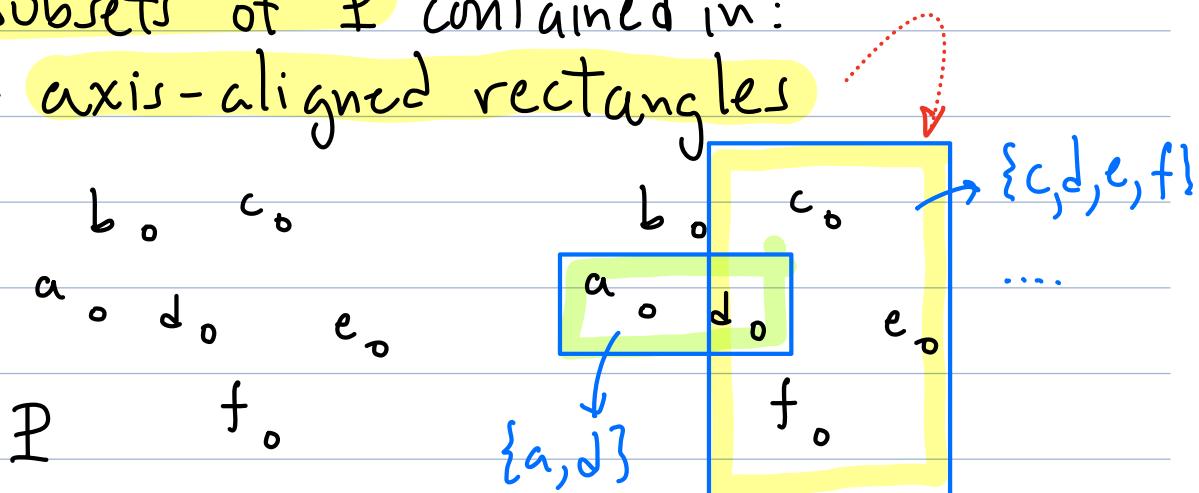


CMSC 754 - Computational Geometry

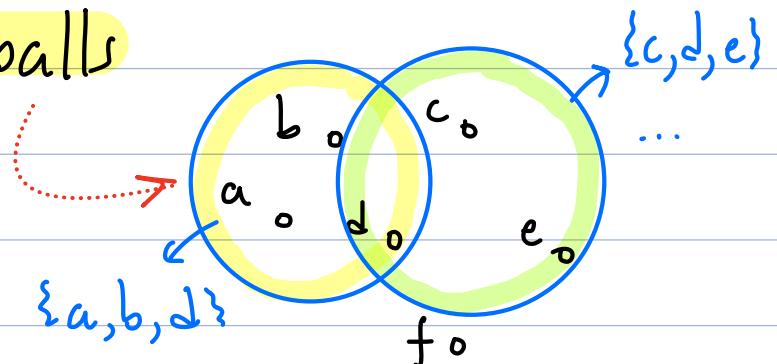
Lecture 19 - Sampling + VC-Dimension

Geometric Set Systems:

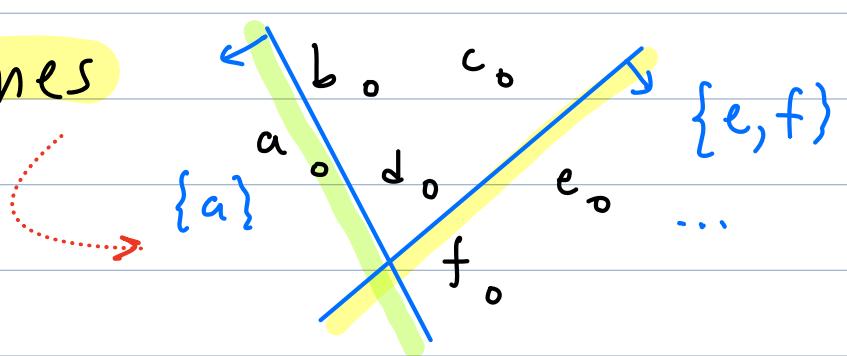
- Many problems involve sets of points that are defined by geometric objects
- Example: Given a set $P \subseteq \mathbb{R}^d$, consider all subsets of P contained in:
 - axis-aligned rectangles



- Euclidean balls



- Halfplanes



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, R) where:

X - domain (a set)

R - ranges - a subset of 2^X

Eg. $X = \{0, 1, 2, 3, \dots\}$

R = all subsets of contiguous values

$\{\{3\}, \{0\}, \{1\}, \{2\}, \dots\}$

$\{0, 1\}, \{1, 2\}, \{2, 3\}, \dots$

$\{0, 1, 2\}, \{1, 2, 3\}, \dots$

\vdots

Restriction: Given $P \subseteq X$, define

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

the restriction of R to P

E.g. $P = \{3, 4, 5\}$

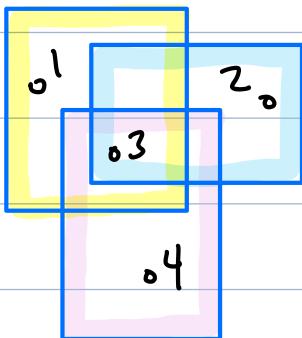
$R_{|P} = \{\{3\}, \{3\}, \{4\}, \{5\}, \{3, 4\}, \{4, 5\}, \{3, 4, 5\}\}$

Geometric Setting:

$(X, \mathcal{R}) : X = \mathbb{R}^2$
 $\mathcal{R} = \text{closed axis-parallel rects}$

Given $P = \{p_1, \dots, p_n\} \subseteq \mathbb{R}^2$

$\mathcal{R}_{|P} = \text{all subsets of } P \text{ defined by containment in rect.}$



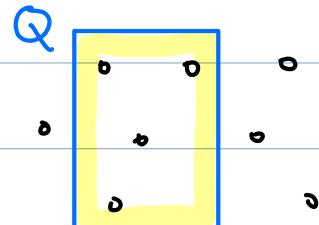
$$\begin{aligned}\mathcal{R}_{|P} = & \emptyset, \{1\}, \dots, \{4\}, \\ & \{1, 2\}, \{1, 3\}, \{2, 3\}, \{2, 4\}, \{3, 4\}, \\ & \{1, 2, 3\}, \{1, 3, 4\}, \{2, 3, 4\}, \\ & \{1, 2, 3, 4\}\end{aligned}$$

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}) + $\varepsilon > 0$

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

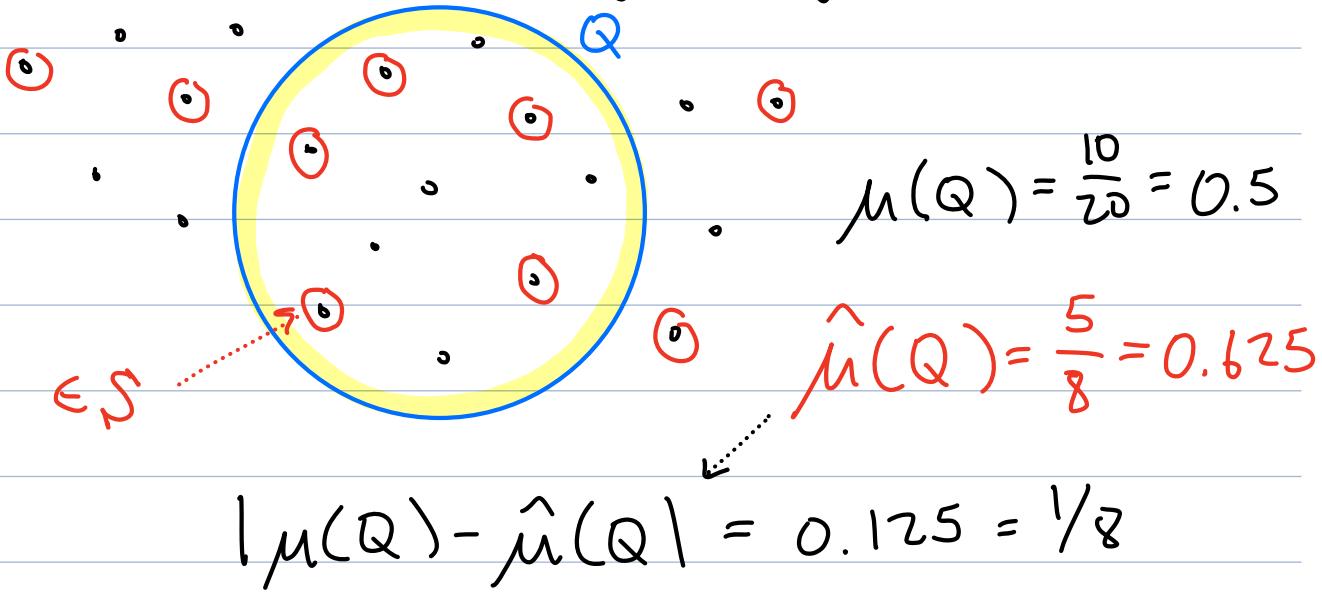
$$|\mu(Q) - \hat{\mu}(Q)| \leq \varepsilon \quad \forall Q \in \mathcal{R}$$

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

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

Intuition:

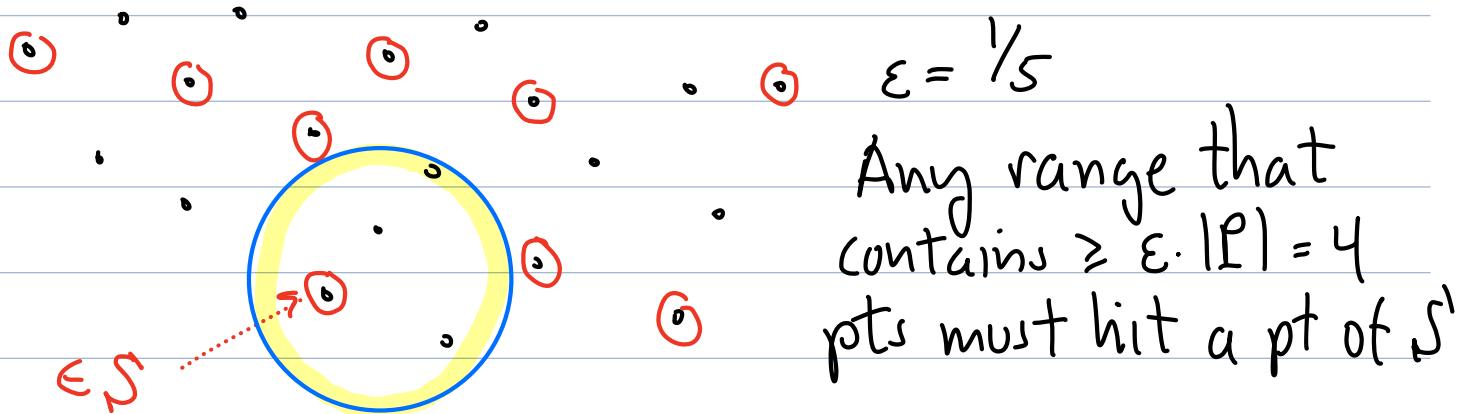
- S is an ϵ -sample if it captures roughly the same proportion of elements for any range



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

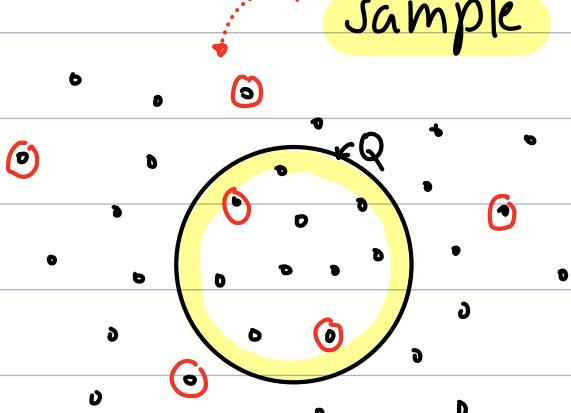
- A range Q is ϵ -heavy if $\mu(Q) \geq \epsilon$

An ϵ -net hits all ϵ -heavy ranges

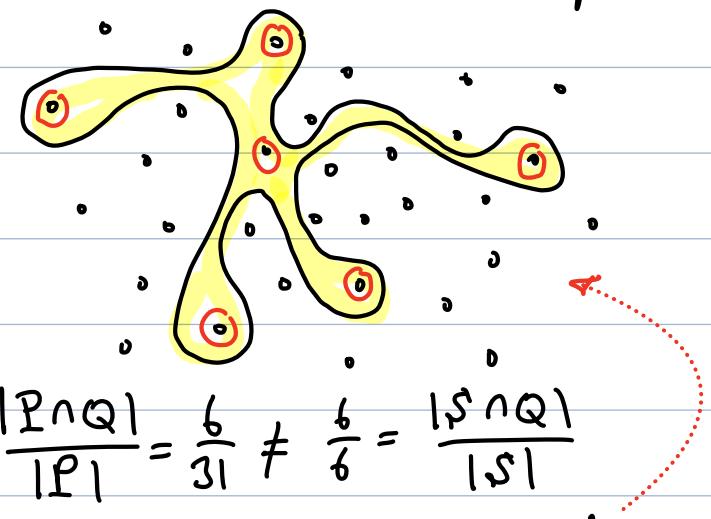


How to construct ε -nets + ε -samples?

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



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



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

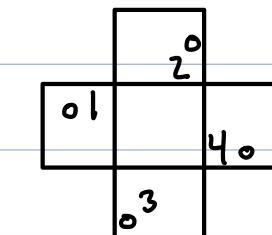
But this fails if we allow very wild range shapes.
How to formally forbid such ranges?

VC-Dimension:

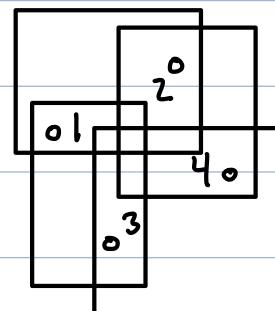
Shattering: A range space (X, R) shatters a pt set P if $R_{|P|} = 2^{|P|}$
(contains all subsets of P)

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

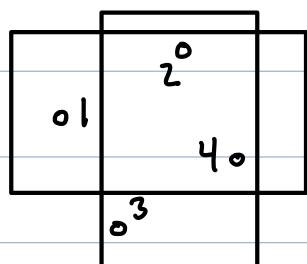
0 1 2
0 1
0 3
0 4



{1,4}, {2,3}

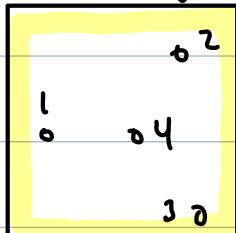


{1,2}, {2,4}, ...



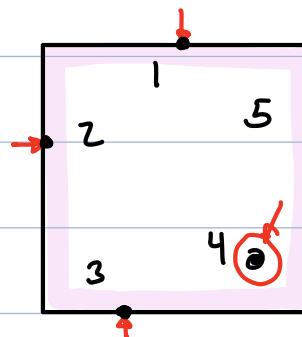
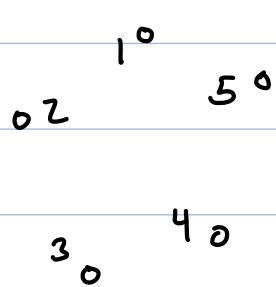
{1,2,4}, {2,3,4}, ...

But they can't shatter everything:



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

... and they can never shatter a set of ≥ 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} .

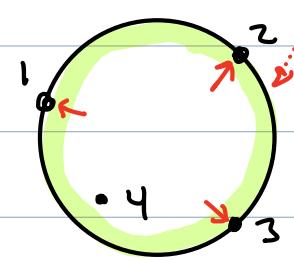
("VC"-Vapnik-Chervonenkis - 1971)

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 = ∞



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: Φ 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:

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

$$\{1\} + \{1, 4\}$$

$$\{\} + \{4\}$$

R_x has:

$$\{2, 3\}$$

$$\{1\}$$

$$\{\}$$

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 $n-1$

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

□

Recall:

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

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

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

ε -net: $S \subseteq P$ is an ε -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
 ε -samples + ε -nets of small size:

ε -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 ε -sample for (P, \mathcal{R}) .

ϵ -Net Theorem: Given range space $(\mathcal{X}, \mathcal{R})$ of $\text{VC-dim } d$, 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 ϵ -net for (P, \mathcal{R}) .

Too many parameters!

tl;dr :

- Constant VC-dim
- Constant prob. of success

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

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

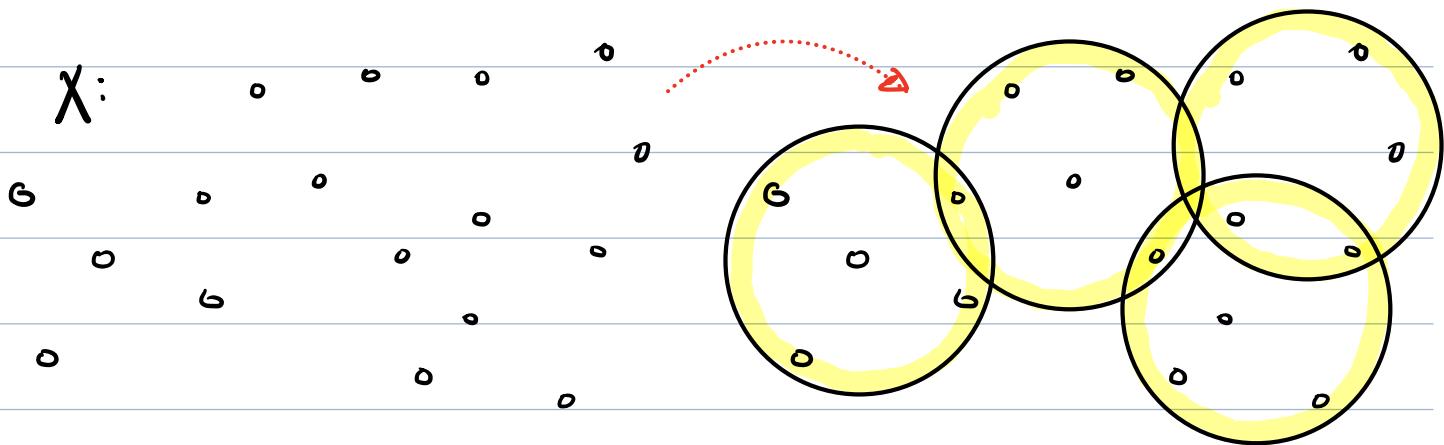
Proofs? See Har-Peled's book

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 $\leq (\ln |X|) \cdot \text{opt}$

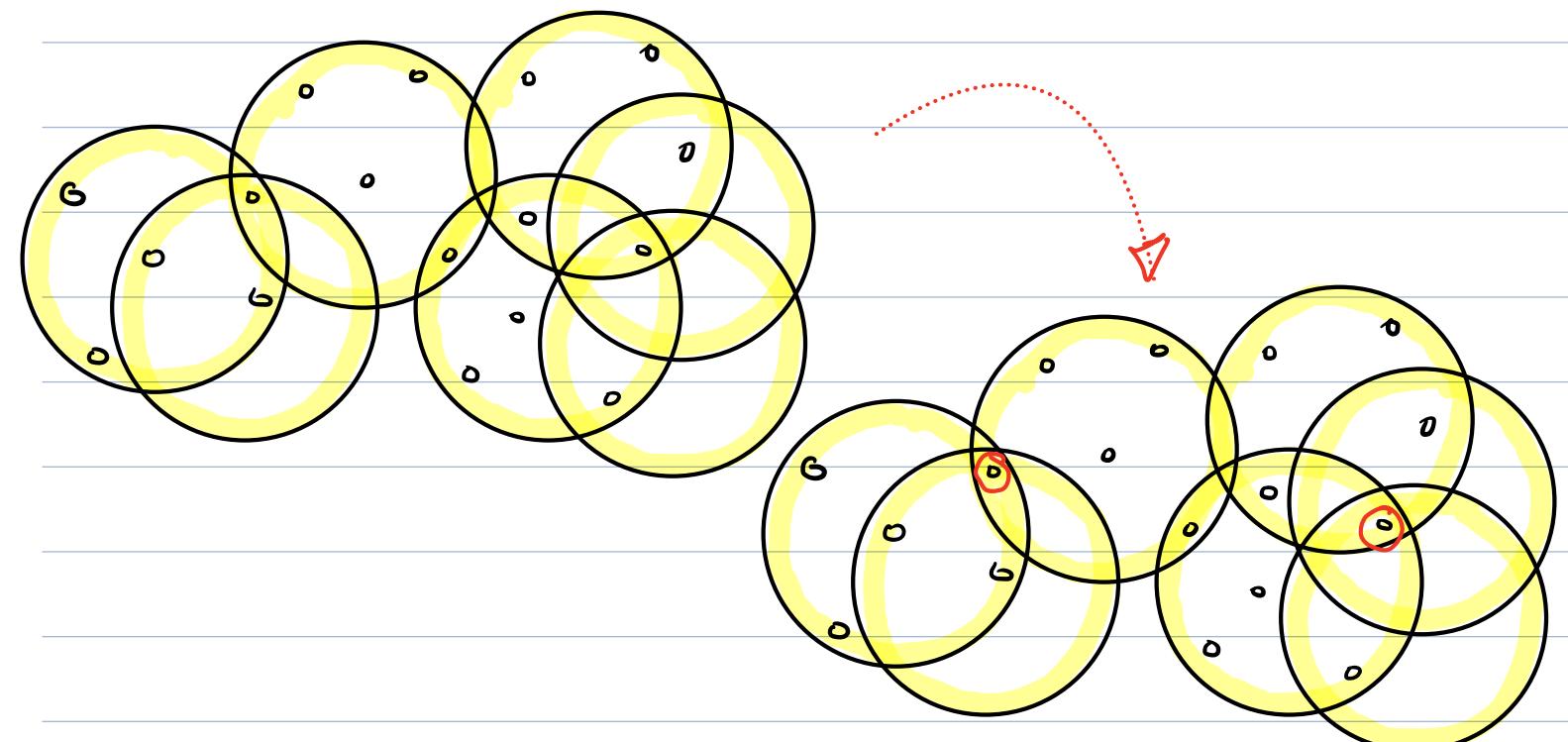
Select set
that covers
the most
uncovered
pts

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)

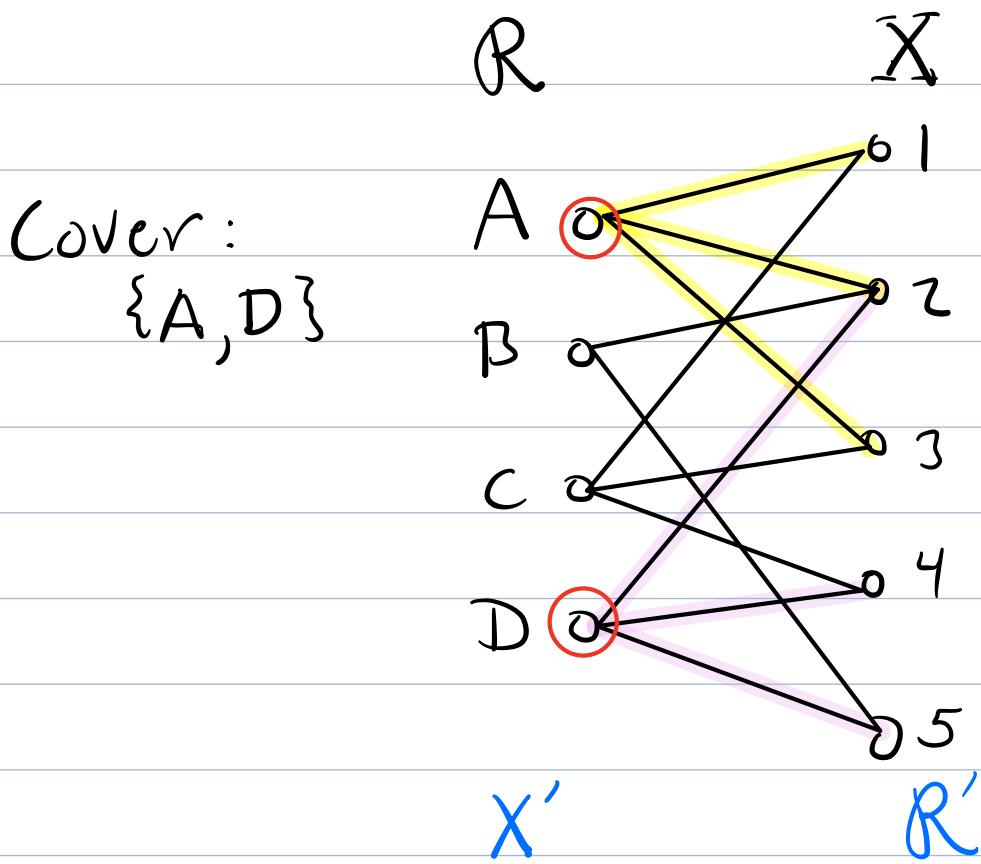
Set cover \leftrightarrow Hitting Set Duality

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

Weighted sampling:

ϵ -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 ϵ -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.

Huh? This can't work!
Just chasing our tail.

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 ϵ -net of size $c \cdot k \cdot \log k$

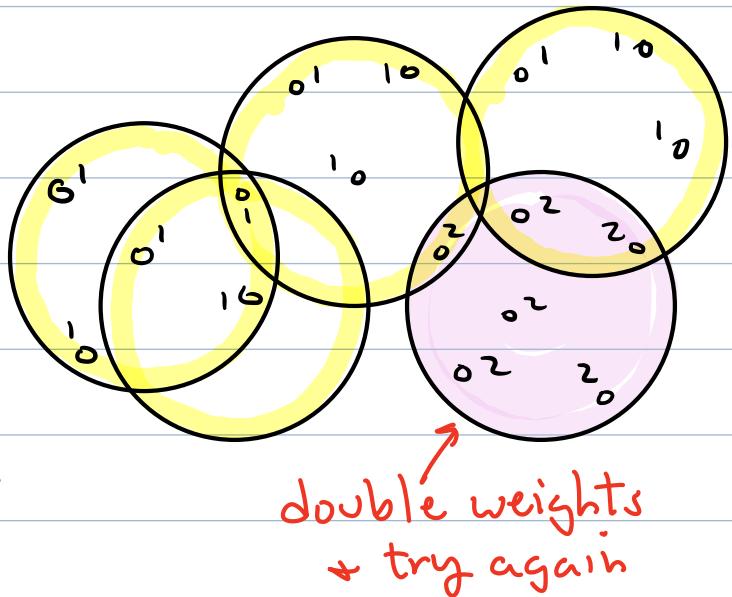
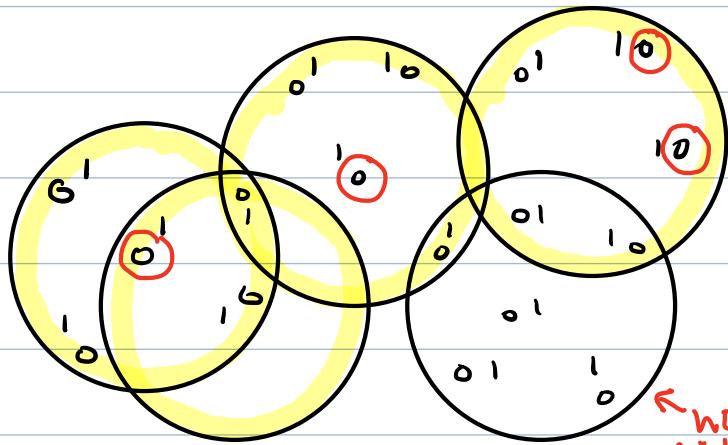
- are all sets of \mathcal{R} hit by S ?

- 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^{th} iteration

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

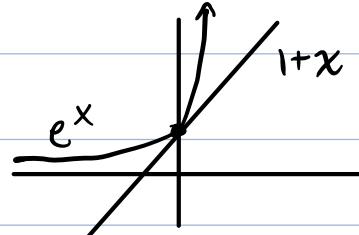
- Note: $\bar{W}_0(X) = |X| = n$

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

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

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

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



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

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^{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 $\varepsilon \leftarrow 1/4k$

$$\Rightarrow k \cdot 2^{i/k} \leq n \cdot e^{i/4k} \quad \text{lg e} < 2$$

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

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

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

(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 $k \leq n$