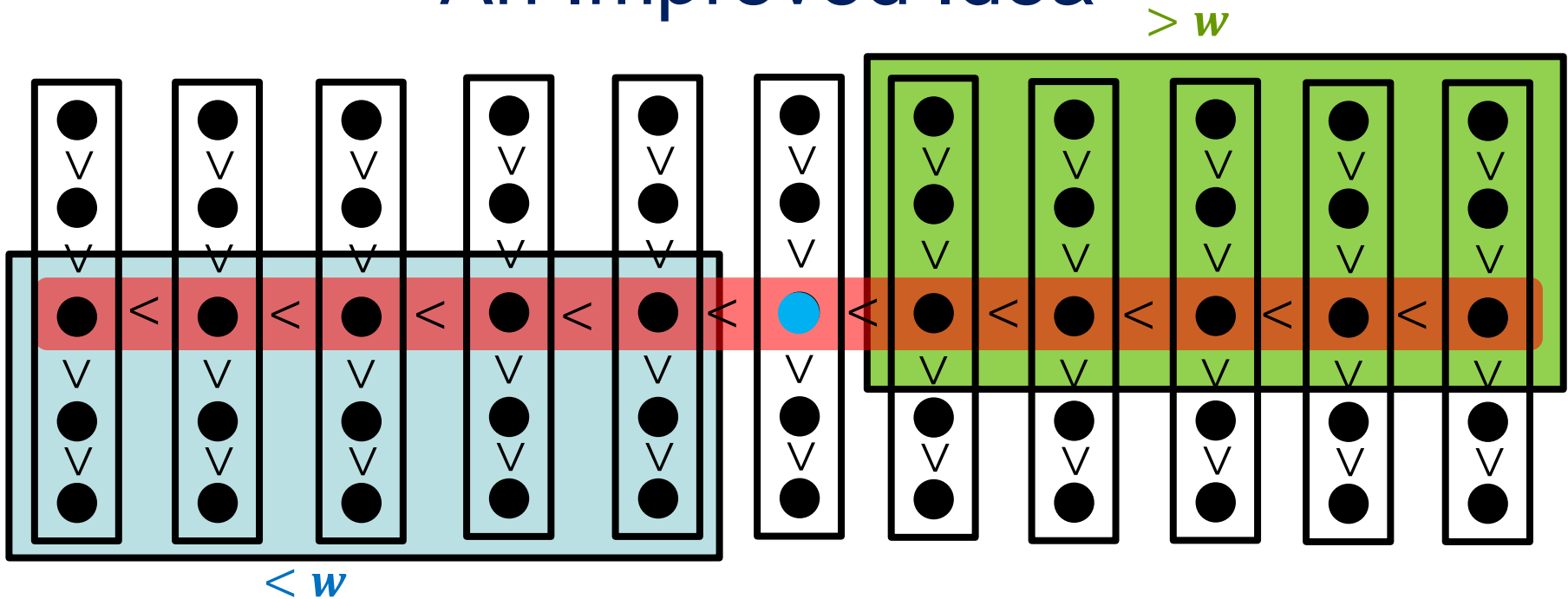


# **CSE 421**

## **Vertex Cover / Set Cover**

Shayan Oveis Gharan

# An Improved Idea



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


- $|S_{<}(w)| \geq 3 \left(\frac{n}{10}\right) = \frac{3n}{10}$
  - $|S_{>}(w)| \geq 3 \left(\frac{n}{10}\right) = \frac{3n}{10}$
- $\frac{3n}{10} \leq |S_{<}(w)|, |S_{>}(w)| \leq \frac{7n}{10}$

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

# An Improved Idea

```
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



# D&C Summary

## Idea:

“Two halves are better than a whole”

- if the base algorithm has super-linear complexity.

“If a little's good, then more's better”

- repeat above, recursively
- Applications: Many.
  - Binary Search, Merge Sort, (Quicksort),
  - Root of a Function
  - Closest points,
  - Integer multiplication
  - Median
  - Matrix Multiplication

# 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.

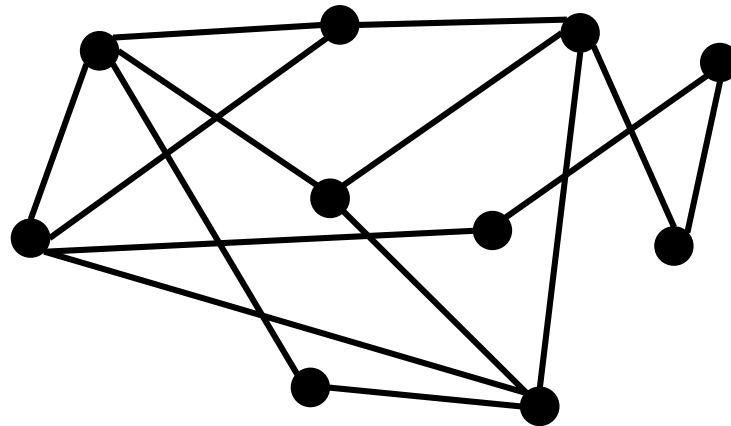
$$\alpha = \frac{\text{Cost of computed solution}}{\text{Cost of the optimum}}$$

**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?

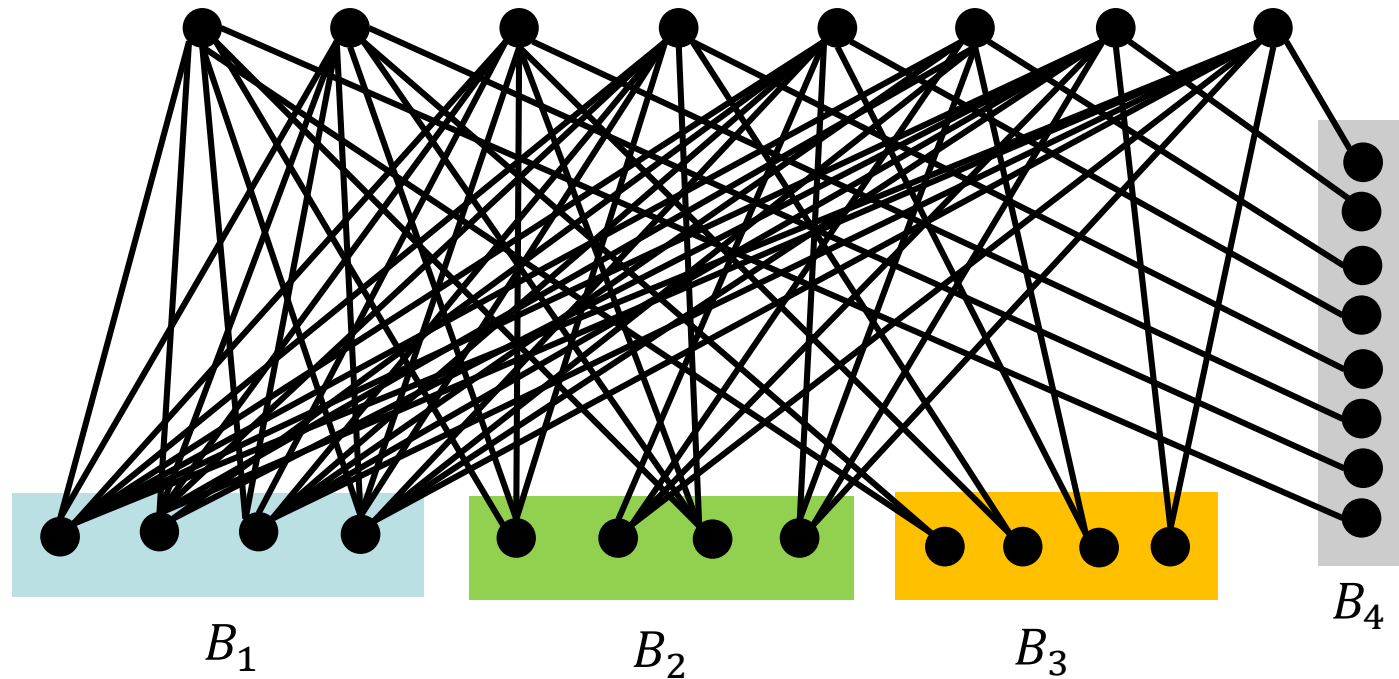
Greedy algorithms are typically used in practice to find a (good) solution to NP-hard problems

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

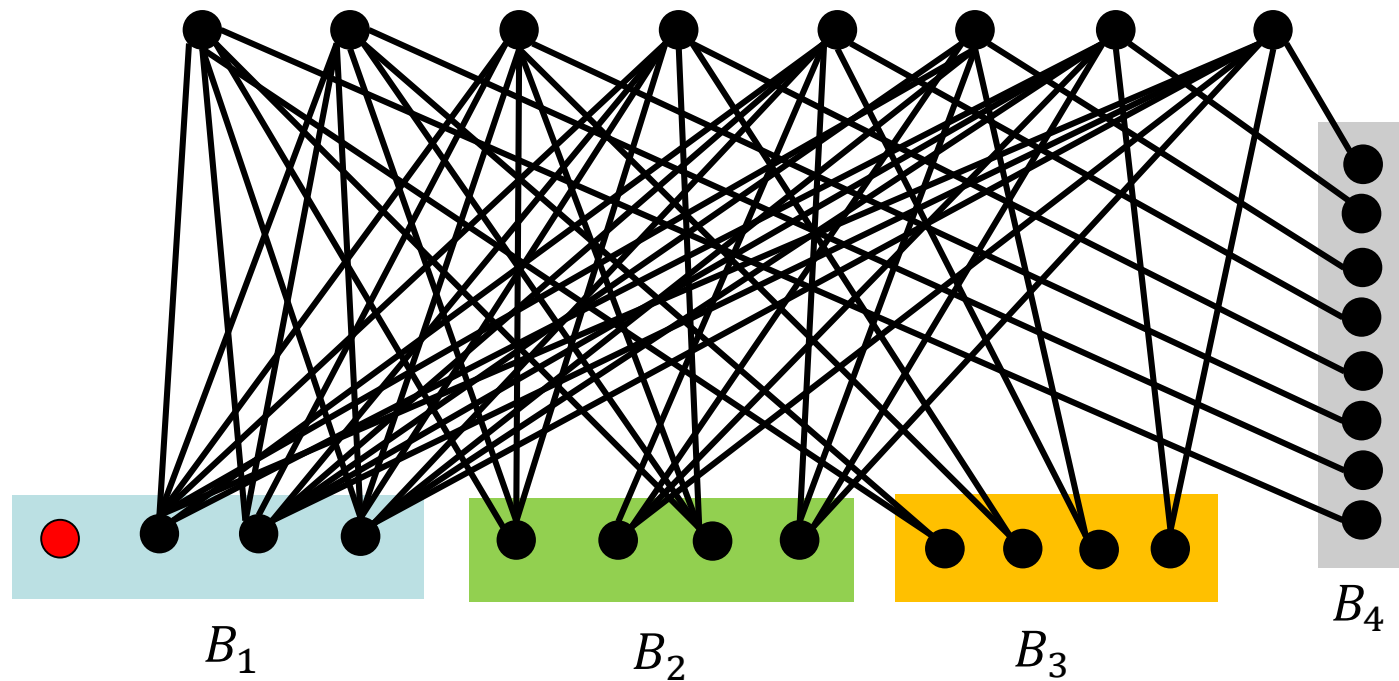
Q: Does this give an optimum solution?

A: No,

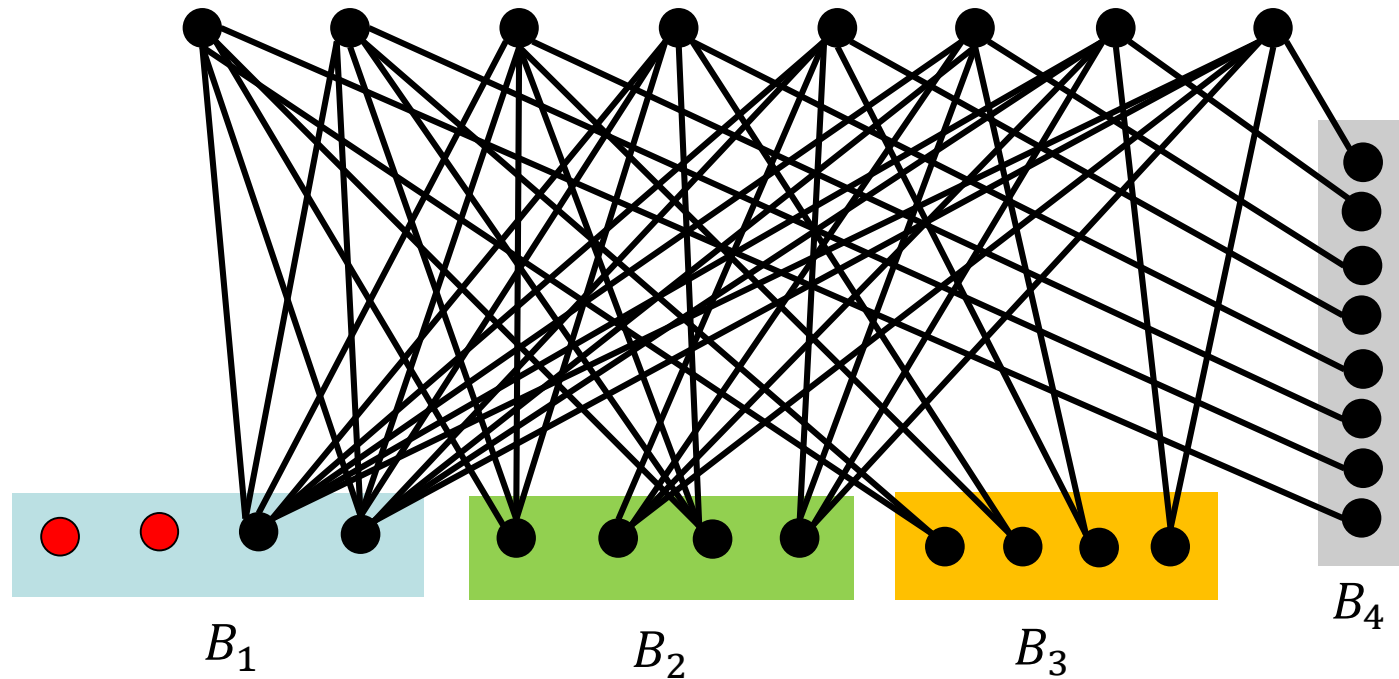
# Greedy (1): Pick vertex that covers the most



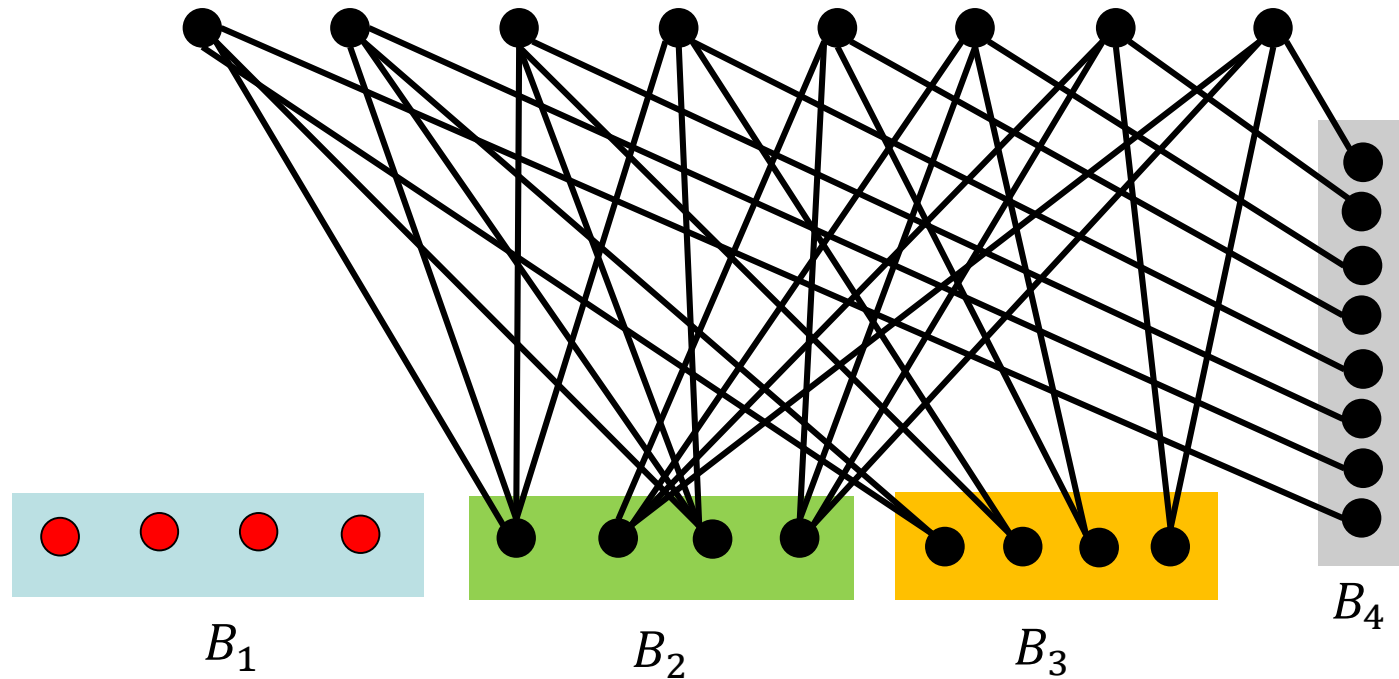
# Greedy (1): Pick vertex that covers the most



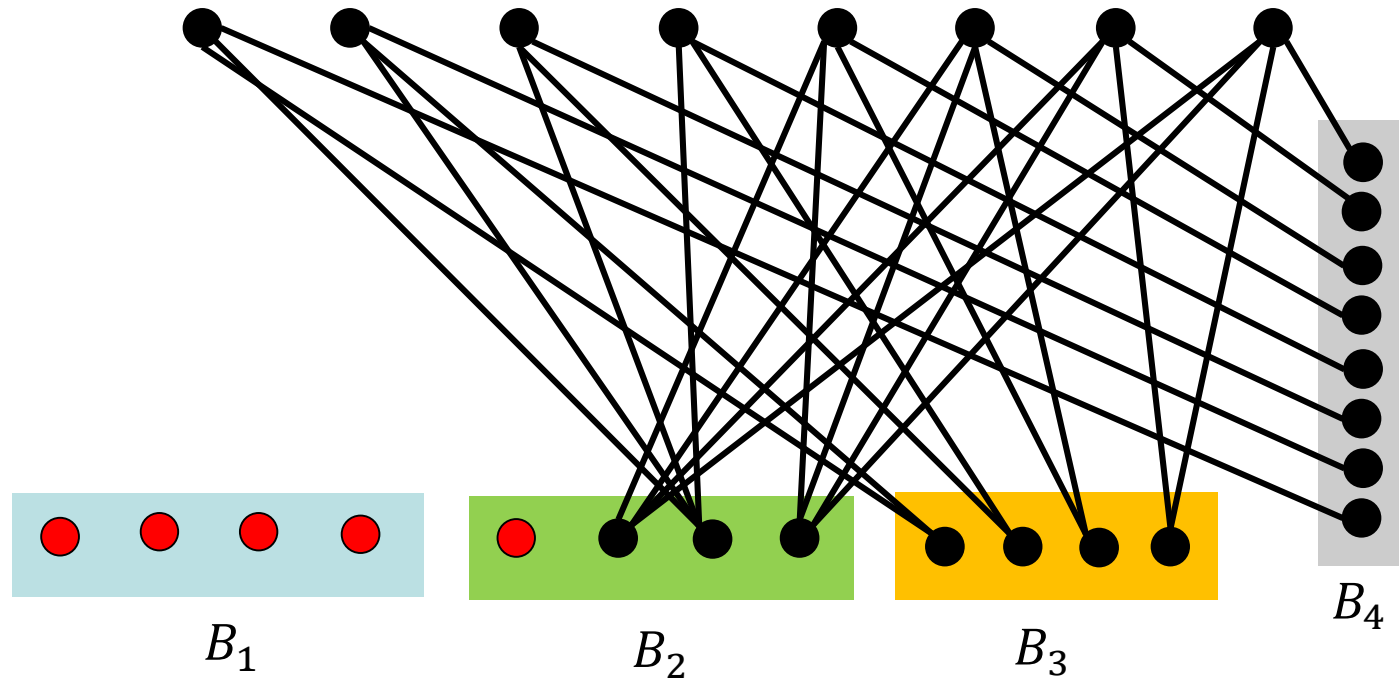
# Greedy (1): Pick vertex that covers the most



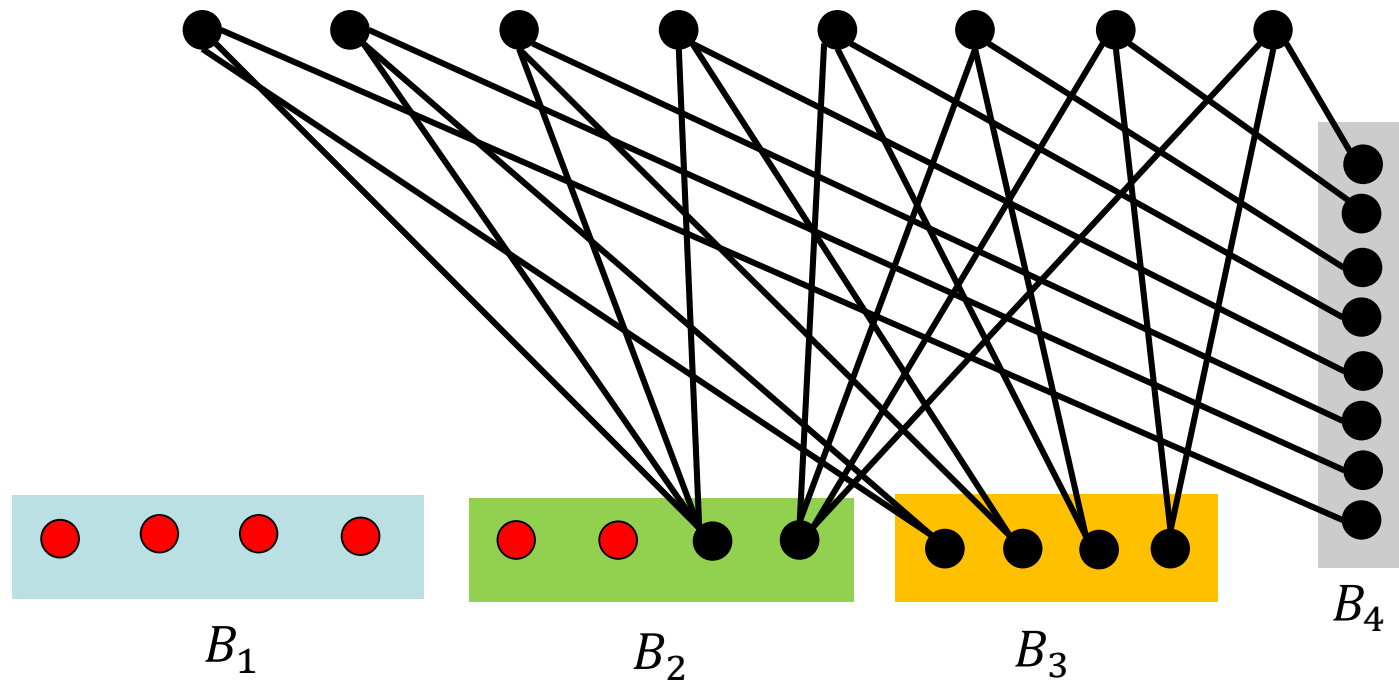
# Greedy (1): Pick vertex that covers the most



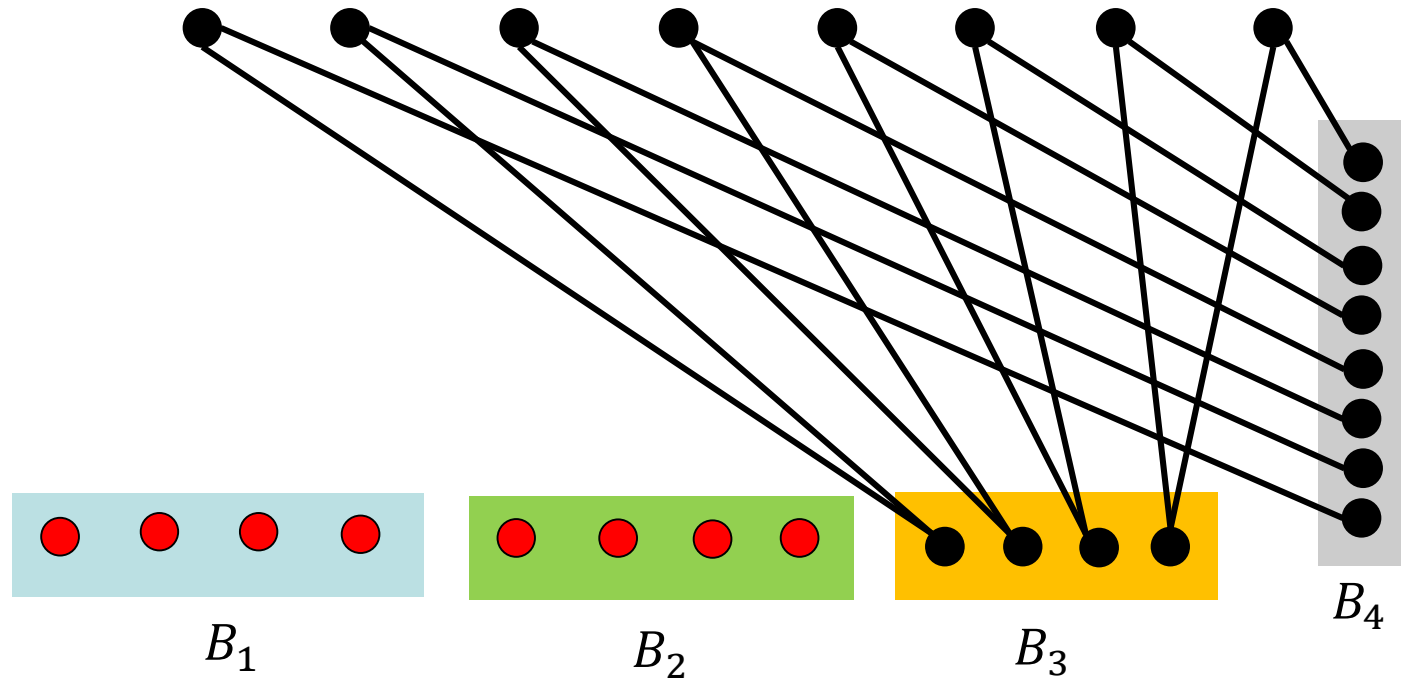
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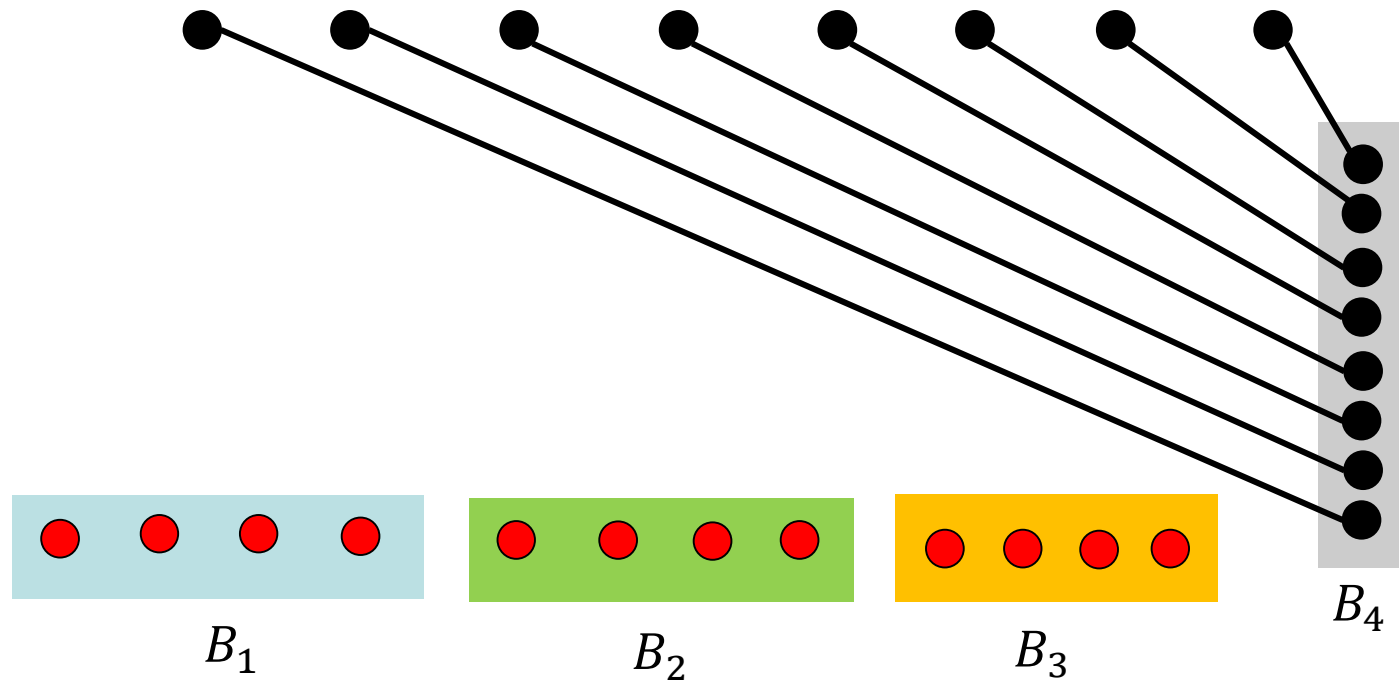


# Greedy (1): Pick vertex that covers the most

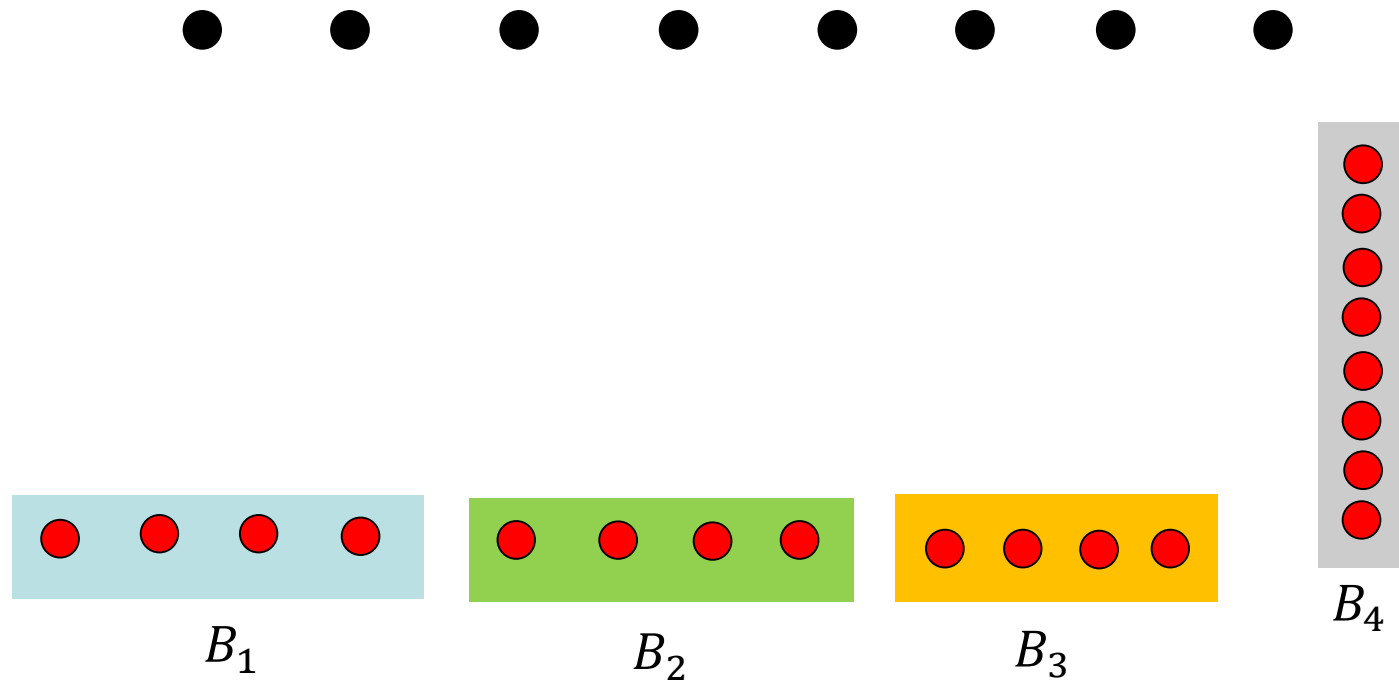




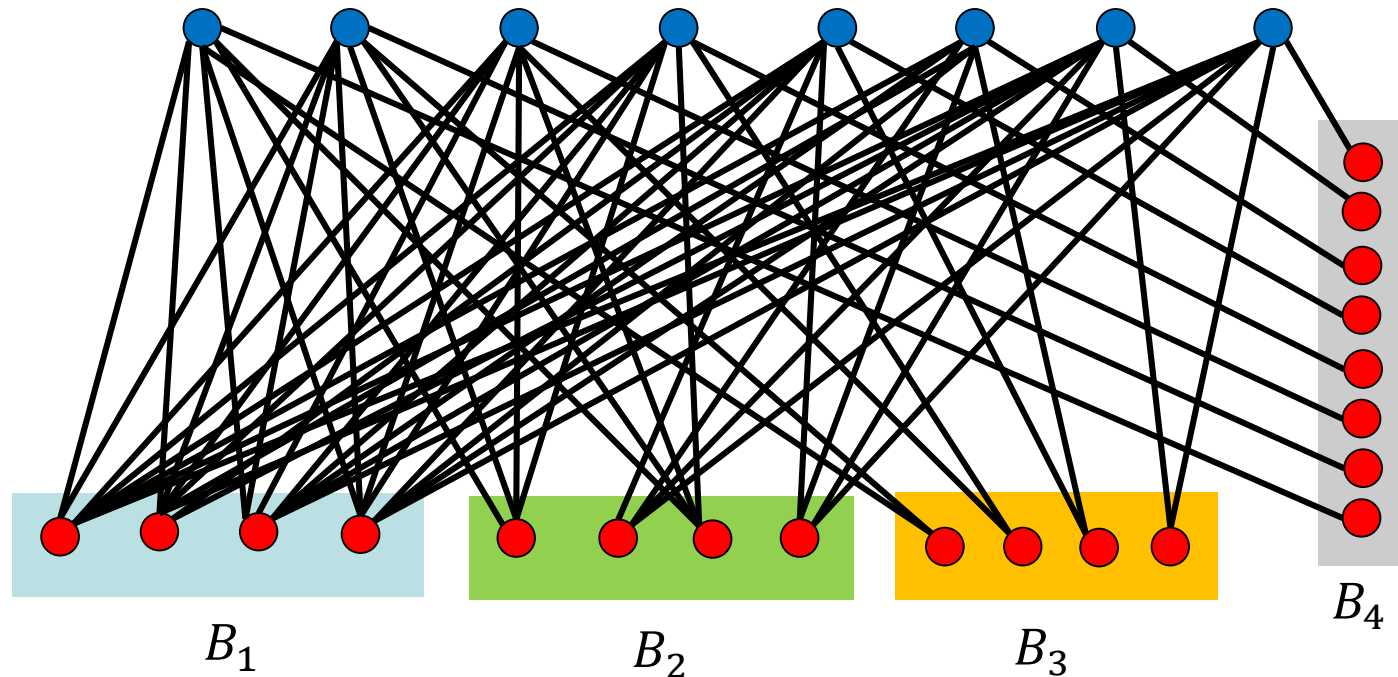
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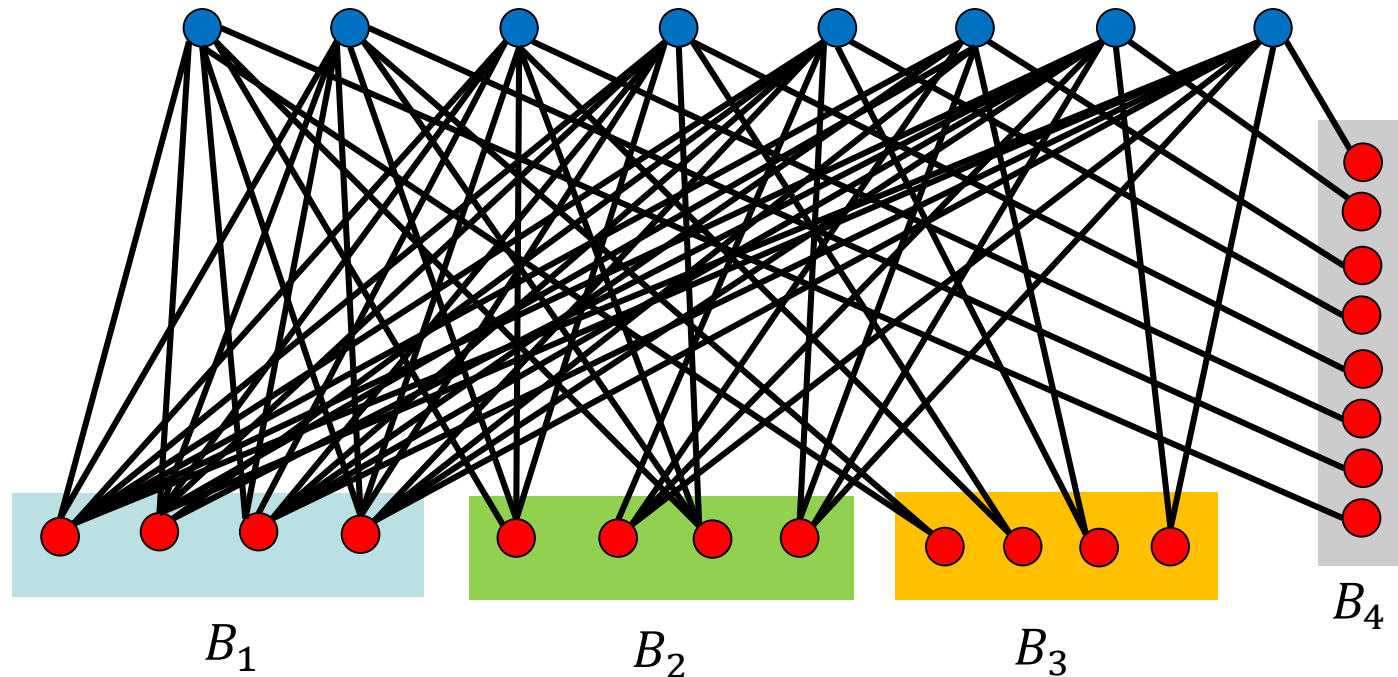
# Greedy (1): Pick vertex that covers the most



Greedy Vertex cover = 20

OPT Vertex cover = 8

# Greedy (1): Pick vertex that covers the most

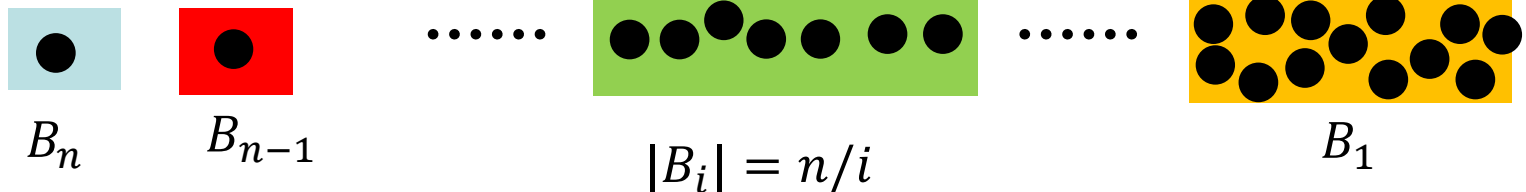
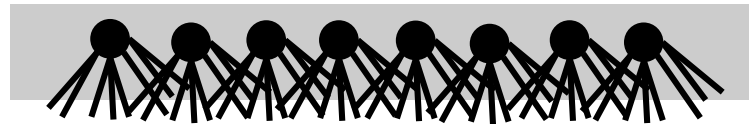


Greedy Vertex cover = 20

OPT Vertex cover = 8

# Greedy (1): Pick vertex that covers the most

$n$  vertices. Each vertex has one edge into each  $B_i$



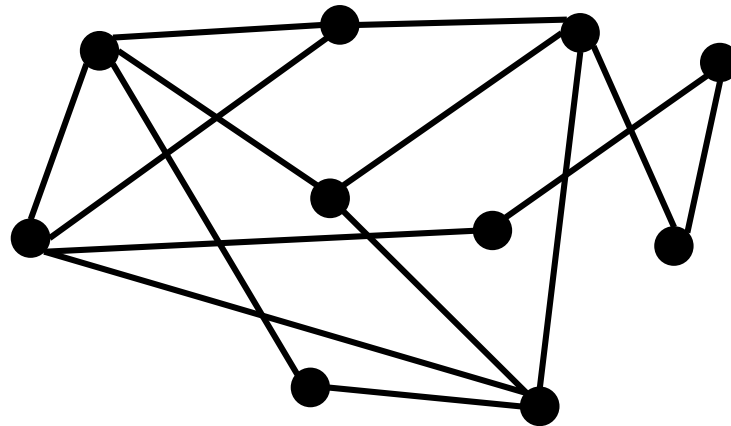
Greedy pick bottom vertices =  $n + \frac{n}{2} + \frac{n}{3} + \dots + 1 \approx n \ln n$

OPT pick top vertices =  $n$

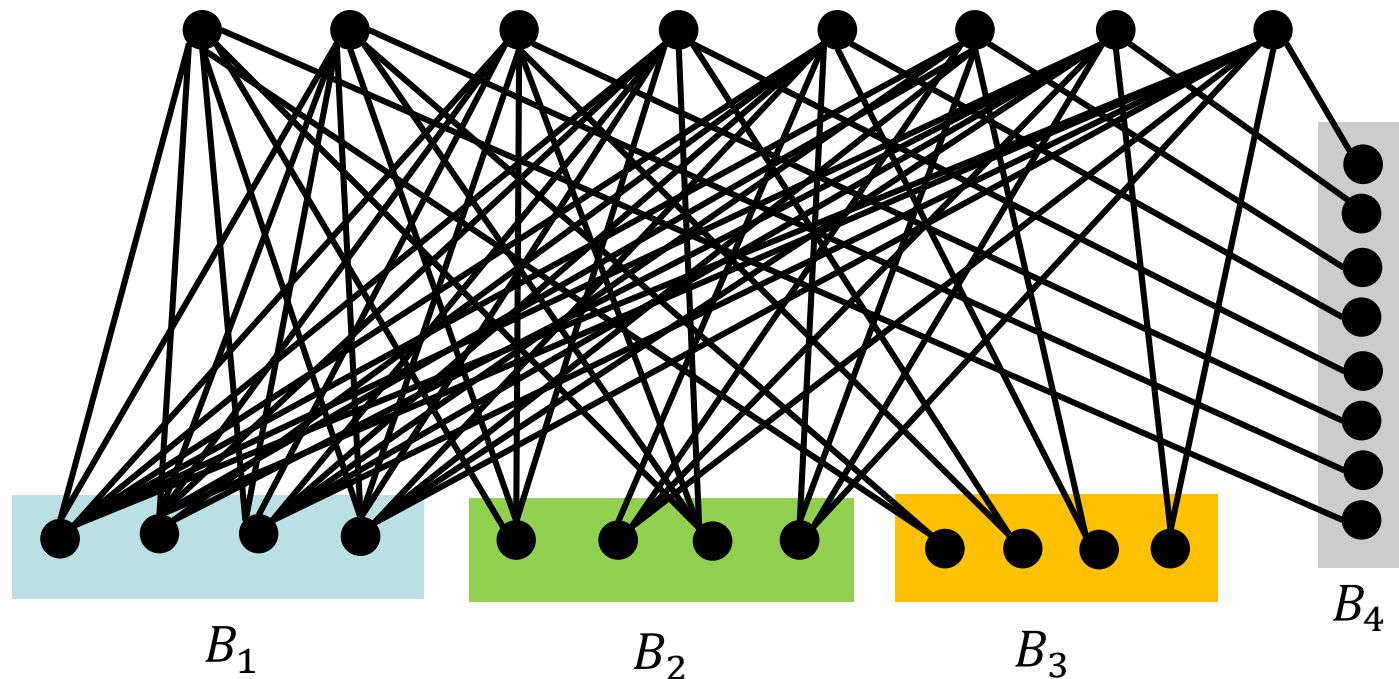
# A Different Greedy Rule

**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.

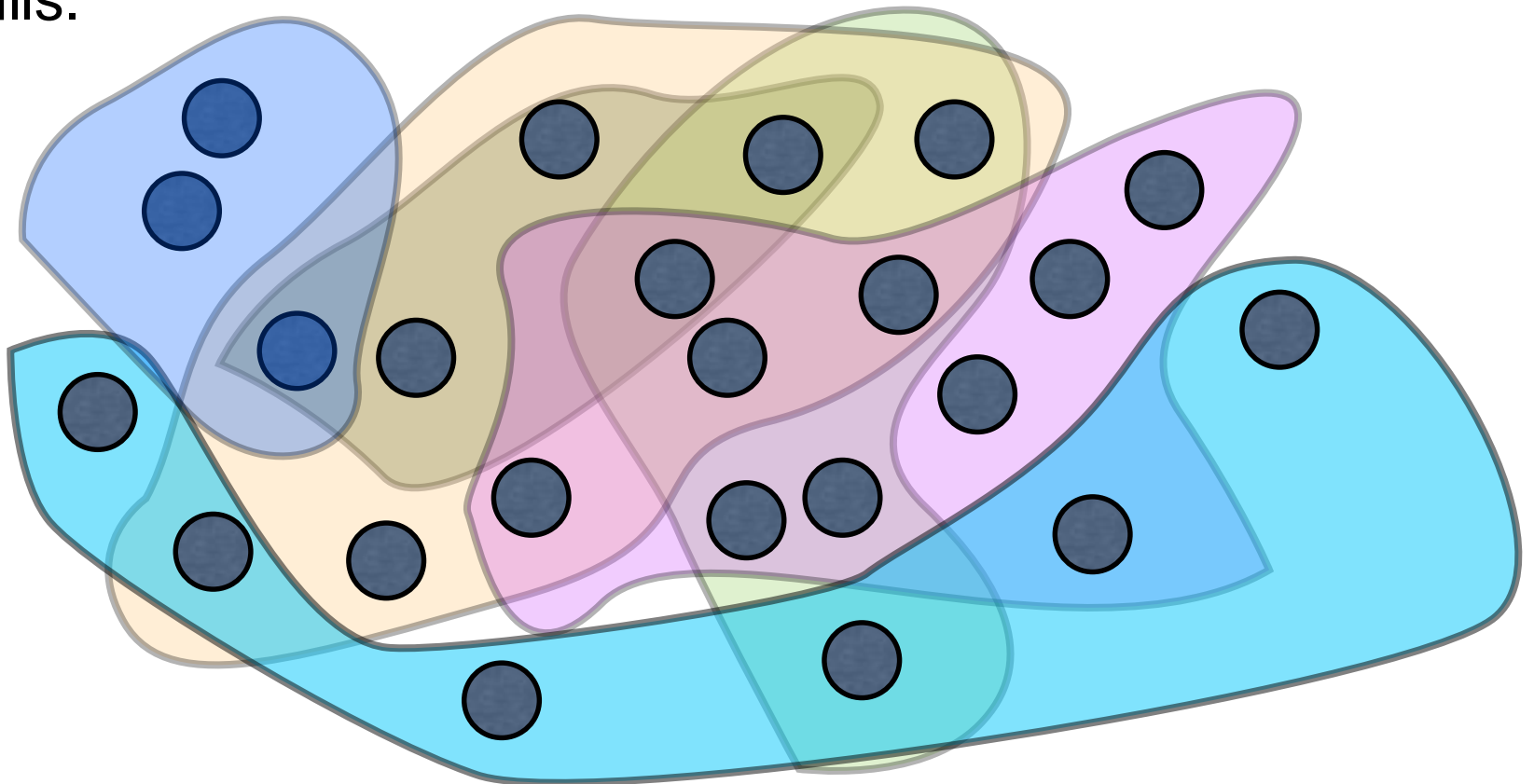


# Set Cover

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.

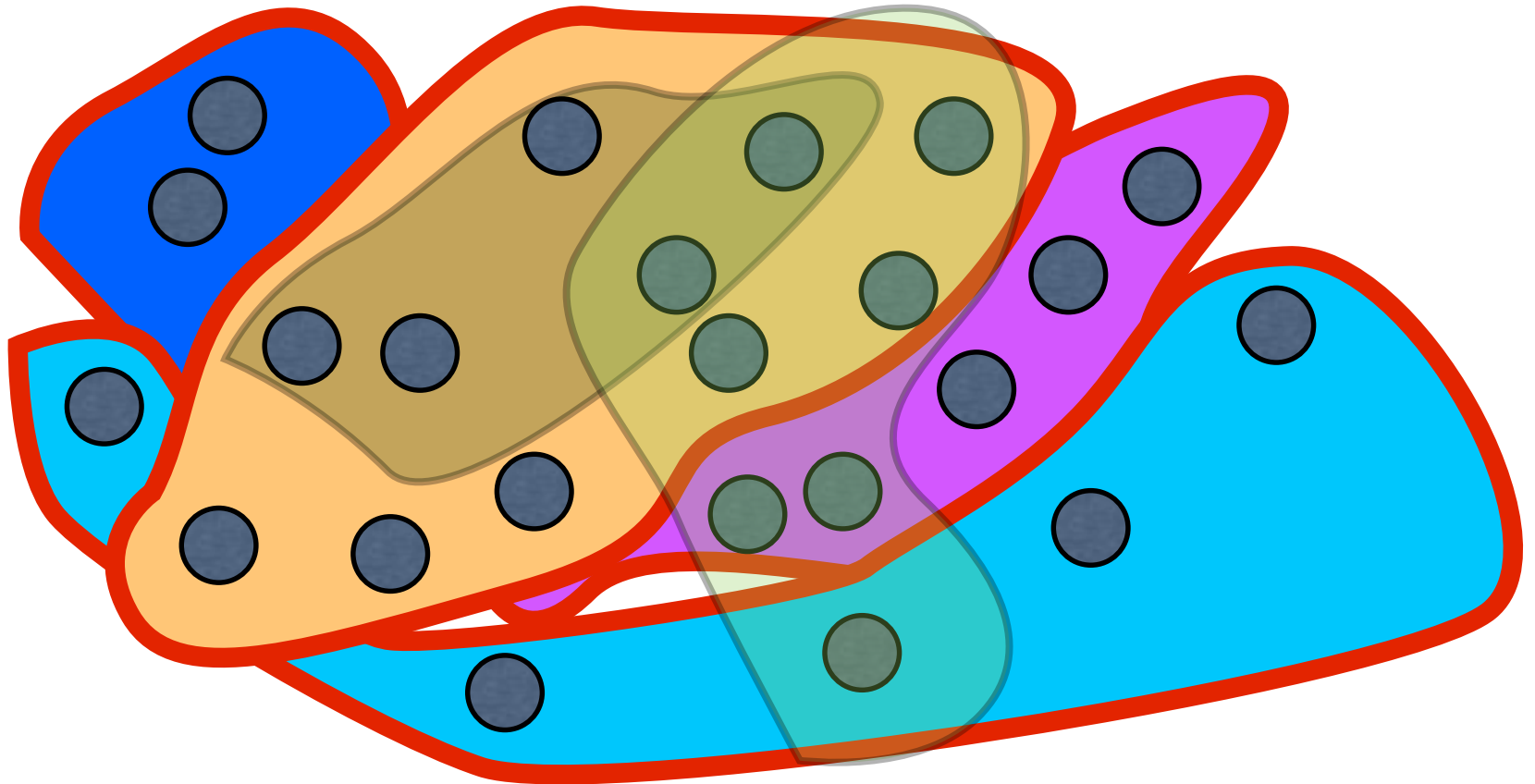


# Set Cover

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

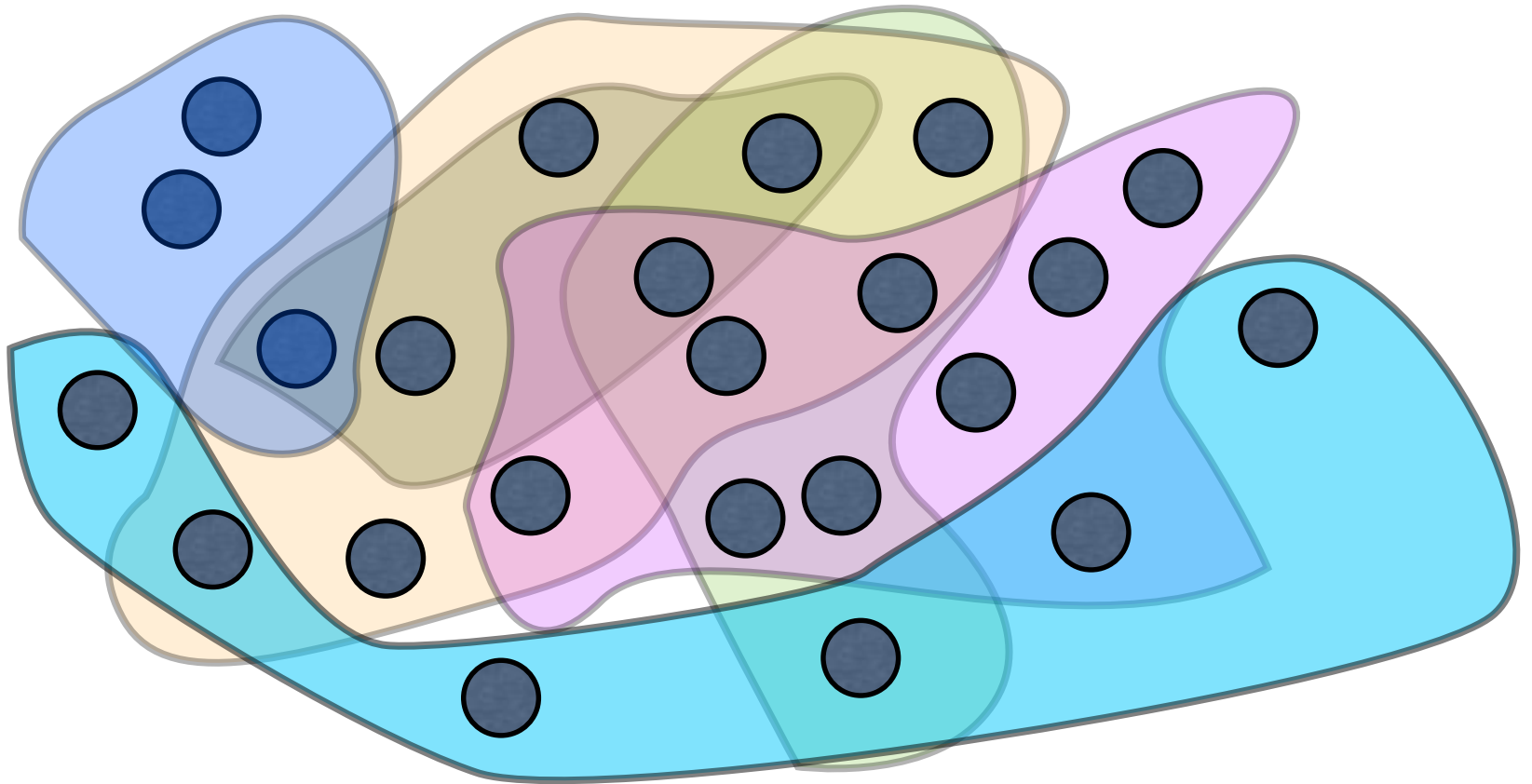
**Goal:** choose minimum number of sets that cover all.

Set cover = 4



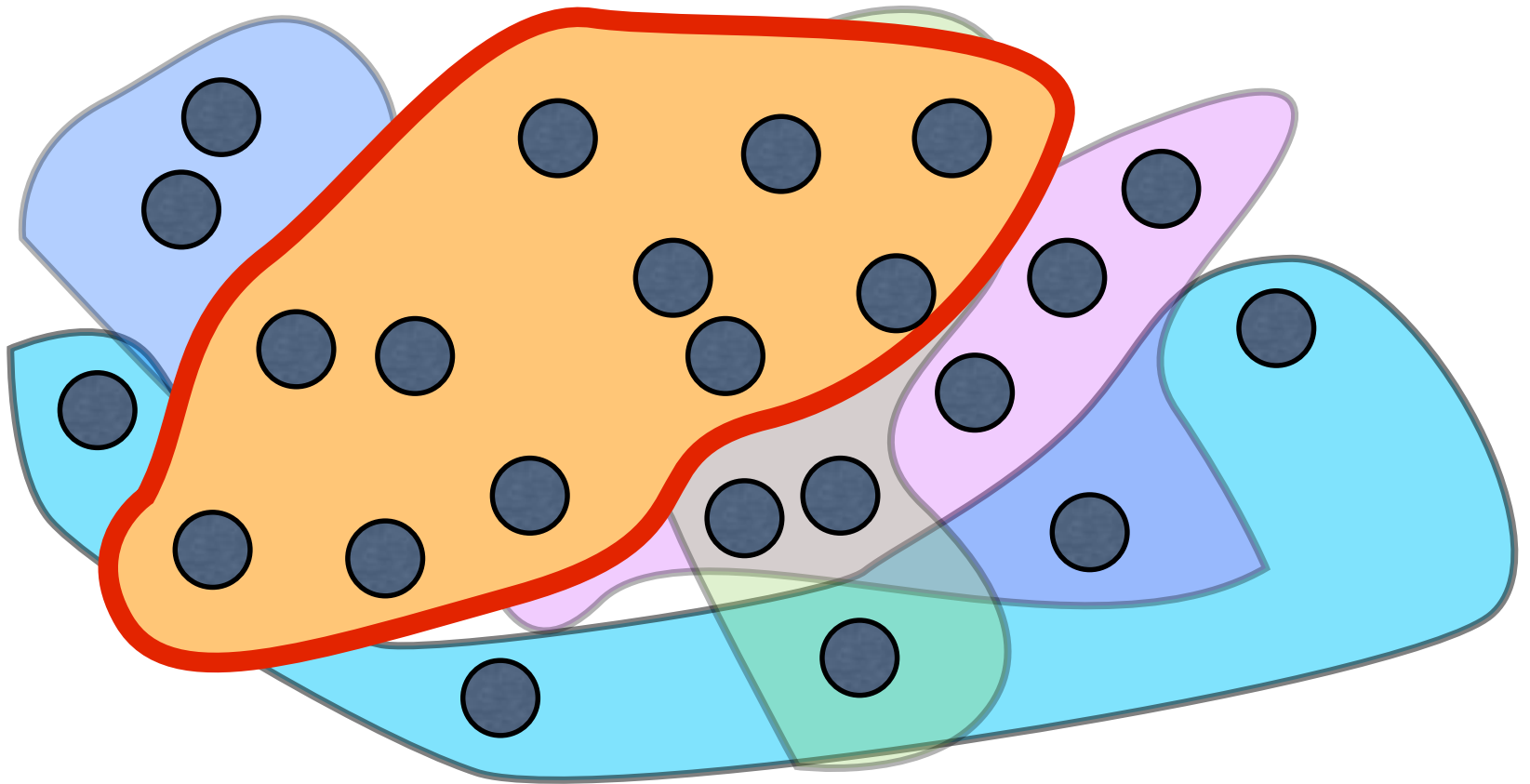
# A Greedy Algorithm

**Strategy:** Pick the set that maximizes # new elements covered



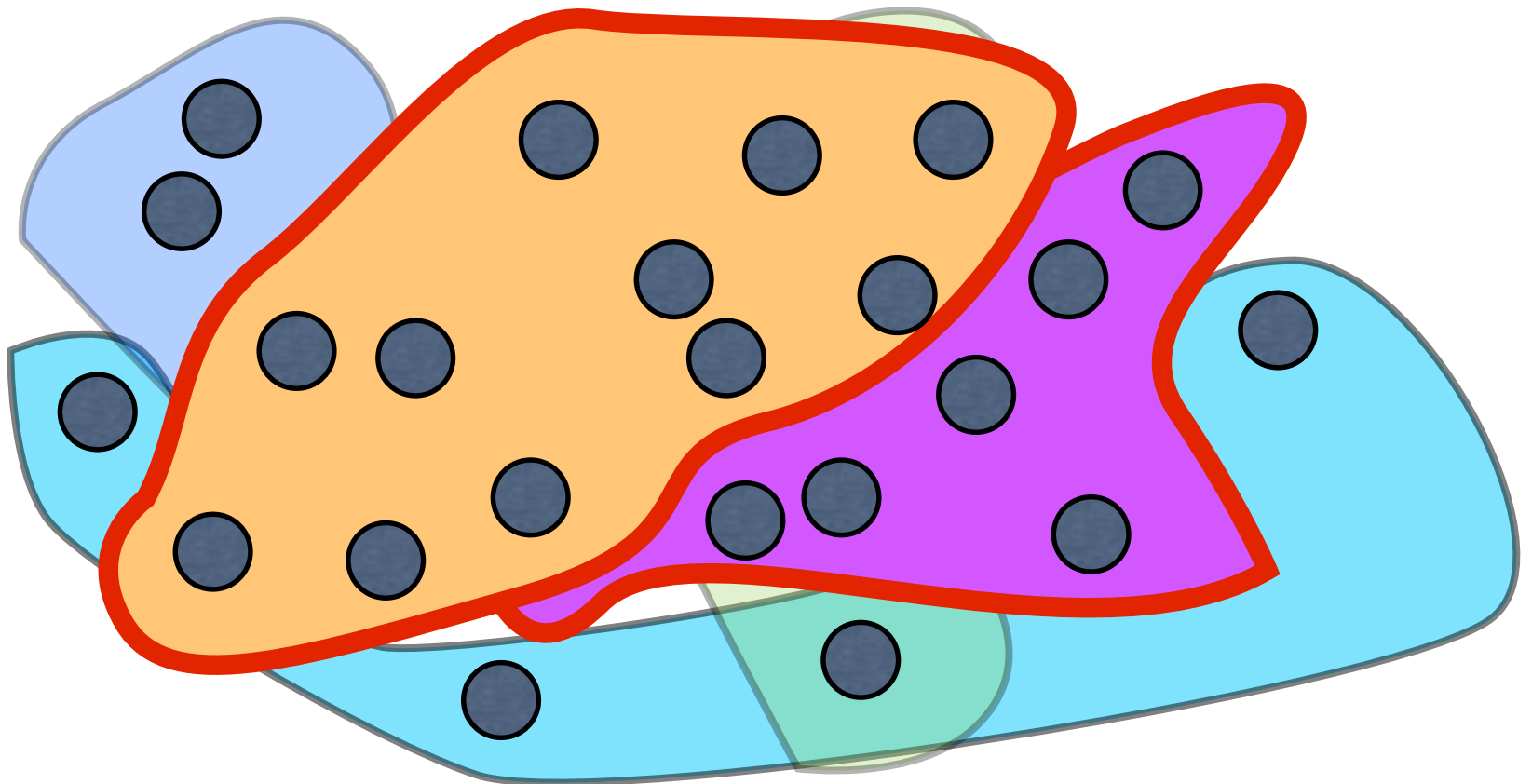
# A Greedy Algorithm

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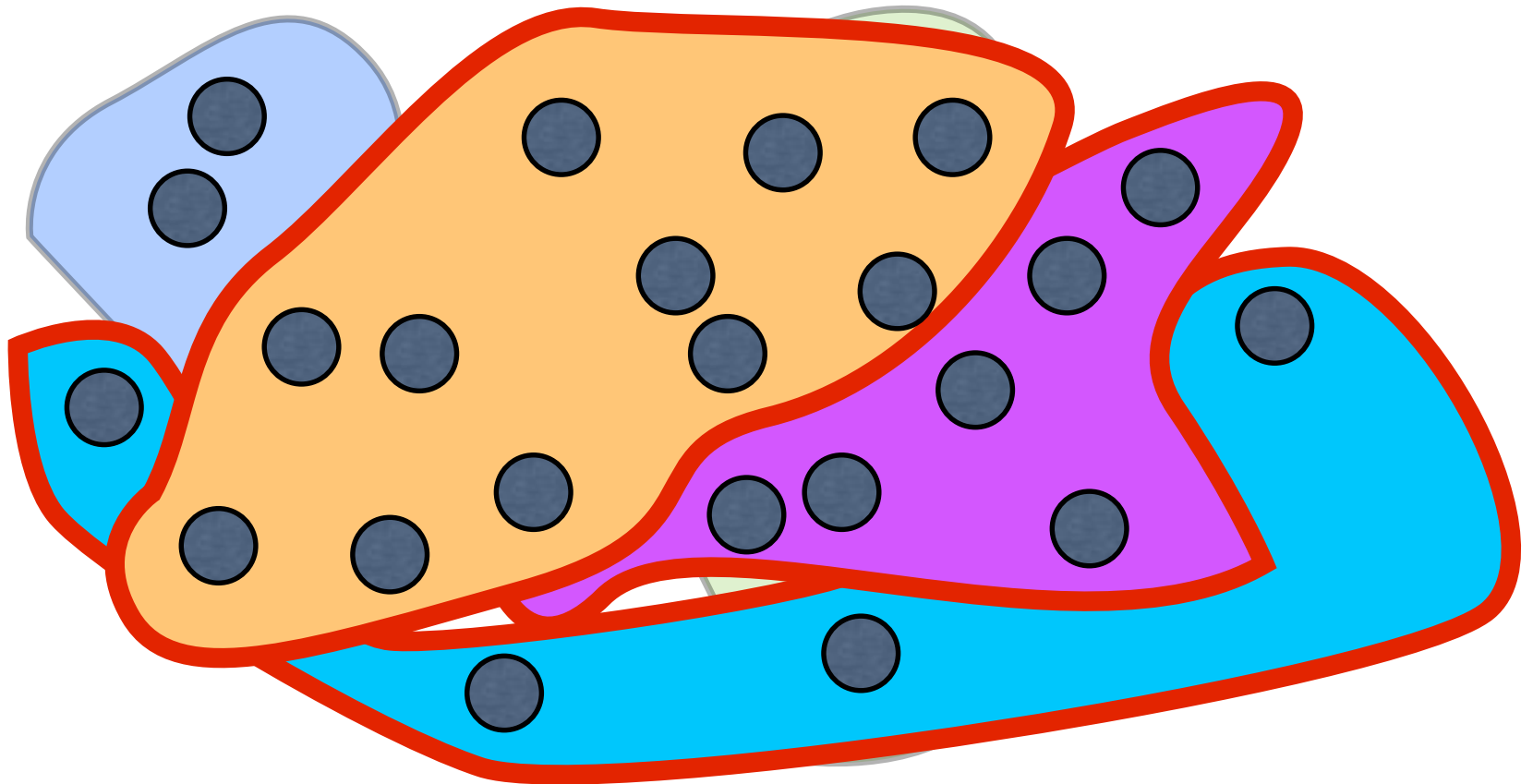
# A Greedy Algorithm

**Strategy:** Pick the set that maximizes # new elements covered



# A Greedy Algorithm

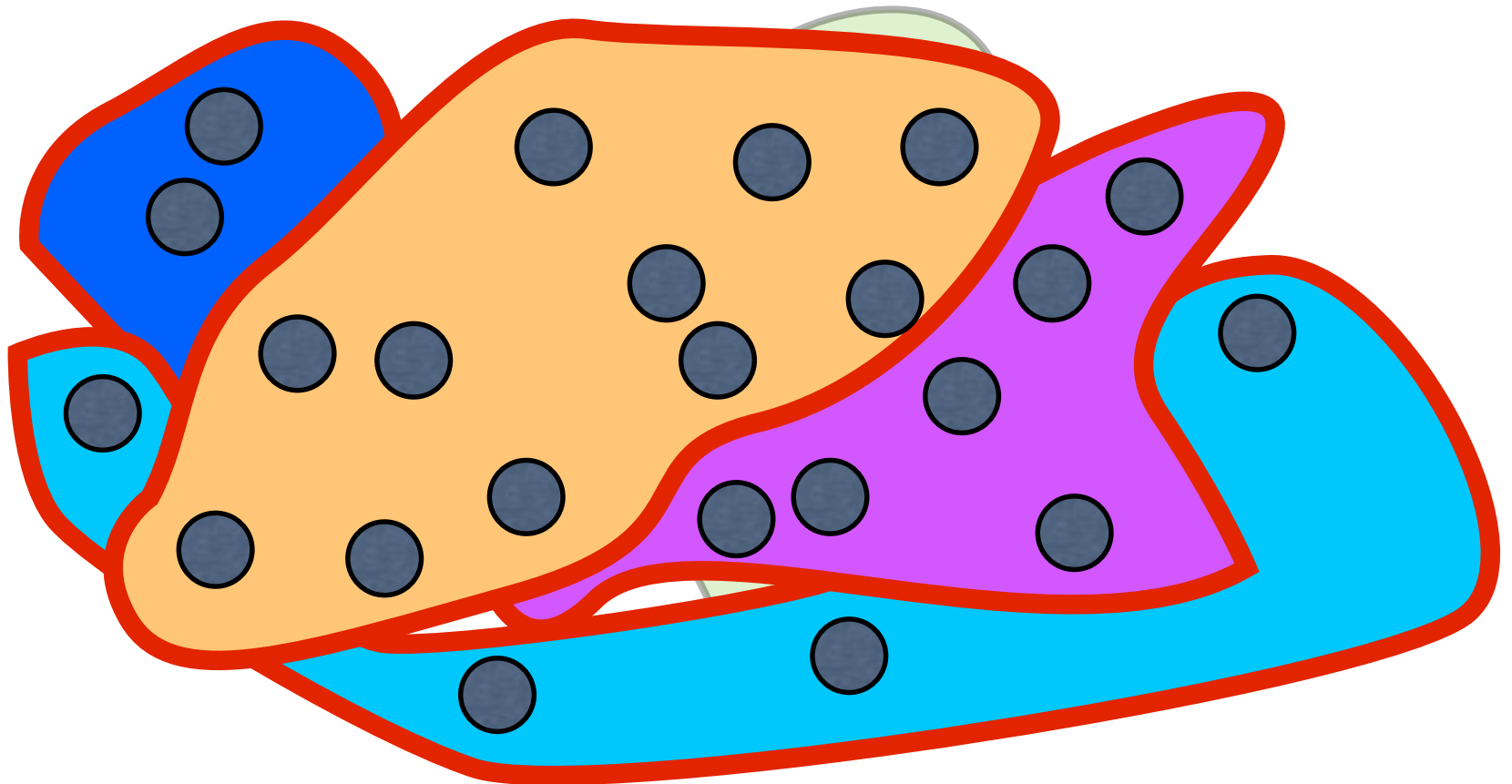
**Strategy:** Pick the set that maximizes # new elements covered



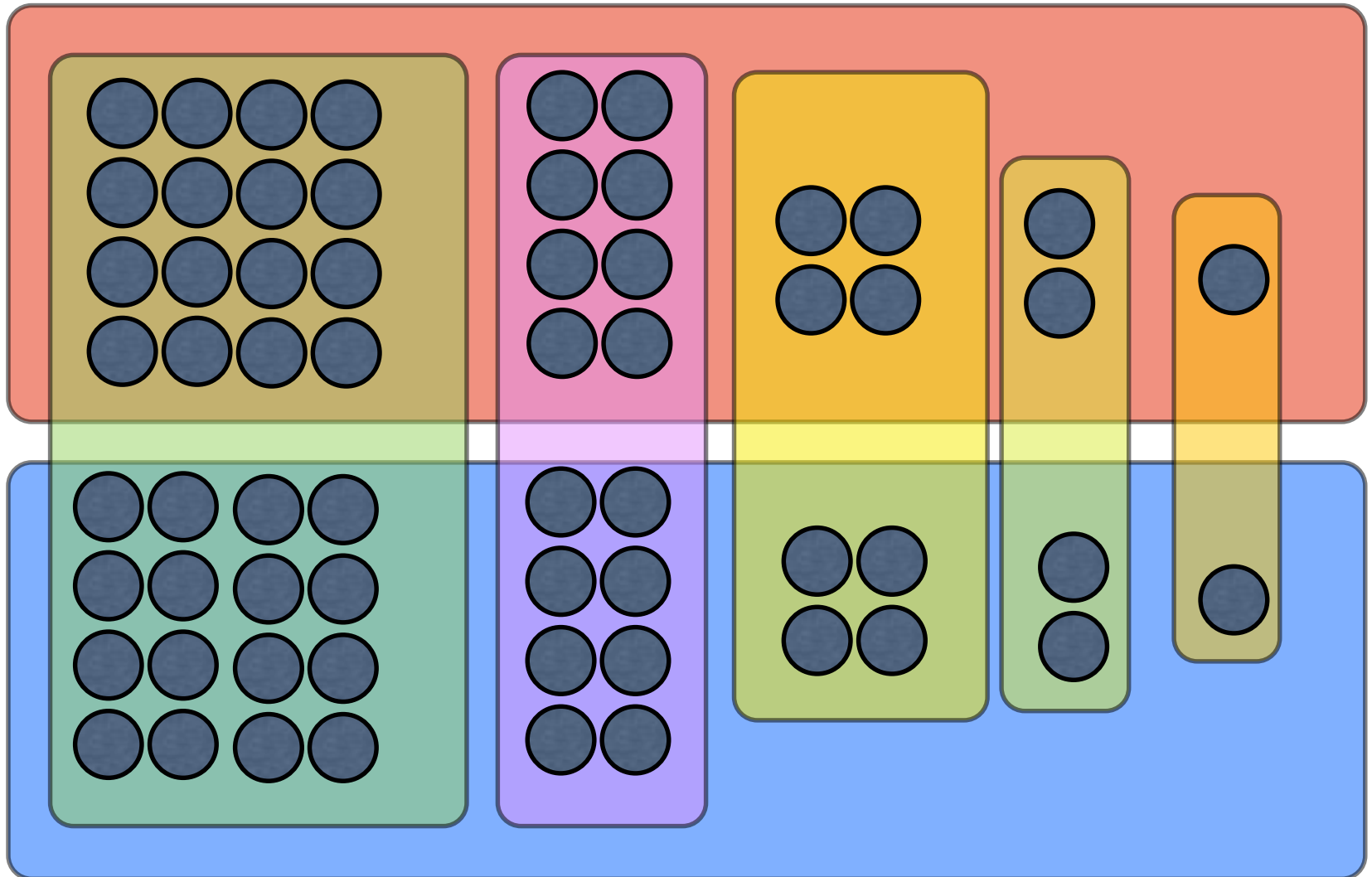
# A Greedy Algorithm

**Strategy:** Pick the set that maximizes # new elements covered

**Thm:** Greedy has  $\ln n$  approximation ratio

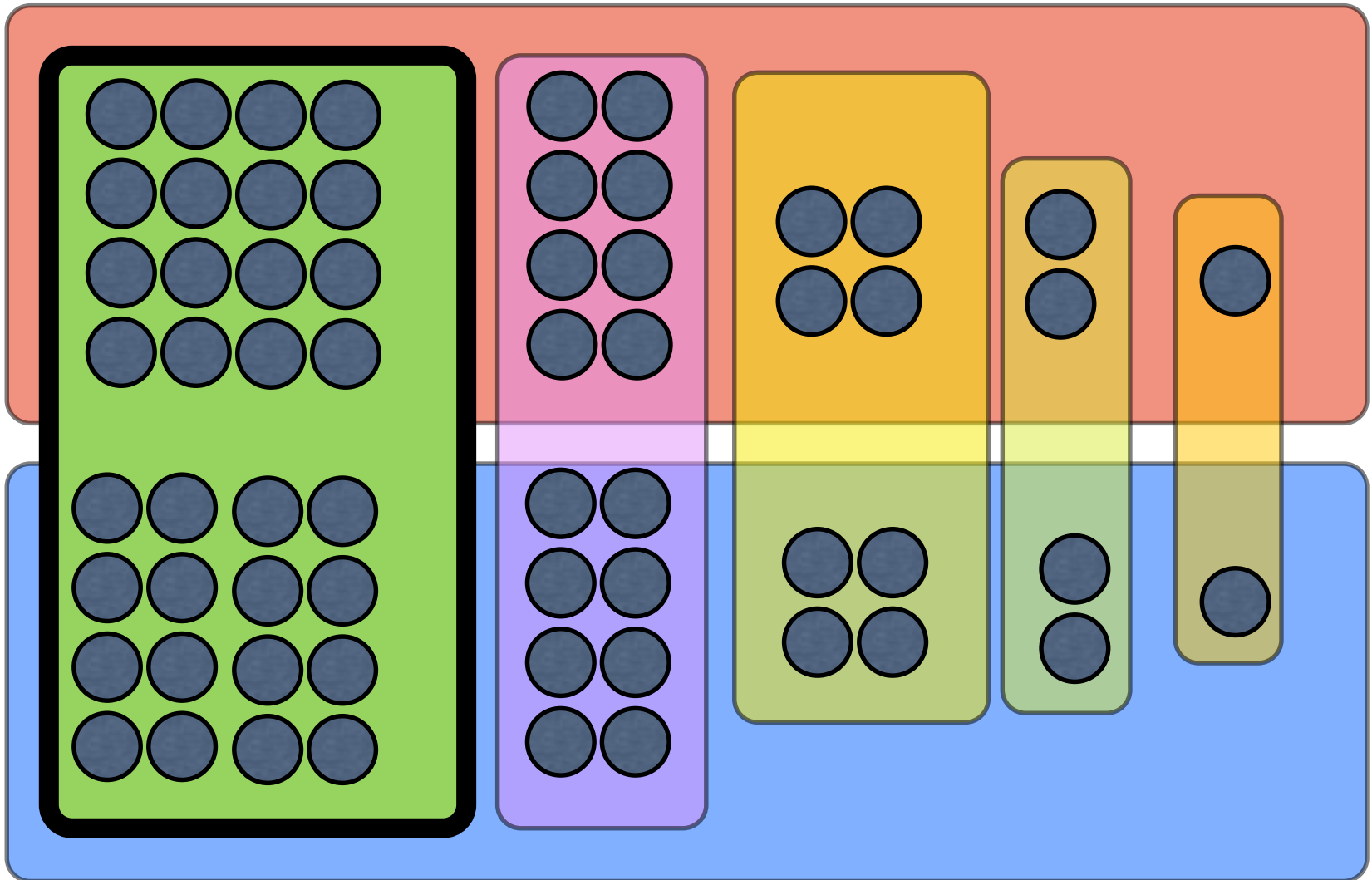


# A Tight Example for Greedy

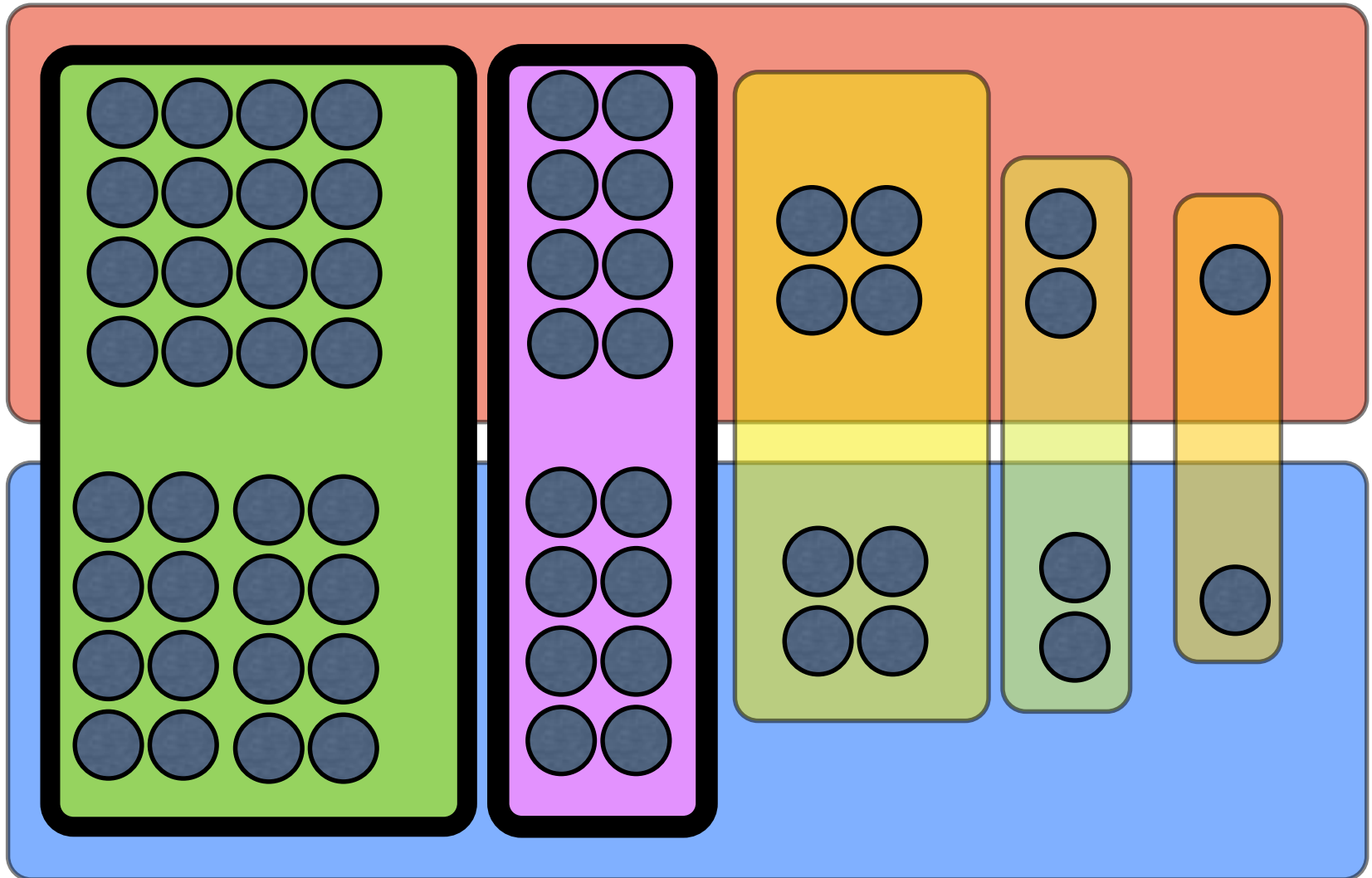




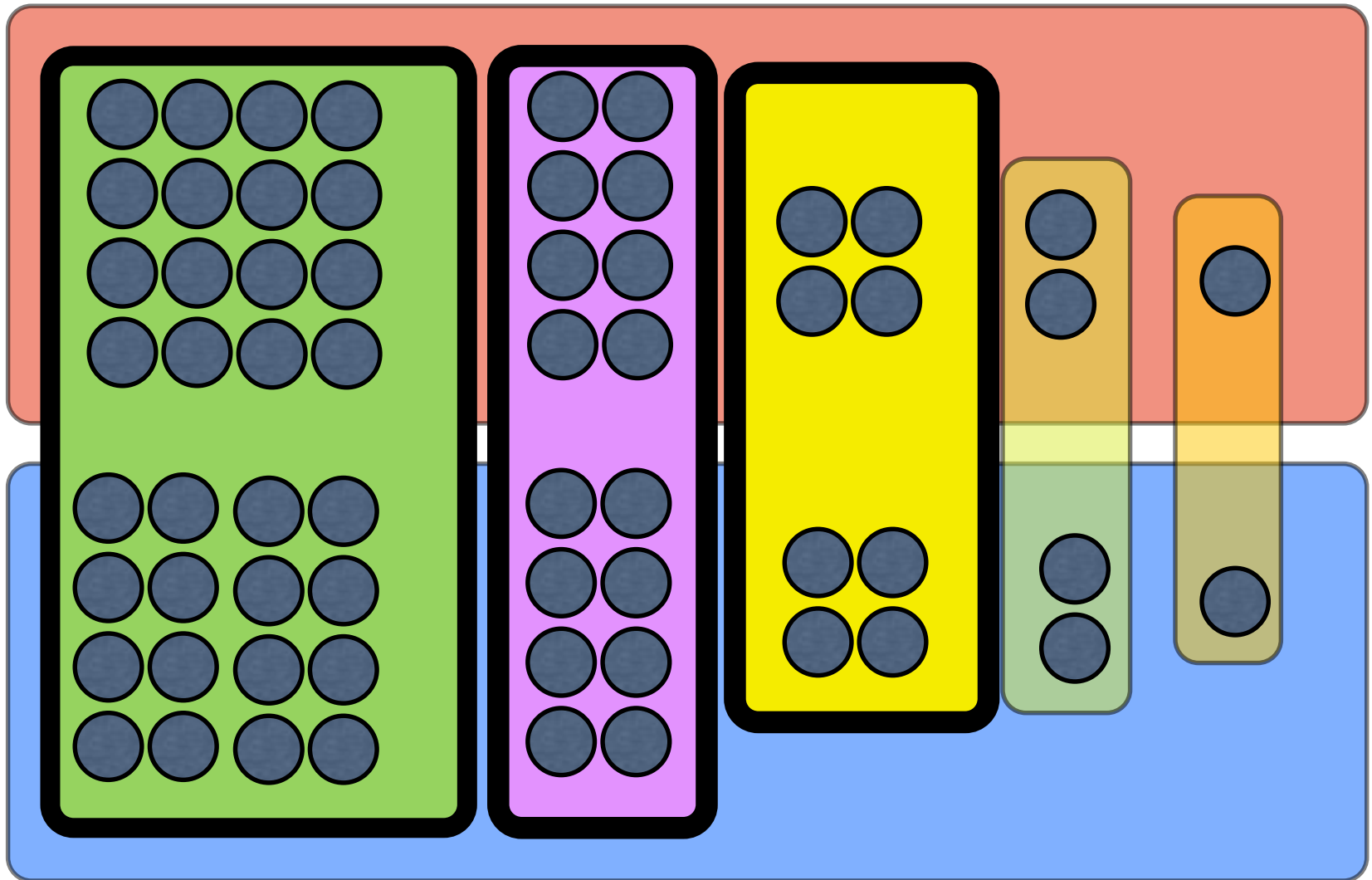
# A Tight Example for Greedy



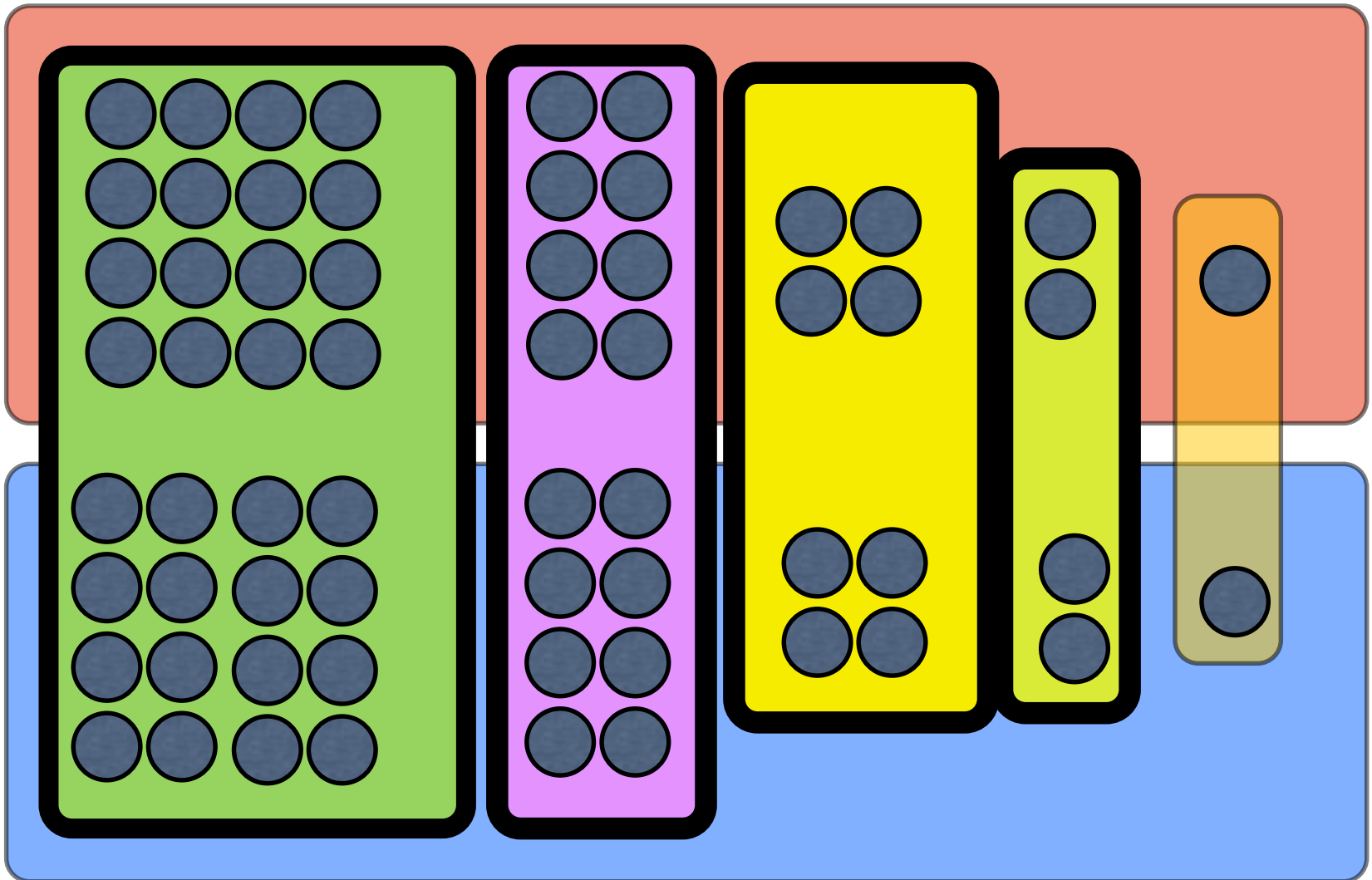
# A Tight Example for Greedy



# A Tight Example for Greedy



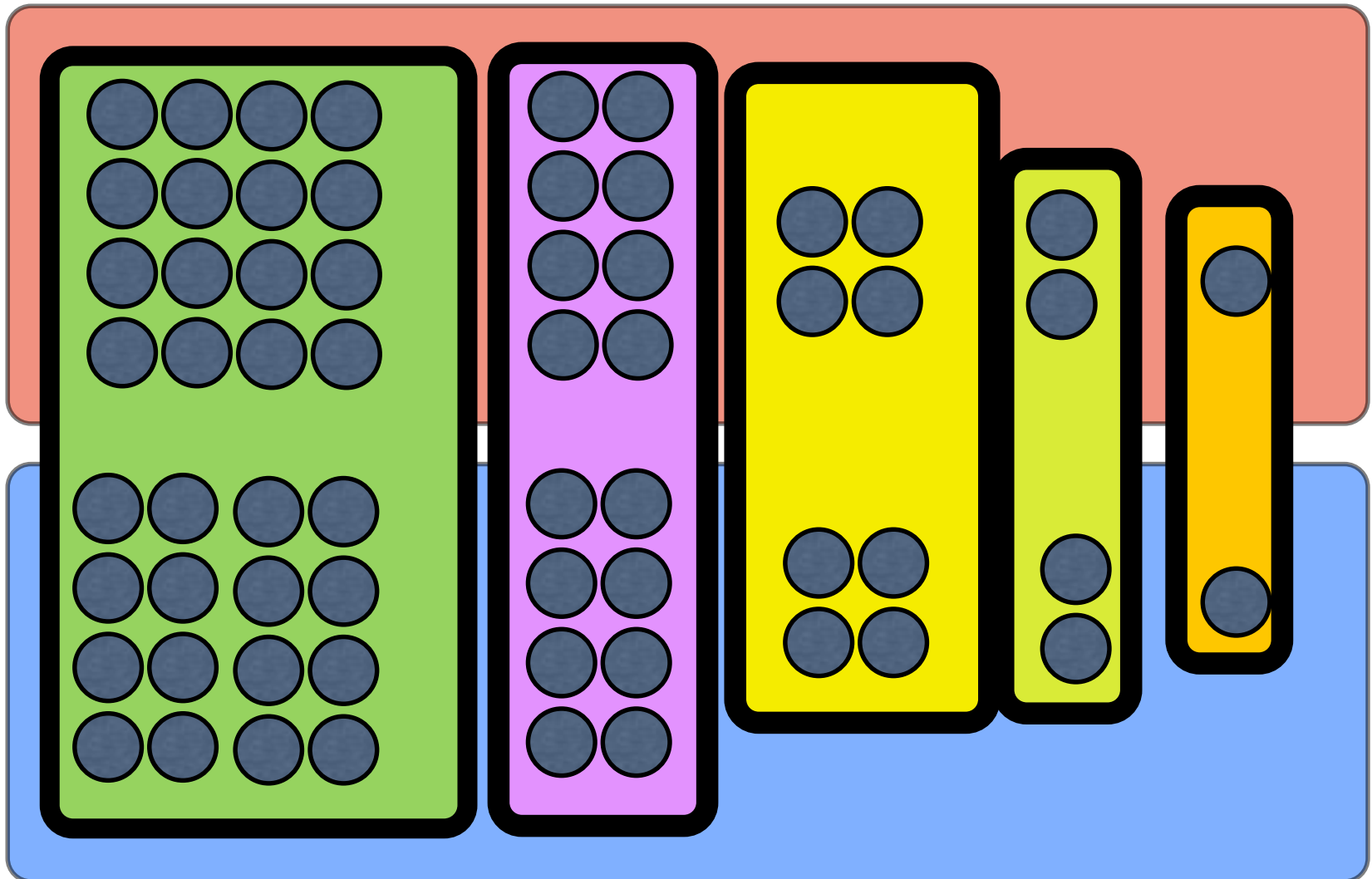
# A Tight Example for Greedy



# 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  $< 1$ .

# Approximation Alg Summary

- To design approximation Alg, always find a way to lower bound OPT
- The best known approximation Alg for vertex cover is the greedy.
  - It has been open for 40 years to obtain a polynomial time algorithm with approximation ratio better than 2
- The best known approximation Alg for set cover is the greedy.
  - It is NP-Complete to obtain better than  $\ln n$  approximation ratio for set cover.