{circles}$ , has VC-dimesion 4.

# **Convex Set System**

- Consider  $(\mathcal{U}, \mathcal{R})$ , where  $\mathcal{U}$  is set of points in the plane, and  $\mathcal{R}$  is family of convex sets.
- Members of  $\mathcal{R}$  are subsets that can be obtained by intersecting  $\mathcal{U}$  with a convex polygon.



Set system of convex polygons

- Any subset  $X \subseteq \mathcal{U}$  can be obtained by intersecting  $\mathcal{U}$  with an appropriate convex polygon.
- Thus, entire set  $\mathcal{U}$  is shattered.
- VC dimension of this set system is  $\infty$ .

#### $\varepsilon$ -Net Theorem

- Suppose  $(\mathcal{U}, \mathcal{R})$  is a set system of VC dimension d, and let  $\varepsilon, \delta$  be real numbers, where  $\varepsilon \in [0, 1]$  and  $\delta > 0$ .
- If we draw

$$O\left(\frac{d}{\varepsilon}\log\frac{d}{\varepsilon} + \frac{1}{\varepsilon}\log\frac{1}{\delta}\right)$$

points at random from  $\mathcal{U}$ , then the resulting set N is an  $\varepsilon$ -net with probability  $\geq \delta$ .

- Size of  $\varepsilon$ -Net is independent of the size of  $\mathcal{U}$ .
- Example: Consider set system of points in the plane with half-space ranges. It has VC-dim = 3. Assuming ε, δ constant, we have an ε-net of O(1) size.

### Consequences

- We will not prove the ε-net theorem, but look at some applications, and prove a related result, bounding the size of the set system.
- Suppose the set system  $(\mathcal{U}, \mathcal{R})$ , where  $|\mathcal{U}| = n$ , has VC dimension d. How many sets can be in the family  $\mathcal{R}$ ?
- Naively, the best one can say is that  $|\mathcal{R}| \leq 2^n$ .
- We will show that

$$|\mathcal{R}| \leq \binom{n}{0} + \binom{n}{1} + \dots + \binom{n}{d} \leq n^d$$

- This is the best bound one can prove in general, but it's not necessarily the best for individual set systems.
- E.g., for points and half-spaces in the plane, this theorem gives  $n^3$ , while we can see that the real bound is  $n^2$ .

- Define  $g(d,n) = \binom{n}{0} + \binom{n}{1} + \dots + \binom{n}{d}$ .
- Proof by induction. Base case trivial: n = d = 0 and  $\mathcal{U} = \mathcal{R} = \emptyset$ .
- Choose an arbitrary point  $x \in U$ , and consider  $U' = U \{x\}$ .
- Let  $\mathcal{R}'$  be the projection of  $\mathcal{R}$  onto  $\mathcal{U}'$ . That is.  $\mathcal{R}' = \{A \cap \mathcal{U}' | A \in \mathcal{R}\}.$
- VC-dim of  $(\mathcal{U}', \mathcal{R}')$  is at most d—if  $\mathcal{R}'$  shatters a (d+1)-size set, so does  $\mathcal{R}$ .
- By induction,  $|\mathcal{R}'| \leq g(d, n-1)$ .



- What's the difference between  $\mathcal{R}$  and  $\mathcal{R}'$ ?
- Two sets  $A, A' \in \mathcal{R}$  map to same set in  $\mathcal{R}'$ only if  $A = A' \cup \{x\}$  and  $x \notin A'$ .
- Define a new set system  $(\mathcal{U}, \mathcal{R}'')$  where

 $\mathcal{R}'' = \{ A' | A' \in \mathcal{R}, \quad x \notin A', \quad A' \cup \{ x \} \in \mathcal{R} \}$ 

- $|\mathcal{R}| = |\mathcal{R}'| + |\mathcal{R}''|$ —sets in  $\mathcal{R}''$  are exactly those that are counted only once in  $\mathcal{R}'$ .
- Claim: VC-dim of  $\mathcal{R}''$  is  $\leq d-1$ .
- We show that whenever  $\mathcal{R}''$  shatters  $Y, \mathcal{R}$  shatters  $Y \cup \{x\}$ .



- Two cases: Consider  $A \subseteq Y \cup \{x\}$ .
  - **1.** If  $A \subseteq Y$ , then since Y is shattered,  $\exists S \in \mathcal{R}''$  so that  $S \cap Y = A$ .
  - **2.** Since  $x \notin S$ , but  $S \in \mathcal{R}$ , it follows that  $S \cap (Y \cup \{x\}) = A$ .
  - **3.** If  $x \in A$ , then  $\exists S \in \mathcal{R}''$  so that  $S \cap Y = A \{x\}.$
  - 4. By definition of  $\mathcal{R}''$ ,  $S \cup \{x\} \in \mathcal{R}$ , and so  $(S \cup \{x\}) \cap (Y \cup \{x\}) = A \cup \{x\} = A$ .
- Thus,  $Y \cup \{s\}$  is shattered.
- Thus, VC-dim of  $\mathcal{R}''$  is at most d-1, and by induction,  $|\mathcal{R}''| \leq g(d-1, n-1)$ .

• Since  $|\mathcal{R}| = |\mathcal{R}'| + |\mathcal{R}''|$ , we have

$$\begin{aligned} |\mathcal{R}| &\leq g(d, n-1) + g(d-1, n-1) \\ &= \sum_{i=0}^{d} \binom{n-1}{i} + \sum_{i=0}^{d-1} \binom{n-1}{i} \\ &= \binom{n-1}{0} + \sum_{i=1}^{d} \left( \binom{n-1}{i} + \binom{n-1}{i-1} \right) \\ &= \binom{n}{0} + \sum_{i=1}^{d} \binom{n}{i} \\ &= g(d, n) \end{aligned}$$

### $\varepsilon$ -Approximation

- Suppose  $(\mathcal{U}, \mathcal{R})$  is a set system of VC dimension d, and let  $\varepsilon, \delta$  be real numbers, where  $\varepsilon \in [0, 1]$  and  $\delta > 0$ .
- A set  $N \subseteq \mathcal{U}$  is called an  $\varepsilon$ -approximation for  $(\mathcal{U}, \mathcal{R})$  if for any  $A \in \mathcal{R}$ ,

$$\left|\frac{|N \cap A|}{|N|} - \frac{|A|}{|\mathcal{U}|}\right| \leq \varepsilon$$

• If we draw

$$O\left(\frac{d}{\varepsilon^2}\log\frac{d}{\varepsilon} + \frac{1}{\varepsilon^2}\log\frac{1}{\delta}\right)$$

points at random from  $\mathcal{U}$ , then the resulting set N is an  $\varepsilon$ -approximation with probability  $\geq \delta$ .

• An  $\varepsilon$ -approximation is also an  $\varepsilon$ -net, but not vice versa.