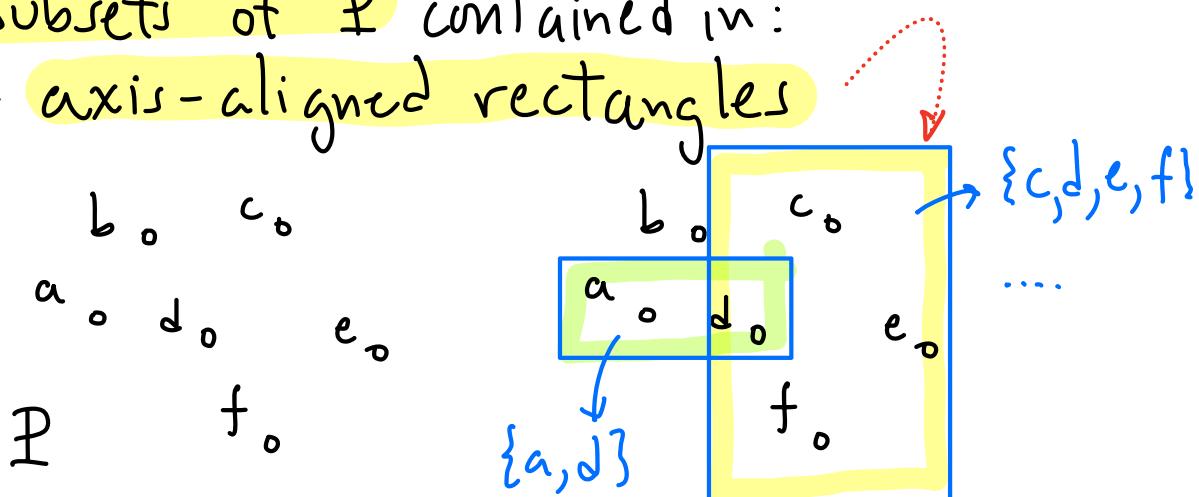


# CMSC 754 - Computational Geometry

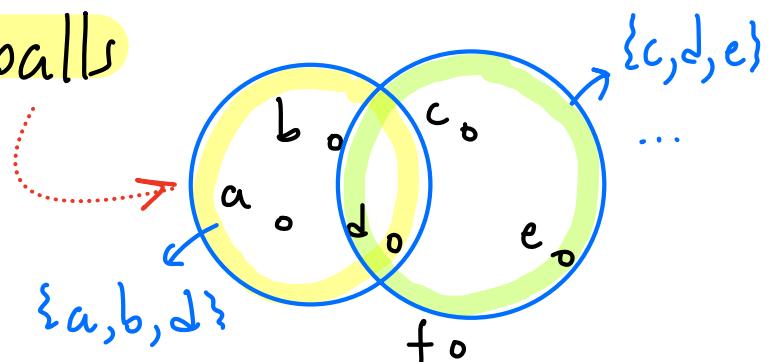
## Lecture 19 - Sampling + VC-Dimension

### Geometric Set Systems:

- Many problems involve sets of points that are defined by geometric objects
- Example: Given a set  $P \subseteq \mathbb{R}^d$ , consider all subsets of  $P$  contained in:
  - axis-aligned rectangles



- Euclidean balls



### Range Space:

Given a set  $P$ , let  $2^P$  denote the power set of  $P$ , consisting of all subsets of  $P$  ( $|2^P| = 2^{|P|}$ )

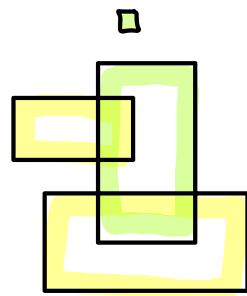
Range space is a pair  $(X, \mathcal{R})$  where:

$X$  - domain (a set)

$\mathcal{R}$  - ranges - a subset of  $2^X$

Eg.  $X = \mathbb{R}^2$

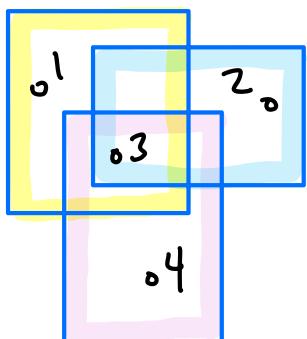
$\mathcal{R}$  = set of all axis-aligned rectangles  
(each is an infinite set)



Restriction: Given  $P \subseteq X$ , define

$$\mathcal{R}_{|P} = \{P \cap Q \mid Q \in \mathcal{R}\}$$

the restriction of  $\mathcal{R}$  to  $P$



$$\mathcal{R}_{|P} = \emptyset, \{1\}, \dots, \{4\}, \dots, \{1, 2, 3, 4\}$$

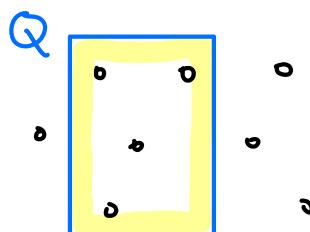
But not:  $\{1, 4\}$  or  $\{1, 2, 4\}$

Range space  $(X, \mathcal{R})$  is discrete if  $|X|$  finite

Given a discrete range space  $(P, \mathcal{R})$

and any  $Q \in \mathcal{R}$  define  $Q$ 's measure

$$\mu(Q) = \frac{|Q \cap P|}{|P|}$$



$$\mu(Q) = \frac{4}{8} = \frac{1}{2}$$

**Sampling:** Rather than deal with entire point set (may be **huge**) we would like a "**good**" sample.

Given  $S \subseteq P$  (presumably  $|S| \ll |P|$ ) define

$$\hat{\mu}_S(Q) = \frac{|Q \cap S|}{|S|}$$

(When  $S$  is clear, we write  $\hat{\mu}(Q)$ )

How good is  $S$  as a sample?

Given a discrete range space  $(P, \mathcal{R})$  +  $\varepsilon > 0$

**$\varepsilon$ -sample:**  $S \subseteq P$  is an  **$\varepsilon$ -sample** if

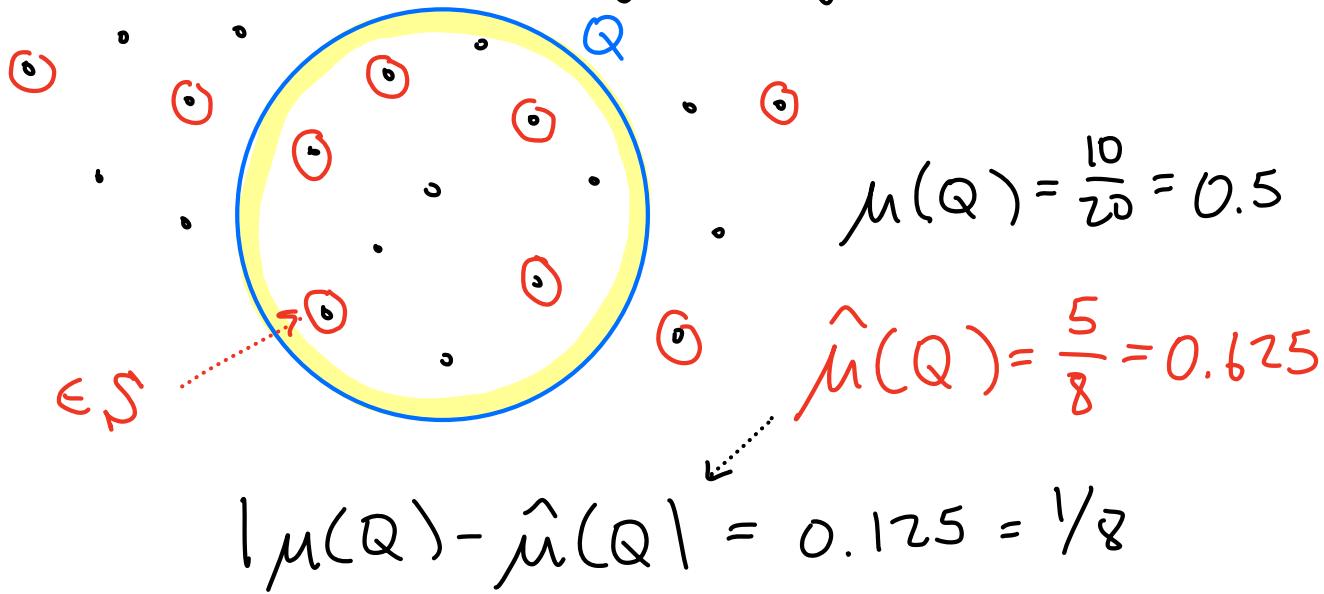
$$|\mu(Q) - \hat{\mu}(Q)| \leq \varepsilon \quad \forall Q \in \mathcal{R}$$

**$\varepsilon$ -net:**  $S \subseteq P$  is an  **$\varepsilon$ -net** if

$$\mu(Q) \geq \varepsilon \Rightarrow S \cap Q \neq \emptyset \quad \forall Q \in \mathcal{R}$$

## Intuition:

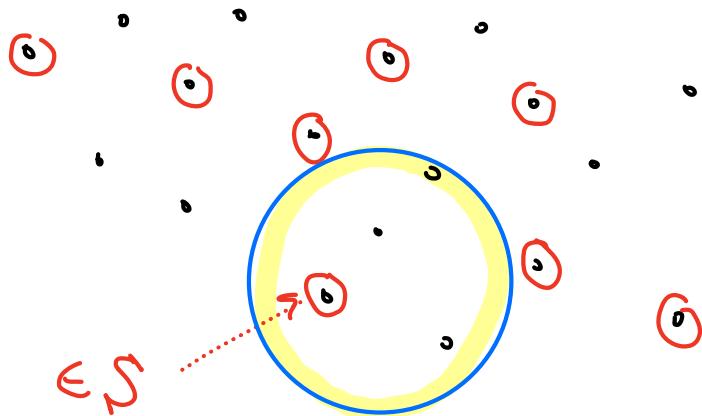
- $S$  is an  $\varepsilon$ -sample if it captures roughly the same proportion of elements for any range



If this holds for all ranges in  $\mathcal{R}$   
 $S$  is a  $\frac{1}{8}$ -sample.

- A range  $Q$  is  $\varepsilon$ -heavy if  $\mu(Q) \geq \varepsilon$

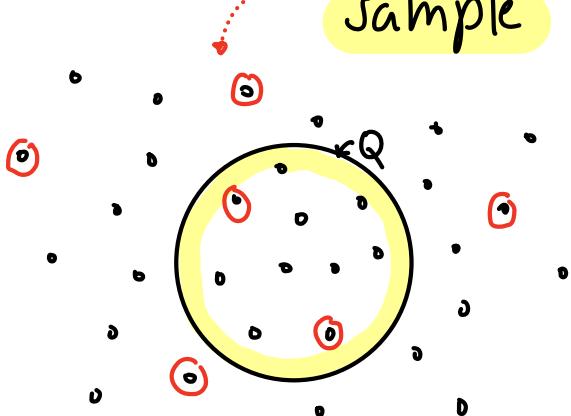
An  $\varepsilon$ -net hits all  $\varepsilon$ -heavy ranges



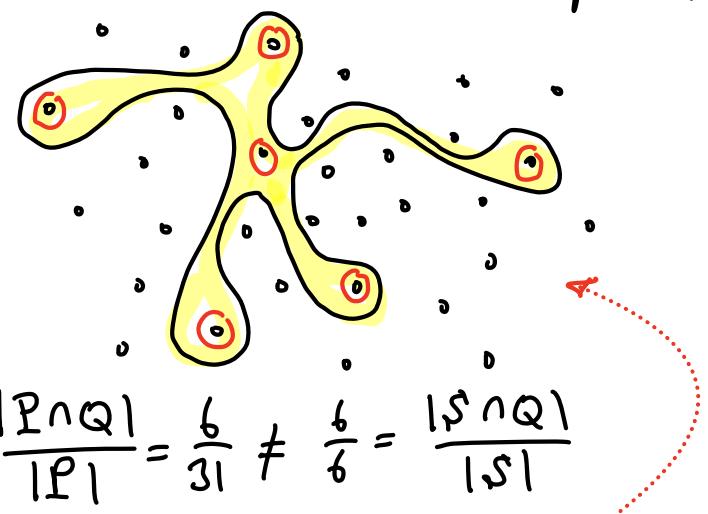
Any range that contains  $\geq \varepsilon \cdot |P| = 4$  pts must hit a pt of  $S'$

## How to construct $\varepsilon$ -nets + $\varepsilon$ -samples?

**Intuition:** Any sufficiently large random sample should work (with some prob.)



$$\frac{|P \cap Q|}{|P|} = \frac{10}{31} \approx \frac{2}{6} = \frac{|S \cap Q|}{|S|}$$



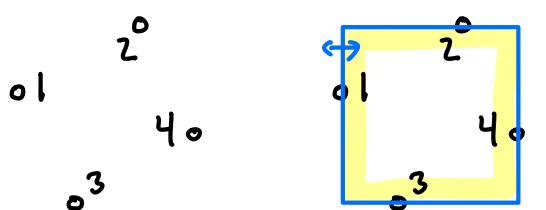
$$\frac{|P \cap Q|}{|P|} = \frac{6}{31} \neq \frac{6}{6} = \frac{|S \cap Q|}{|S|}$$

But this fails if we allow very wild range shapes.  
How to formally forbid such ranges?

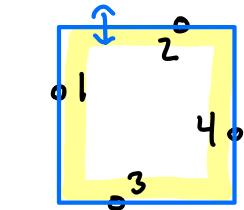
## VC-Dimension:

**Shattering:** A range space  $(X, \mathcal{R})$  shatters a pt set  $P$  if  $\mathcal{R}_{|P|} = 2^{|P|}$   
(contains all subsets of  $P$ )

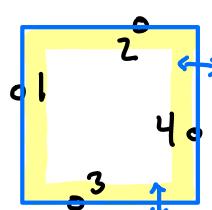
E.g. Axis-aligned rectangles shatter the pt set below:



Can include or exclude 1

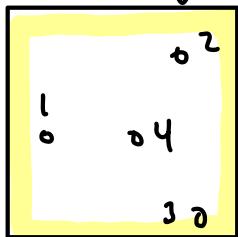


Can include or exclude 2



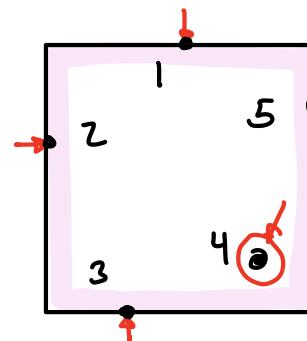
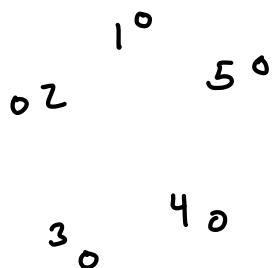
Same for 3+4

But they can't shatter everything:



Any rect. containing 1,2,3 must contain 4

... and they can never shatter a set of  $\geq 5$



Any rect that contains the 1,2,3,5 must contain 4

Def: The VC-dimension of a range space  $(X, \mathcal{R})$  is the size of the largest pt set shattered by  $\mathcal{R}$ .

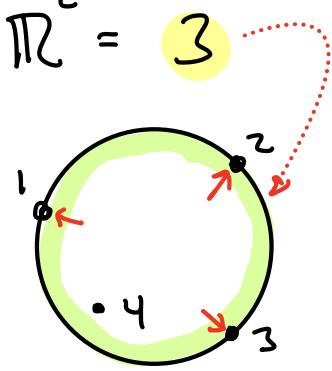
("VC" - Vapnik-Chervonenkis - 1971)

Examples:

→ VC-dim of axis-aligned rects in  $\mathbb{R}^2$  = 4

→ VC-dim of Euclidean disks in  $\mathbb{R}^2$  = 3

→ VC-dim of simple polygons in  $\mathbb{R}^2$  =  $\infty$



Intuitively: Range spaces of constant VC-dim have a constant num. of degrees of freedom

Sauer's Lemma: If  $(X, \mathcal{R})$  is a range space of VC-dim  $d$  in  $|X| = n$ , then

$$|\mathcal{R}| = \mathcal{O}(n^d)$$

More precisely:

$$|\mathcal{R}| \leq \Phi_d(n)$$

where:

$$\Phi_d(n) = \binom{n}{0} + \binom{n}{1} + \dots + \binom{n}{d}$$

Observe:  $\Phi$  satisfies the recurrence:

$$\Phi_d(n) = \Phi_d(n-1) + \Phi_{d-1}(n-1)$$

↳(Exercise)

Proof: (of Sauer's Lemma) Induction on  $d+n$ .

Basis:  $n=0$  or  $d=0$  - trivial  $\mathcal{R}=\{\emptyset\}$

Step: Fix any  $x \in X$

Consider two new range spaces:  
over  $X \setminus \{x\}$

$R_x = \{ Q \setminus \{x\} : Q \cup \{x\} \in R \text{ & } Q \setminus \{x\} \in R \}$   
 ↪ Pairs that differ only on  $x$

$R \setminus \{x\} = \{ Q \setminus \{x\} : Q \in R \}$   
 ↪ Just remove  $x$

Example:  $X = \{1, 2, 3, 4\}$  let  $x = 4$

Suppose  $R$  has:  $R_x$  has:

$\{2, 3\} + \{2, 3, 4\}$	$\longrightarrow$	$\{2, 3\}$
$\{1\} + \{1, 4\}$	$\longrightarrow$	$\{1\}$
$\{\} + \{4\}$	$\longrightarrow$	$\{\}$

and  $R$  has:  $\{1, 3\}$  but not  $\{1, 3, 4\}$   
 $\{2, 4\}$  but not  $\{2\}$

Then:  $R_x = \{\{\}, \{1\}, \{2, 3\}\}$

$R \setminus \{x\} = \{\{\}, \{1\}, \{2, 3\}, \{1, 3\}, \{2\}\}$

Observe:

- $|R| = |R_x| + |R \setminus \{x\}|$
- $R_x$  has VC-dim  $d-1$
- Both over domain of size  $n-1$

$$\Rightarrow |R| \leq \sum_{d=1}^n (n-1) + \sum_d (n-1) = \sum_d (n)$$
□

Recall:

Given a discrete range space  $(P, \mathcal{R})$  +  $\varepsilon > 0$

$\varepsilon$ -sample:  $S \subseteq P$  is an  $\varepsilon$ -sample if

$$|\mu(Q) - \hat{\mu}(Q)| \leq \varepsilon \quad \forall Q \in \mathcal{R}$$

$\varepsilon$ -net:  $S \subseteq P$  is an  $\varepsilon$ -net if

$$\mu(Q) \geq \varepsilon \Rightarrow S \cap Q \neq \emptyset \quad \forall Q \in \mathcal{R}$$

Range spaces of low VC-dimension have  
 $\varepsilon$ -samplers +  $\varepsilon$ -nets of small size:

$\varepsilon$ -Sample Theorem: Given range space  $(X, \mathcal{R})$  of  $\text{VC-dim } d$ , let  $P$  be finite subset of  $X$ . There exists constant  $c$  s.t. with probability  $\geq 1 - \varphi$ , a random sample of  $P$  of size  $\geq$

$$\frac{c}{\varepsilon^2} \left( d \cdot \log \frac{d}{\varepsilon} + \log \frac{1}{\varphi} \right)$$

is an  $\varepsilon$ -sample for  $(P, \mathcal{R})$ .

$\epsilon$ -Net Theorem: Given range space  $(\mathcal{X}, \mathcal{R})$  of VC-dim  $d$ , let  $P$  be finite subset of  $\mathcal{X}$ . There exists constant  $c$  s.t. with probability  $\geq 1 - \varphi$ , a random sample of  $P$  of size  $\geq$

$$\frac{c}{\epsilon} \left( d \log \frac{1}{\epsilon} + \log \frac{1}{\varphi} \right)$$

is an  $\epsilon$ -net for  $(P, \mathcal{R})$ .

Too many parameters!  $\therefore$

tl;dr : - Constant VC-dim  
- Constant prob. of success

Size of  $\epsilon$ -sample is  $\mathcal{O}\left(\frac{1}{\epsilon^2} \log \frac{1}{\epsilon}\right)$

$\epsilon$ -net is  $\mathcal{O}\left(\frac{1}{\epsilon} \log \frac{1}{\epsilon}\right)$

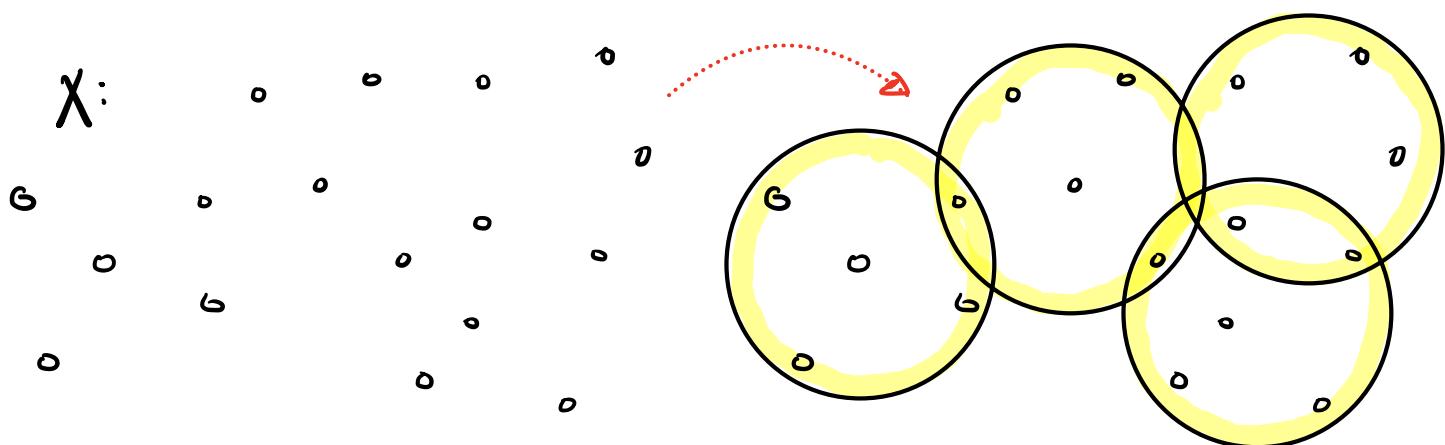
Proofs? See Har-Peled's book

## Application: Geometric Set Cover

Given a pt set  $X$  + a collection of sets  $R$  over  $X$ , a **cover** is a collection of sets from  $R$  that contain every pt of  $X$

E.g.  $X$  is a set of  $n$  pts in  $\mathbb{R}^d$

$R$  = set of all unit Euclidean balls in  $\mathbb{R}^d$



**Set cover Problem:** Given  $X$  and  $R$ , find the **smallest** cover of  $X$

- Set cover is **NP-hard**
- No known constant factor approximation
- Simple greedy algorithm computes a cover of size  $\leq (\ln |X|) \cdot \text{opt}$

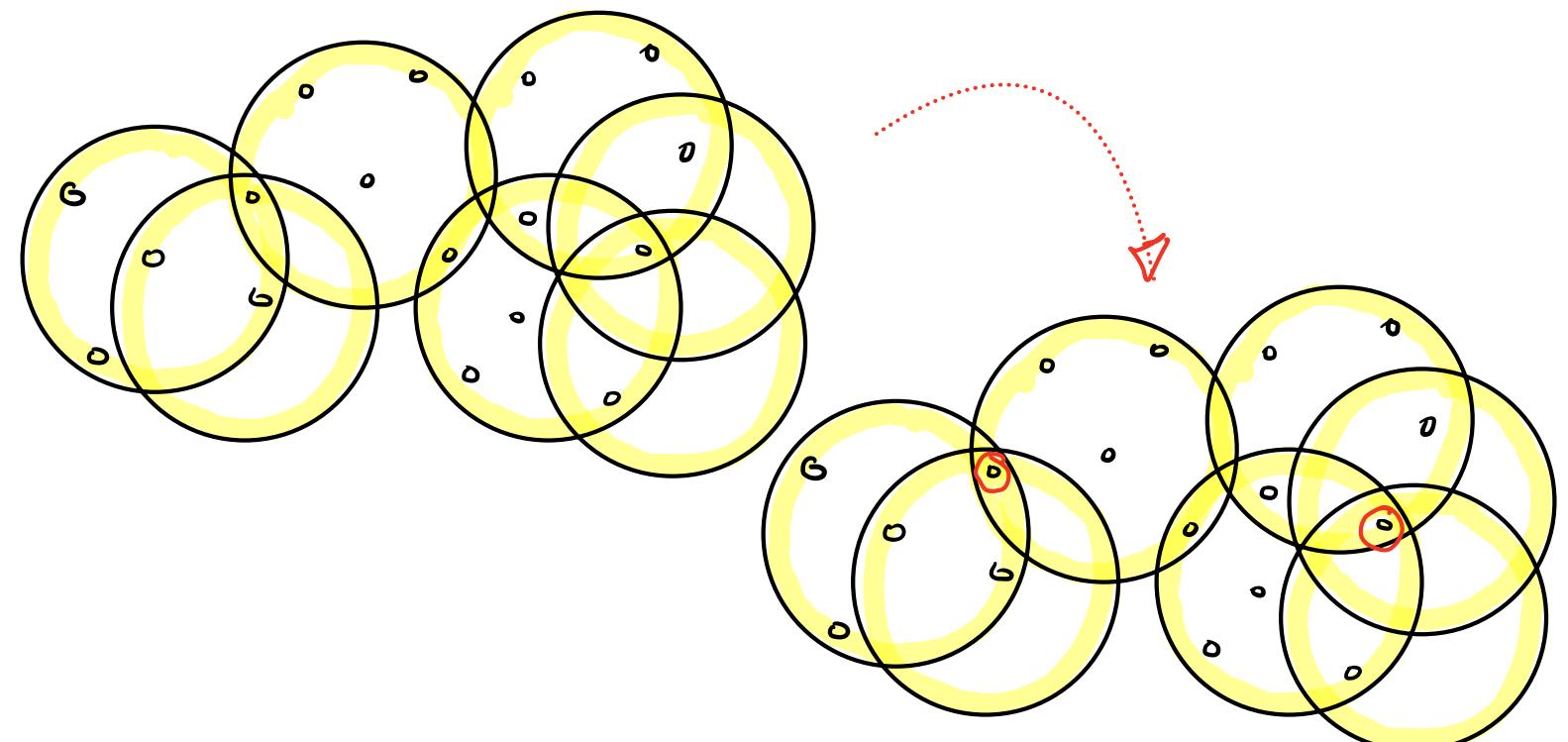
Select set that covers the most uncovered pts

We'll show that if  $(X, \mathcal{R})$  is a set system of constant VC-dimension, it is possible to compute an approx. solution of size  $\leq (\log k) \cdot \text{opt}$

where  $k$  is number of sets in opt. cover  
(Note  $k < |X|$ , so this is always better)

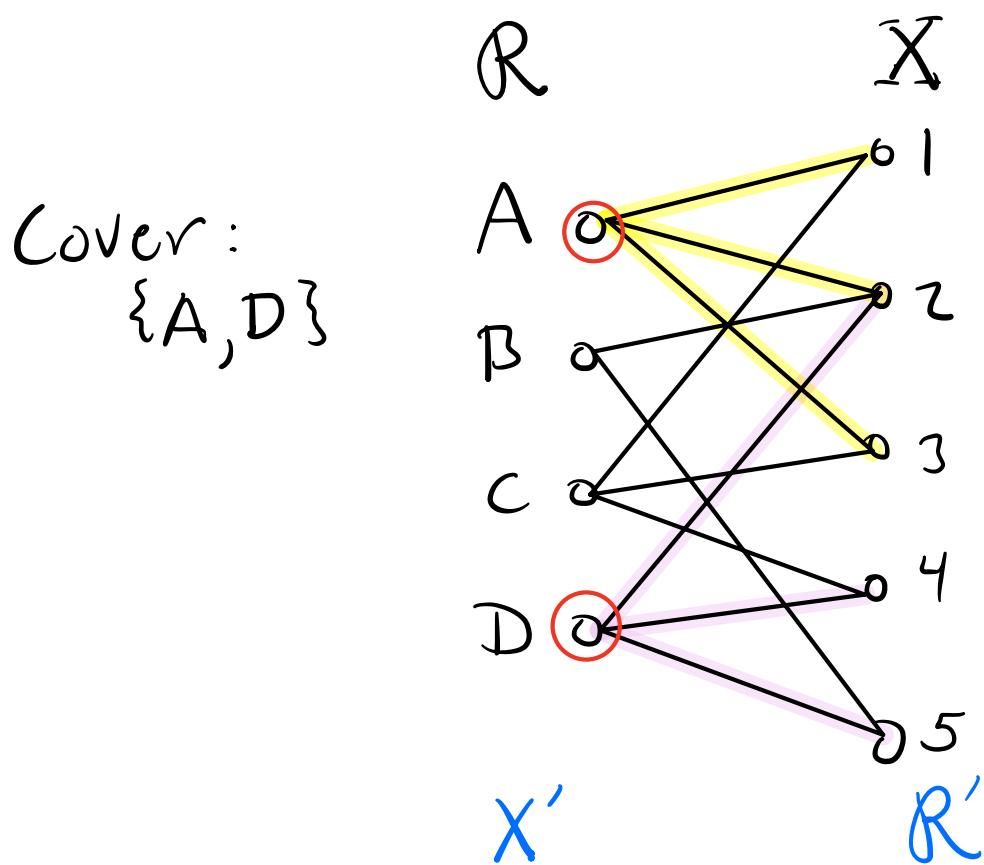
## Set cover $\leftrightarrow$ Hitting Set Duality

**Hitting Set:** Given a collection of sets  $\mathcal{R}$  over some domain  $X$ , a **hitting set** is a subset of  $X$  such that every set of  $\mathcal{R}$  contains at least one of them.



Set cover + hitting set are the same problem in disguise

E.g.  $A = \{1, 2, 3\}$   $B = \{2, 5\}$   
 $C = \{1, 3, 4\}$   $D = \{2, 4, 5\}$



Let's reinterpret: sets  $\rightarrow X'$ ; pts  $\rightarrow R'$

1:  $\{A, C\}$  2:  $\{A, B, D\}$  3:  $\{A, C\}$  4:  $\{C, D\}$  5:  $\{B, D\}$

Hitting set:  $\{A, D\}$

Obs:  $(X, R)$  has set cover of size  $k$  iff  $(X', R')$  has hitting set of size  $k$

**Theorem:** Given a set system  $(X, \mathcal{R})$  of constant VC-dimension, in polynomial time it is possible to compute a hitting set of size  $\tilde{O}(k^* \log k^*)$  where  $k^*$  = size of optimal hitting set.

**Note:** A set has constant VC-dim iff its dual has constant VC-dim.

## Iterative Reweighting:

**Weighted  $\varepsilon$ -Nets:** Given a set system  $(X, \mathcal{R})$  where each  $x \in X$  has a positive weight  $w(x)$ . Let  $w(X)$  be total weight:

$$w(X) = \sum_{x \in X} w(x)$$

A set  $S \subseteq X$  is an  $\varepsilon$ -net if

$$\forall Q \subseteq \mathcal{R} \text{ if } \frac{w(Q \cap S)}{w(Q)} \geq \varepsilon \text{ then } Q \cap S \neq \emptyset$$

Standard  $\varepsilon$ -net  $\equiv$  all pts have  $w(x) = 1$

## Weighted sampling:

$\epsilon$ -Net Theorem still holds, but rather than random sample of size  $O(\frac{1}{\epsilon} \log \frac{1}{\epsilon})$  sample each point with probability proportionate to its weight to get a set of this size.

## Iterative Reweighting:

- Guess the size  $k$  of opt hitting set (binary search to get best  $k$ )
- Set all weights to 1
- Repeat:
  - $S \leftarrow$  weighted  $\epsilon$ -net of  $X$
  - Is this a hitting set? Yes  $\rightarrow$  success
  - No? Find any set  $Q \subseteq R$  not hit
    - + double weights of all  $x \in Q$
  - Too many iterations? Fail  $\rightarrow$  try larger  $k$

Intuition: If we fail to hit we double weights of unhit object - more likely to hit next time.

Why it works: Critical items (in opt. solution) increase in weight rapidly - eventually they are all selected.

# Algorithm: Given $(X, \mathcal{R})$

for  $k = 1, 2, 4, \dots, 2^i, \dots$  until success

// Guess that  $\exists$  hitting set of size  $k$

-  $\forall x \in X$  set  $w(x) \leftarrow 1$

- Set  $\epsilon \leftarrow \frac{1}{4k}$

(for suitable const.  $c$ )

- Repeat until success or  $2k \cdot \lg \frac{n}{\epsilon k}$

iterations

-  $S \leftarrow$  wgt  $\epsilon$ -net of size  $c \cdot k \cdot \log k$

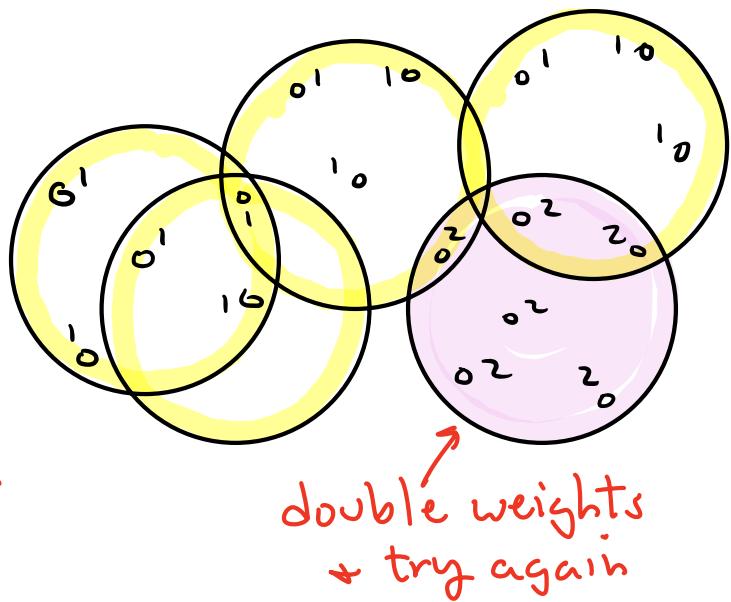
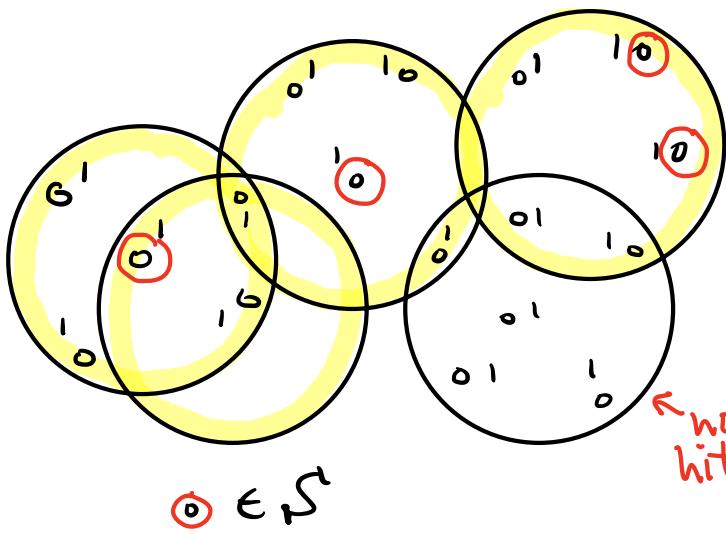
- are all sets of  $\mathcal{R}$  hit by  $S$ ?

- yes  $\rightarrow$  return with success!

- no  $\rightarrow$  find any set  $Q \in \mathcal{R}$

not hit

$\forall x \in Q, w(x) \leftarrow 2 \cdot w(x)$



Why this works? Assume  $k$  is correct

- Since opt hitting set hits all sets, at least one point of opt doubles in weight
- Weight of opt hitting set grows exponentially fast
- Total weight of pt set grows much more slowly
- Soon, opt hitting set's weight is so high we must sample it.

Lemma: If  $(X, R)$  has hitting set of size  $k$ , then the repeat-loop has success within  $2k \cdot \lg^n/k$  iterations. ( $\lg = \log_2$ )

Proof: Let  $n = |X|$   $m = |R|$

- Let  $H$  be hitting set of size  $k$

$\bar{W}_i(X) =$  total weight after  $i^{\text{th}}$  iteration

$\bar{W}_i(H) =$  weight of  $H$

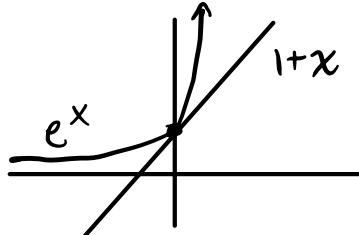
- Note:  $\bar{W}_0(X) = |X| = n$

- Since  $S$  is an  $\varepsilon$ -net, if we fail to hit a set  $Q$ , then  $w_i(Q) < \varepsilon \bar{W}_i(X)$

$$\Rightarrow \bar{W}_i(X) = \bar{W}_{i-1}(X) + \omega_{i-1}(Q) \\ \leq \bar{W}_{i-1}(X) + \varepsilon \cdot \bar{W}_{i-1}(X) \\ = (1 + \varepsilon) \bar{W}_{i-1}(X)$$

$$\Rightarrow \bar{W}_i(X) \leq (1 + \varepsilon)^2 \bar{W}_{i-2}(X) \\ \leq (1 + \varepsilon)^3 \bar{W}_{i-3}(X) \\ \vdots \\ \leq (1 + \varepsilon)^i \bar{W}_0(X) = (1 + \varepsilon)^i \cdot n$$

Fact:  $1+x \leq e^x$



$$\Rightarrow \bar{W}_i(X) \leq n \cdot e^{i \cdot \varepsilon}$$

Since  $H$  hits all sets, it hits  $Q$

$\Rightarrow$  in each (unsuccessful) iteration, at least one element of  $H$  doubles

$\Rightarrow$  growth rate of  $\bar{W}_i(H)$  is slowest if all its members double at same rate (Jensen's Ineq.)

$\Rightarrow$  After  $i^{\text{th}}$  iteration, each of the  $k$  elements of  $H$  doubled  $i/k$  times

$$\Rightarrow \bar{W}_i(H) \geq k \cdot 2^{i/k}$$

Since  $H \subseteq X$ , we know  $\overline{W}_i(H) \leq \overline{W}_i(X)$

$$\Rightarrow k \cdot 2^{i/k} \leq n \cdot e^{i \cdot \varepsilon}$$

Recall, we set  $\varepsilon \leftarrow 1/4k$

$$\Rightarrow k \cdot 2^{i/k} \leq n \cdot e^{i/4k}$$

$$\begin{aligned} \Rightarrow \lg k + \frac{i}{k} &\leq \lg n + \frac{i}{4k} \\ &\leq \lg n + \frac{i}{2k} \end{aligned}$$

$$\Rightarrow \frac{i}{k} - \frac{i}{2k} = \frac{i}{2k} \leq \lg n - \lg k = \lg \frac{n}{k}$$

$$\Rightarrow \text{No. of iterations } i \leq 2k \cdot \lg \frac{n}{k}$$

(If we exceed this number, we know  $|H| > k$ , and we fail)

□

Total time:

$$(2k \cdot \log \frac{n}{k}) \cdot [(k \cdot \log k) + m \cdot k]$$

$$= O(n^2 \cdot m \cdot \log n)$$

since  $k \leq n$