

## Question 2 (cont'd)

Now we will define the subproblems differently.  
Assume the  $v_i$ s are positive integers.

For  $i = 0, 1, \dots, n$ , let  $V_i = \sum_{t=1}^i v_t$ . ( $V_0 = 0$ .)

For  $i = 0, 1, \dots, n$ , and  $v = 0, 1, \dots, V_i$ ,

$W(i, v)$  = the **minimum weight** of a subset of items  $\{1, 2, \dots, i\}$  whose **value** is  $\geq v$ .

Total value of items  $1, \dots, i$

Compare to the subproblems we defined before:

For  $i = 0, 1, \dots, n$ , and  $c = 0, 1, \dots, C$ ,

$K(i, c)$  = the **maximum value** of a subset of items  $\{1, 2, \dots, i\}$  whose **weight** is  $\leq c$ .

# Question 2 (cont'd)

For  $i = 0, 1, \dots, n$ , and  $v = 0, 1, \dots, V_i$ ,

$W(i, v)$  = the minimum weight of a subset of items  $\{1, 2, \dots, i\}$  whose value is  $\geq v$ .

- Give a recursive formula to compute the subproblems.
- Describe your DP algorithm in pseudocode.
- Analyze the running time of your algorithm.
- Modify the algorithm to find the actual set of items of maximum value whose weight does not exceed the knapsack capacity  $C$ .

# Question 2 — answer (cont'd)

- Recursive formula to compute the subproblems.

Case 1:  $v > V_{i-1}$ . (Lightest set of items of value  $\geq v$  must use item  $i$ .)

$$W(i, v) = W(i - 1, \max(0, v - v_i)) + w_i$$

Case 2:  $v \leq V_{i-1}$ . (Lightest set of items of value  $\geq v$  may or may not use item  $i$ .)

$$W(i, v) = \min(W(i - 1, v), \\ W(i - 1, \max(0, v - v_i)) + w_i)$$

# Question 2 — answer (cont'd)

- Describe your DP algorithm in pseudocode.

```
V[0] := 0; for i := 1 to n do V[i] := V[i - 1] + v_i
for i := 0 to n do W[i, 0] := 0
for v := 1 to V[n] do W[0, v] := 0
for i := 1 to n do
  for v := 1 to V[i] do
    if v > V[i - 1] then
      W[i, v] := W[i - 1, max(0, v - v_i)] + w_i
    else
      W[i, v] := min(W[i - 1, v],
                    W[i - 1, max(0, v - v_i)] + w_i)
return max{v: W[n, v] ≤ C}
```

# Question 2 — answer (cont'd)

- Analyze the running time of your algorithm.

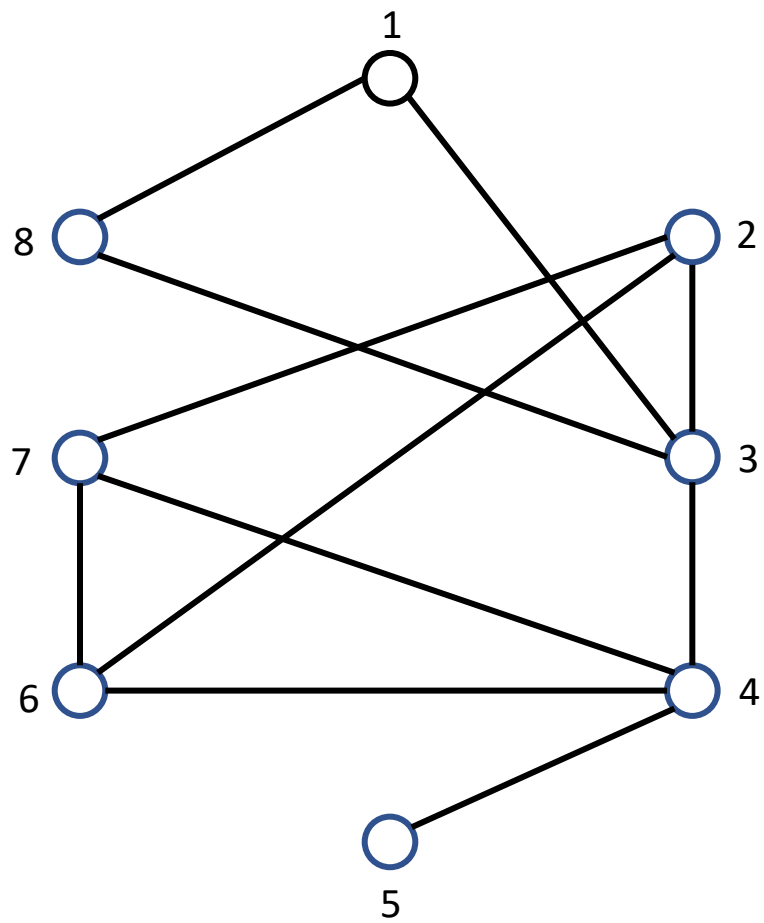
$$\Theta\left(n \cdot \sum_{i=1}^n v_i\right)$$

This is pseudopolynomial.

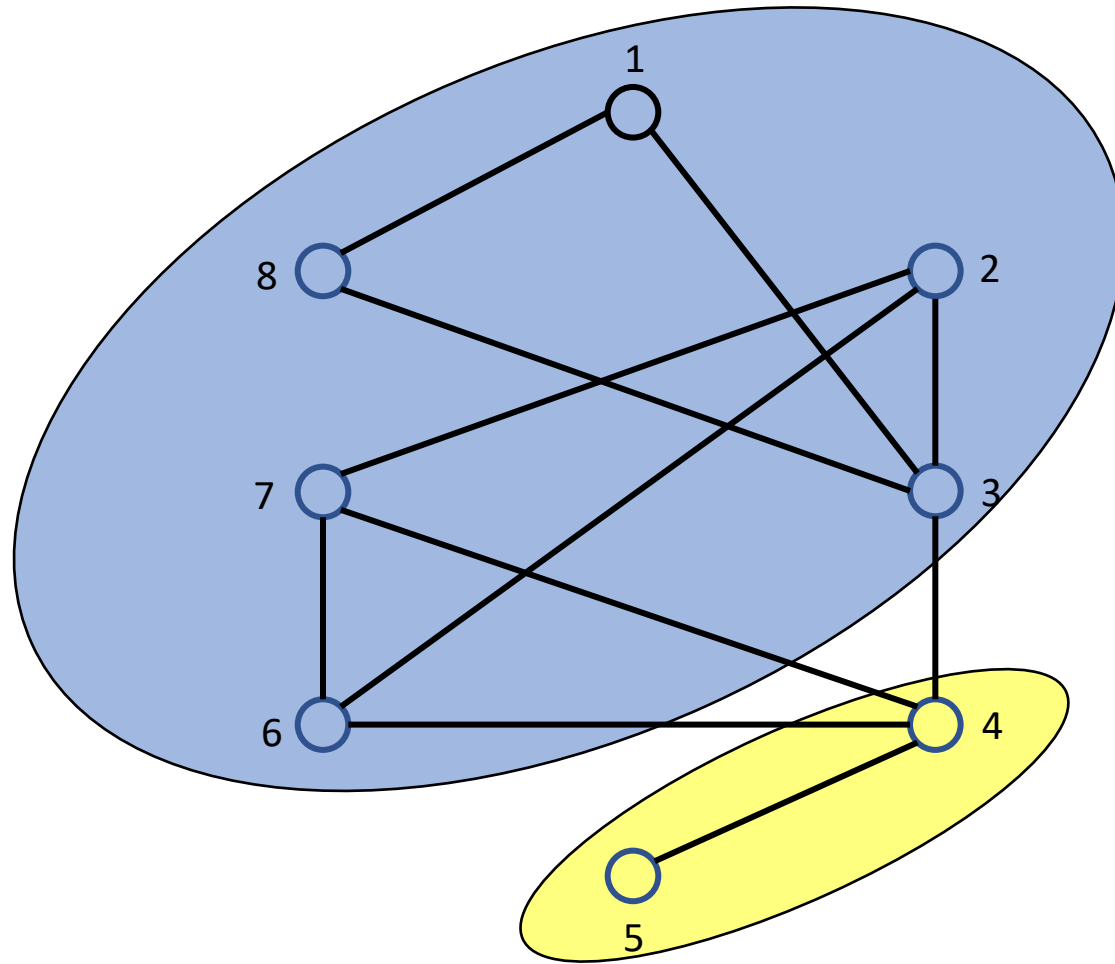
This version of the algorithm (with subproblems based on value rather than weight) is the basis for a **polynomial-time approximation algorithm** for knapsack that we will see at the end of the course.

Approximation local search algorithm for max cut

Graph with 8 nodes  
and 11 edges

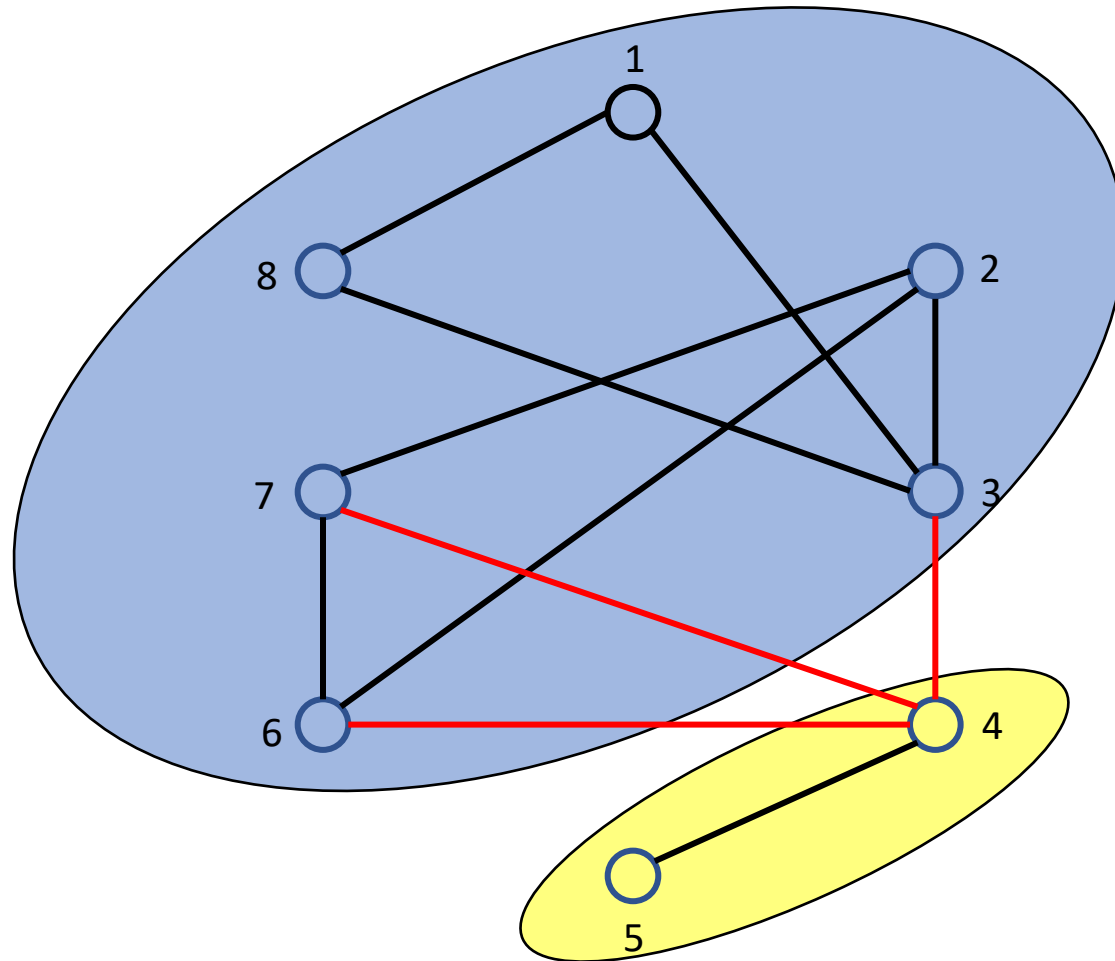


A cut of  
the graph



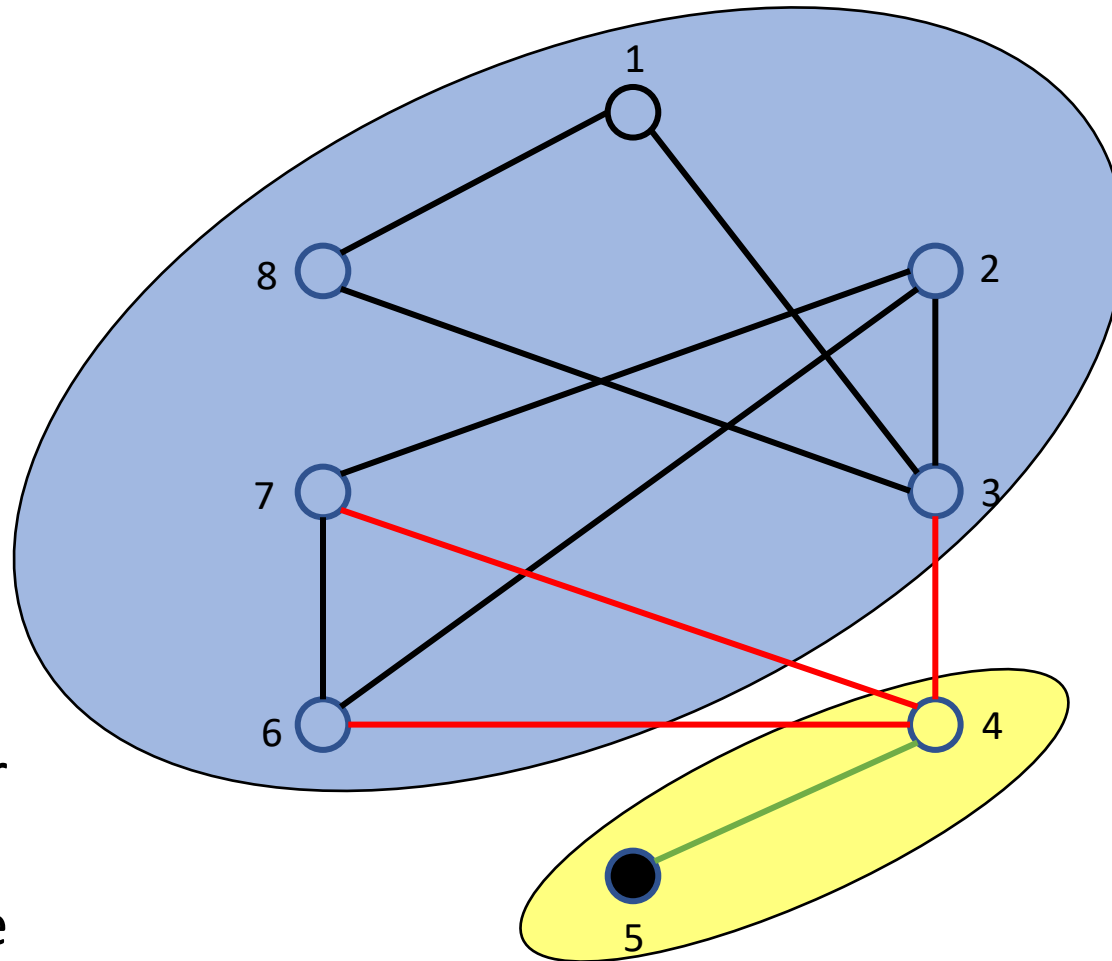


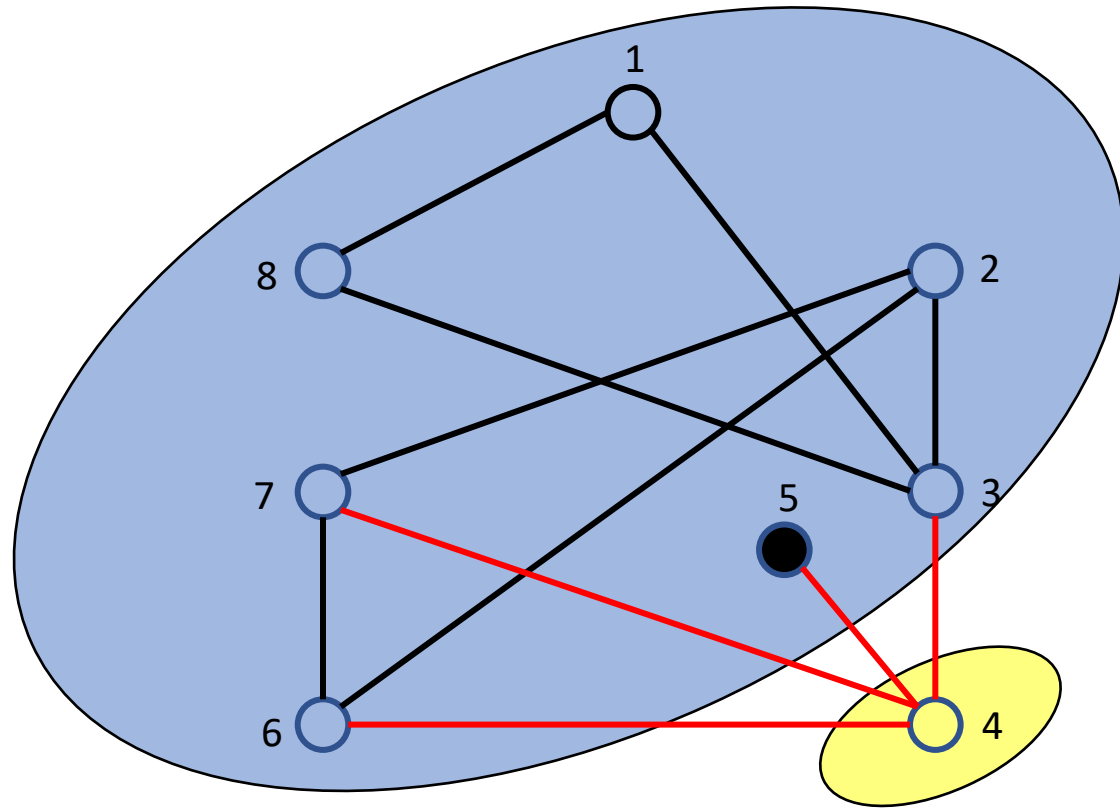
This cut has  
3 **cross** edges



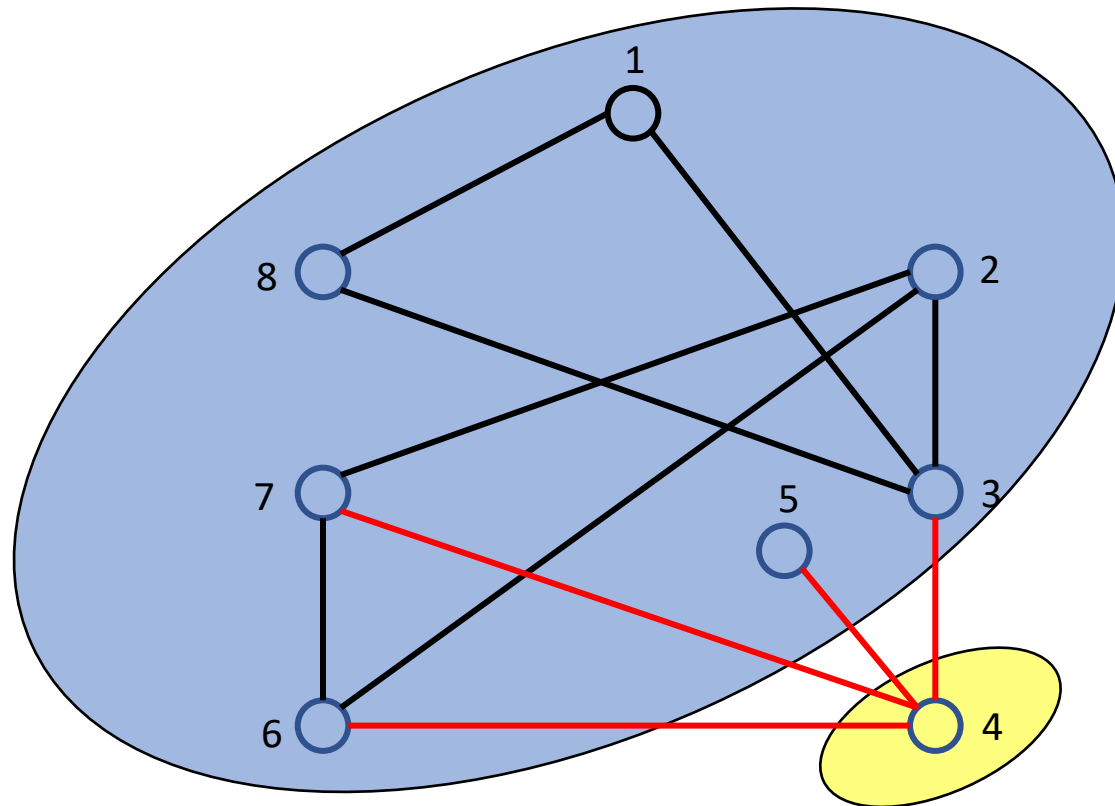
Node 5 has more **internal** edges (1) than **cross** edges (0).

Increase the number of cross edges by moving it to the blue side.



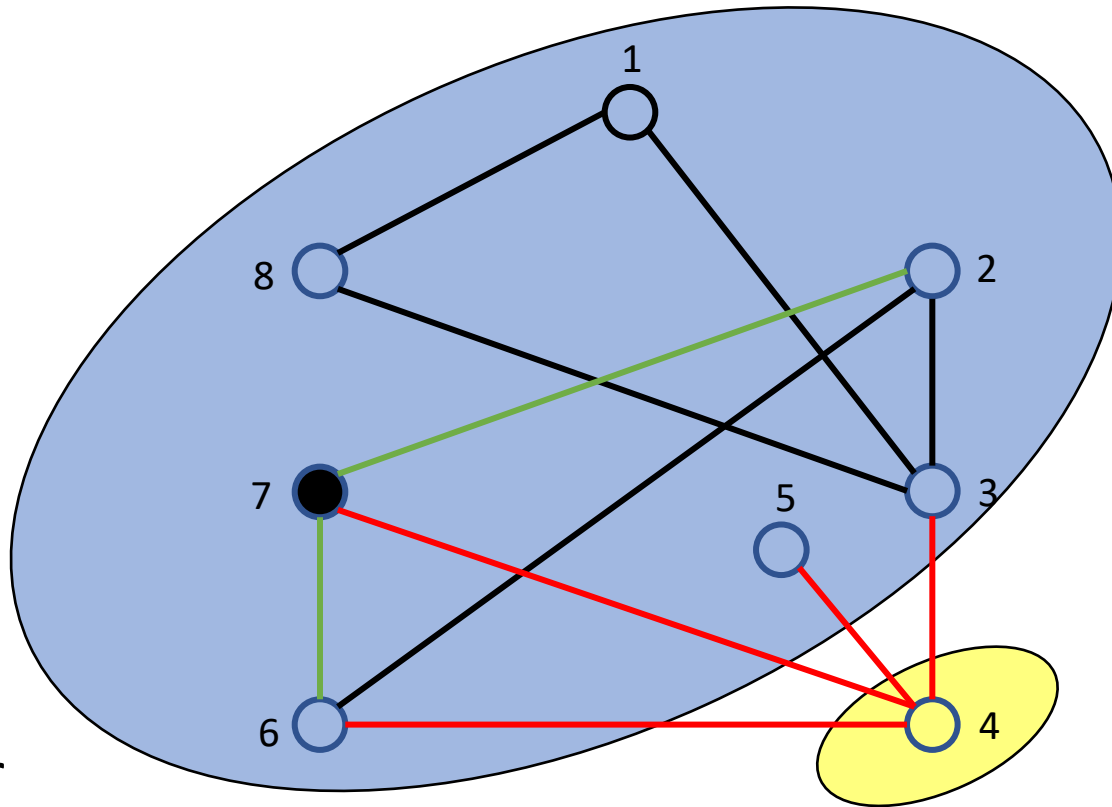


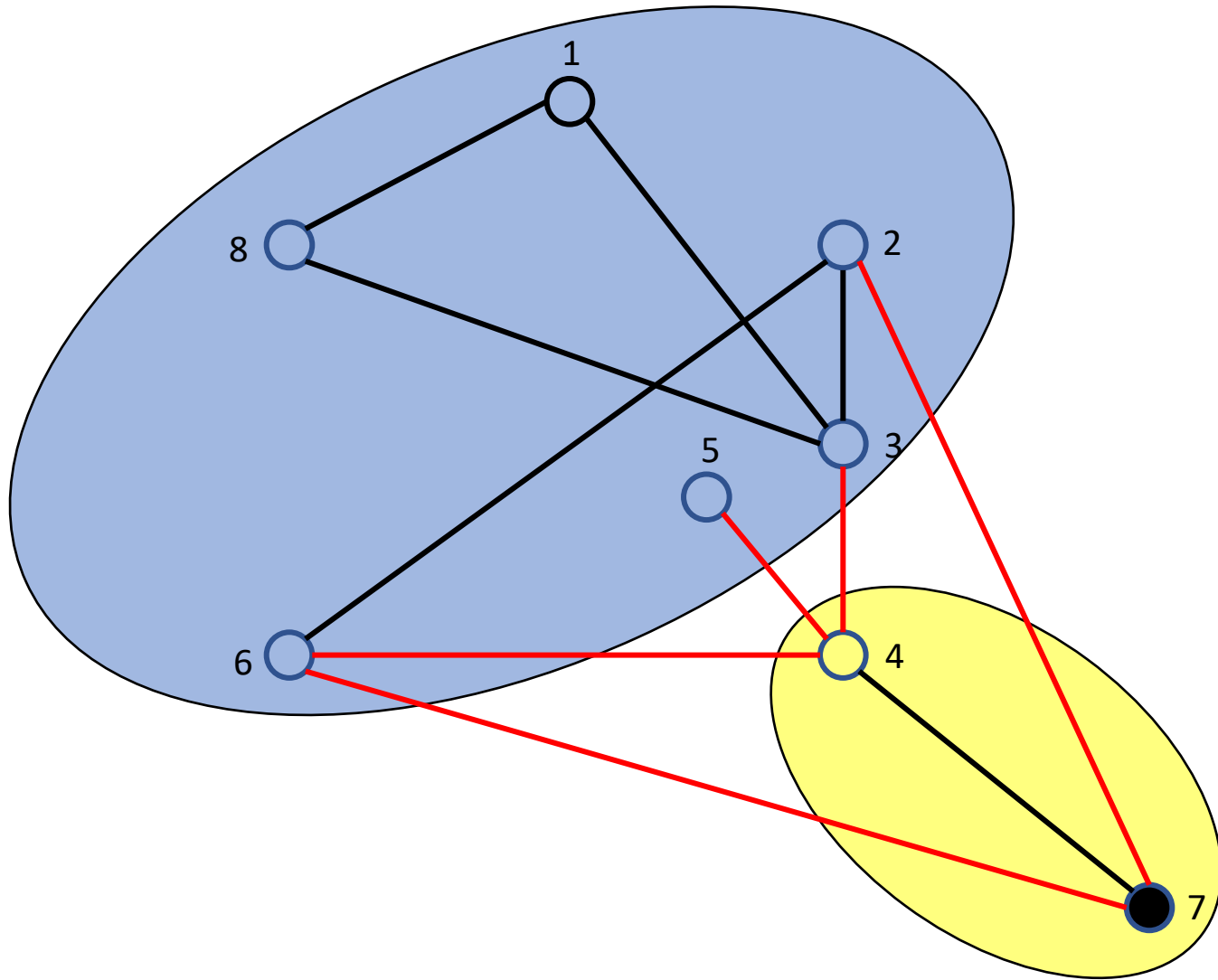
This cut has  
4 **cross** edges



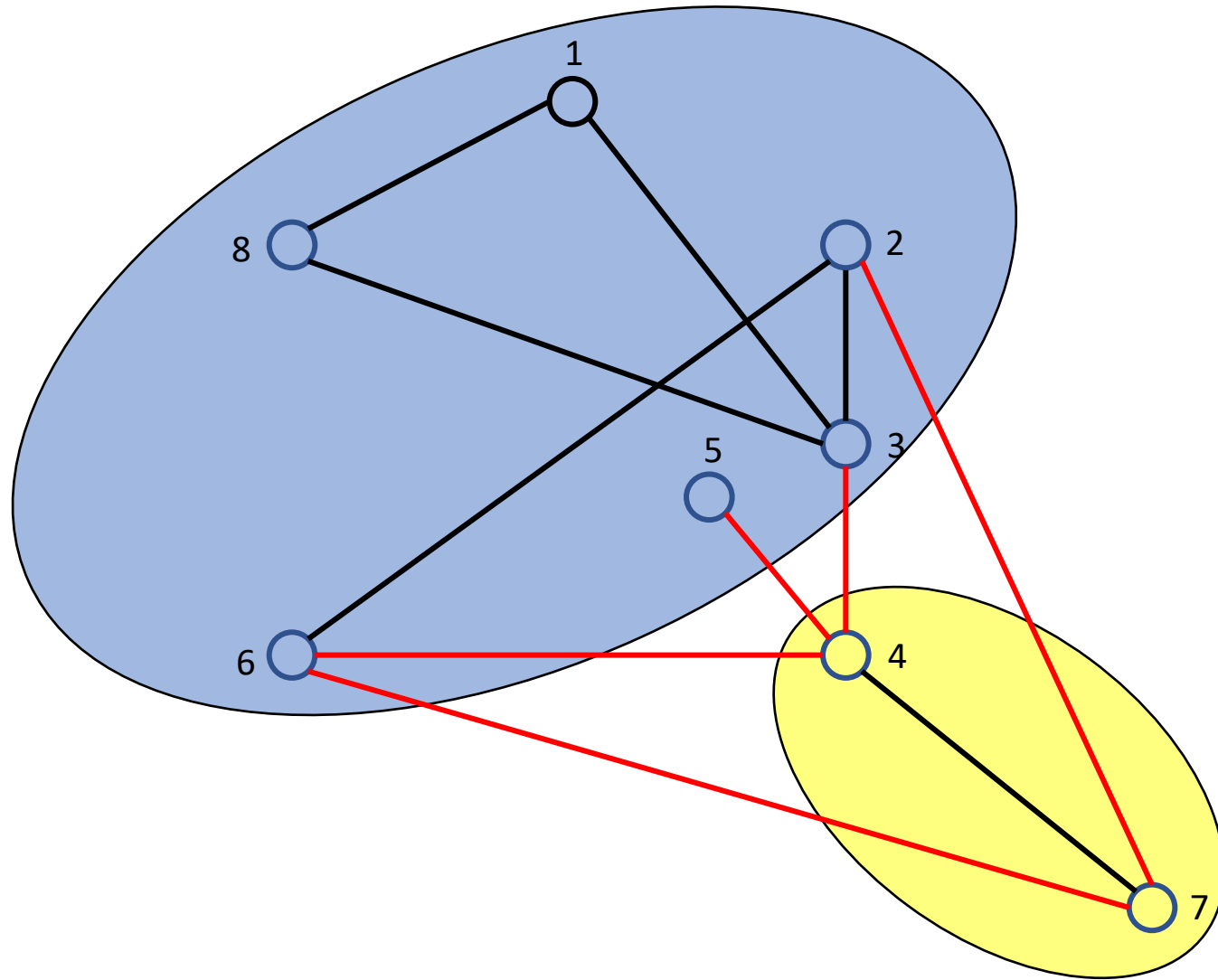
Node 7 has more **internal** edges (2) than **cross** edges (1).

Increase the number of cross edges by moving it to the yellow side.



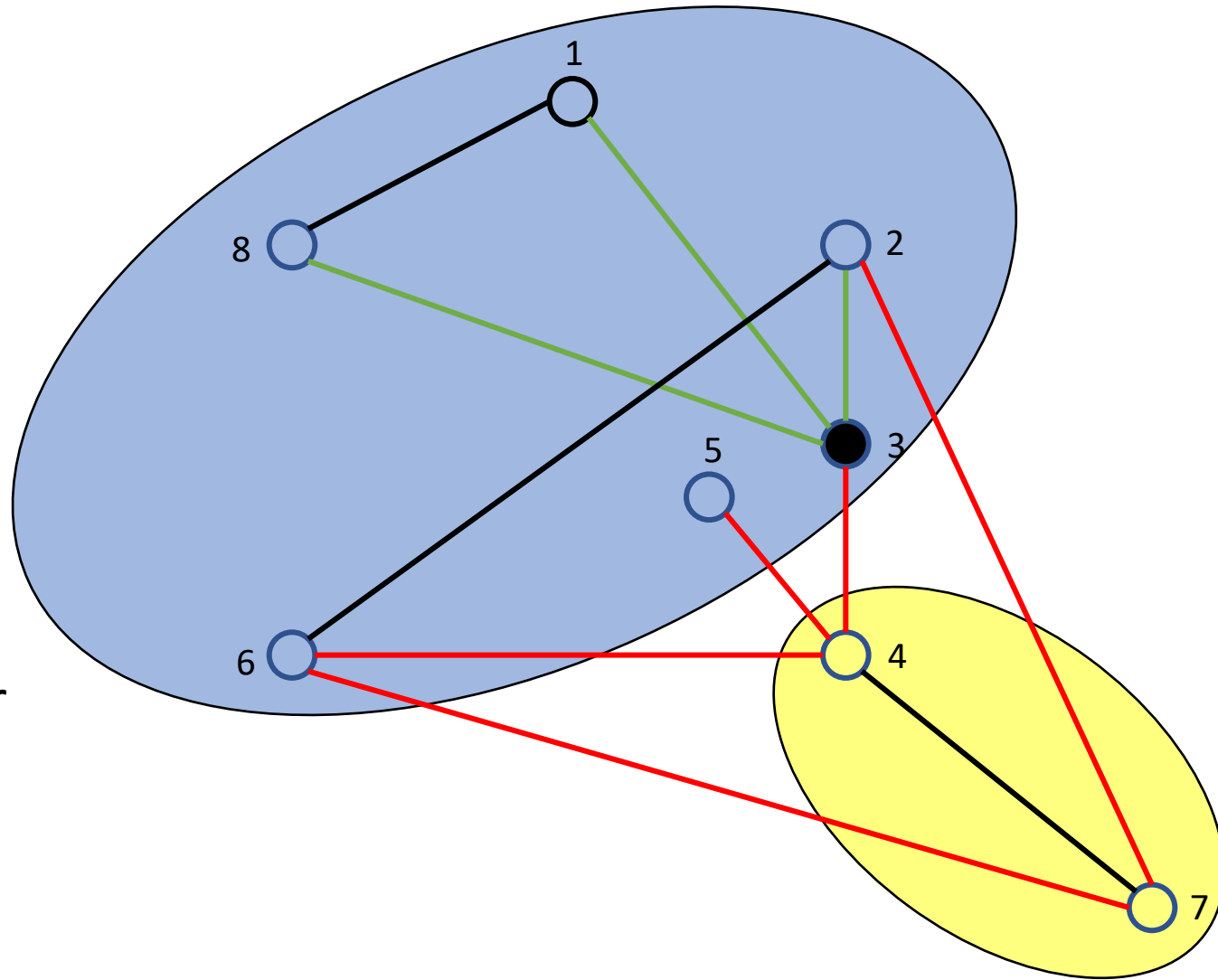


This cut has  
5 **cross** edges

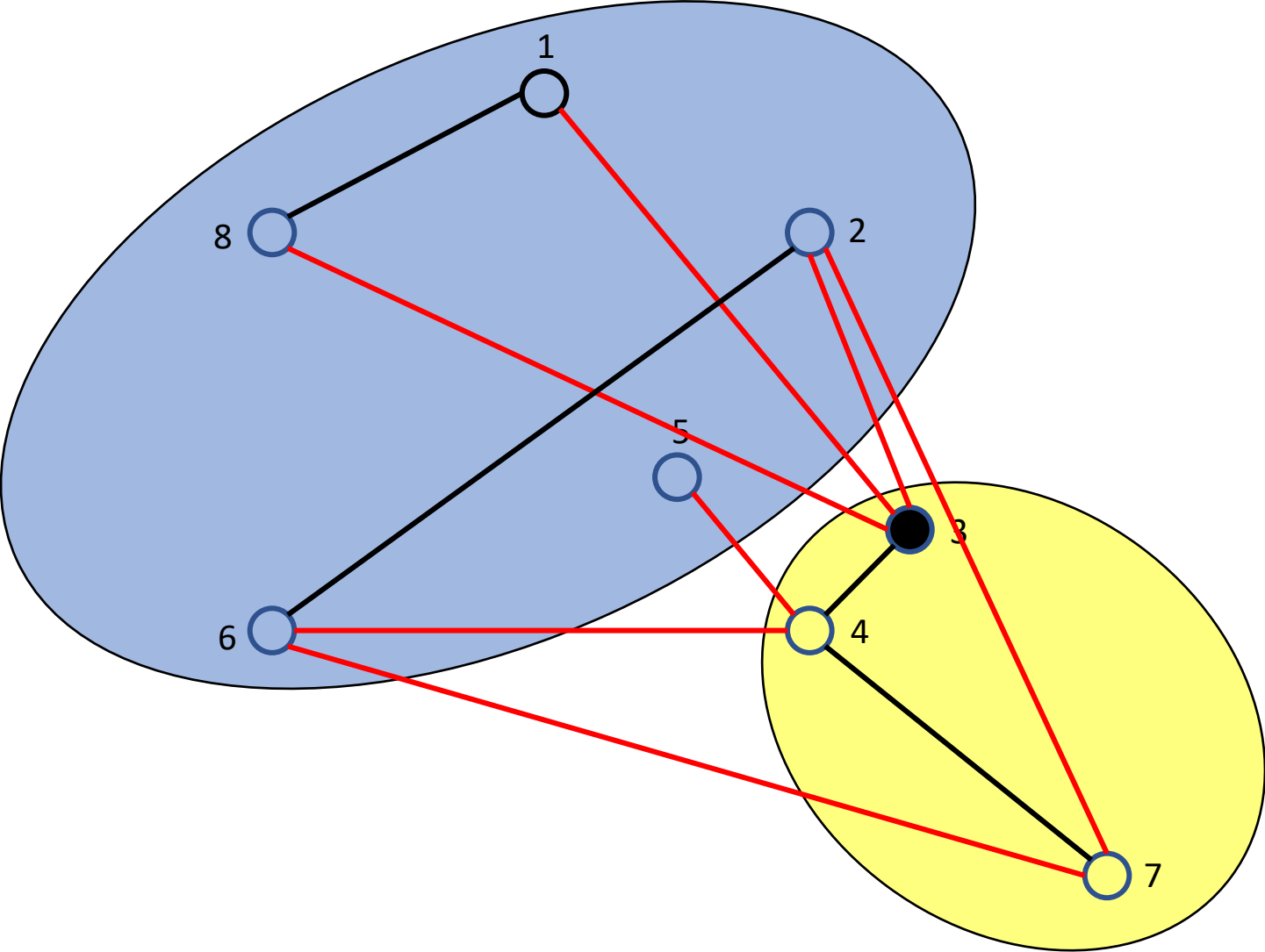


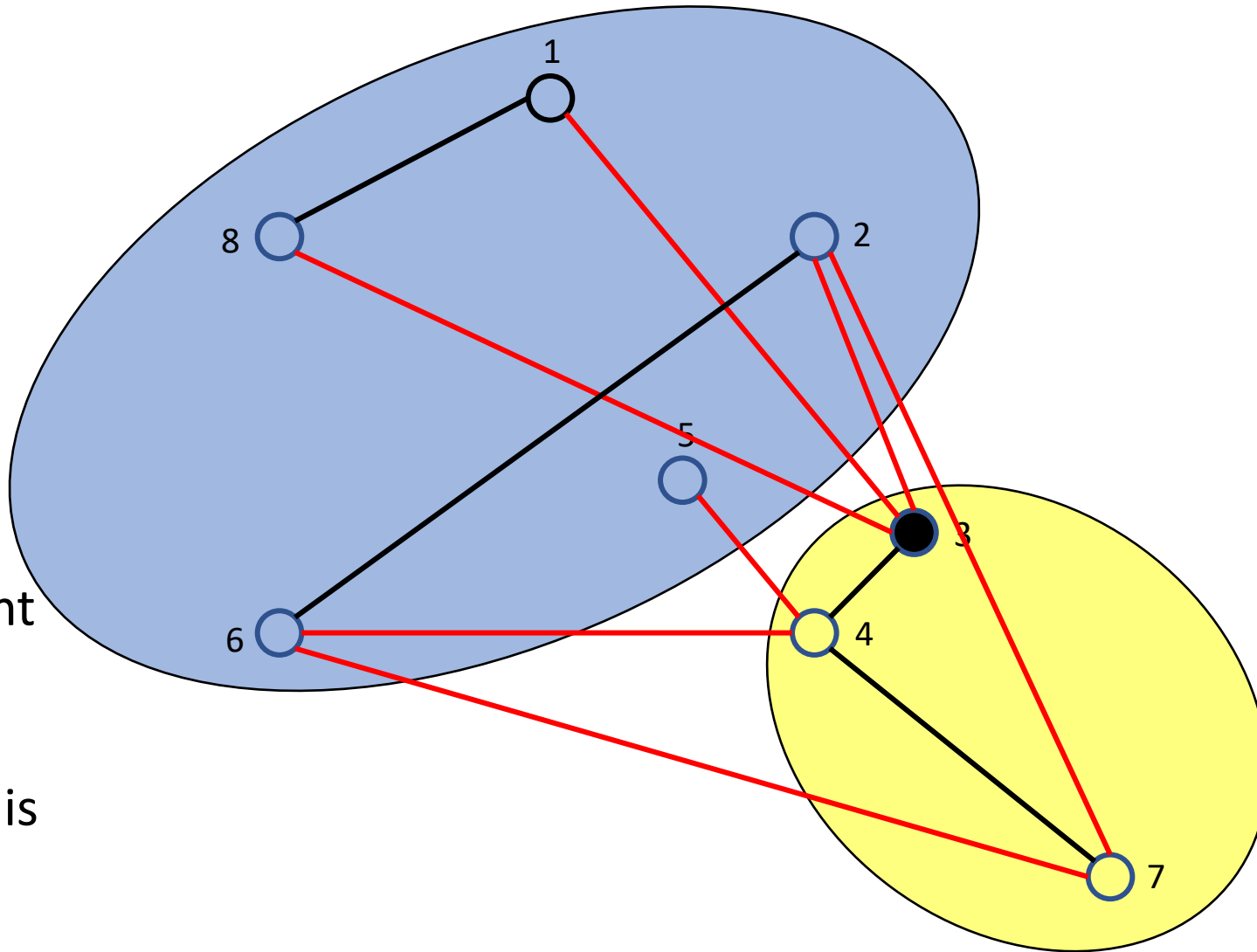
Node 3 has more **internal** edges (3) than **cross** edges (1).

Increase the number of cross edges by moving it to the yellow side.









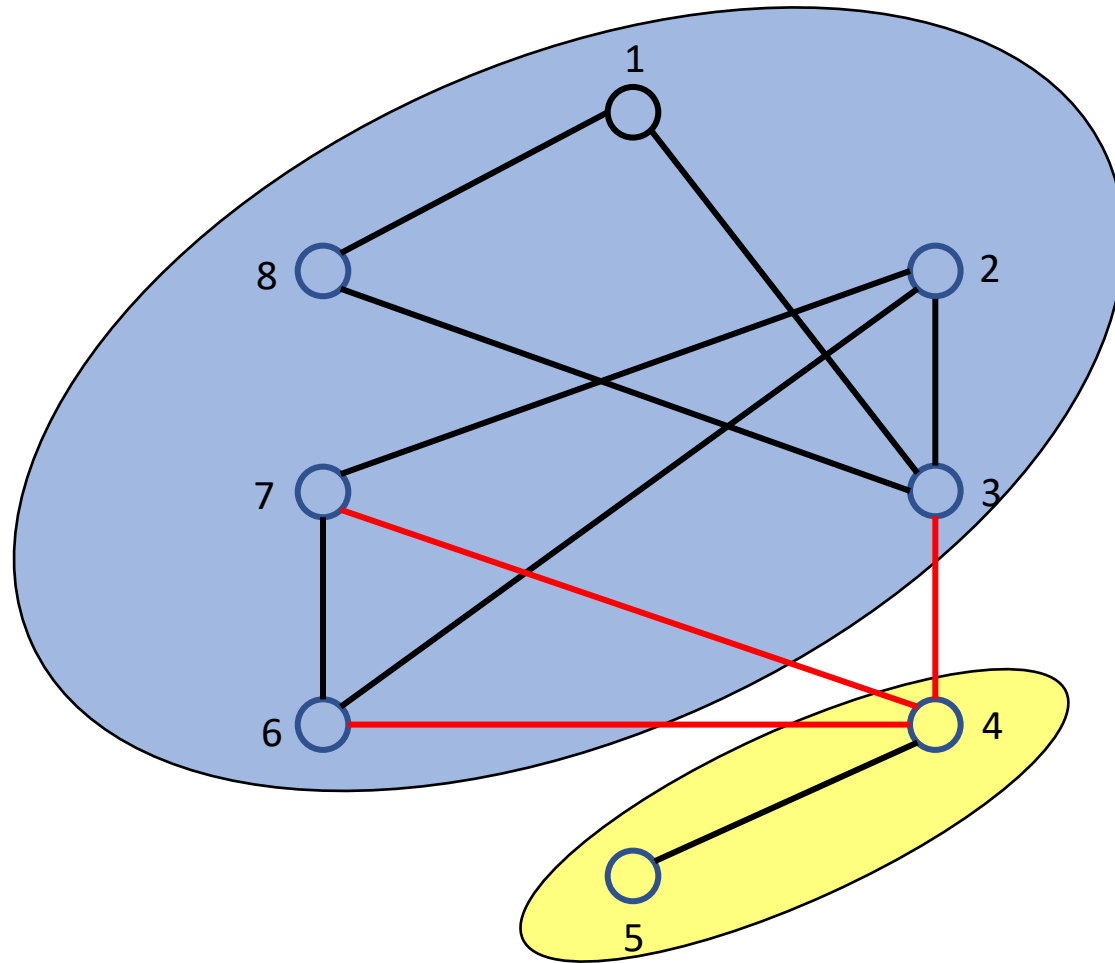
This cut has  
7 **cross** edges.

No local improvement  
is possible.

But as we will see, it is  
not a max cut!

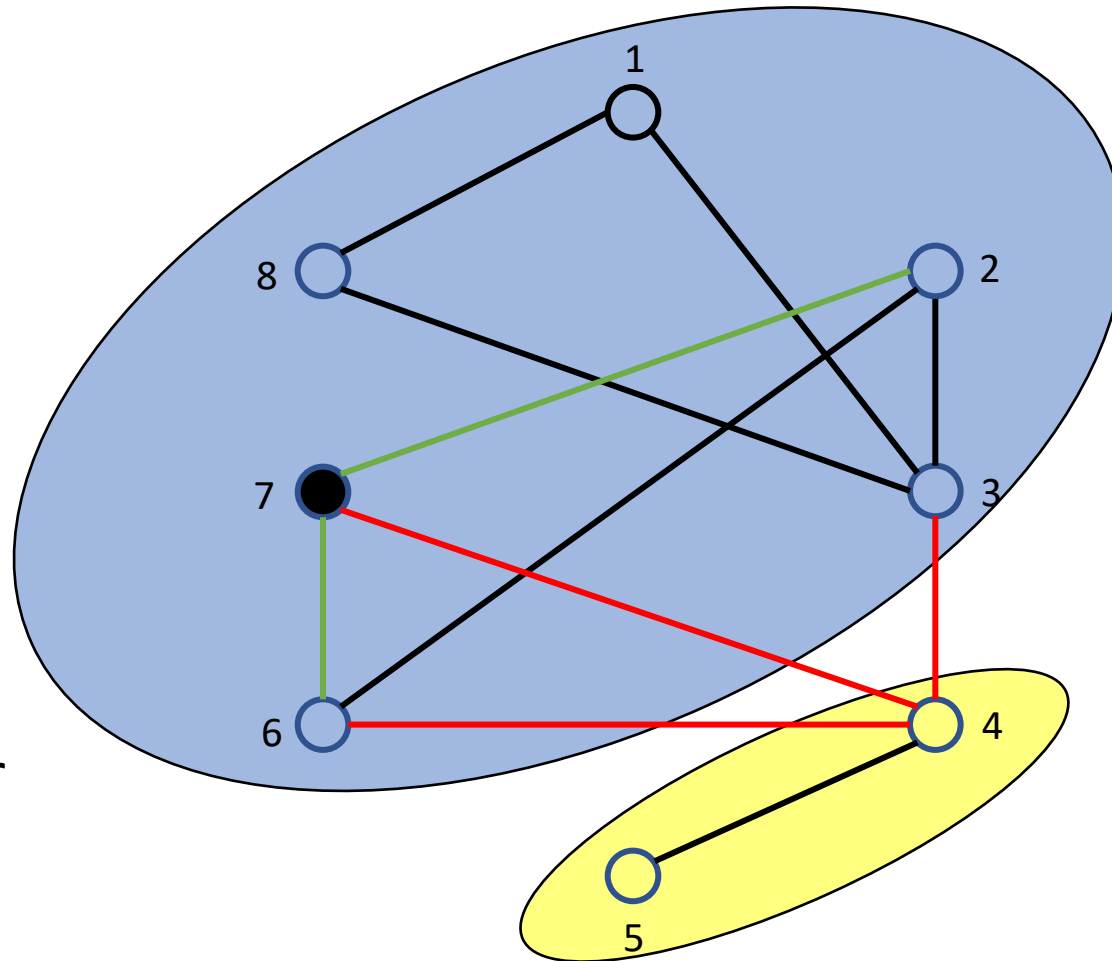
Back to the original cut with 3 **cross** edges.

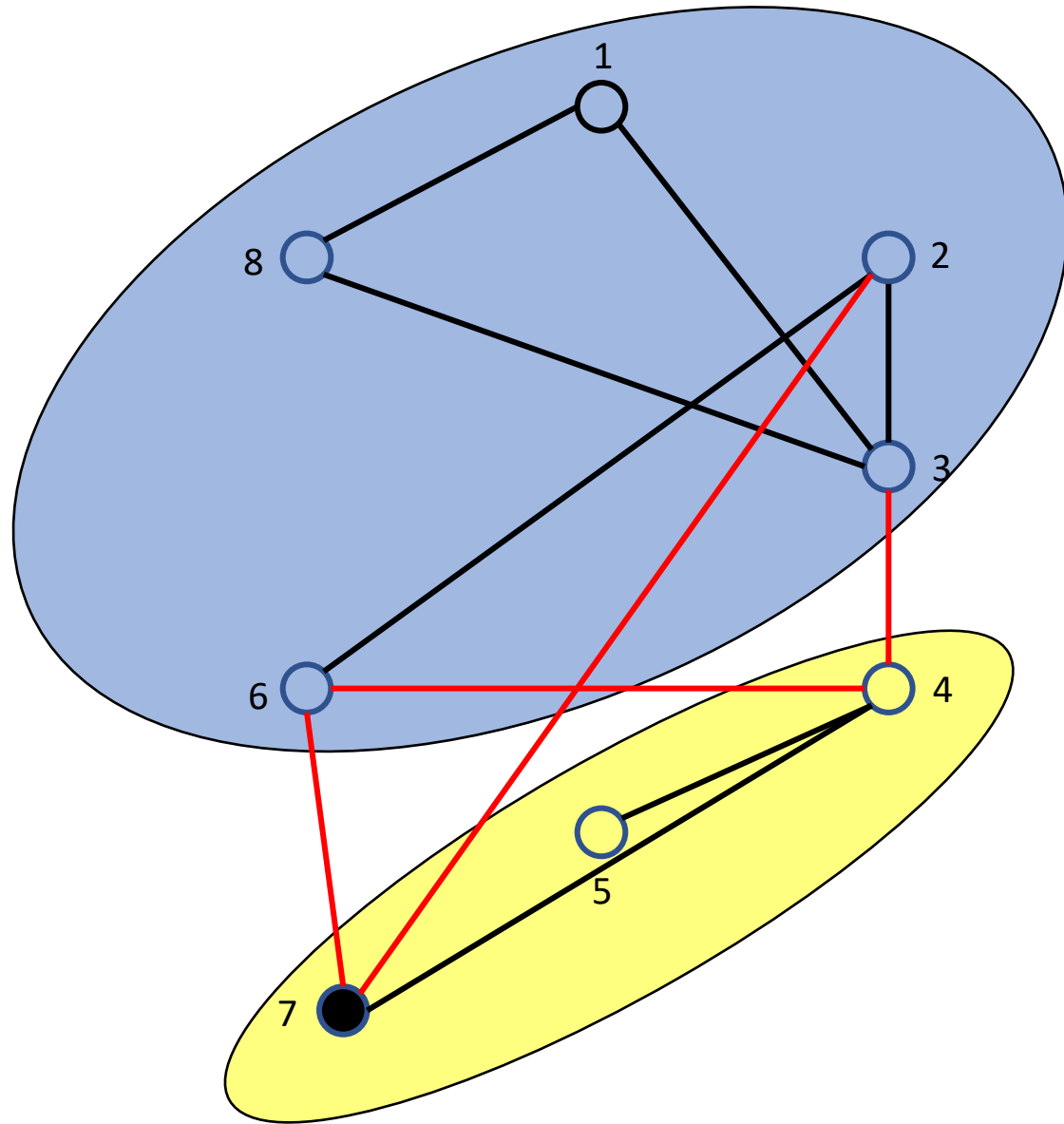
Now move nodes in a different order.



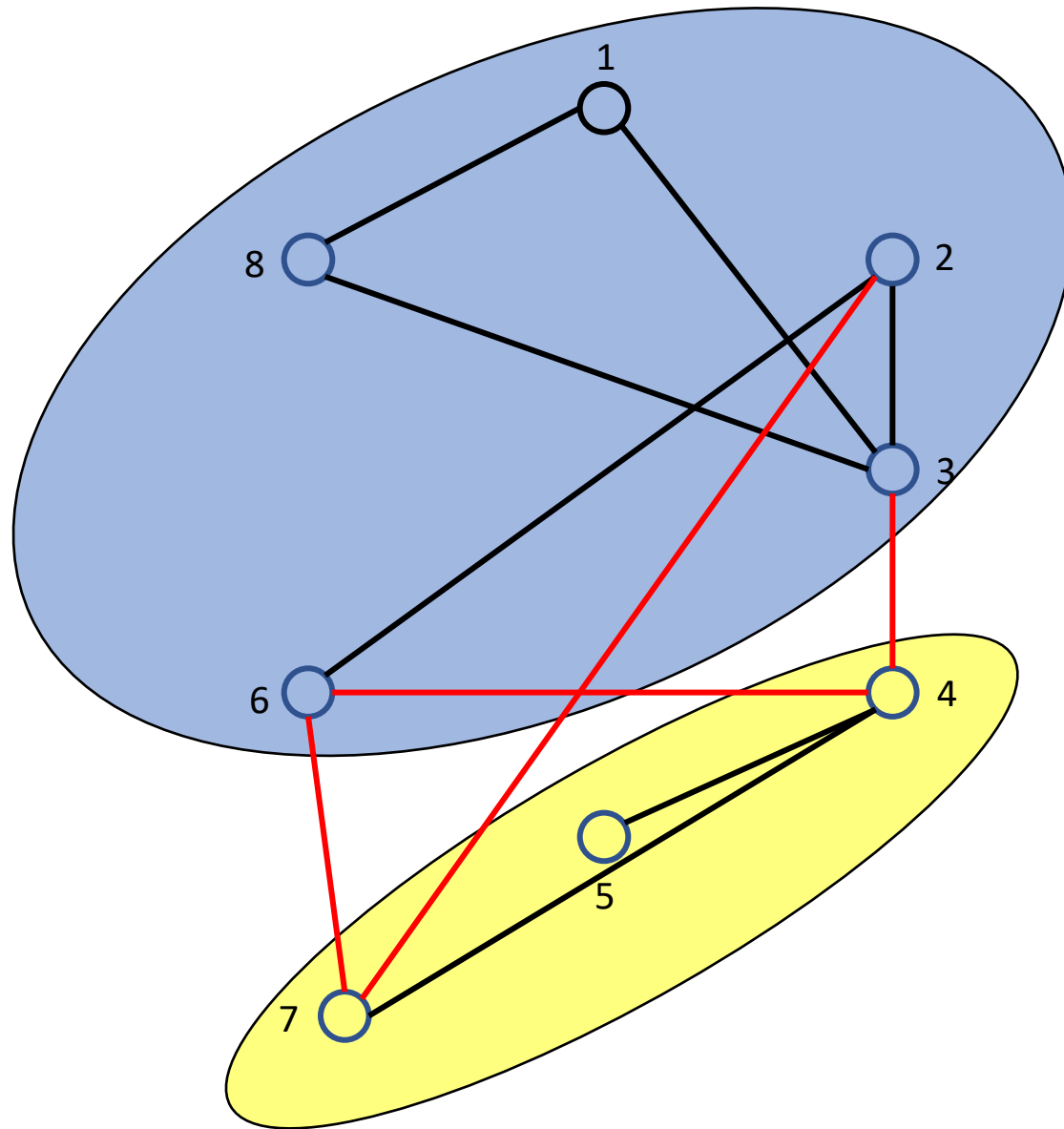
Node 7 has more **internal** edges (2) than **cross** edges (1).

Improve the number of cross edges by moving it to the yellow side (instead of moving node 5 to the blue side, as before).



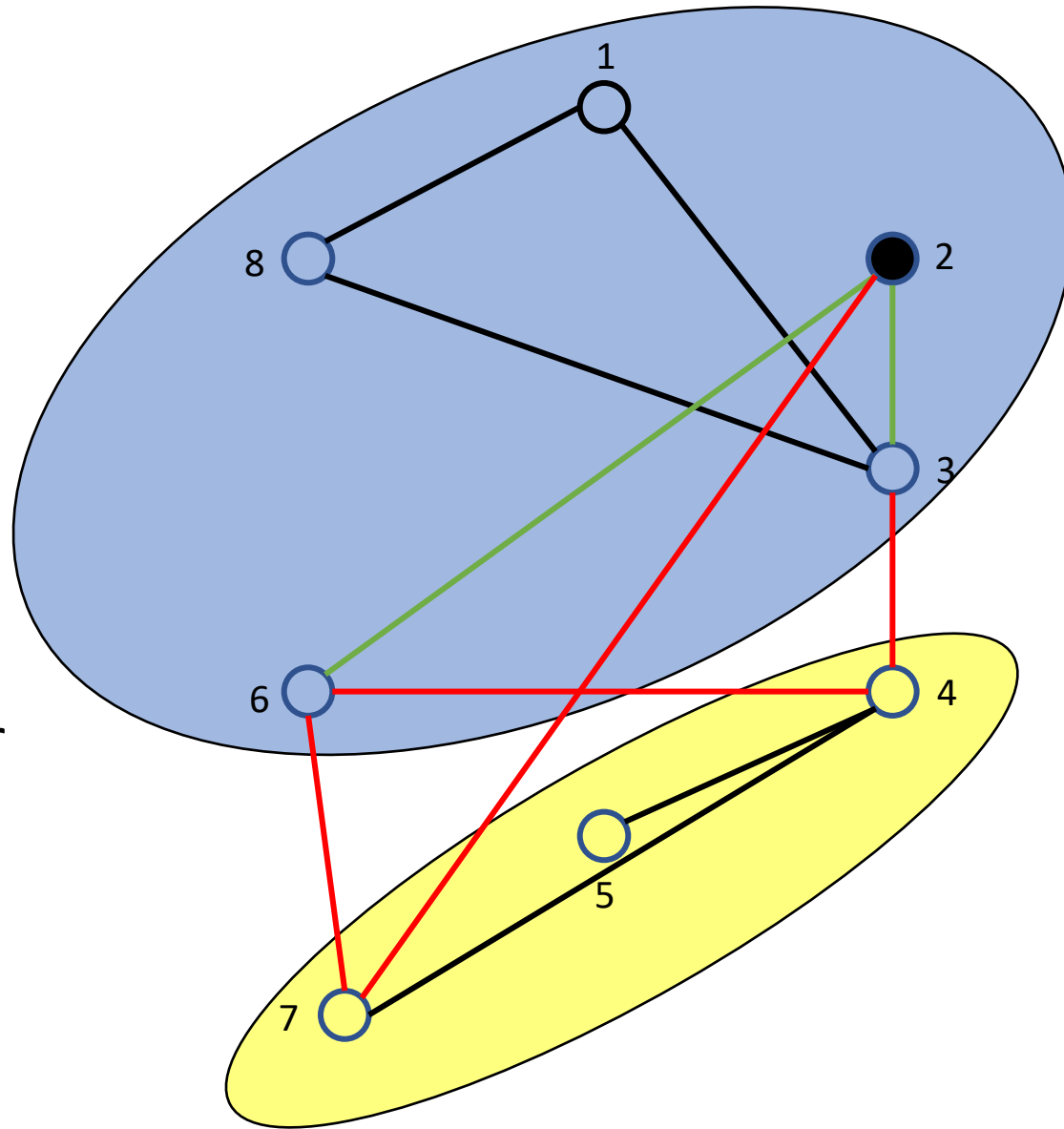


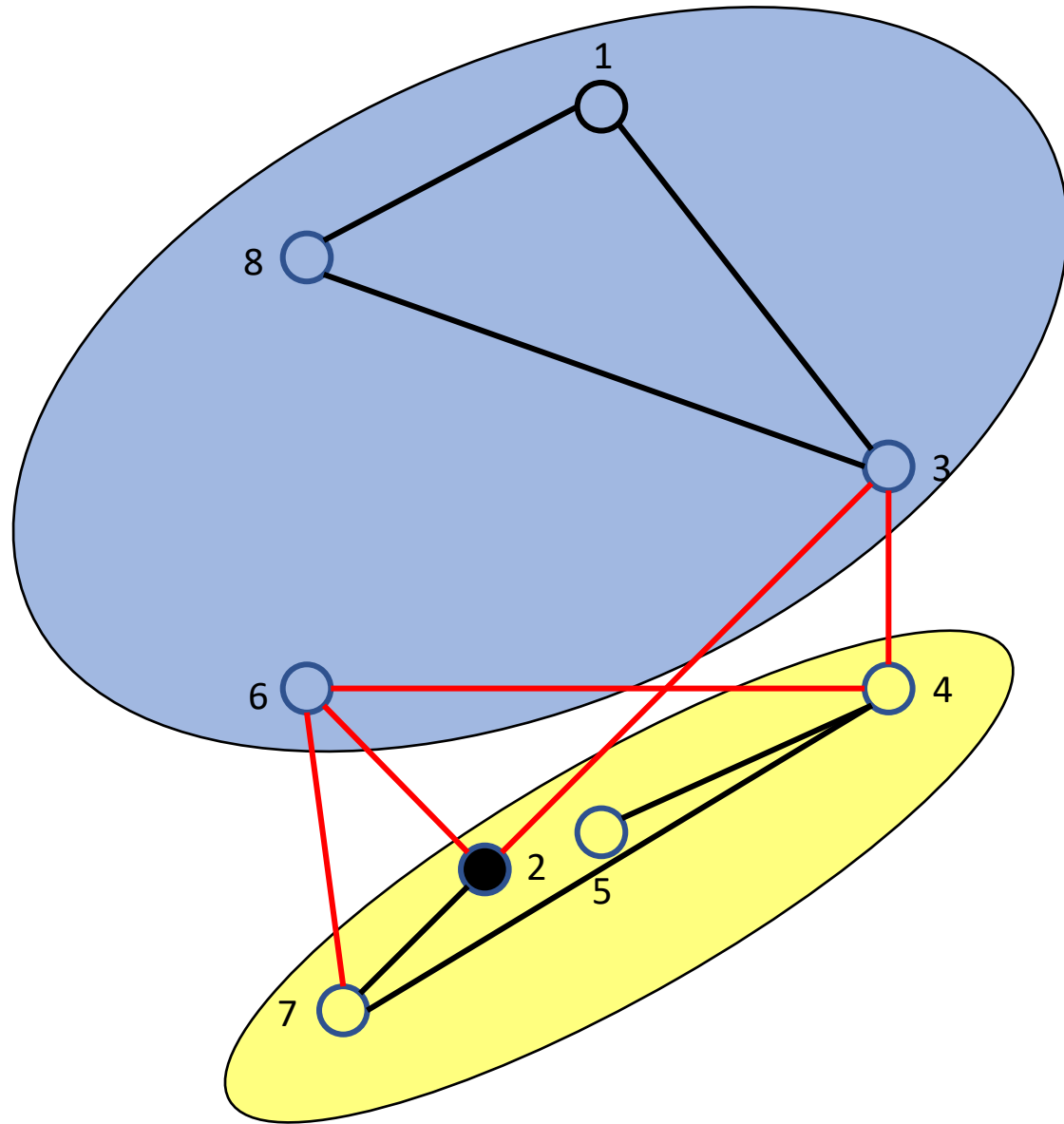
This cut has  
4 **cross** edges



Node 2 has more **internal** edges (2) than **cross** edges (1).

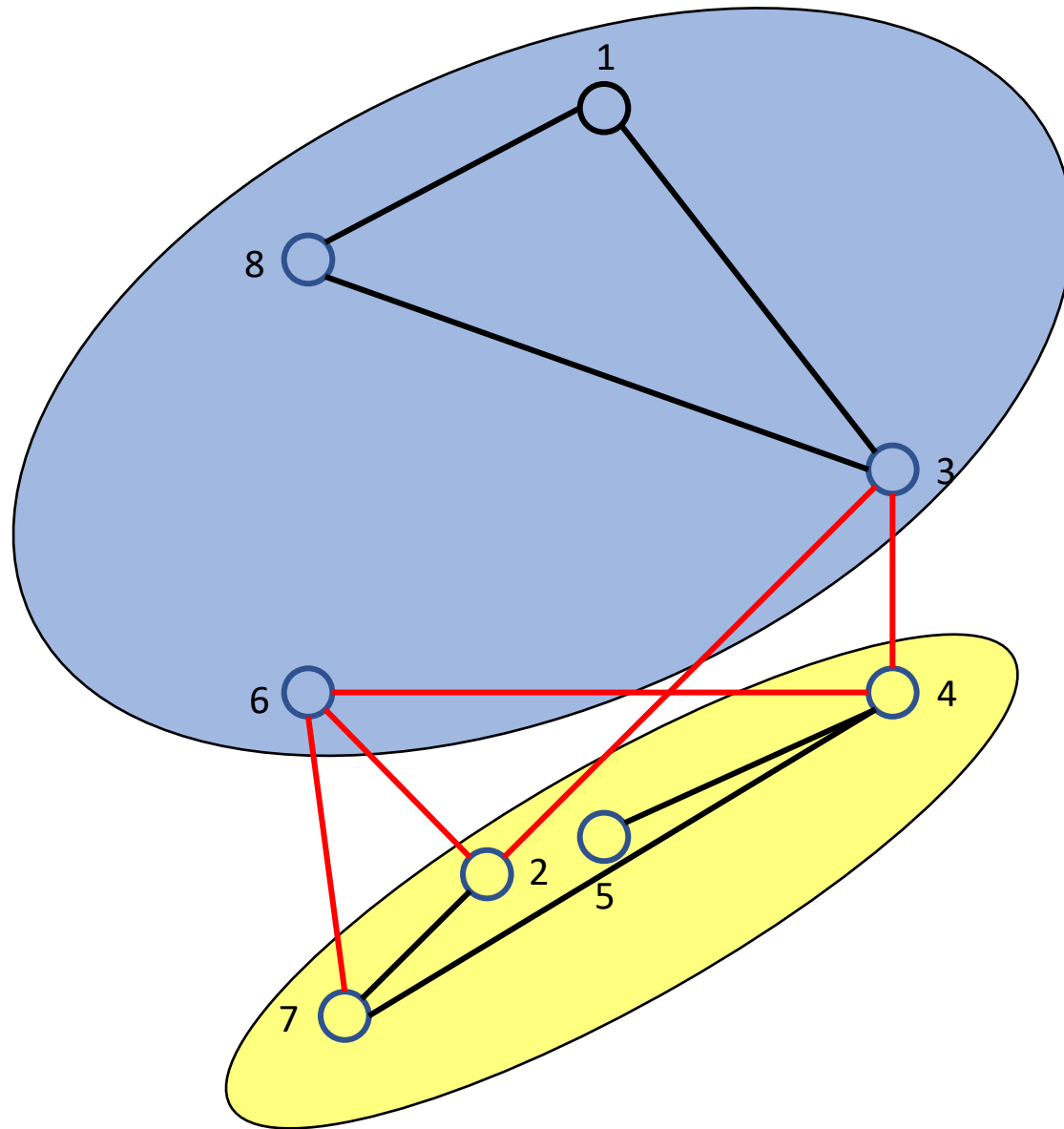
Improve the number of cross edges by moving it to the yellow side.







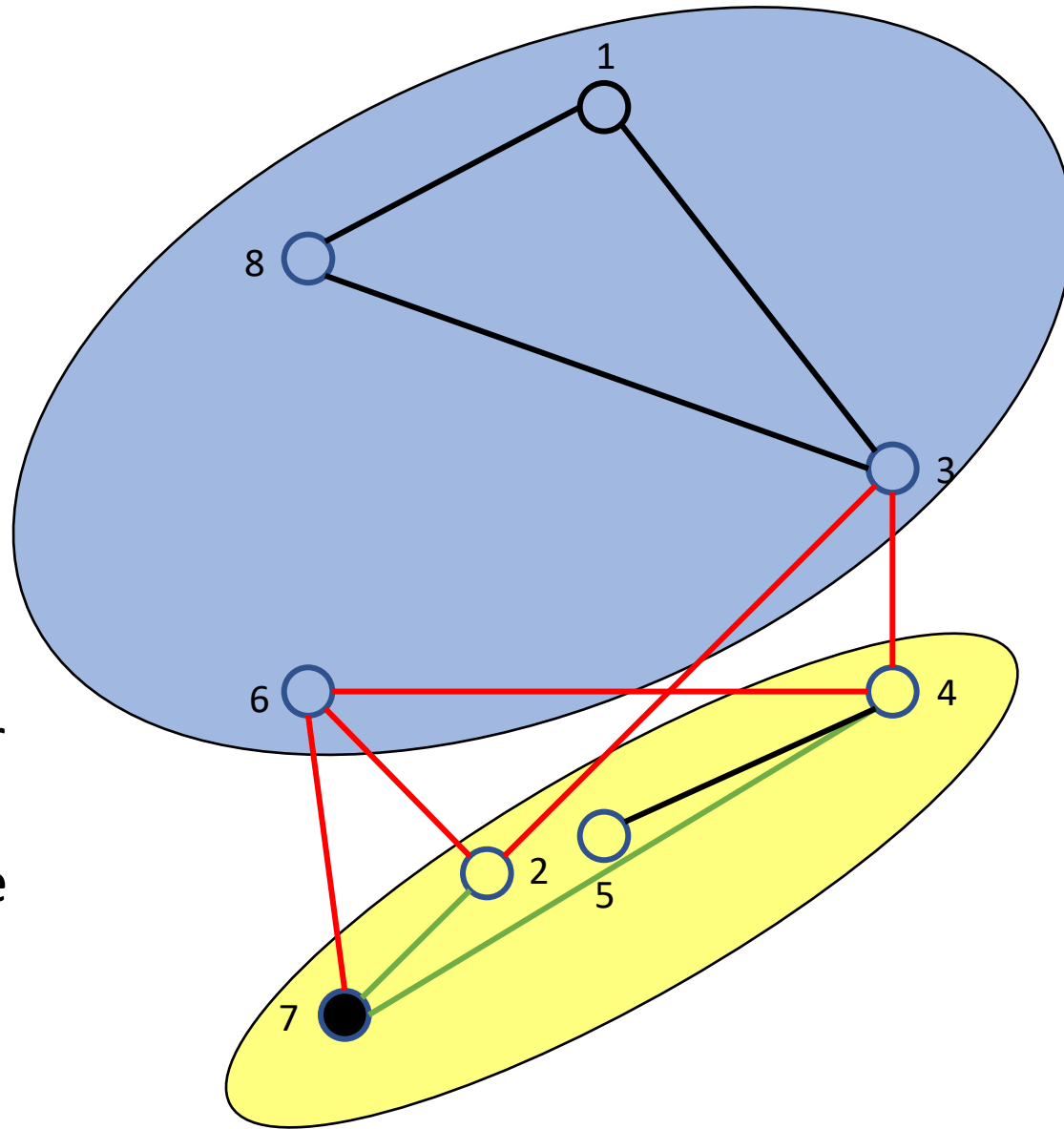
This cut has  
5 **cross** edges

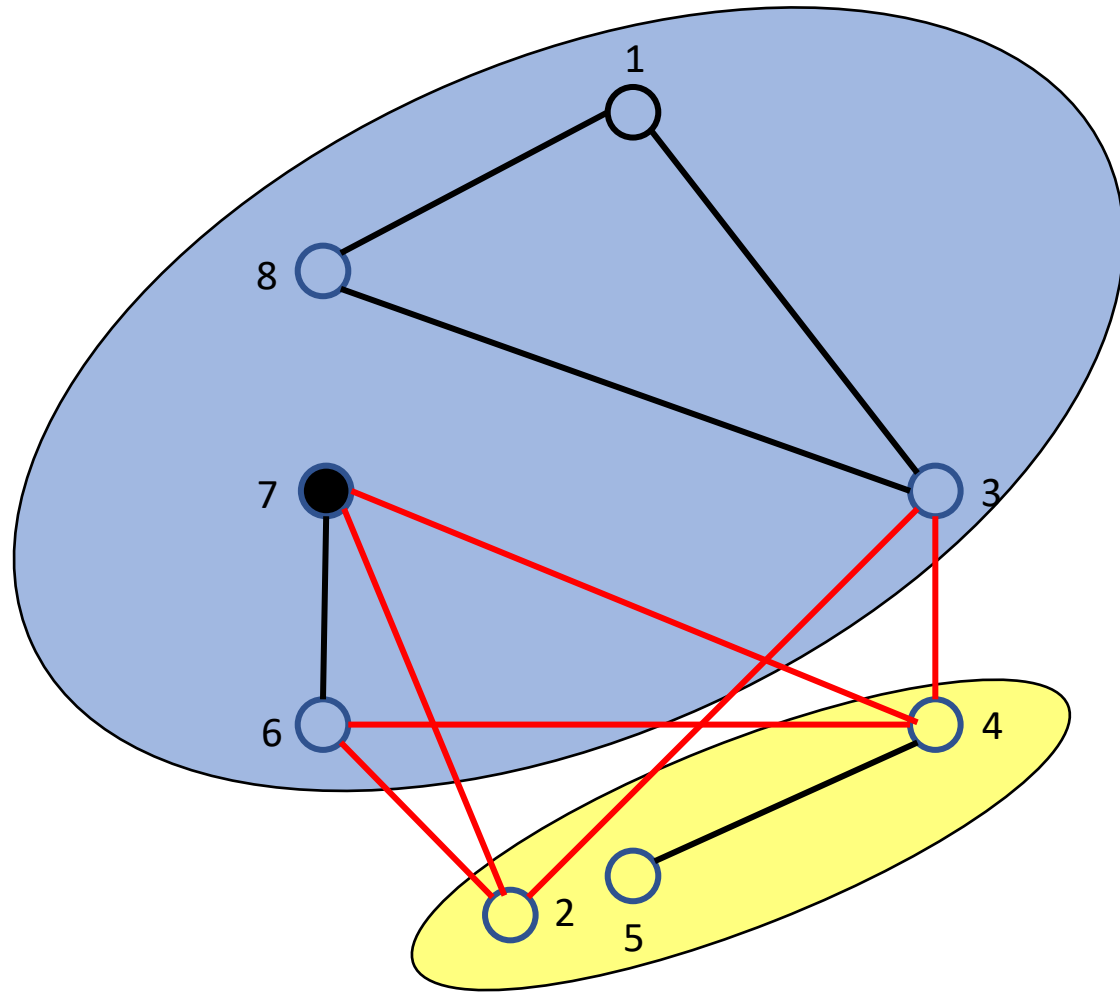


Node 7 has more **internal** edges (2) than **cross** edges (1).

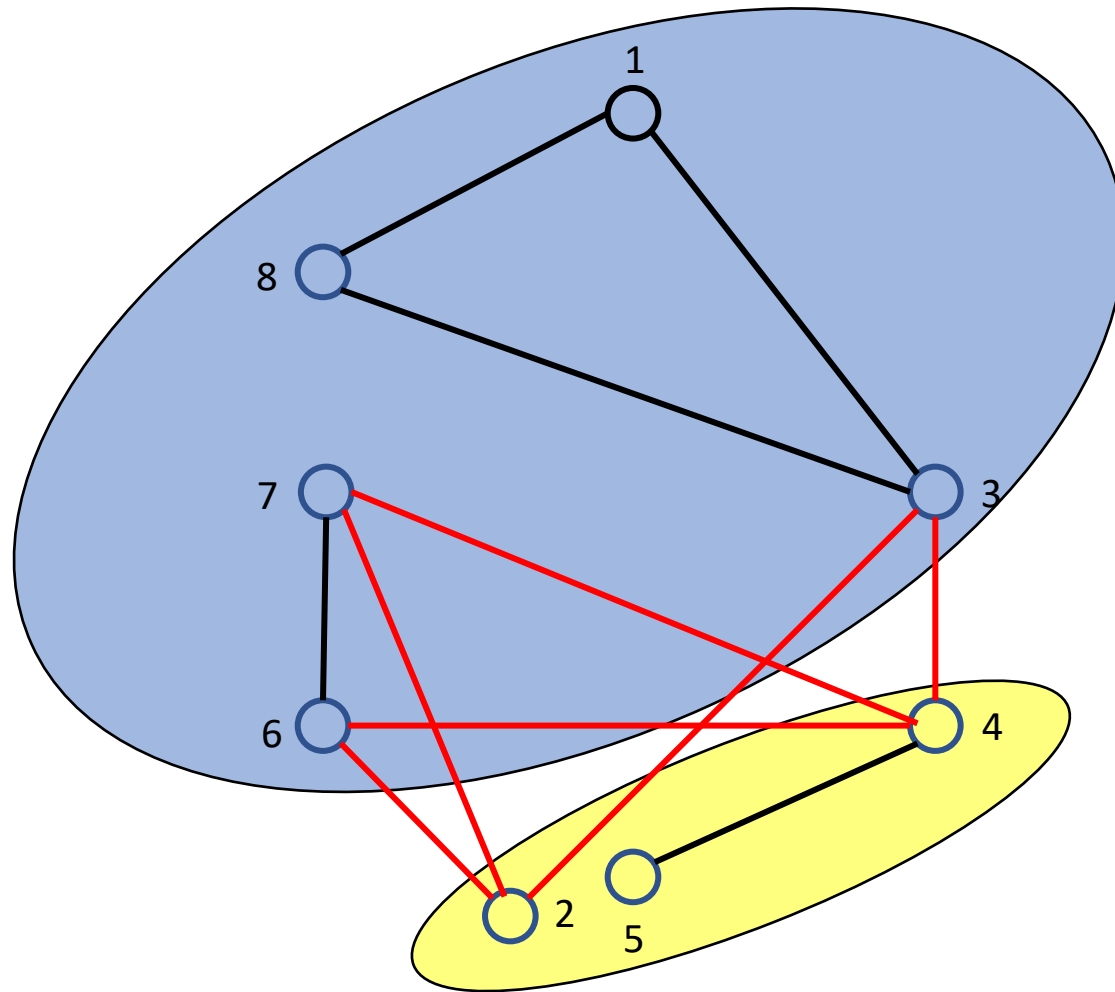
Improve the number of cross edges by moving it to the blue side.

NB: Moving back!



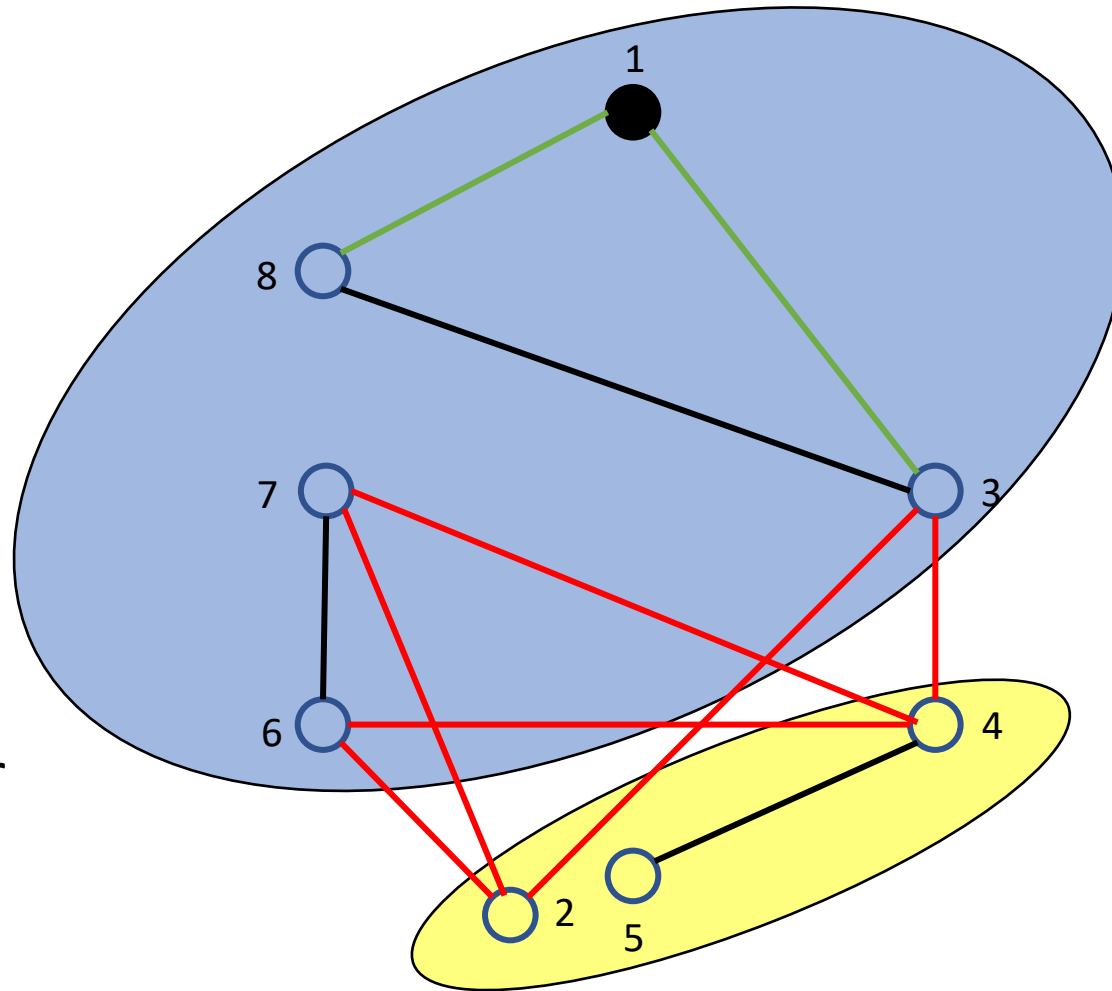


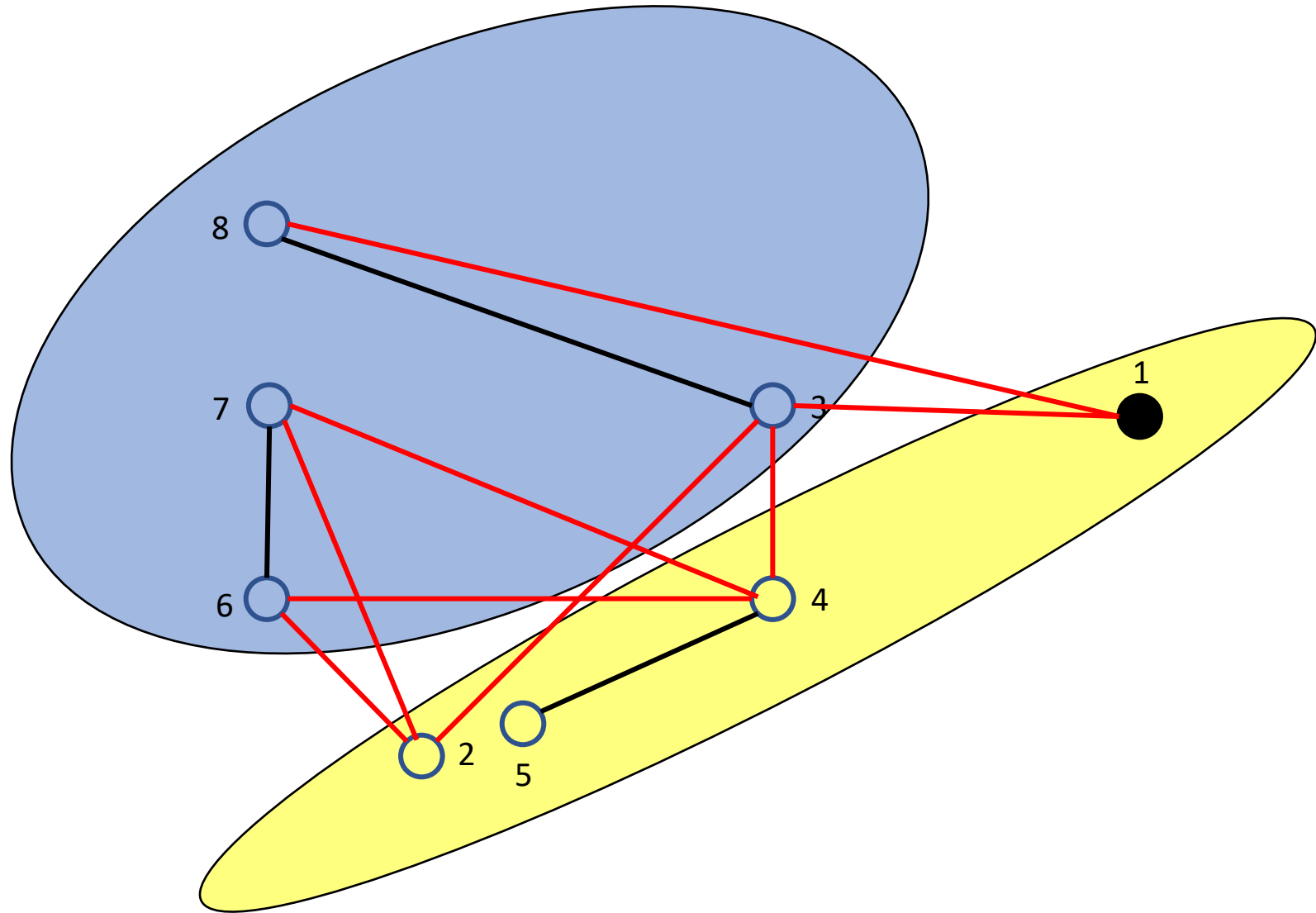
This cut has  
6 **cross** edges



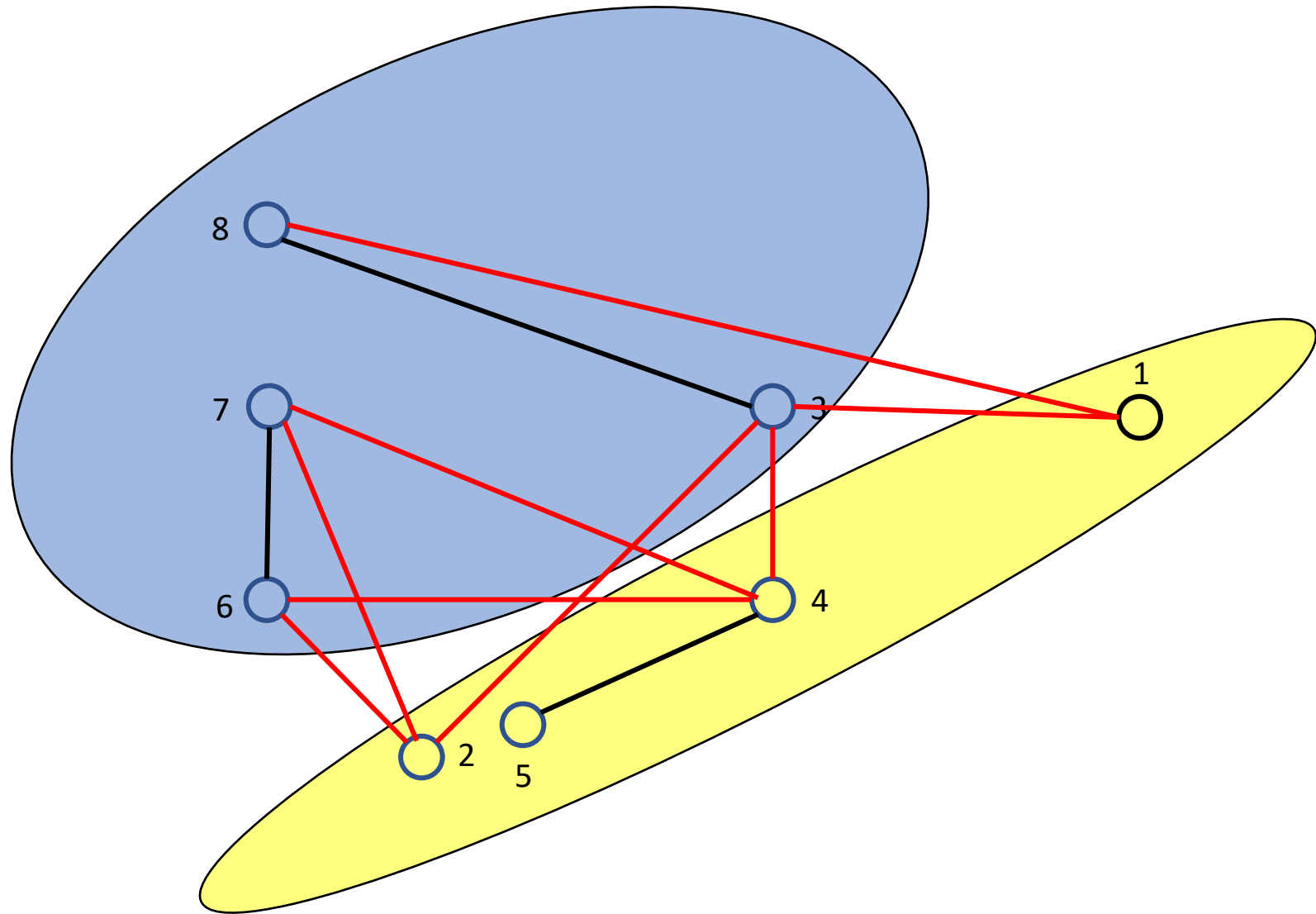
Node 1 has more **internal** edges (2) than **cross** edges (0).

Improve the number of cross edges by moving it to the yellow side.



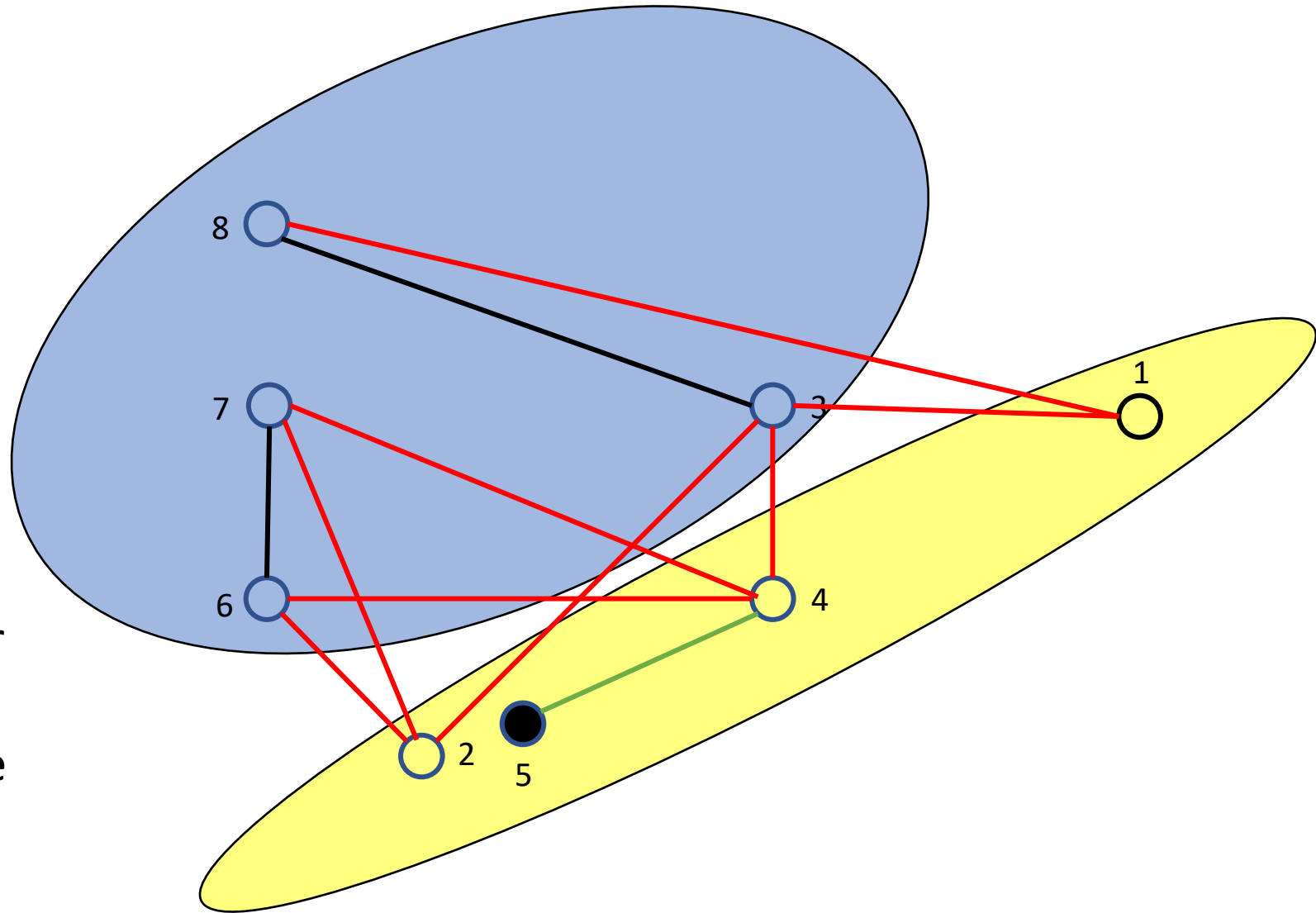


This cut has  
8 **cross** edges

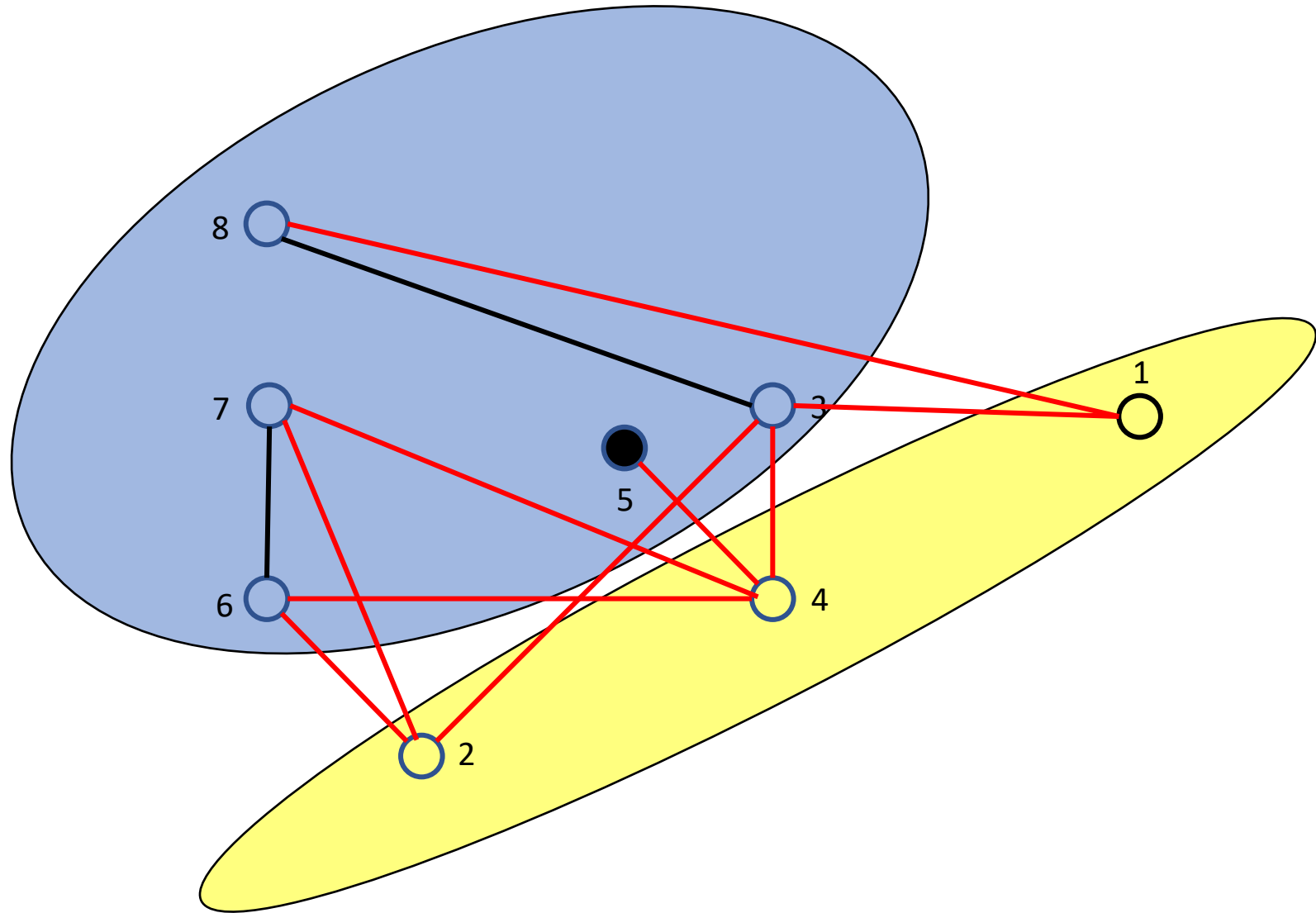


Node 5 has more **internal** edges (1) than **cross** edges (0).

Improve the number of cross edges by moving it to the blue side.





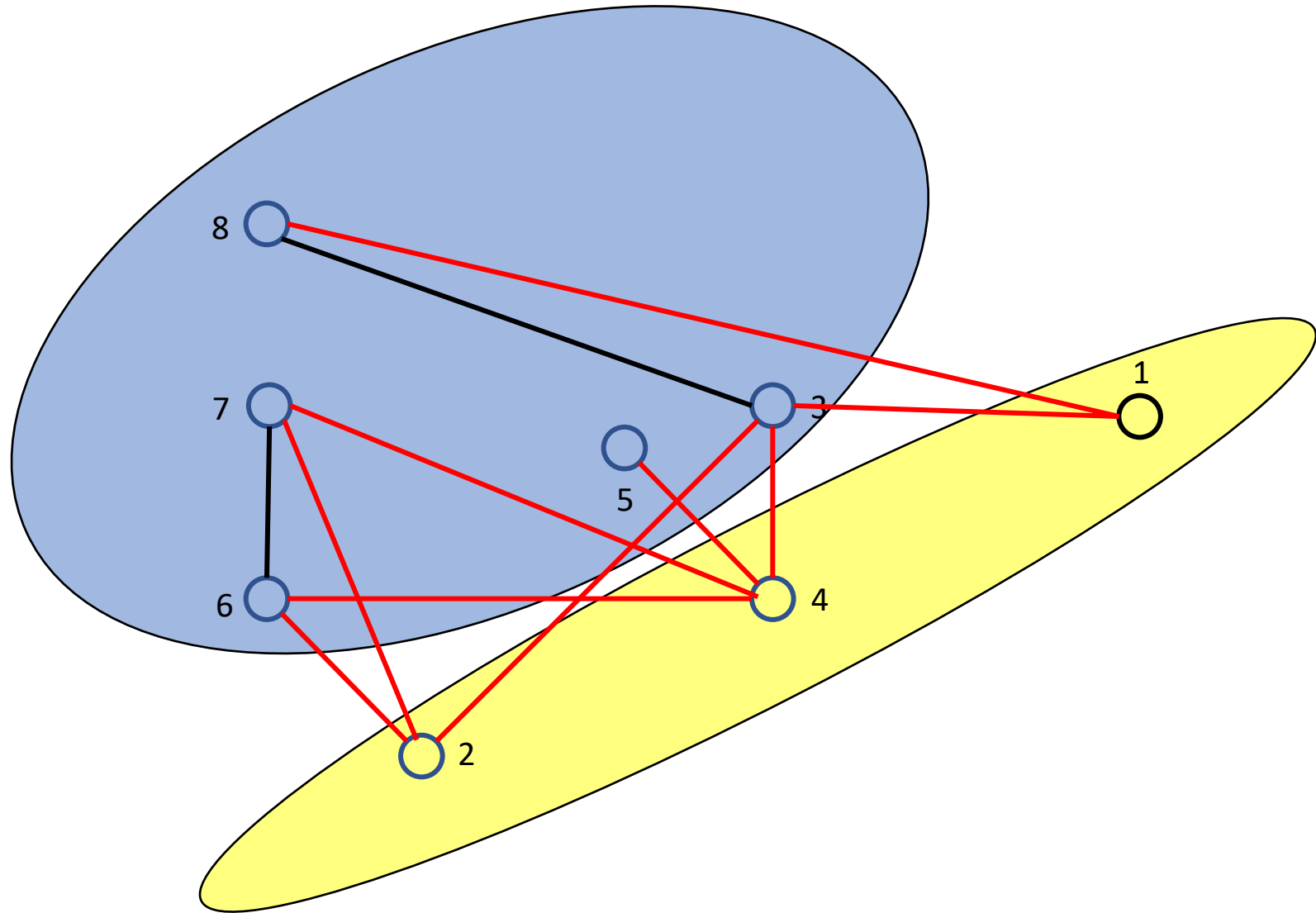


This cut has  
9 **cross** edges.

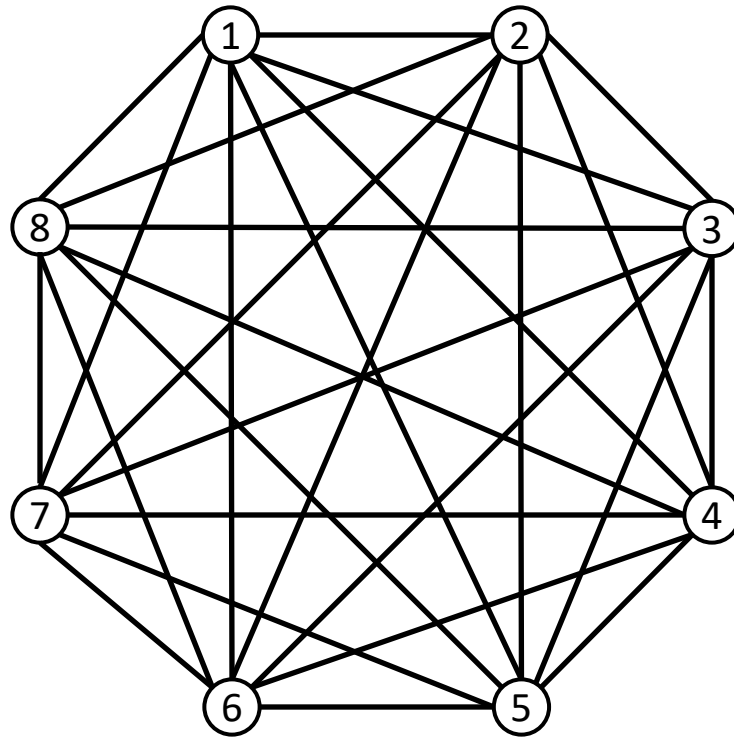
This is a max cut.

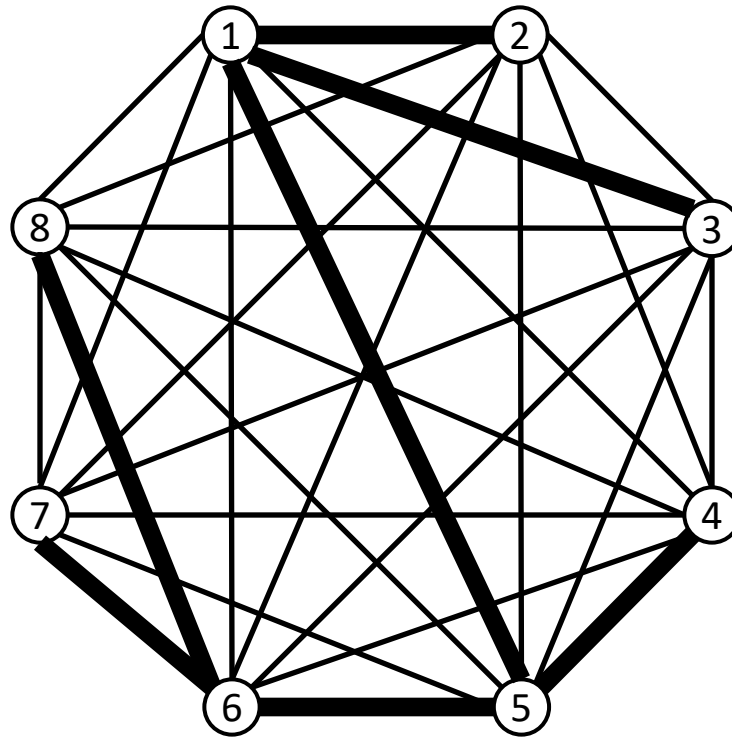
Why?

Hint: Disjoint  
triangles 1,3,8 and  
4,6,7.

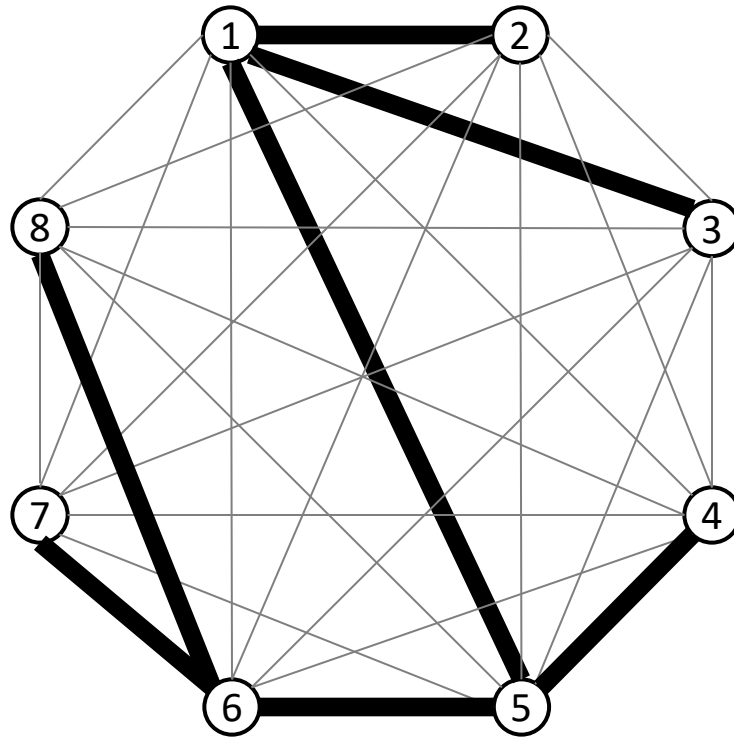


# Approximation algorithm for metric TSP

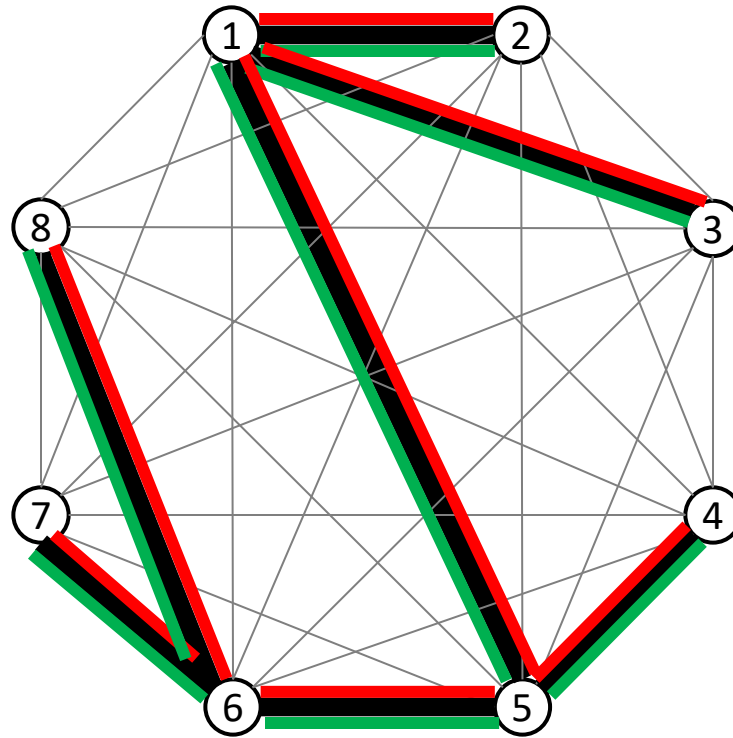




Step 1: Find a MST of the graph

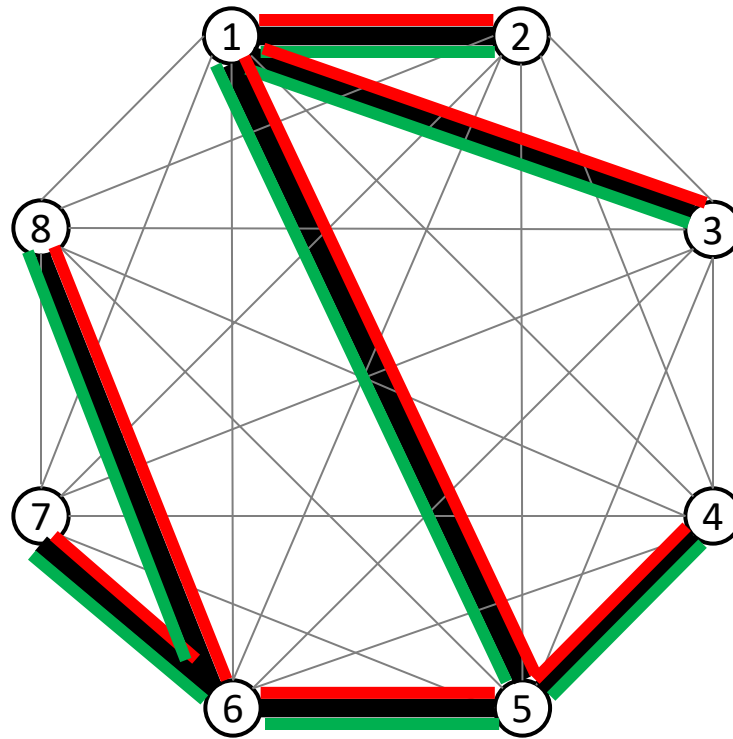


Step 1: Find a MST of the graph



Step 2: Do a DFS of the MST

(each edge of the MST is visited twice: once when **discovered** and once again when **backtracking**)

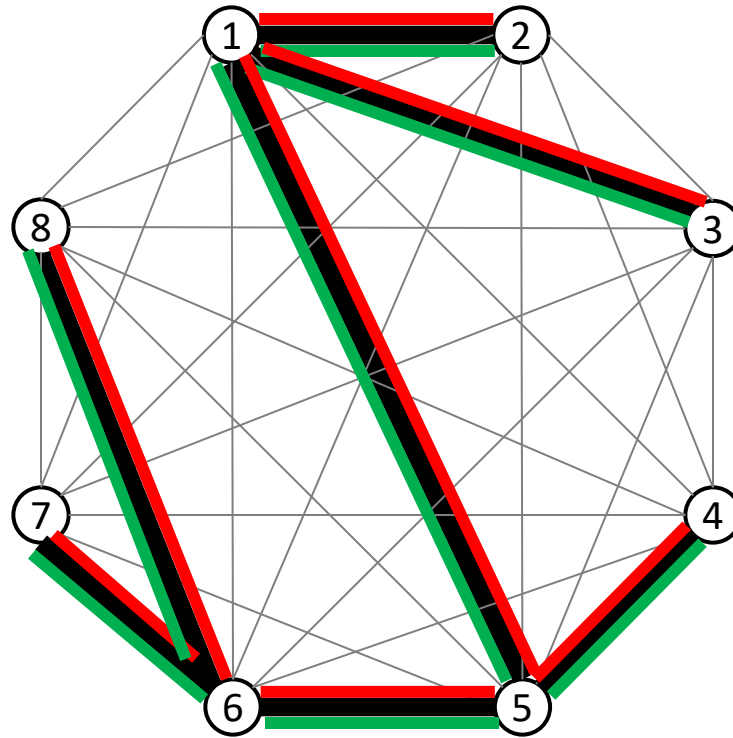


1, 2, 1, 3, 1, 5, 4, 5, 6, 8, 6, 7, 6, 5, 1

**Step 2: Do a DFS of the MST**

Record the sequence of nodes in the order visited





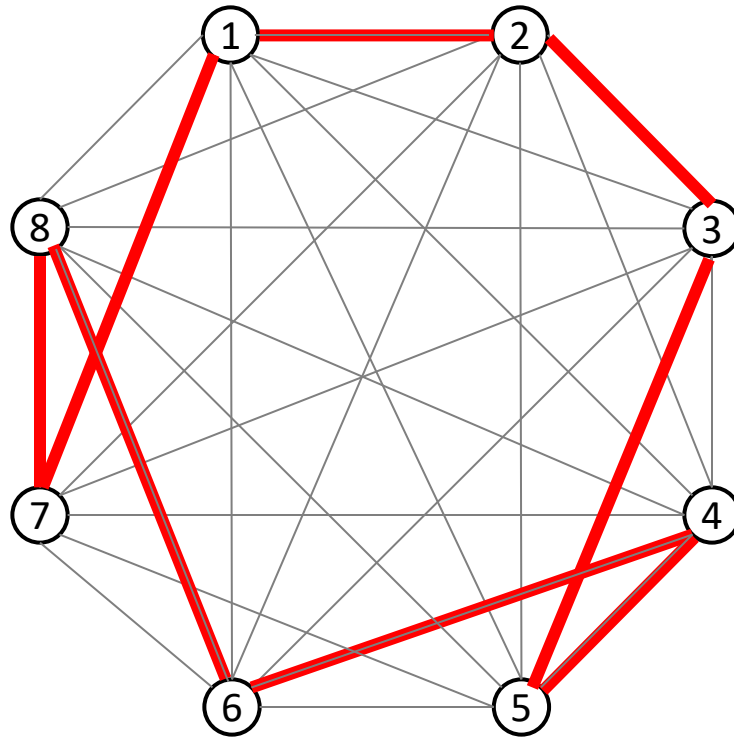
1, 2, 1, 3, 1, 5, 4, 5, 6, 8, 6, 7, 6, 5, 1

"Trail" produced by the DFS of the MST

1, 2, 3, 5, 4, 6, 8, 7, 1

Tour produced by the algorithm

Step 3: Keep only the first occurrence of each node, then back to the first node



1, 2, 3, 5, 4, 6, 8, 7, 1

Tour produced by the algorithm

**Step 3: This is the algorithm's tour**  
(its cost is at most twice the cost of the optimal tour)

# Metric TSP approximation algorithm

1. Find a MST  $T^*$  of the graph
2. Do a DFS of  $T^*$
3.  $S' :=$  sequence of nodes in the order visited by the DFS  
#  $S'$  is not a tour
4.  $S :=$  subsequence of  $S'$  containing only the first occurrence of each node, followed by the first node  
#  $S$  is a tour
5. return  $S$