

CSE 421

Approximation Alg

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An Idea

Choose a number w from x_1, \dots, x_n

Define

- $S_{<}(w) = \{x_i: x_i < w\}$
- $S_{=}(w) = \{x_i: x_i = w\}$
- $S_{>}(w) = \{x_i: x_i > w\}$

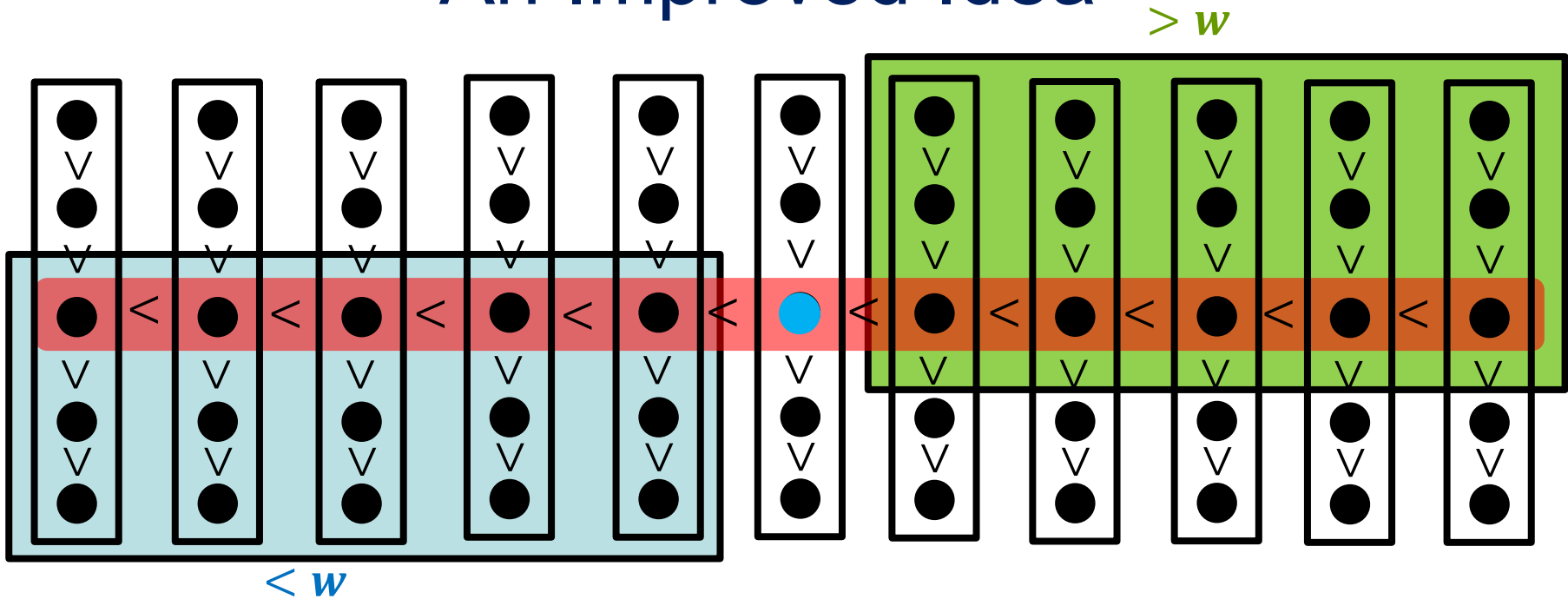
Can be computed in linear time

Solve the problem recursively as follows:

- If $k \leq |S_{<}(w)|$, output $Sel(S_{<}(w), k)$
- Else if $k \leq |S_{<}(w)| + |S_{=}(w)|$, output w
- Else output $Sel(S_{>}(w), k - |S_{<}(w)| - |S_{=}(w)|)$

Ideally want $|S_{<}(w)|, |S_{>}(w)| \leq n/2$. In this case ALG runs in $O(n) + O\left(\frac{n}{2}\right) + O\left(\frac{n}{4}\right) + \dots + O(1) = O(n)$.

An Improved Idea



Partition into $n/5$ sets. Sort each set and set $w = \text{Sel}(\text{midpoints}, n/10)$

- $|S_{<}(w)| \geq 3 \left(\frac{n}{10}\right) = \frac{3n}{10}$
- $|S_{>}(w)| \geq 3 \left(\frac{n}{10}\right) = \frac{3n}{10}$

$\frac{3n}{10} \leq |S_{<}(w)|, |S_{>}(w)| \leq \frac{7n}{10}$

$$T(n) = T\left(\frac{n}{5}\right) + T\left(\frac{7n}{10}\right) + O(n) \Rightarrow T(n) = O(n)$$


Median Algorithm

```

Sel(S, k) {
  n ← |S|
  If (n < ??) return ??
  Partition S into n/5 sets of size 5
  Sort each set of size 5 and let M be the set of medians, so
  |M|=n/5
  Let w=Sel(M,n/10)
  For i=1 to n{
    If  $x_i < w$  add x to  $S_{<}(w)$ 
    If  $x_i > w$  add x to  $S_{>}(w)$ 
    If  $x_i = w$  add x to  $S_{=}(w)$ 
  }
  If ( $k \leq |S_{<}(w)|$ )
    return Sel( $S_{<}(w)$ , k)
  else if ( $k \leq |S_{<}(w)| + |S_{=}(w)|$ )
    return w;
  else
    return Sel( $S_{>}(w)$ ,  $k - |S_{<}(w)| - |S_{=}(w)|$ )
}

```

We can maintain each set in an array



Approximation Algorithms

How to deal with NP-complete Problem

Many of the important problems in real world are NP-complete.

SAT, Set Cover, Graph Coloring, TSP, Max IND Set, Vertex Cover, ...

So, we cannot find optimum solutions in polynomial time.

What to do instead?

- Find optimum solution of special cases (e.g., random inputs)
- Find near optimum solution in the worst case

Approximation Algorithm

Polynomial-time Algorithms with a guaranteed approximation ratio.

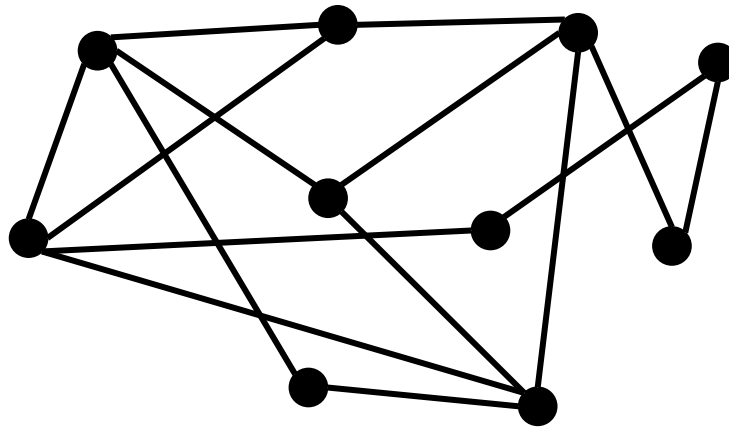
$$\alpha = \frac{\text{Cost of computed solution}}{\text{Cost of the optimum}}$$

worst case over all instances.

Goal: For each NP-hard problem find an approximation algorithm with the best possible approximation ratio.

Vertex Cover

Given a graph $G=(V,E)$, Find smallest set of vertices touching every edge



Greedy Algorithm?

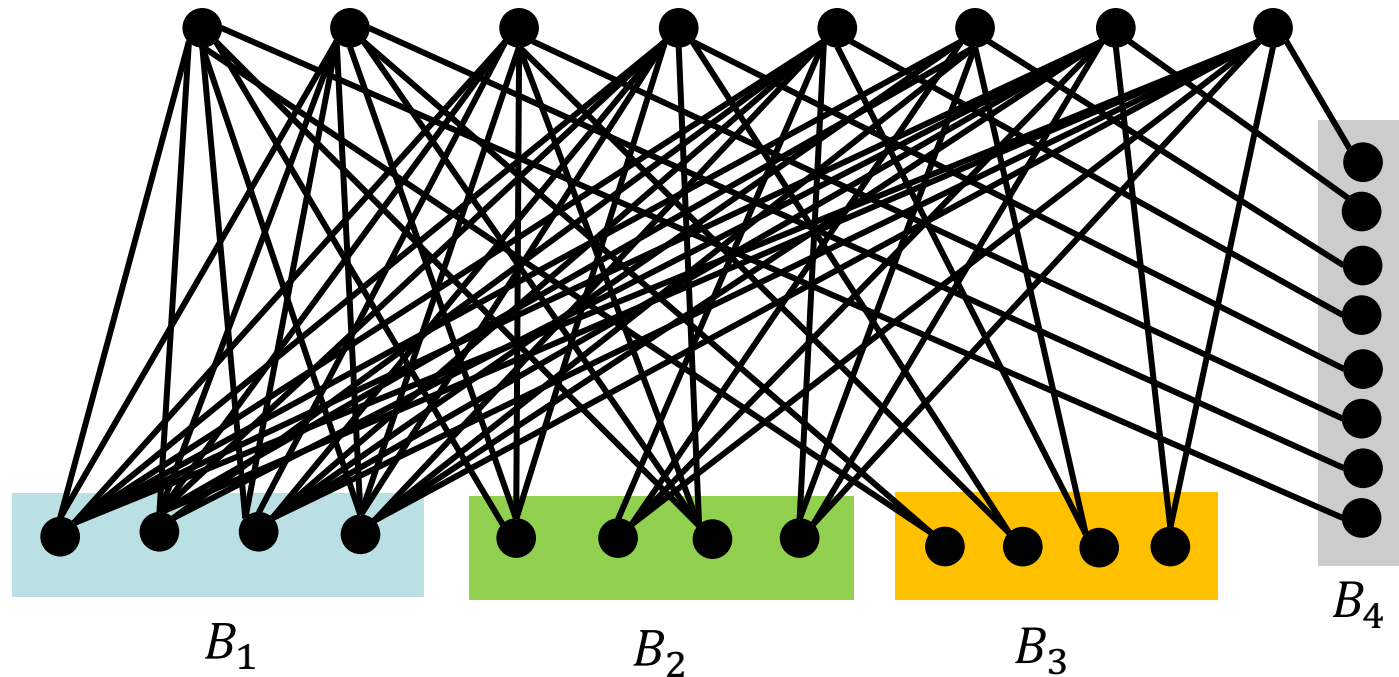
Greedy algorithms are typically used in practice to find a (good) solution to NP-hard problems

Strategy (1): Iteratively, include a vertex that covers most new edges

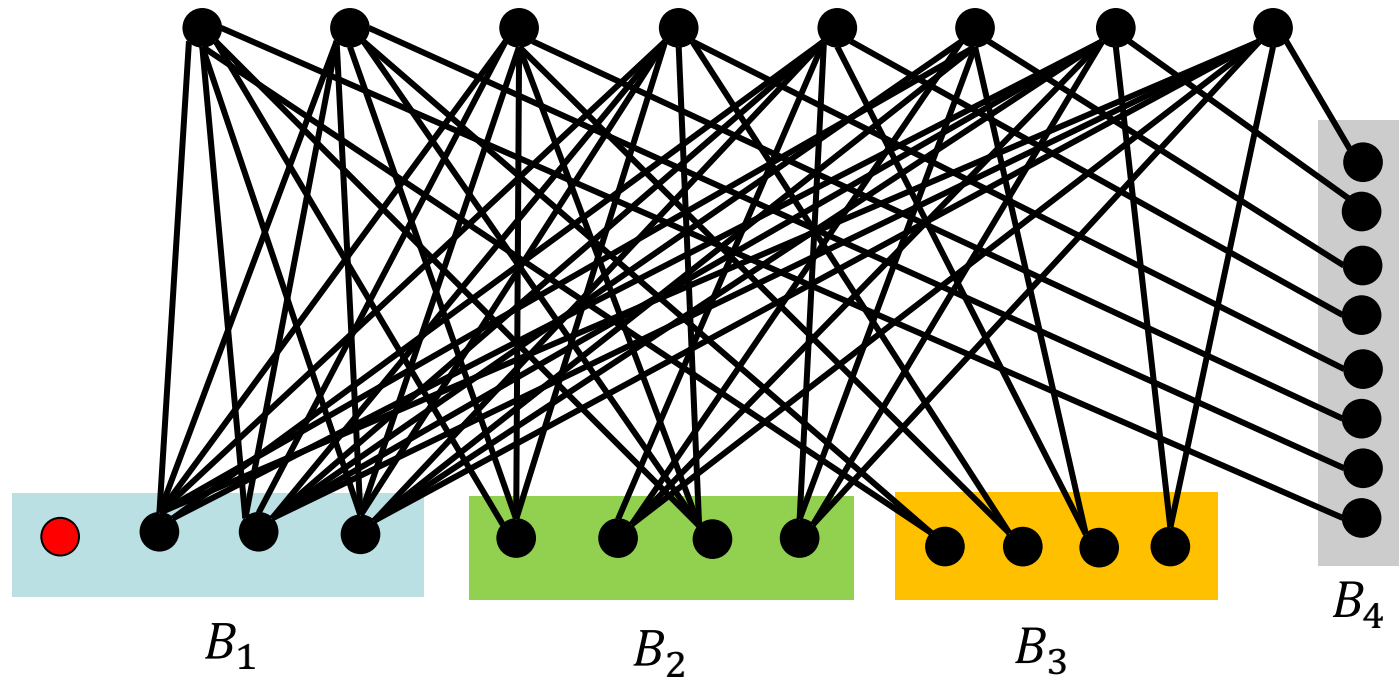
Q: Does this give an optimum solution?

A: No,

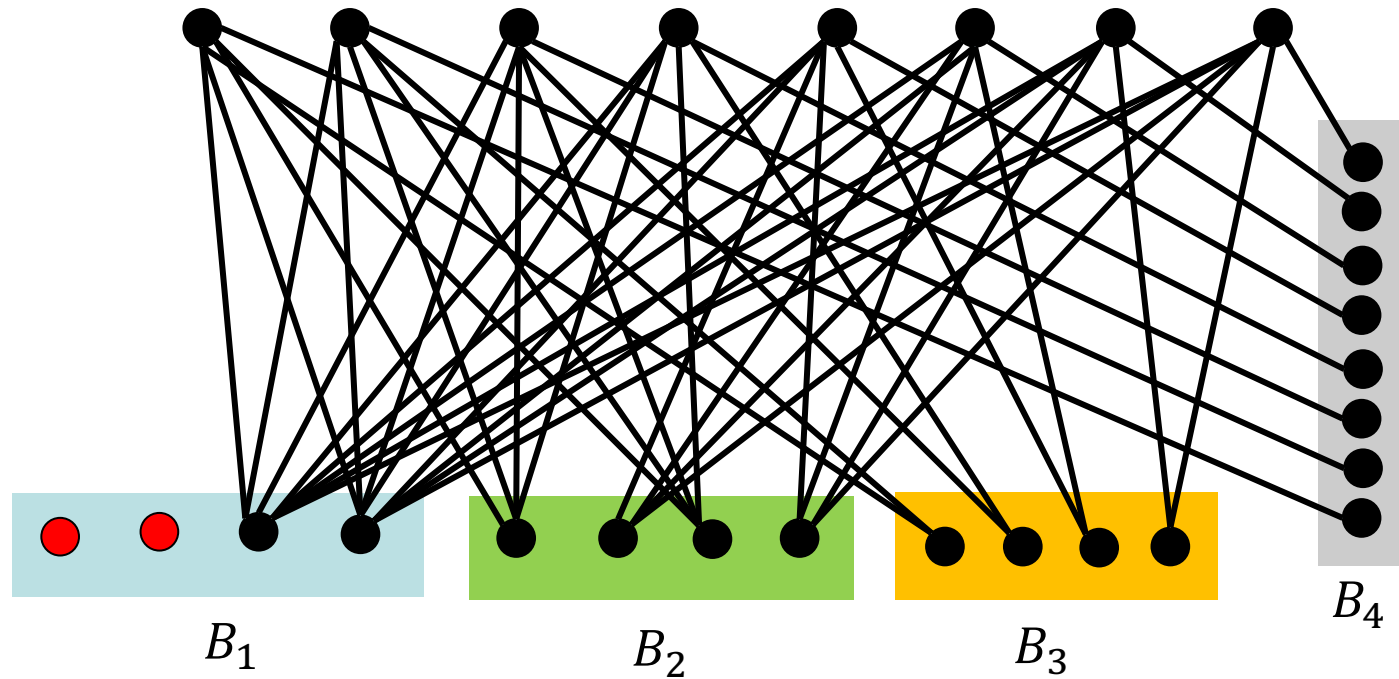
Greedy (1): Pick vertex that covers the most



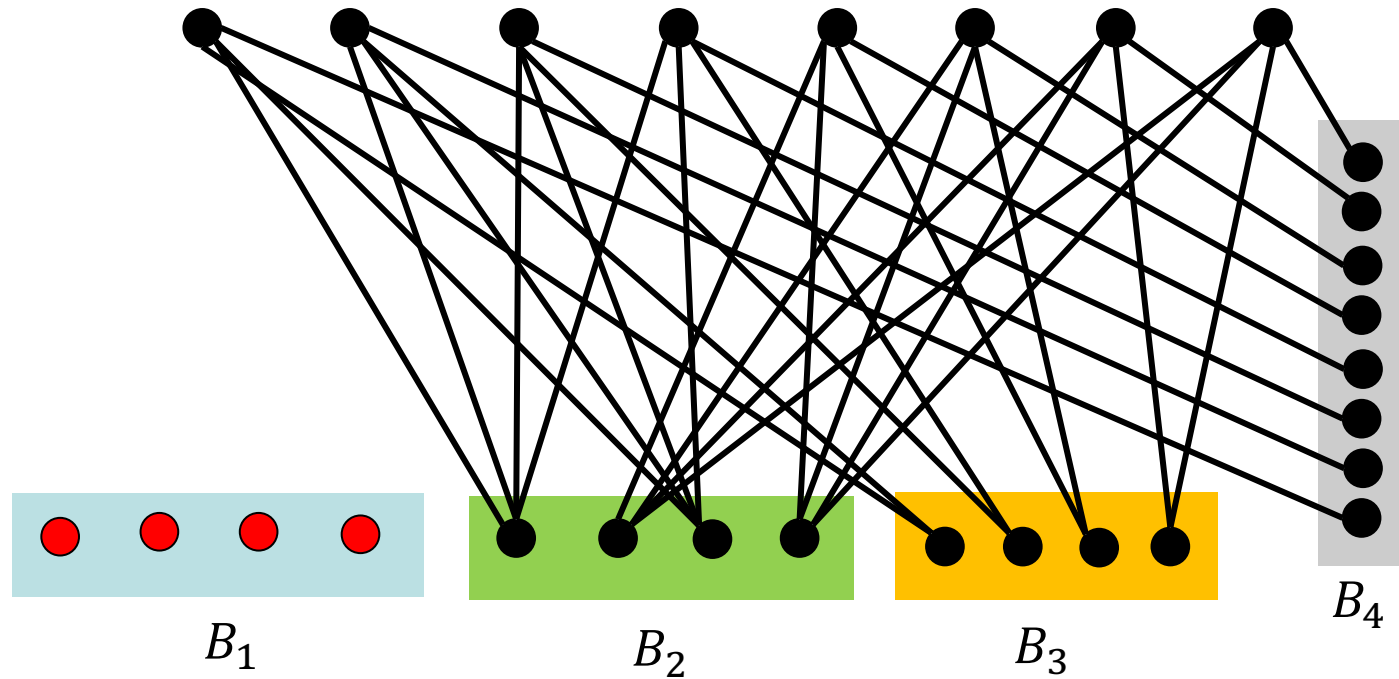
Greedy (1): Pick vertex that covers the most



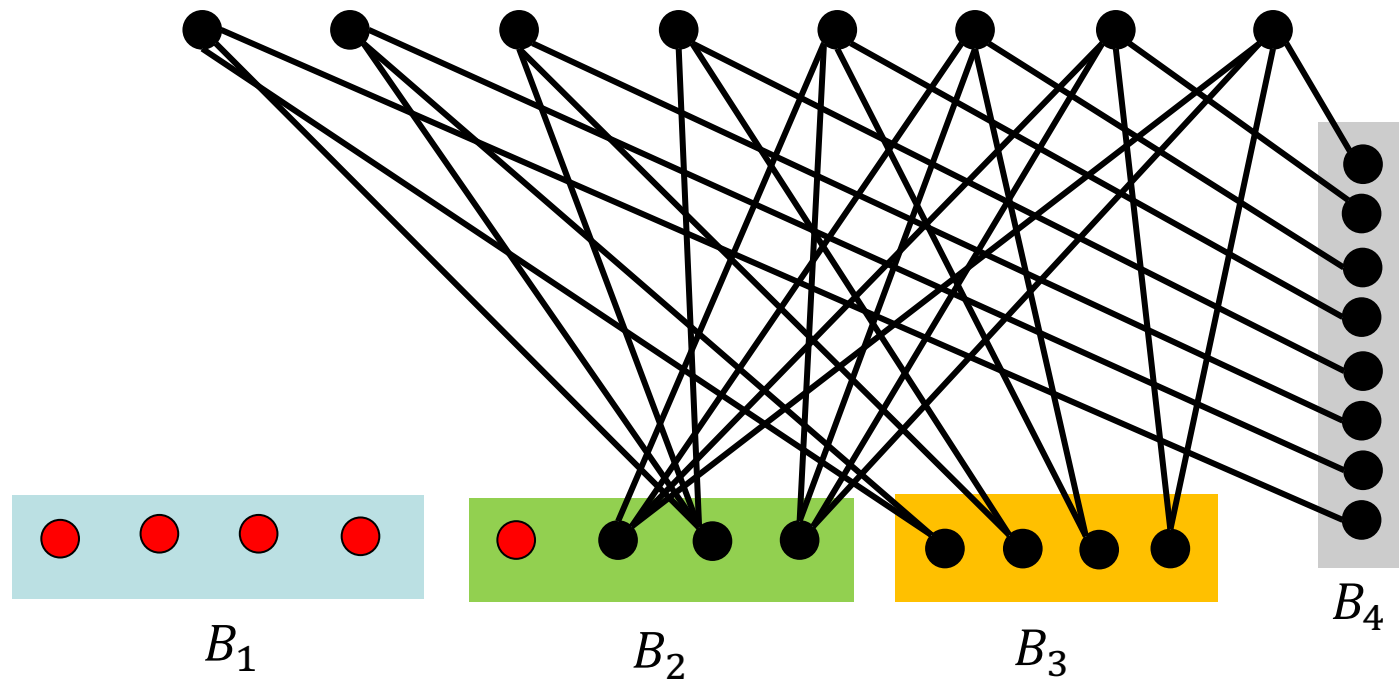
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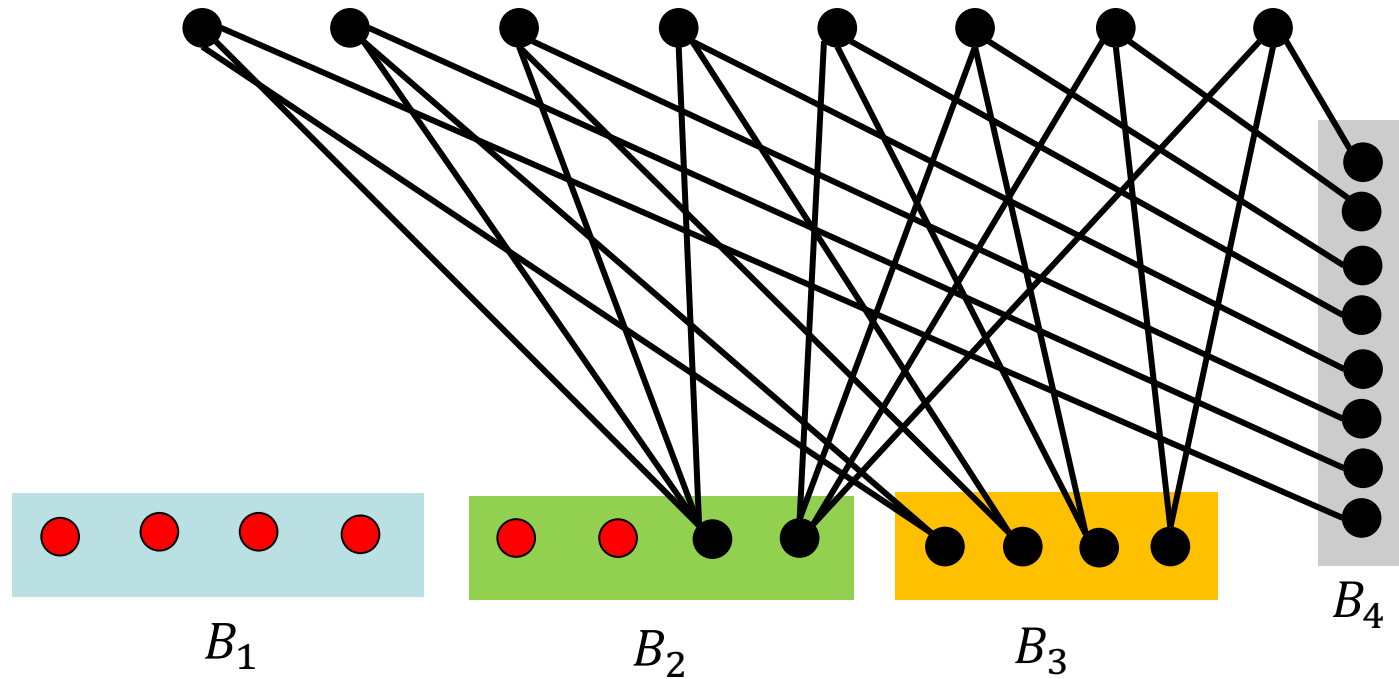
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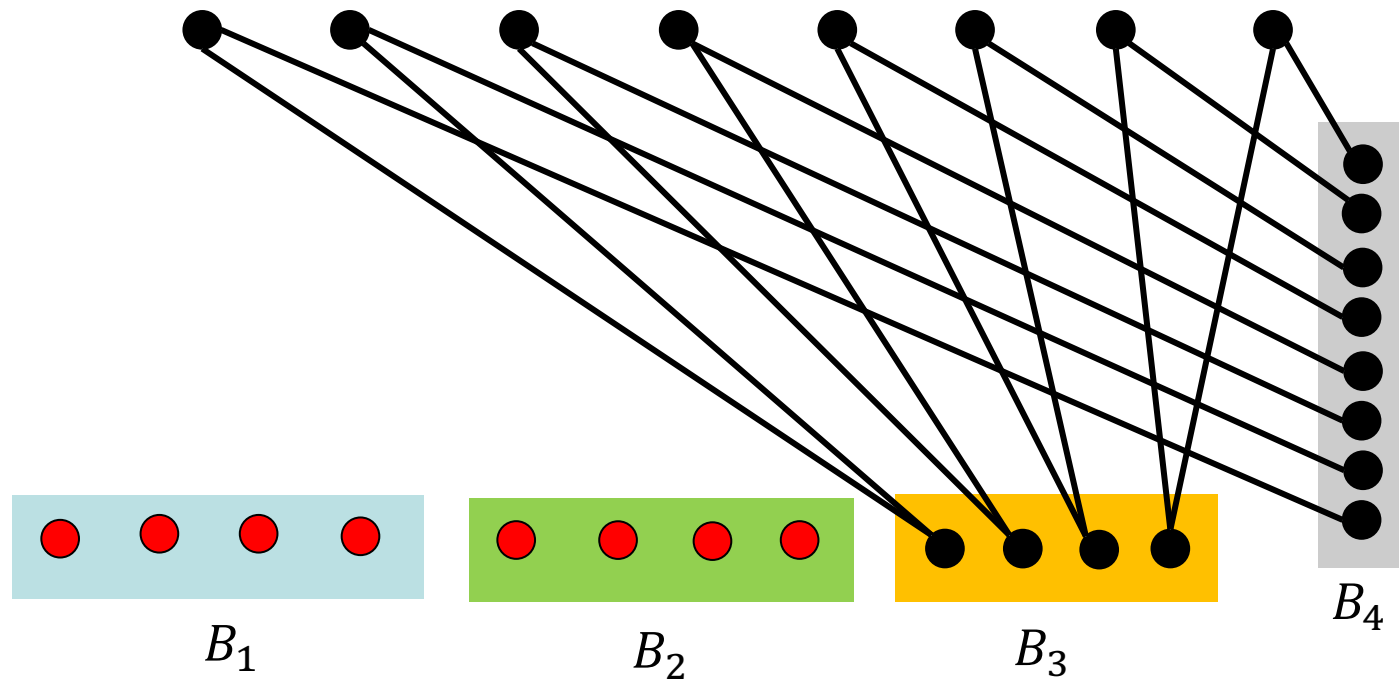
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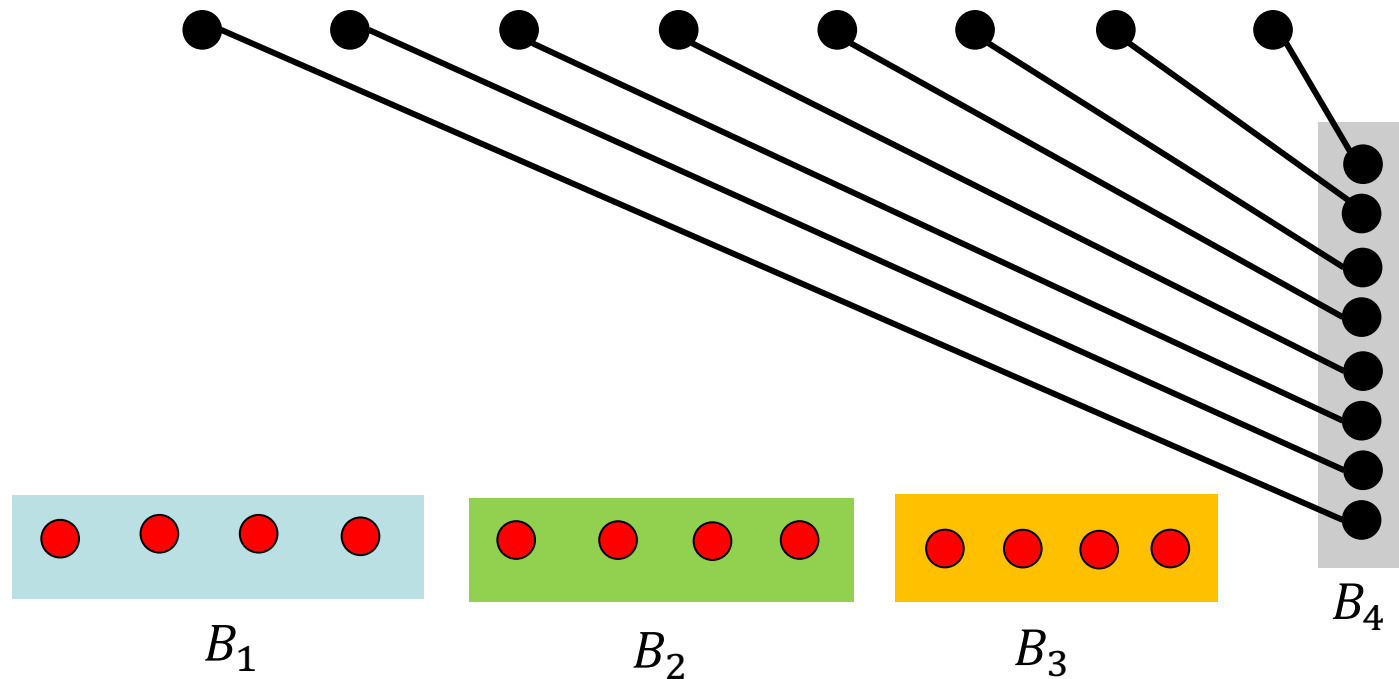
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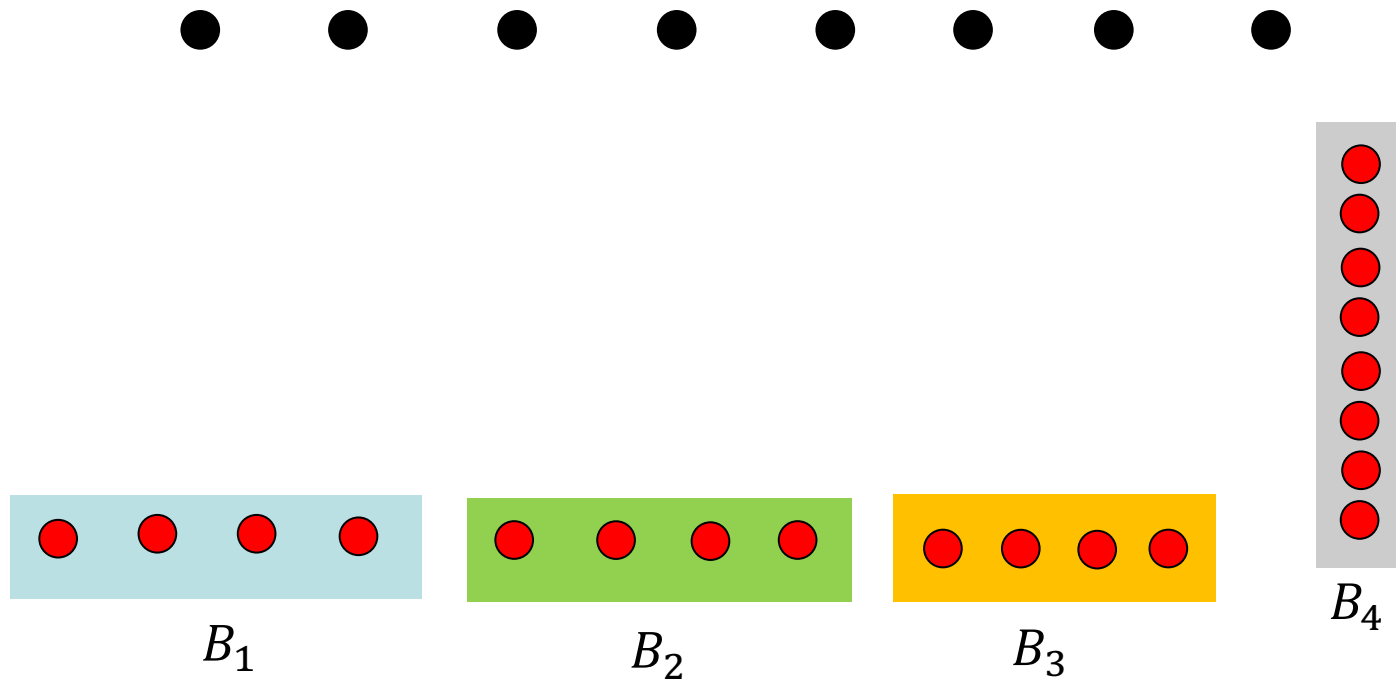
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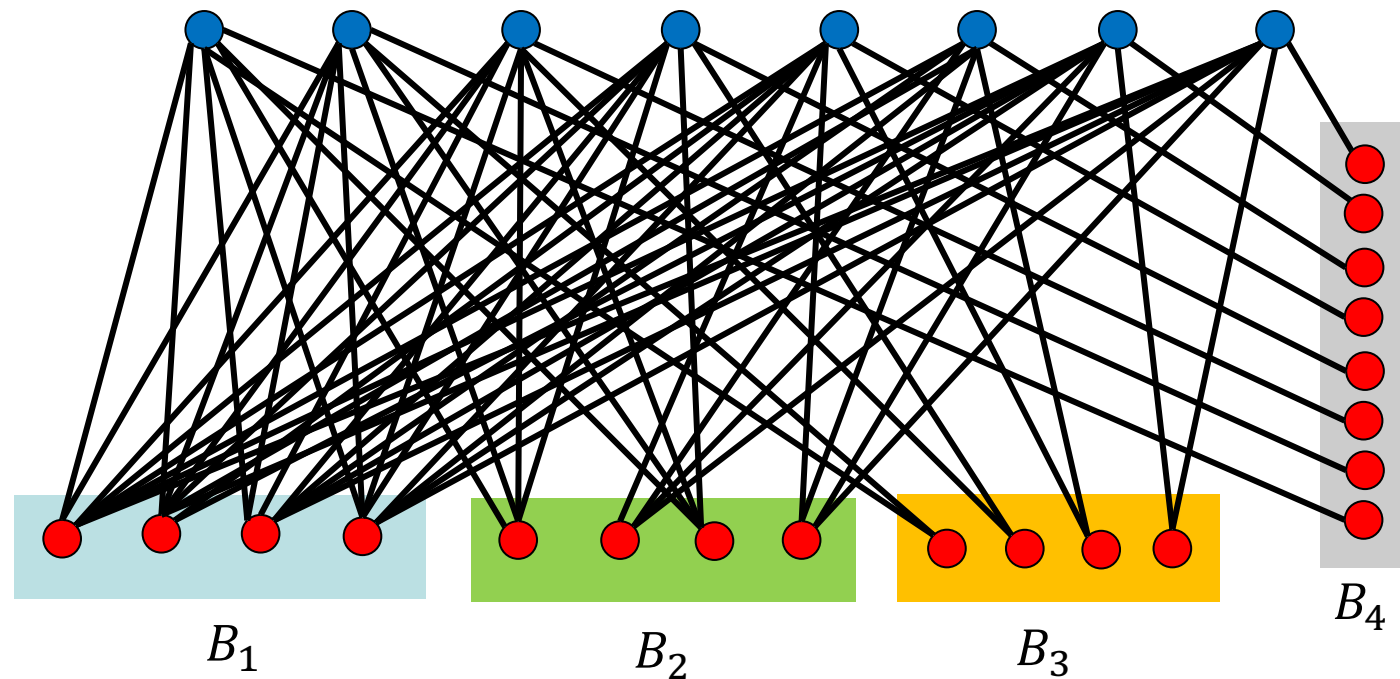
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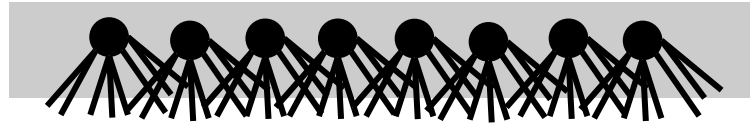


Greedy Vertex cover = 20

OPT Vertex cover = 8

Greedy (1): Pick vertex that covers the most

n vertices. Each vertex has one edge into each B_i



B_n



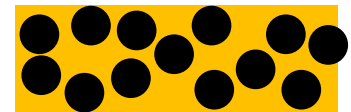
B_{n-1}

.....



$|B_i| = n/i$

.....



B_1

Each vertex in B_i has i edges to top

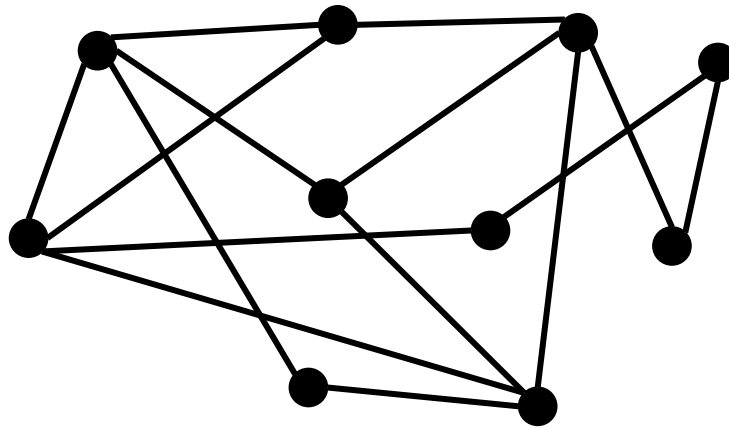
Greedy pick bottom vertices = $n + \frac{n}{2} + \frac{n}{3} + \dots + 1 \approx n \ln n$

OPT pick top vertices = n

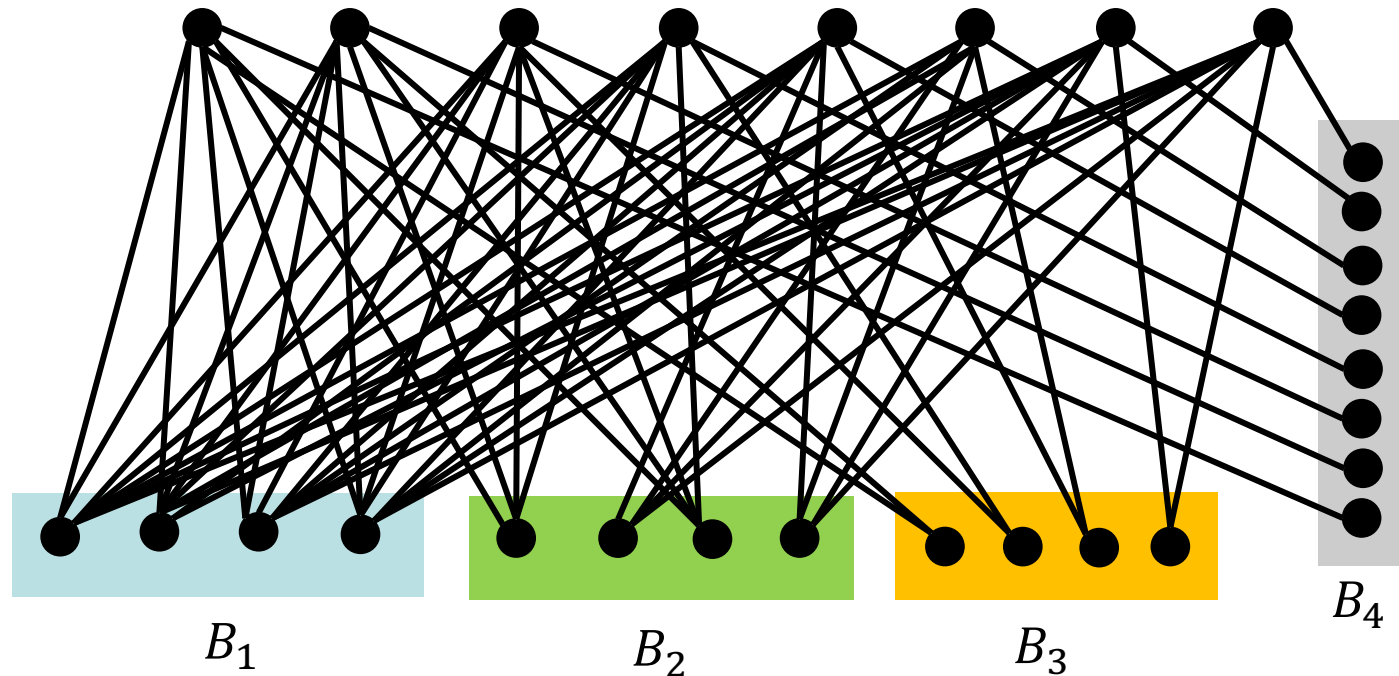
A Different Greedy Rule

Greedy 2: Iteratively, pick **both endpoints** of an uncovered edge.

Vertex cover = 6



Greedy 2: Pick Both endpoints of an uncovered edge



Greedy vertex cover = 16

OPT vertex cover = 8

Greedy (2) gives 2-approximation

Thm: Size of greedy (2) vertex cover is at most twice as big as size of optimal cover

Pf: Suppose Greedy (2) picks endpoints of edges e_1, \dots, e_k . Since these edges do not touch, every valid cover must pick one vertex from each of these edges!

i.e., $OPT \geq k$.

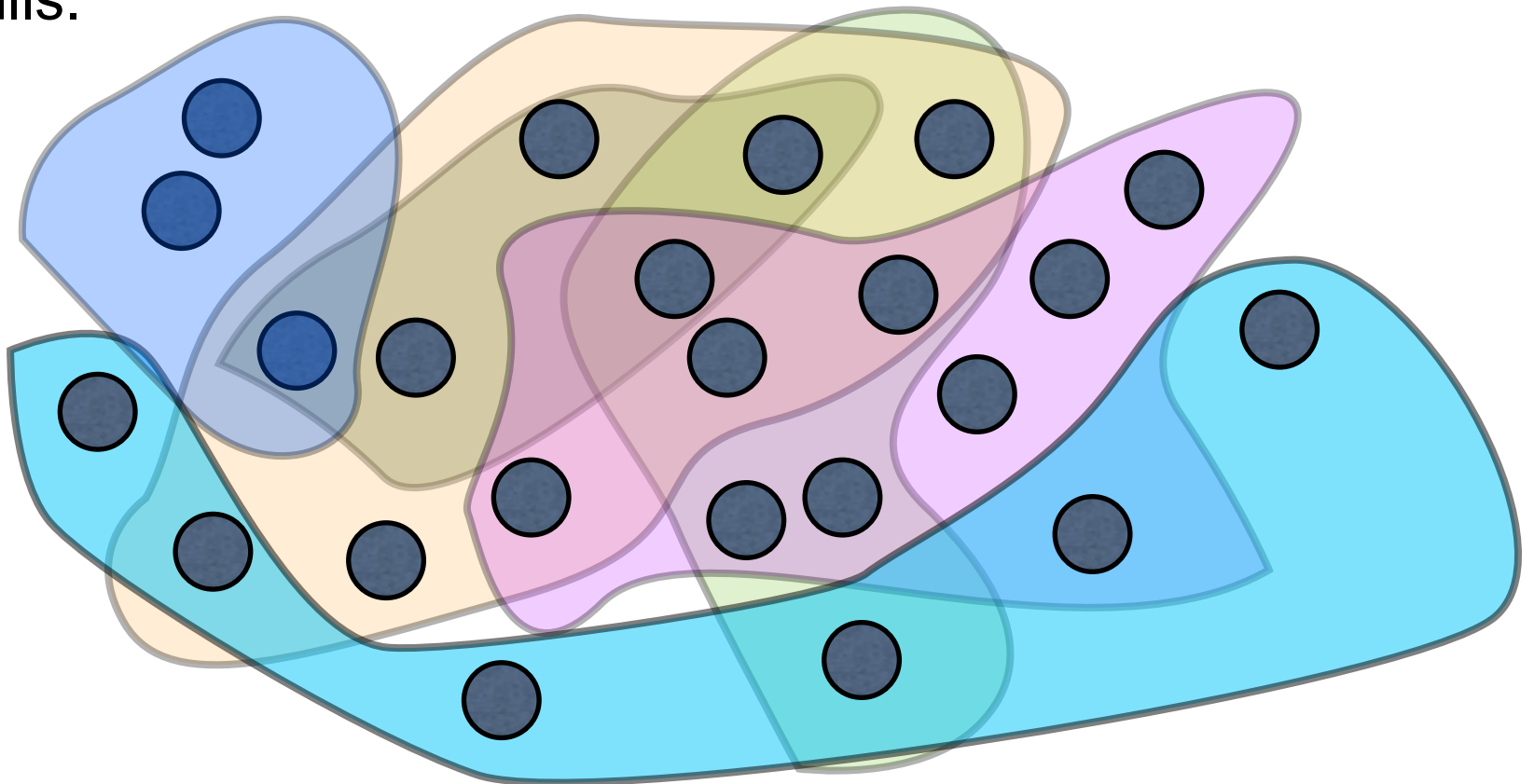
But the size of greedy cover is $2k$. So, Greedy is a 2-approximation.

Set Cover

Given a number of sets on a ground set of elements,

Goal: choose minimum number of sets that cover all.

e.g., a company wants to hire employees with certain skills.

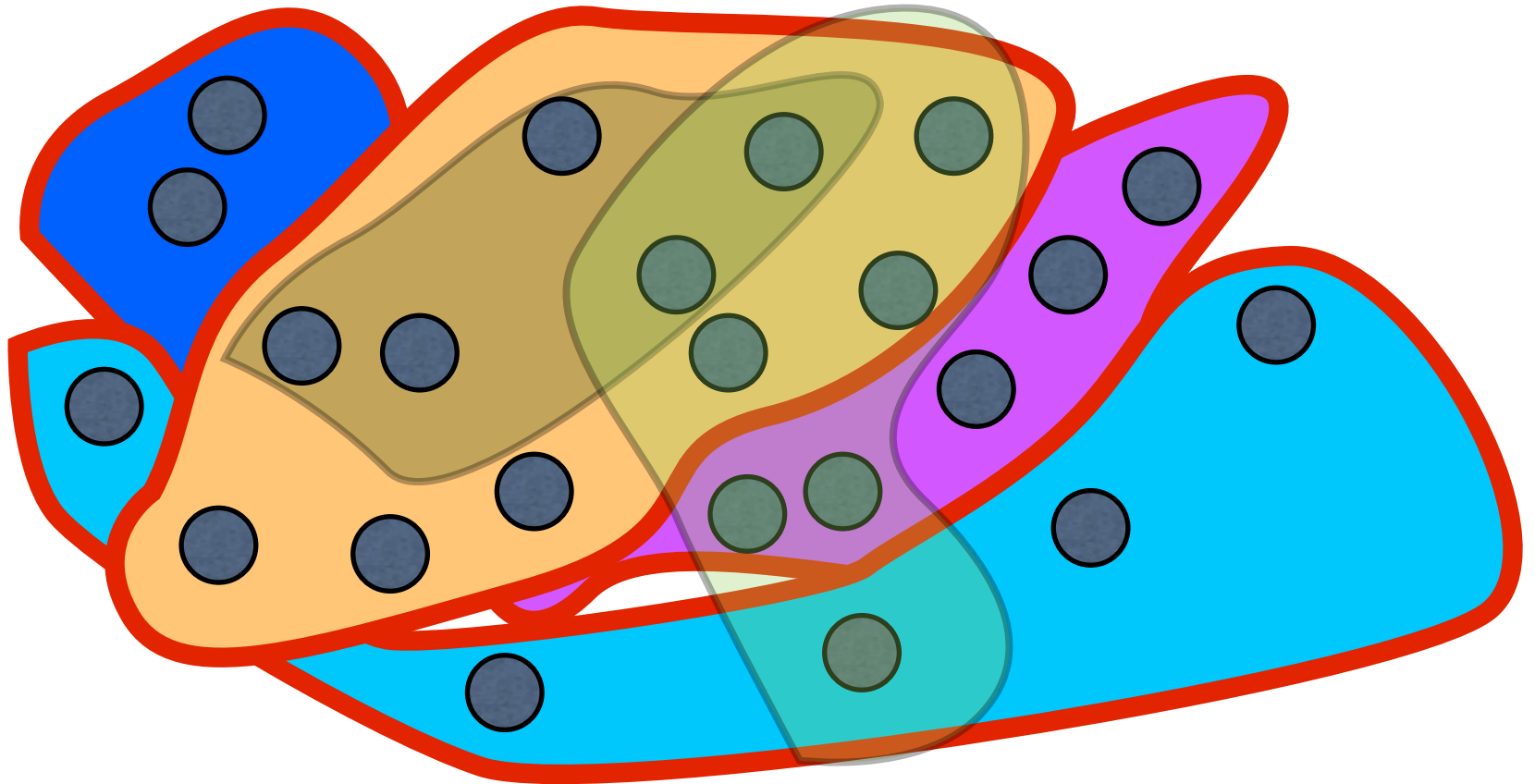


Set Cover

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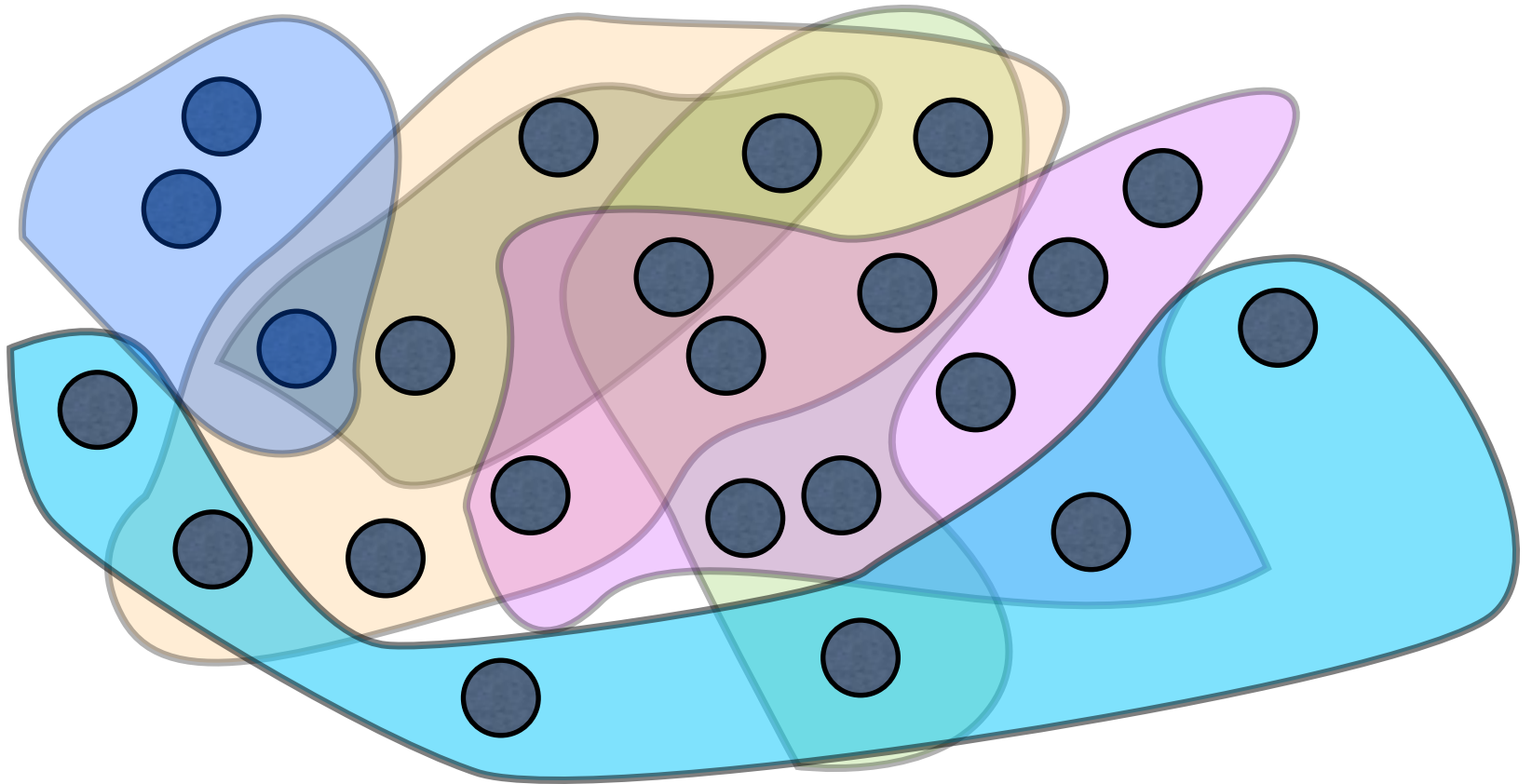
Goal: choose minimum number of sets that cover all.

Set cover = 4



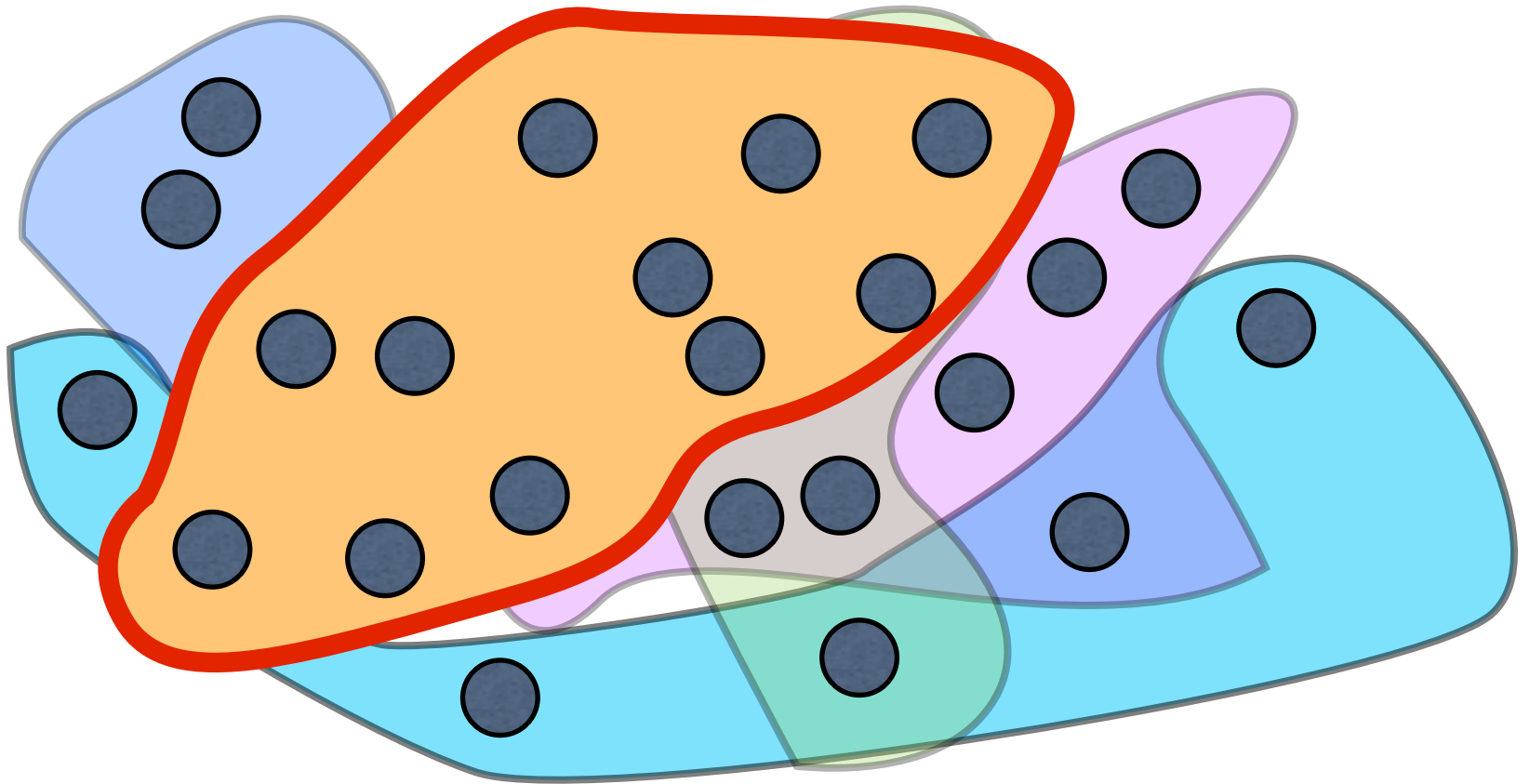
A Greedy Algorithm

Strategy: Pick the set that maximizes # new elements covered



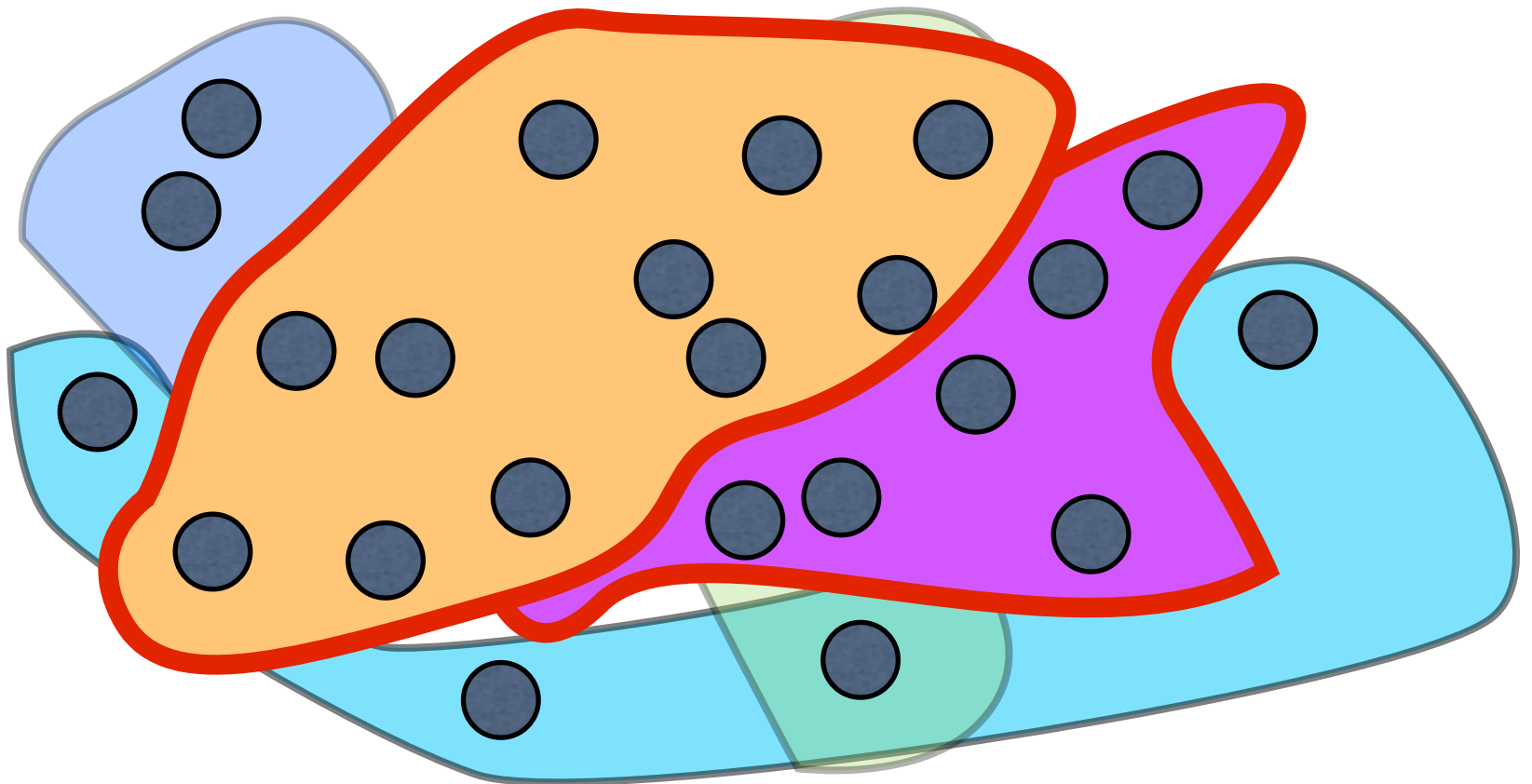
A Greedy Algorithm

Strategy: Pick the set that maximizes # new elements covered



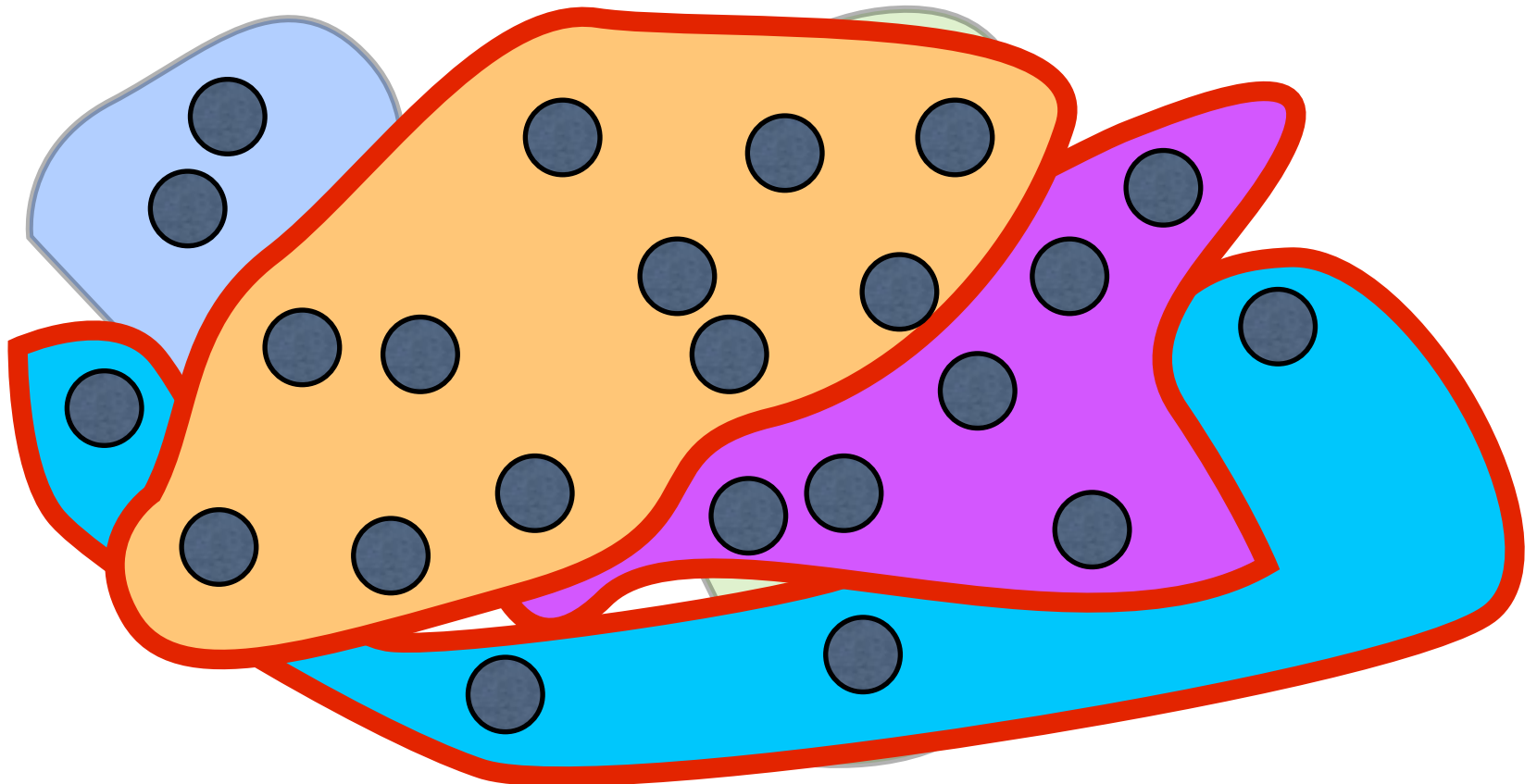
A Greedy Algorithm

Strategy: Pick the set that maximizes # new elements covered



A Greedy Algorithm

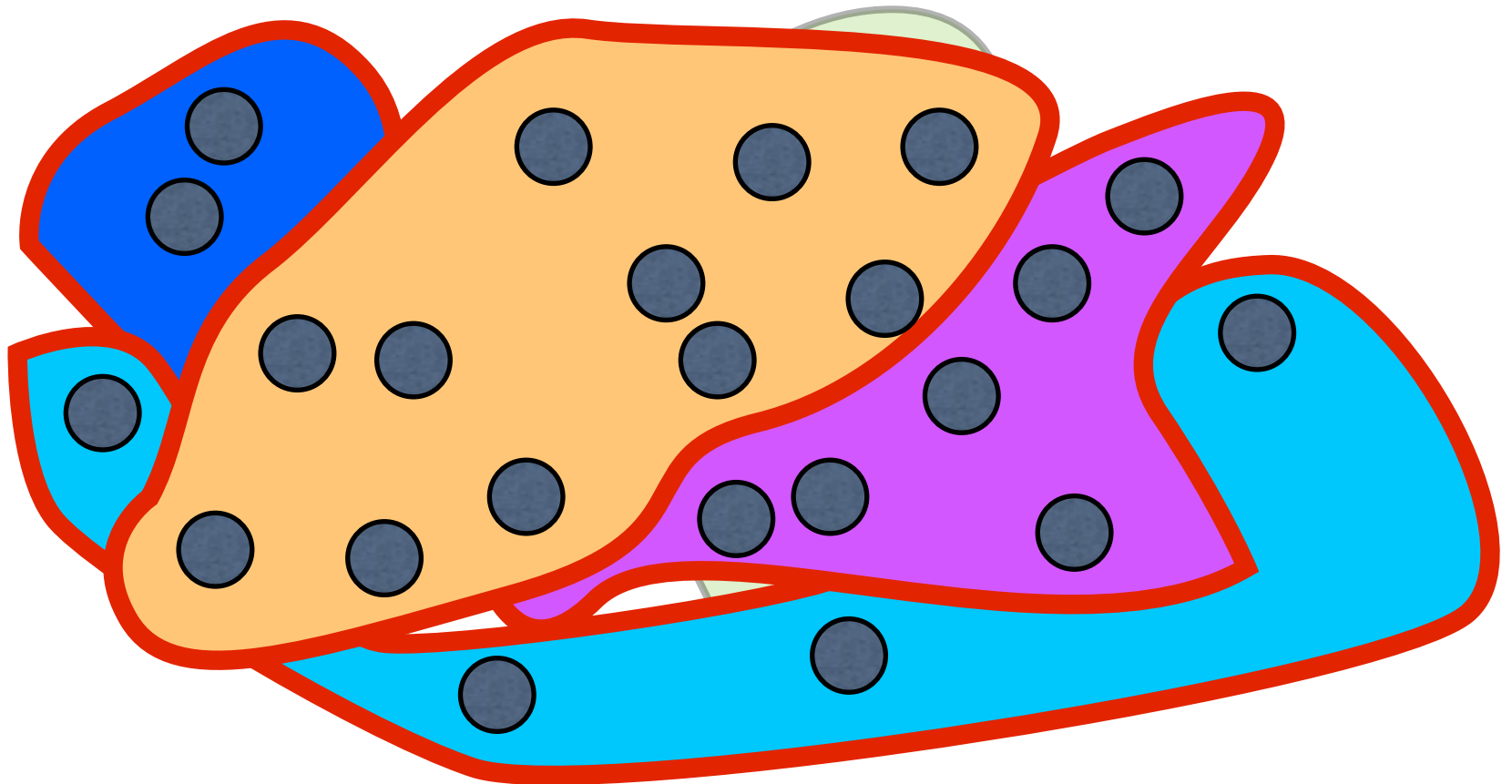
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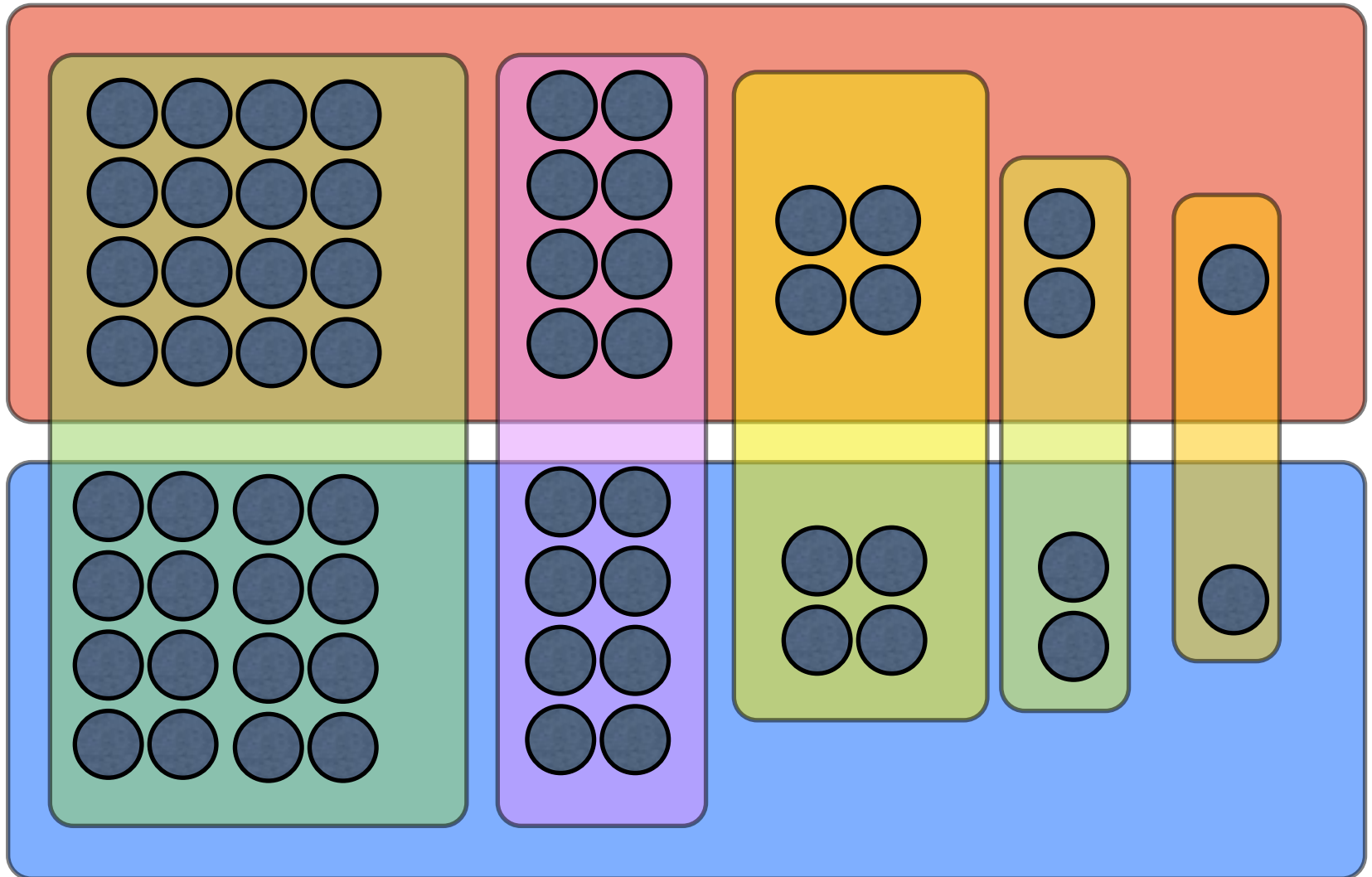
A Greedy Algorithm

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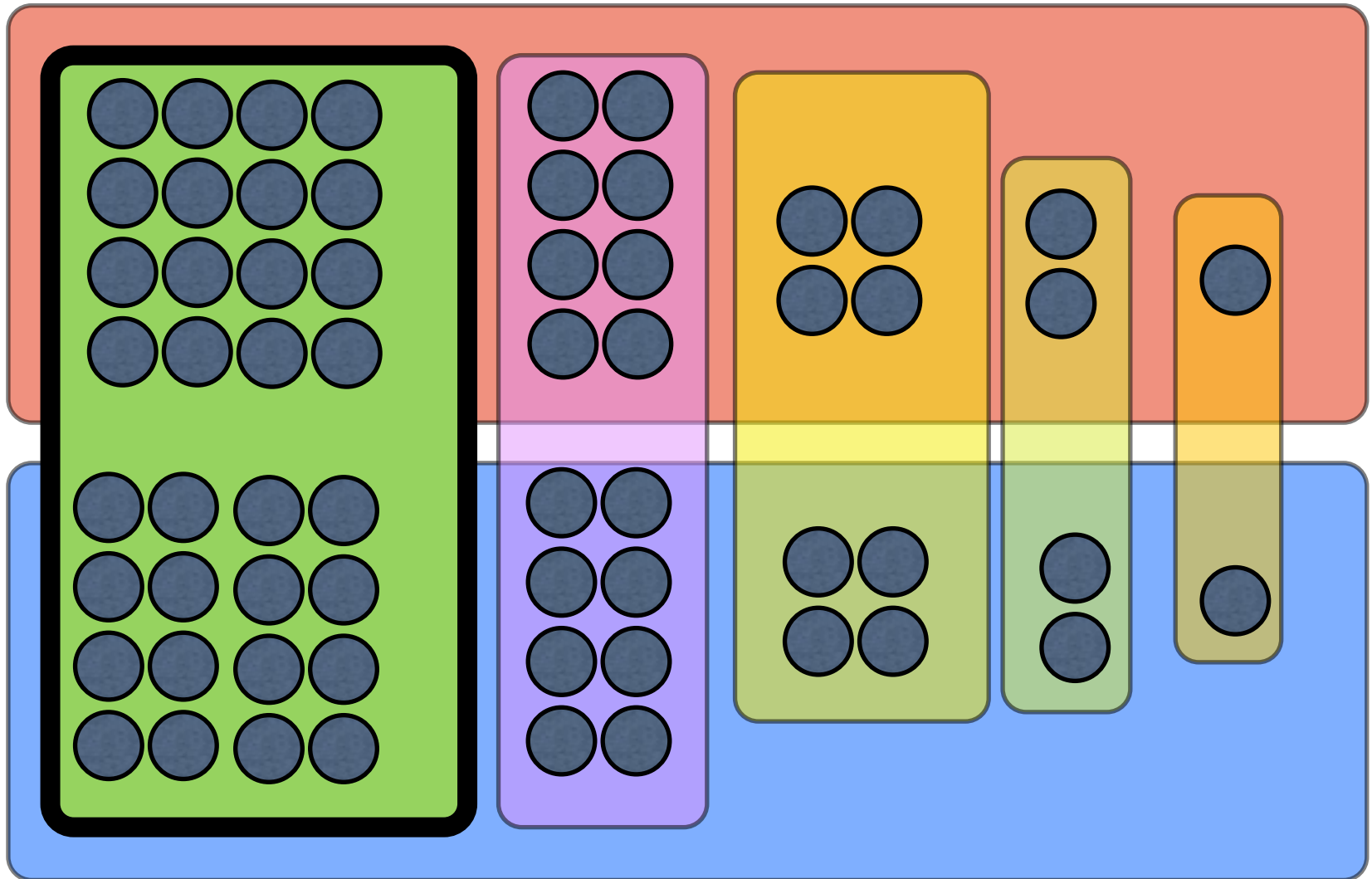
Thm: Greedy has $\ln n$ approximation ratio



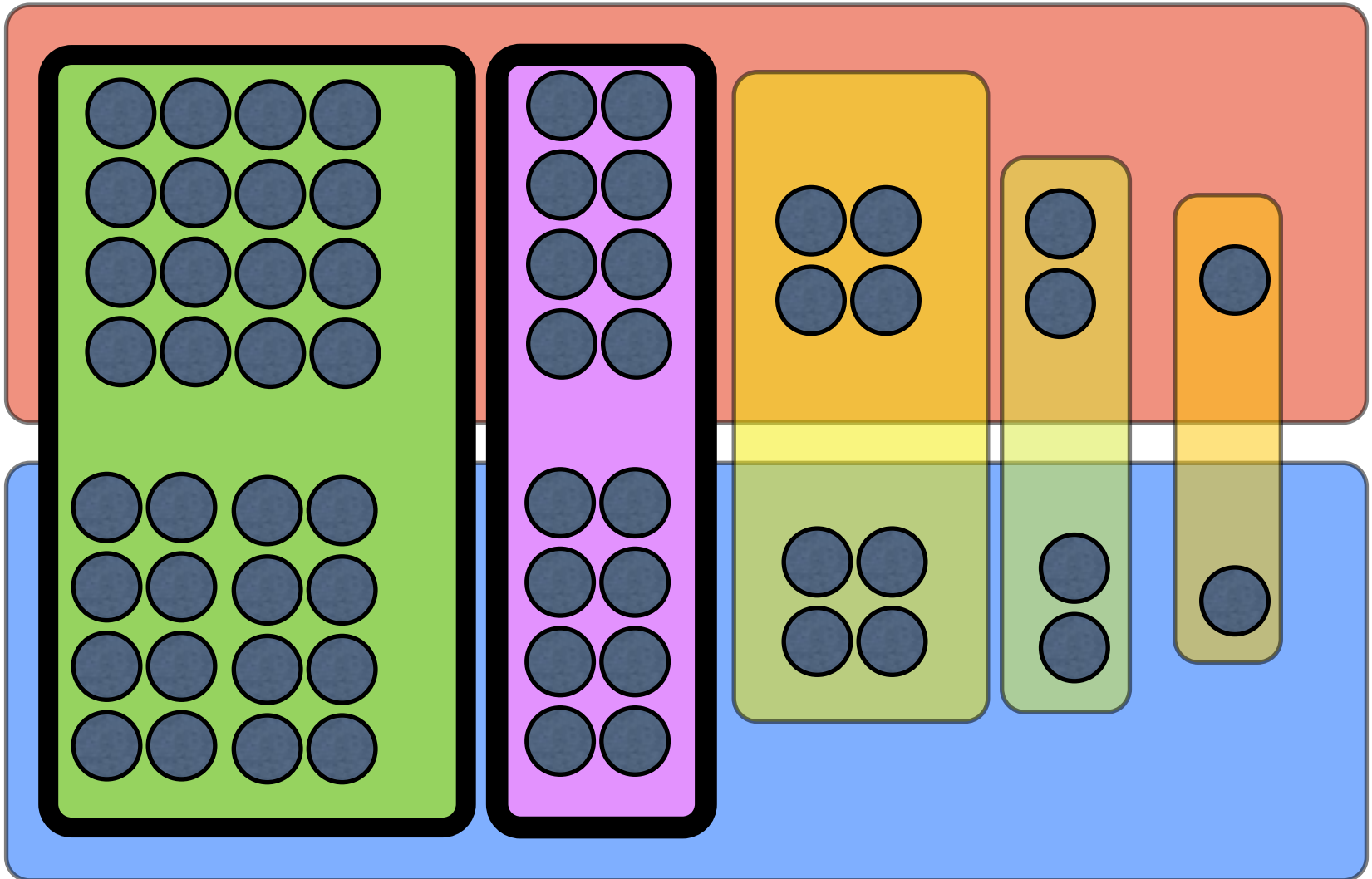
A Tight Example for Greedy



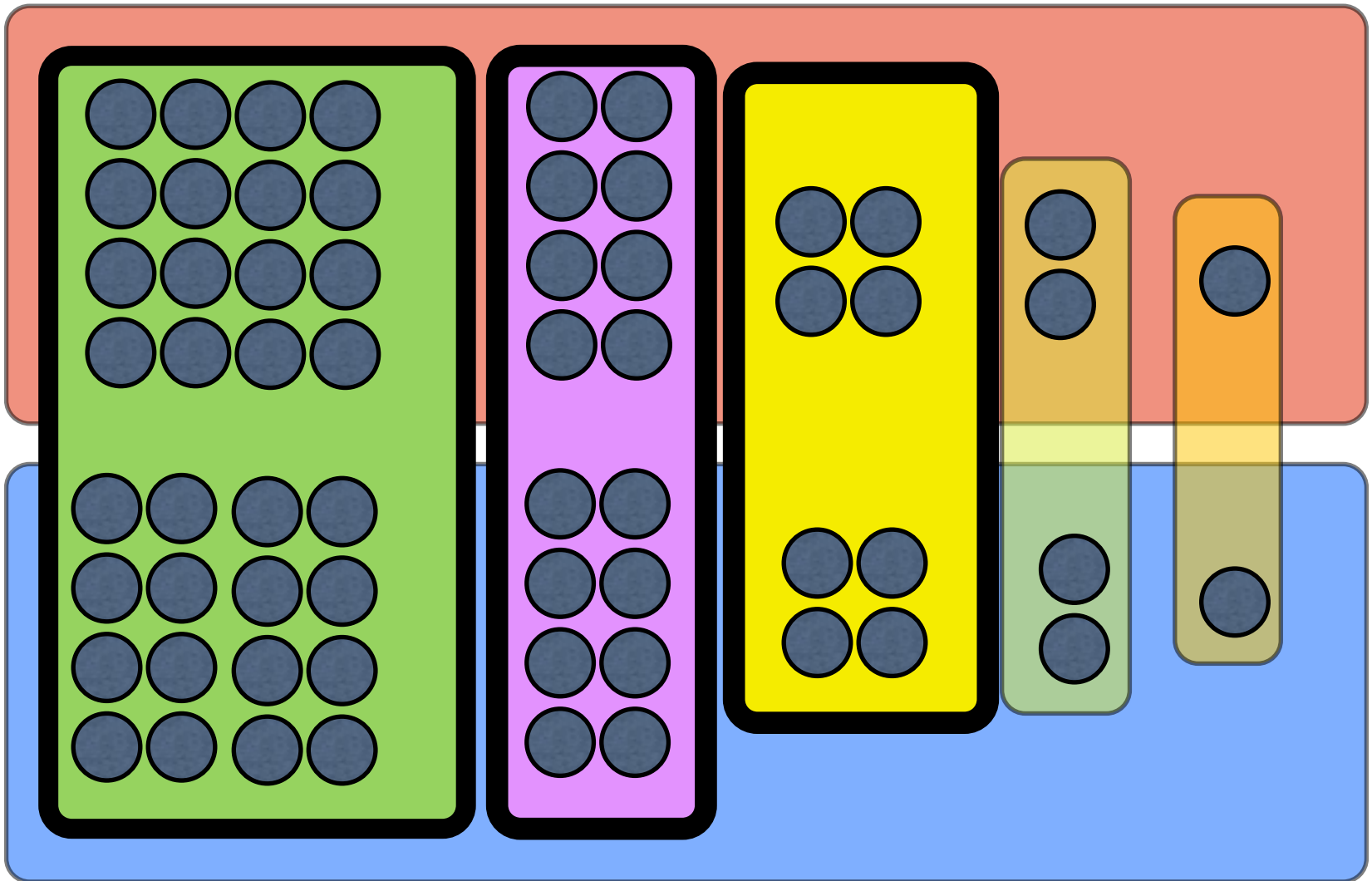
A Tight Example for Greedy



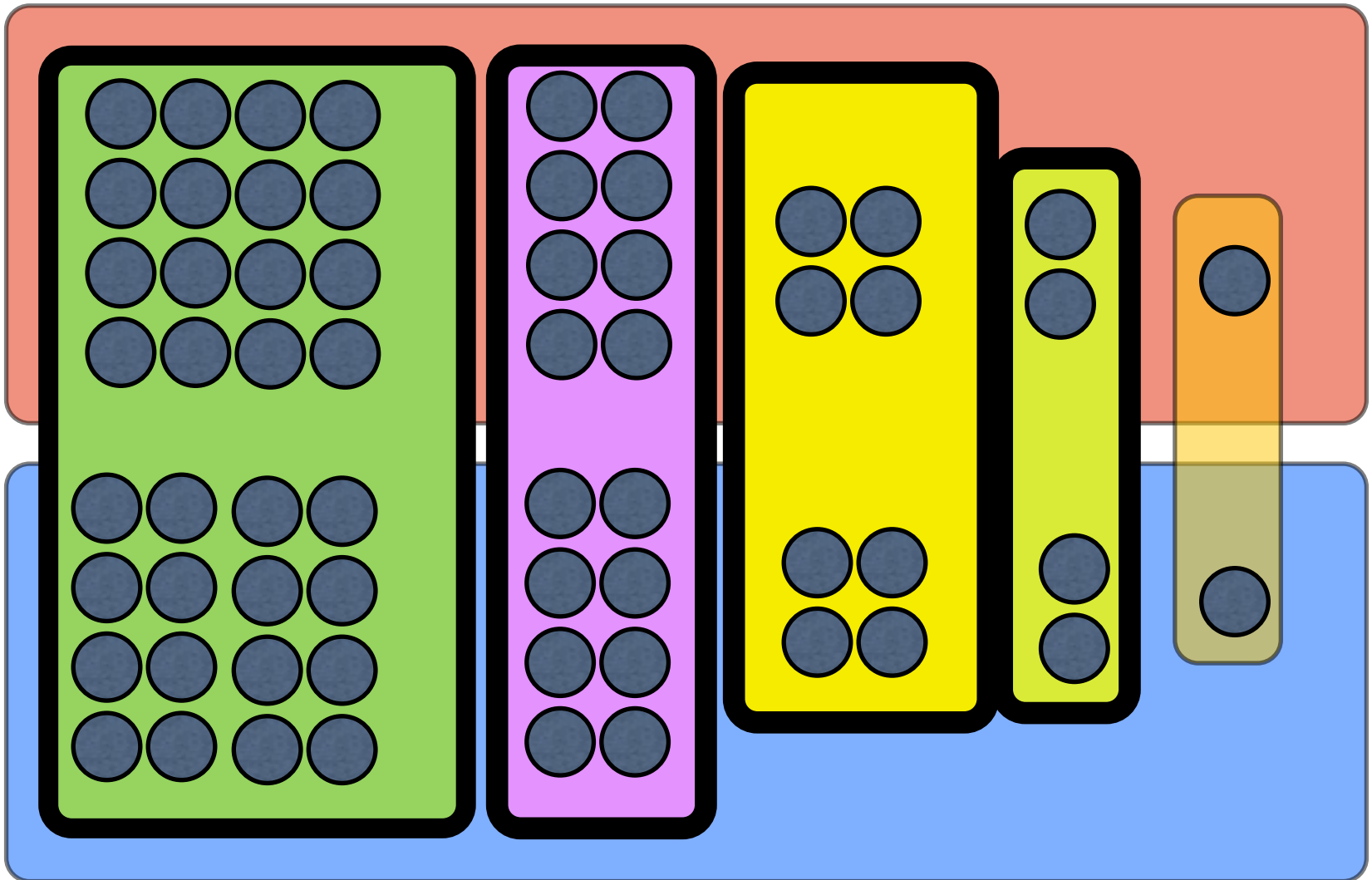
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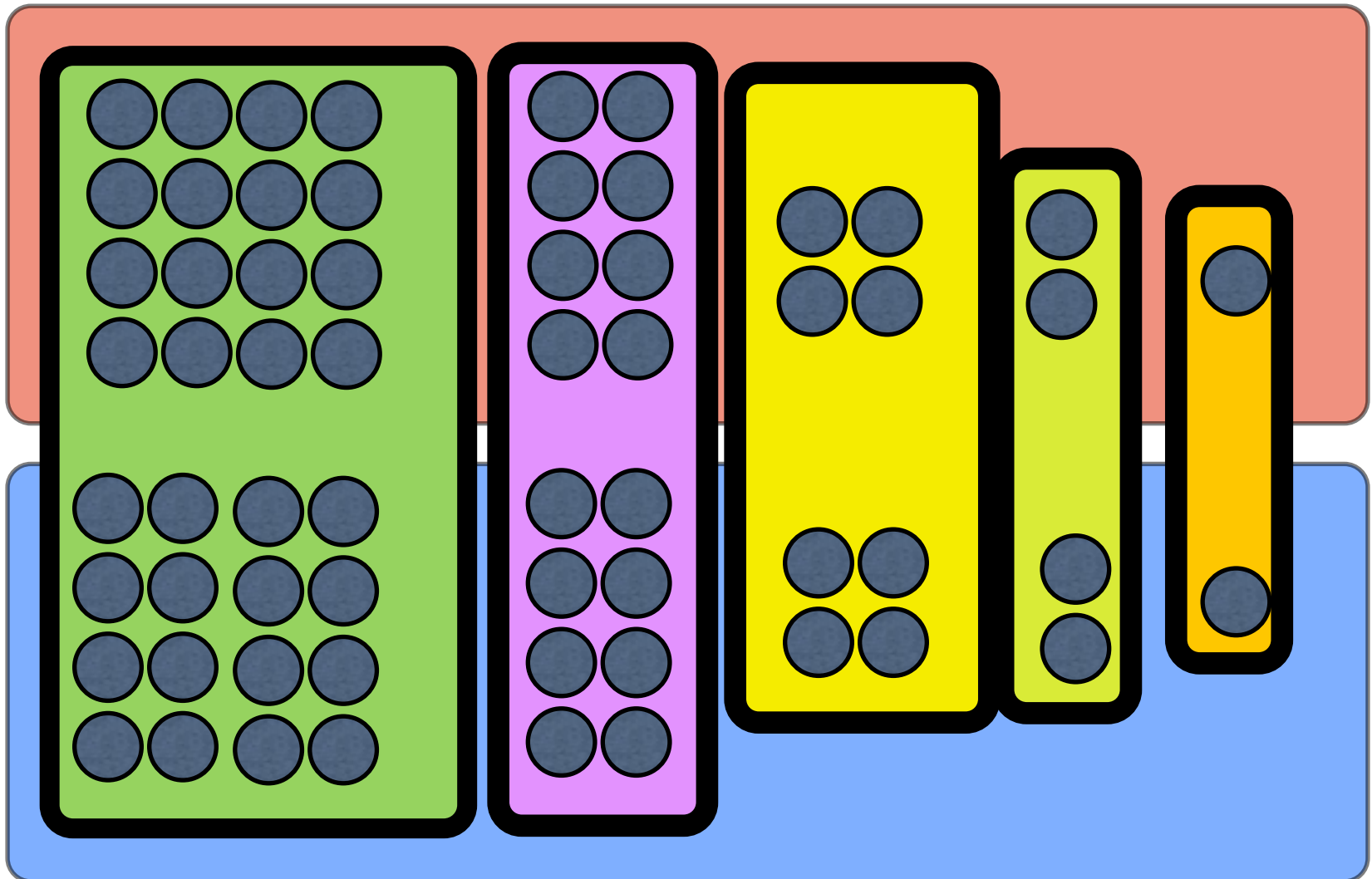
A Tight Example for Greedy



A Tight Example for Greedy

Greedy = 5

OPT = 2



Greedy Gives $O(\log(n))$ approximation

Thm: If the best solution has k sets, greedy finds at most $k \ln(n)$ sets.

Pf: Suppose $OPT=k$

There is set that covers $1/k$ fraction of remaining elements, since there are k sets that cover all remaining elements.

So **in each step**, algorithm will cover $1/k$ fraction of remaining elements.

#elements uncovered after t steps

$$\leq n \left(1 - \frac{1}{k}\right)^t \leq n e^{-\frac{t}{k}}$$

So after $t = k \ln n$ steps, # uncovered elements < 1 .