



The k clique community finding algorithm

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Ignace Agbogba
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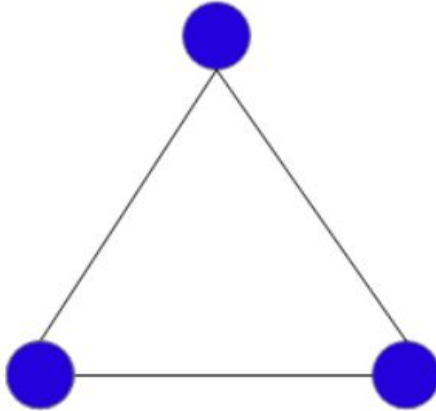


Description of the algorithm

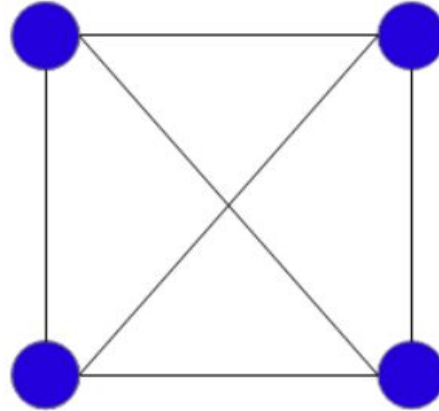
What are k-cliques?



1-clique



3-clique

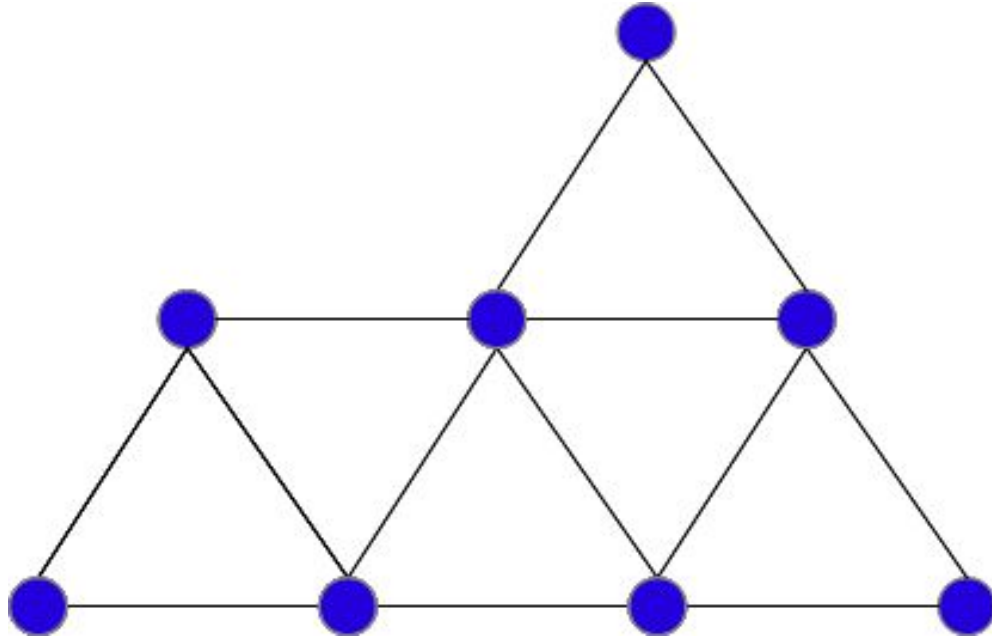


4-clique

Description of the algorithm

What is a k -clique community?

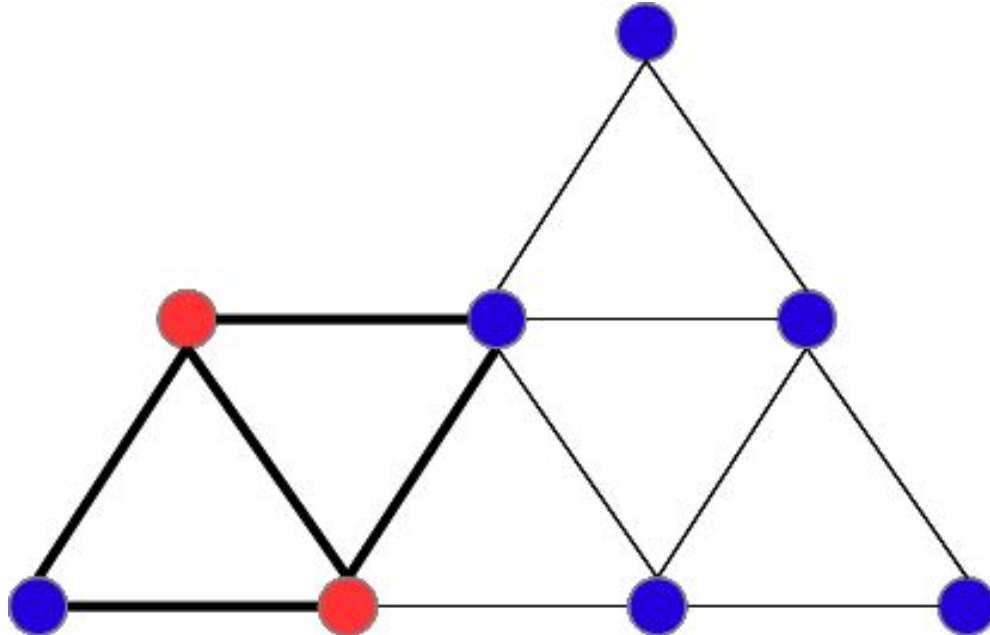
Union of k -cliques reached by adjacent k -cliques



Description of the algorithm

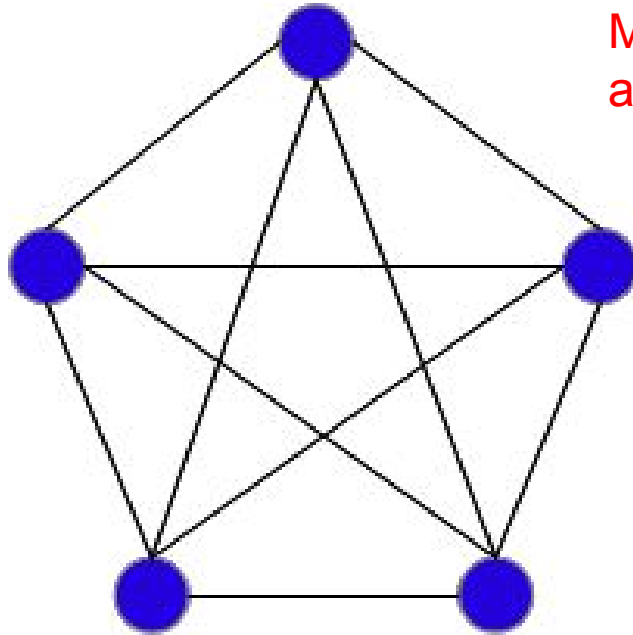
What is a k -clique community?

Union of k -cliques reached by adjacent k -cliques



Description of the algorithm

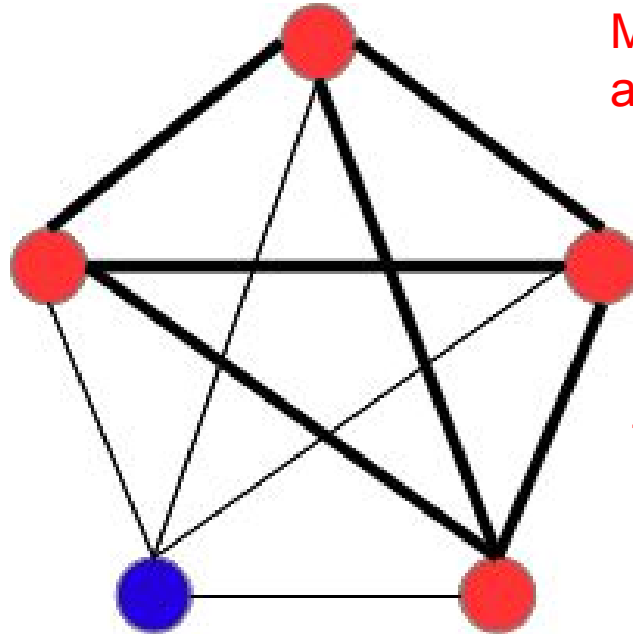
From cliques to k-cliques



Maximal complete subgraphs
are called cliques

Description of the algorithm

From cliques to k-cliques

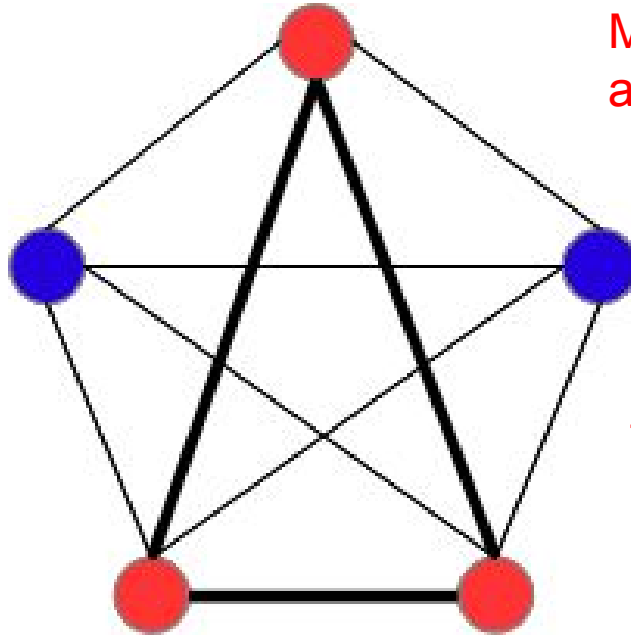


Maximal complete subgraphs
are called cliques

k-cliques can be subsets of
larger complete subgraphs

Description of the algorithm

From cliques to k-cliques

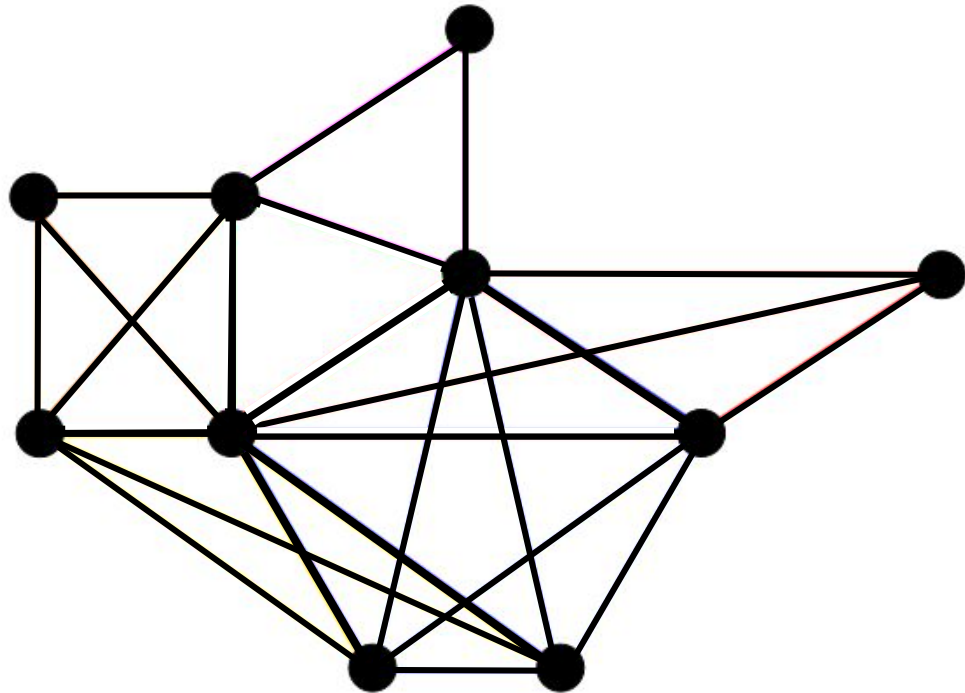


Maximal complete subgraphs
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k-cliques can be subsets of
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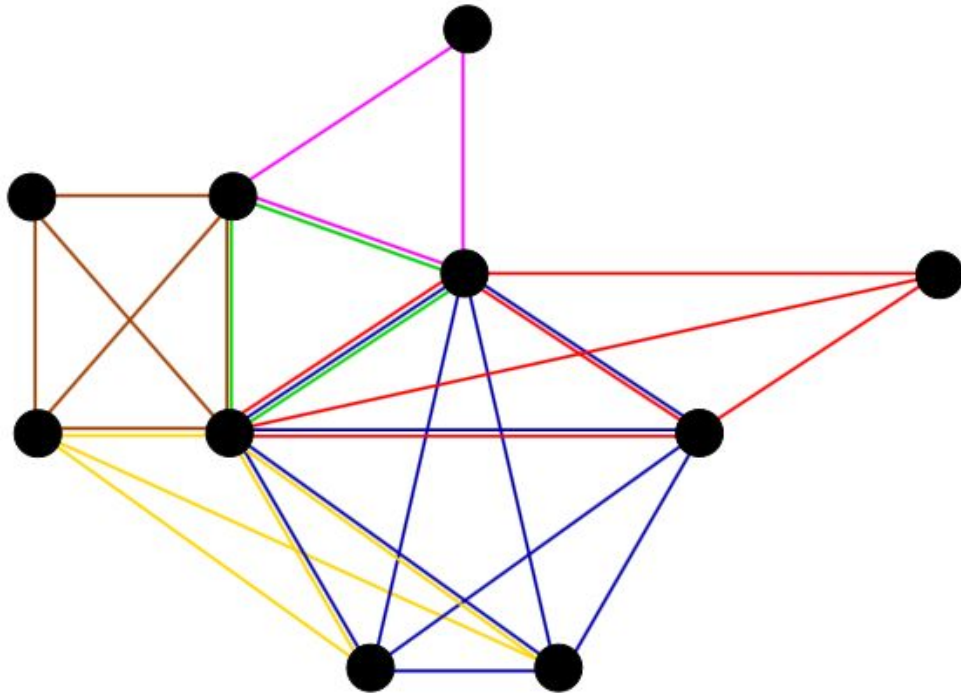
Description of the algorithm

Applying the algorithm



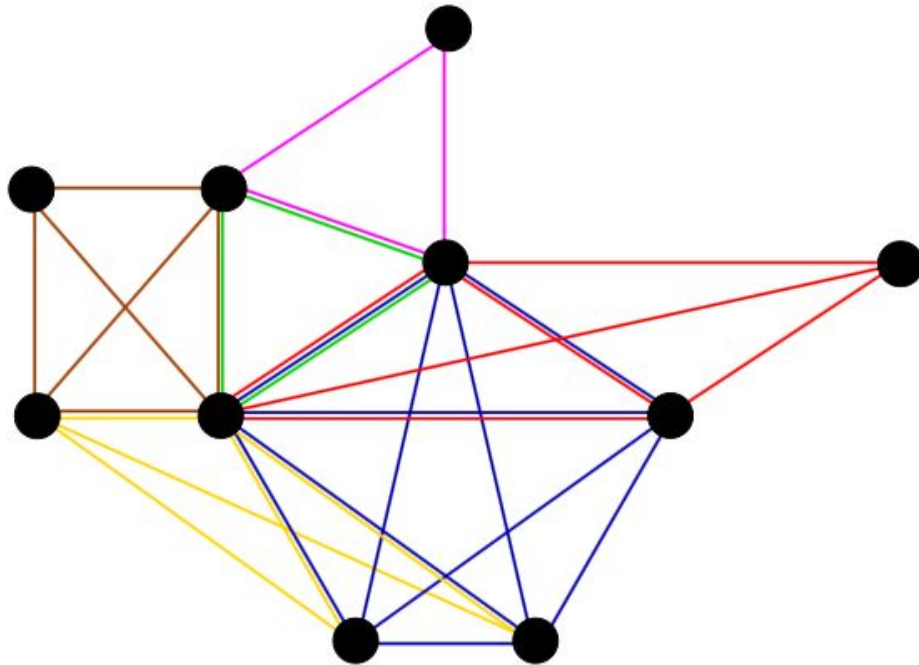
Description of the algorithm












1 - get the cliques



Description of the algorithm

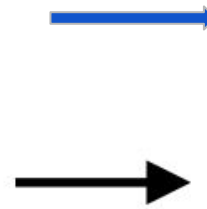
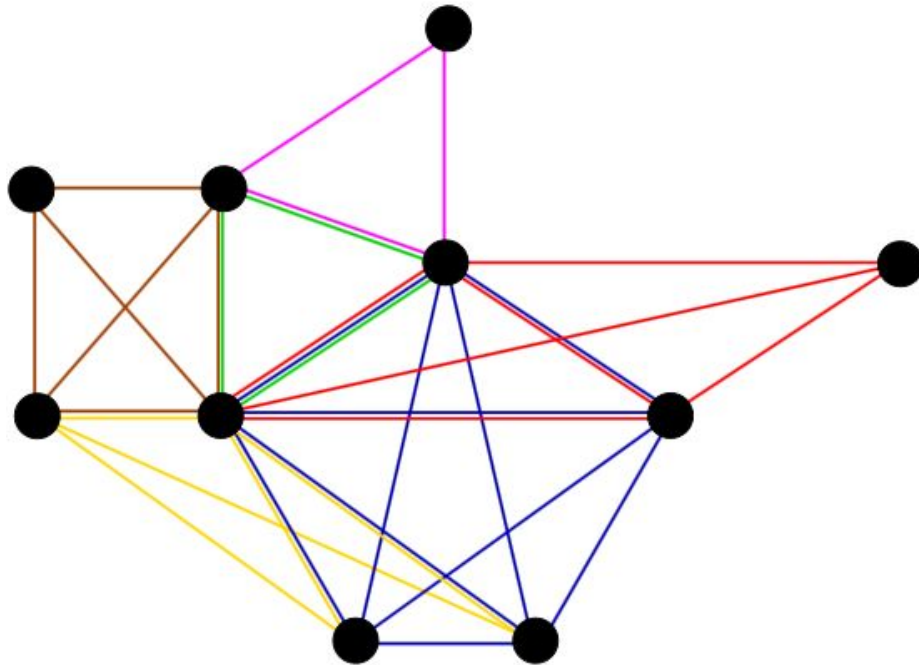
2 - get the overlap matrix













						
	5	3	2	1	3	1
	3	4	2	1	1	1
	2	2	3	2	1	2
	1	1	2	3	0	1
	3	1	1	0	4	2
	1	1	2	1	2	4

Description of the algorithm


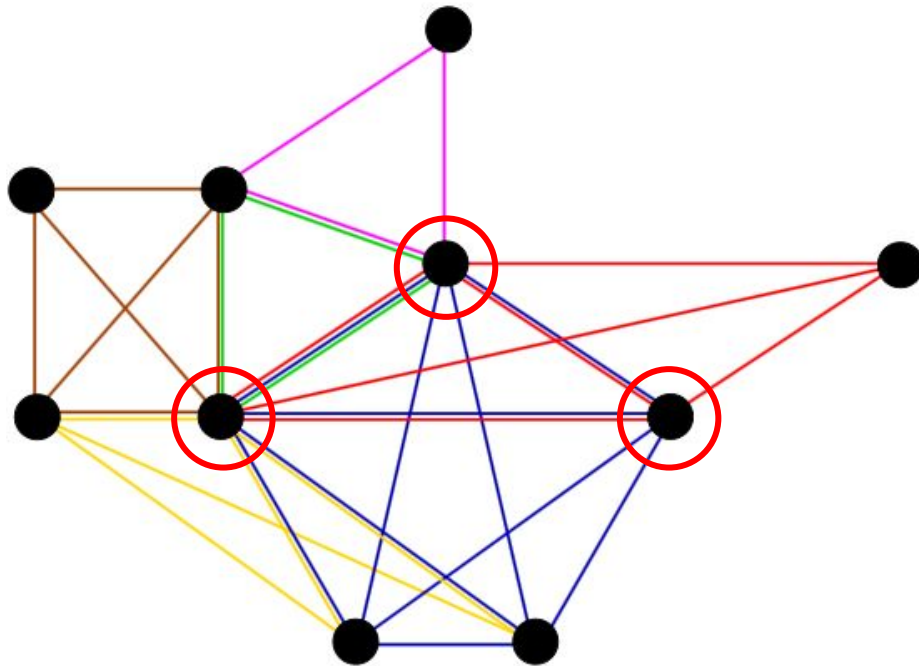
2 - get the overlap matrix















						
	5	3	2	1	3	1
	3	4	2	1	1	1
	2	2	3	2	1	2
	1	1	2	3	0	1
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Description of the algorithm

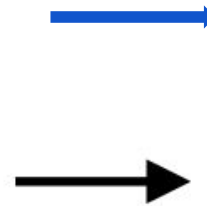
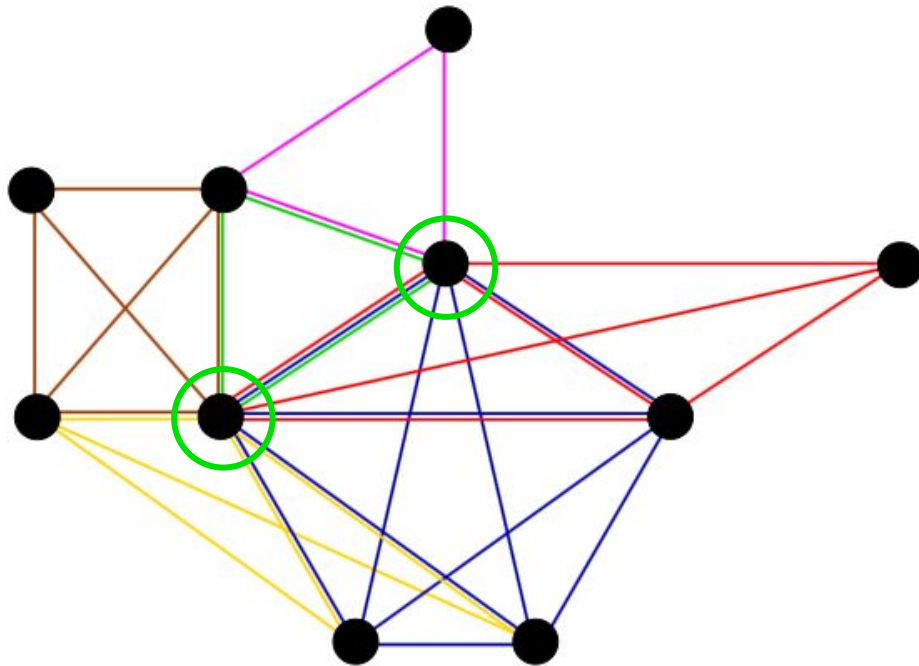
2 - get the overlap matrix











						
	5	3	2	1	3	1
	3	4	2	1	1	1
	2	2	3	2	1	2
	1	1	2	3	0	1
	3	1	1	0	4	2
	1	1	2	1	2	4

Description of the algorithm

2 - get the overlap matrix

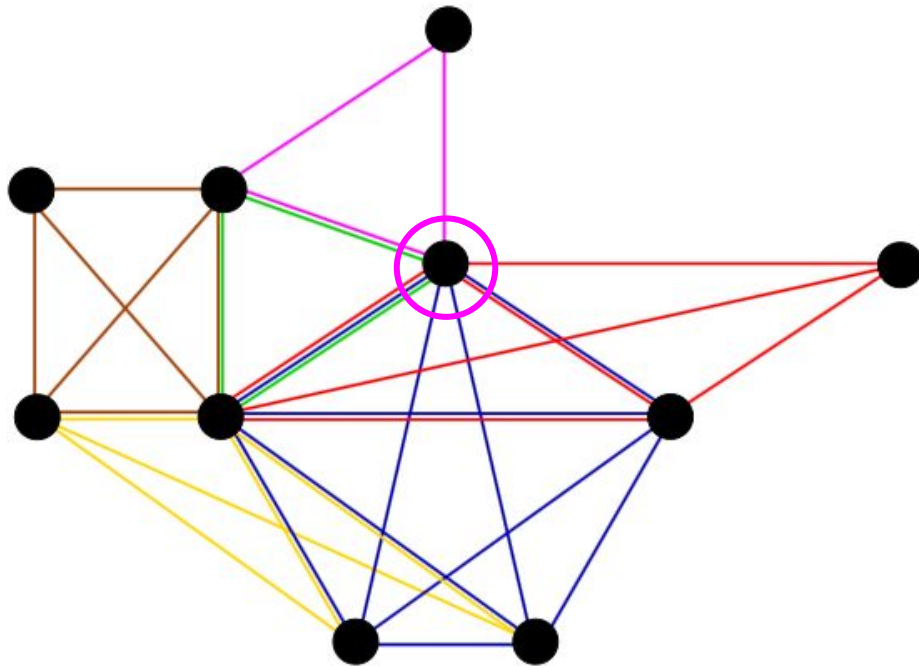












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	5	3	2	1	3	1
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	1	1	2	3	0	1
	3	1	1	0	4	2
	1	1	2	1	2	4

Description of the algorithm

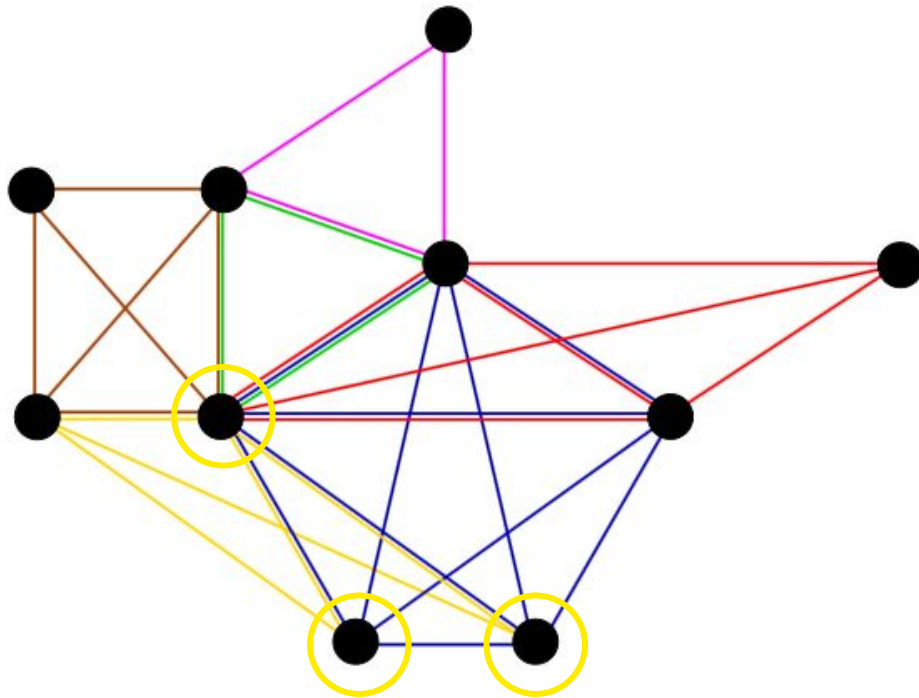
2 - get the overlap matrix












						
	5	3	2	1	3	1
	3	4	2	1	1	1
	2	2	3	2	1	2
	1	1	2	3	0	1
	3	1	1	0	4	2
	1	1	2	1	2	4

Description of the algorithm

2 - get the overlap matrix

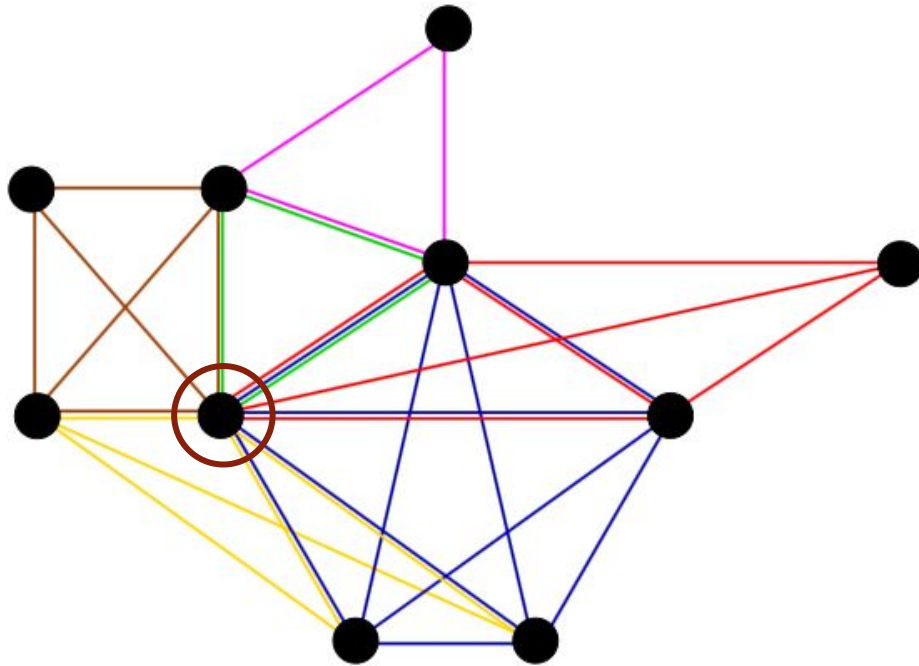










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	5	3	2	1	3	1
	3	4	2	1	1	1
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	1	1	2	3	0	1
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Description of the algorithm

2 - get the overlap matrix



						
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	2	2	3	2	1	2
	1	1	2	3	0	1
	3	1	1	0	4	2
	1	1	2	1	2	4



Description of the algorithm


3 - get the communities matrix

Delete diagonal elements that are smaller than k

Delete off-diagonal elements that are smaller than $k-1$

	Blue	Red	Green	Pink	Yellow	Brown
Blue	5	3	2	1	3	1
Red	3	4	2	1	1	1
Green	2	2	3	2	1	2
Pink	1	1	2	3	0	1
Yellow	3	1	1	0	4	2
Brown	1	1	2	1	2	4













$k=4$

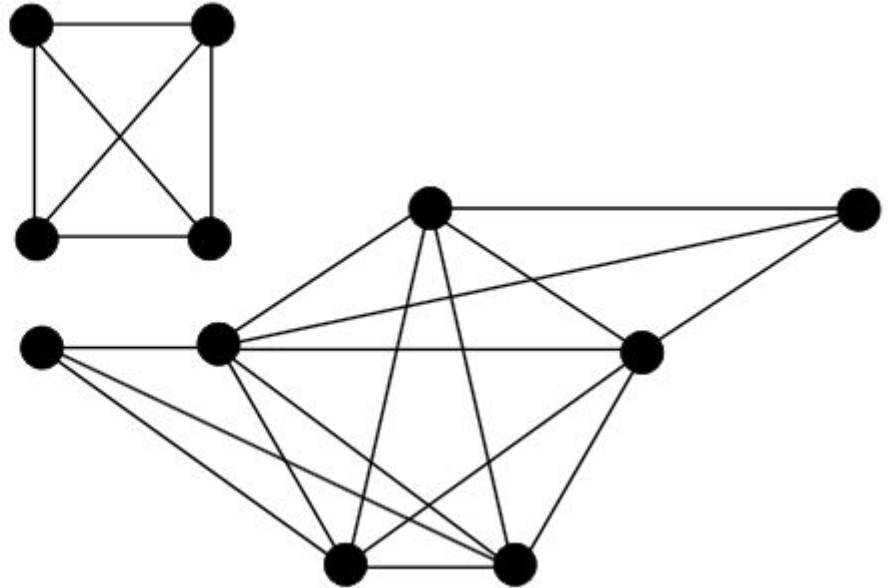


	Blue	Red	Green	Pink	Yellow	Brown
Blue	1	1	0	0	1	0
Red	1	1	0	0	0	0
Green	0	0	0	0	0	0
Pink	0	0	0	0	0	0
Yellow	1	0	0	0	1	0
Brown	0	0	0	0	0	1

Description of the algorithm













4 - get the k-clique communities

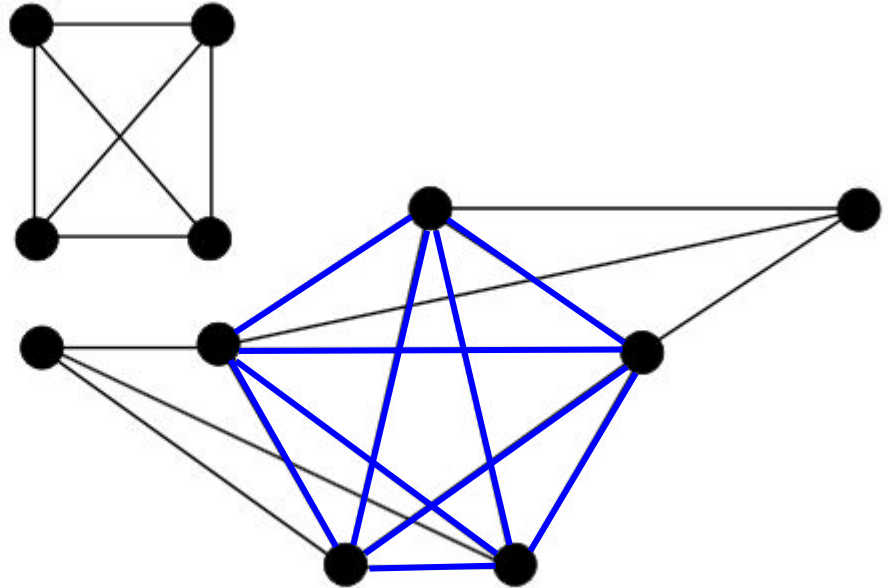
						
	1	1	0	0	1	0
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	0	0	0	0	0	0
	0	0	0	0	0	0
	1	0	0	0	1	0
	0	0	0	0	0	1



Description of the algorithm













4 - get the k-clique communities

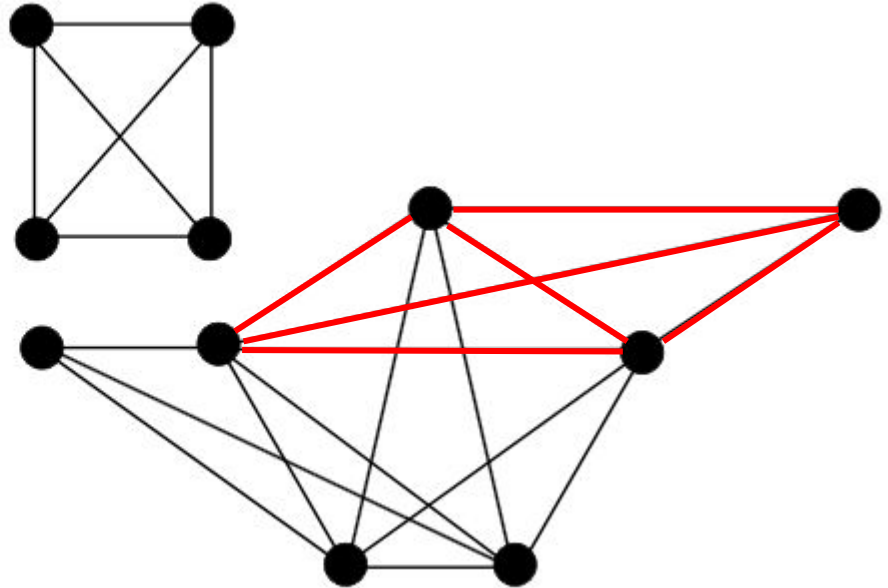
						
	1	1	0	0	1	0
	1	1	0	0	0	0
	0	0	0	0	0	0
	0	0	0	0	0	0
	1	0	0	0	1	0
	0	0	0	0	0	1



Description of the algorithm













4 - get the k-clique communities

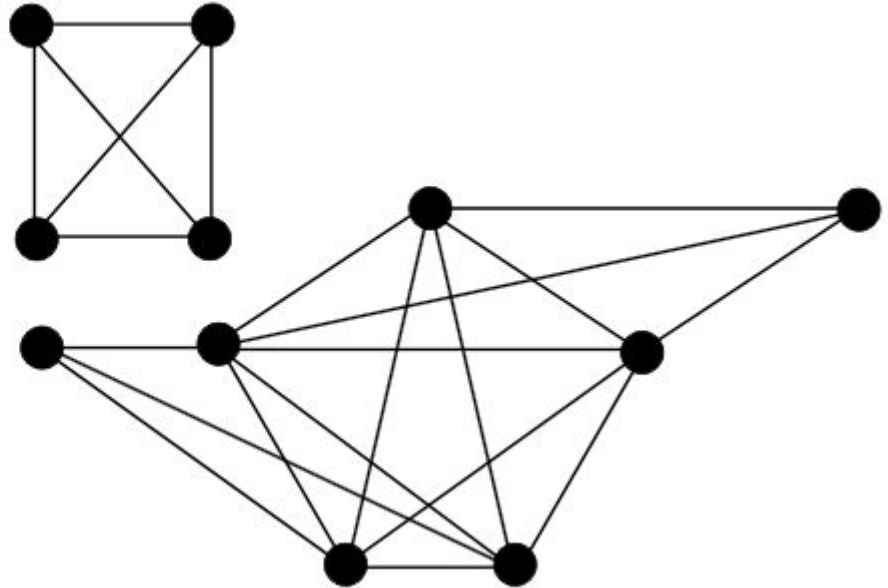
						
	1	1	0	0	1	0
	1	1	0	0	0	0
	0	0	0	0	0	0
	0	0	0	0	0	0
	1	0	0	0	1	0
	0	0	0	0	0	1



Description of the algorithm













4 - get the k-clique communities

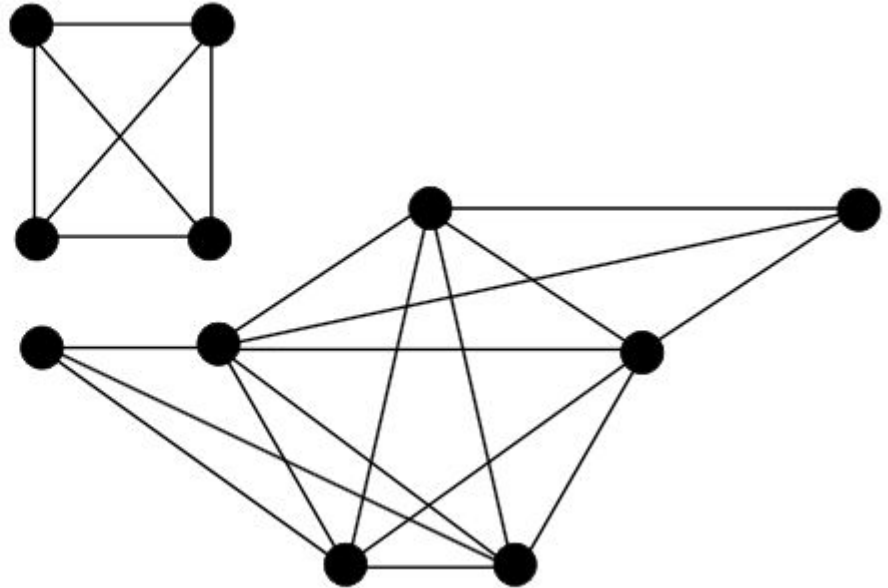
						
	1	1	0	0	1	0
	1	1	0	0	0	0
	0	0	0	0	0	0
	0	0	0	0	0	0
	1	0	0	0	1	0
	0	0	0	0	0	1



Description of the algorithm













4 - get the k-clique communities

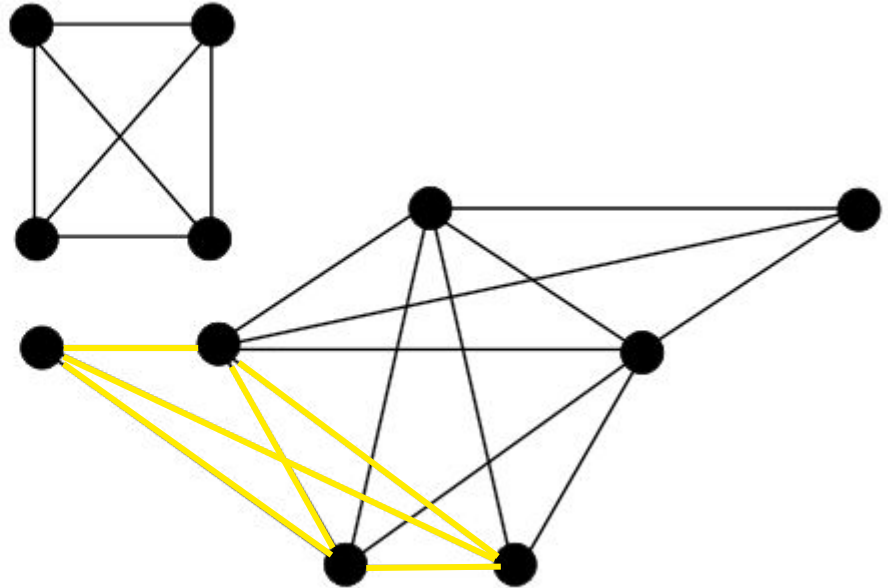
						
	1	1	0	0	1	0
	1	1	0	0	0	0
	0	0	0	0	0	0
	0	0	0	0	0	0
	1	0	0	0	1	0
	0	0	0	0	0	1



Description of the algorithm













4 - get the k-clique communities

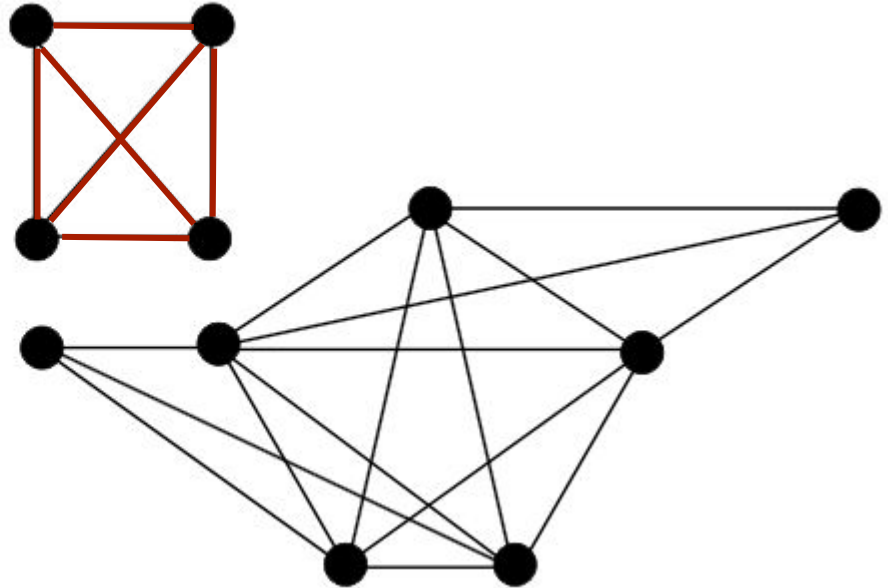
						
	1	1	0	0	1	0
	1	1	0	0	0	0
	0	0	0	0	0	0
	0	0	0	0	0	0
	1	0	0	0	1	0
	0	0	0	0	0	1



Description of the algorithm

4 - get the k-clique communities

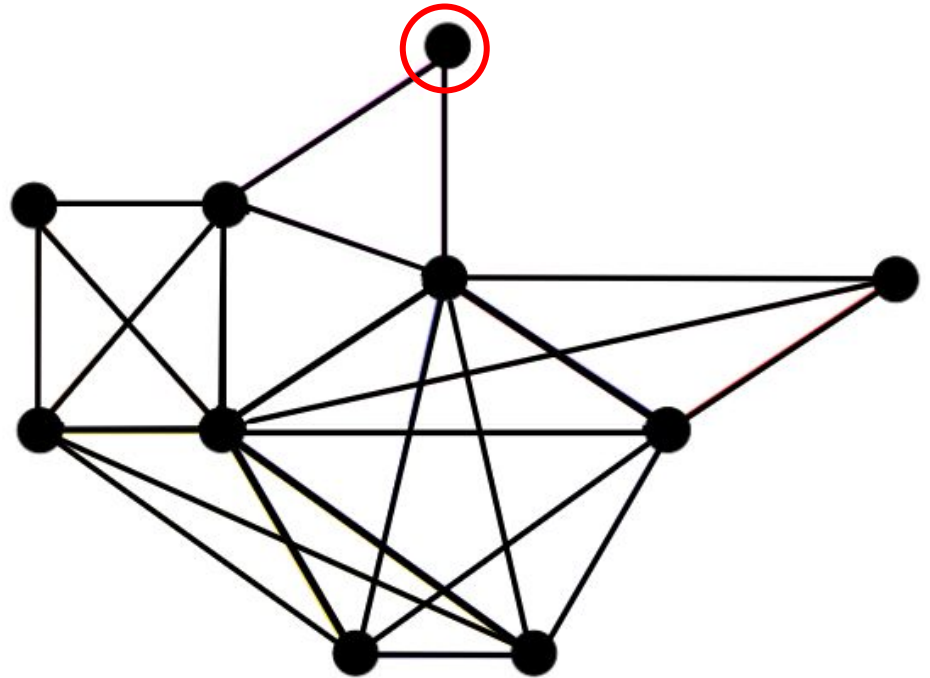
						
	1	1	0	0	1	0
	1	1	0	0	0	0
	0	0	0	0	0	0
	0	0	0	0	0	0
	1	0	0	0	1	0
	0	0	0	0	0	1



Description of the algorithm

Important to notice:

Not all vertices will necessarily
be selected to a community

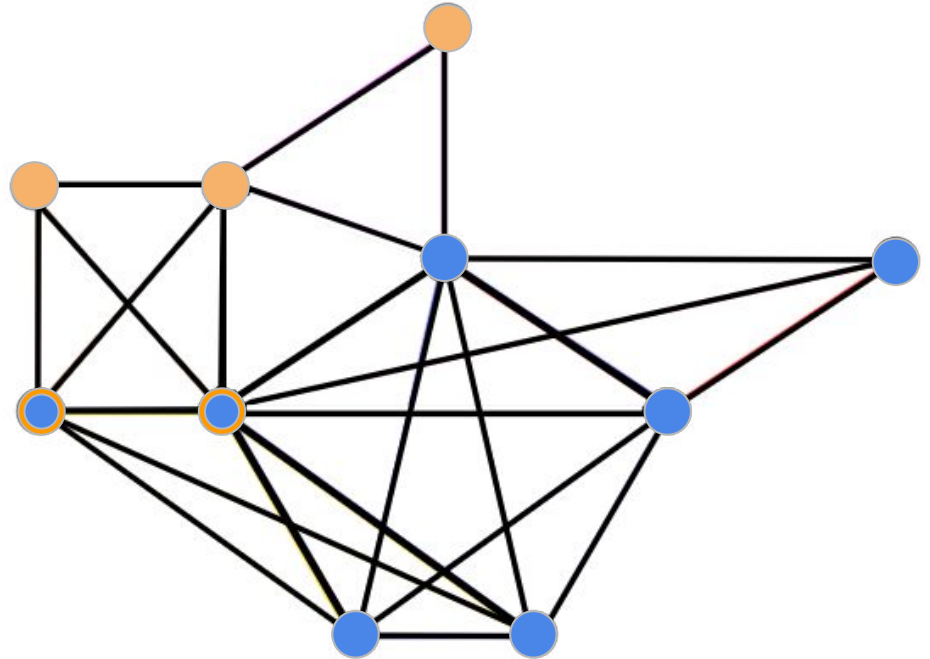


Description of the algorithm

Important to notice:

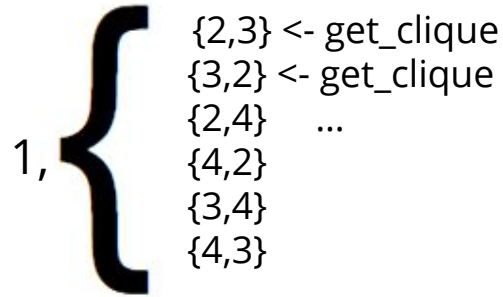
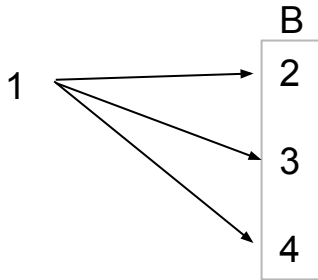
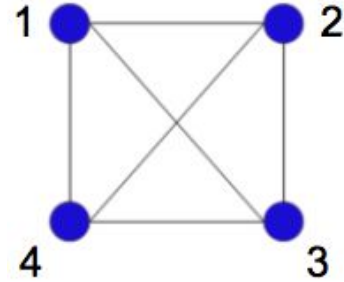
Not all vertices will necessarily
be selected to a community

Communities might overlap



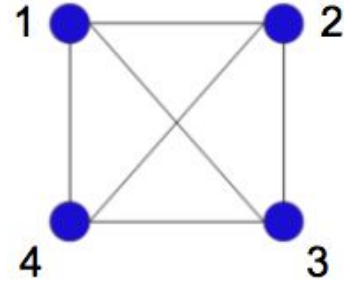
Description of the code

```
get_k_clique(k, graph):  
    clique = []  
  
    for node in graph:  
        B = get_neighbours(node, graph)  
        for subset in itertools.permutations(B, k-1):  
            clique += get_clique(node, k, list(subset), graph)
```



Description of the code

```
get_clique(node,k,B,graph):  
    A = set()  
    A.add(node)  
    while len(B) > 0 and len(A) < k:  
        n = B.pop(0)  
        A.add(n)  
        B = list(set(B).intersection(get_neighbours(n, graph)))  
        if k == len(A):  
            return " ".join(sorted(A))  
    return []
```



init: A = [1] B= [2, 3] node = 1

it 1: A = [1,2] B= [~~2~~, 3] (B = [3] \cap [1,3,4])

it 2: A = [1,2,3] B= [~~2~~, ~~3~~] (B = [] \cap [])

size_of(A) == K \rightarrow return "1 2 3"

Complexity of the algorithm

Finding the full set of cliques on a graph:

Non-polynomial problem

Getting the overlap matrix:

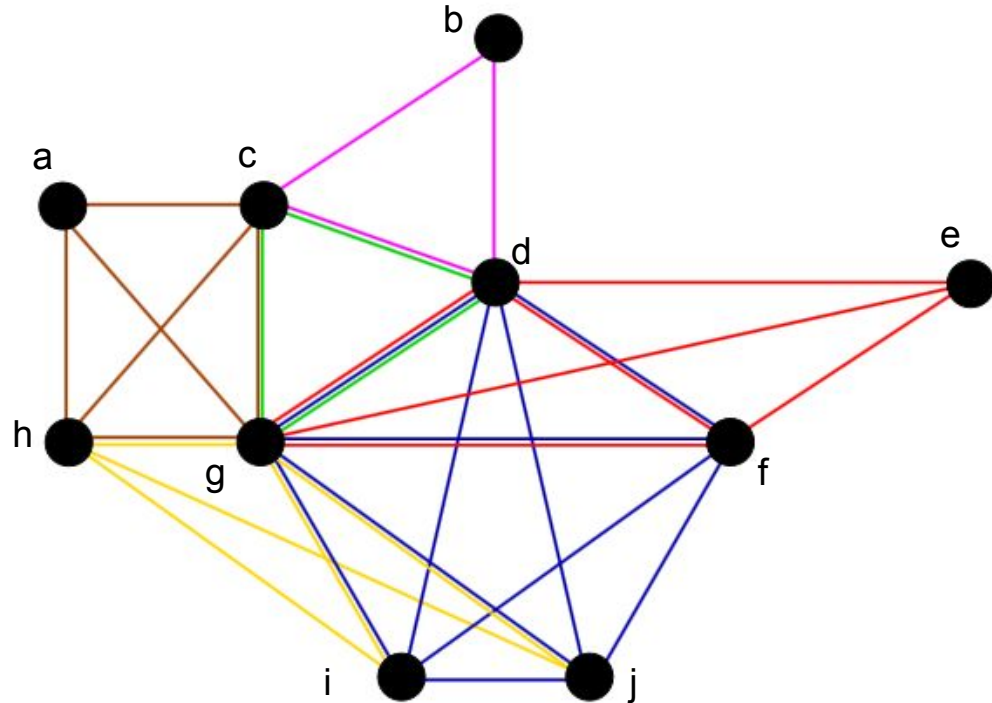
Compare each clique $(c) = O(c^2)$

Getting the k-cliques community:

Check half of the elements in the matrix $= O(c^2)$

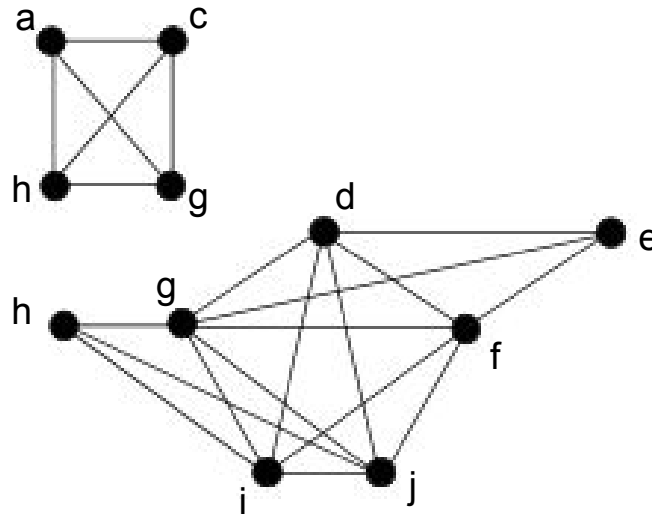
Datasets used - example 1

- First example



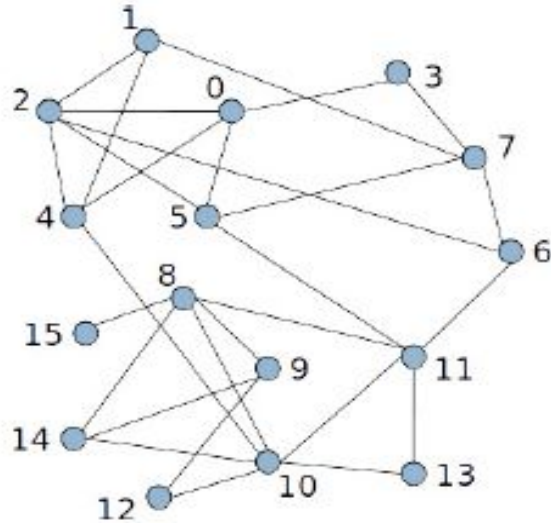
Datasets used - example 1

- Value of K : 4
- final result



Datasets used - example 2

Let's consider the following graph :



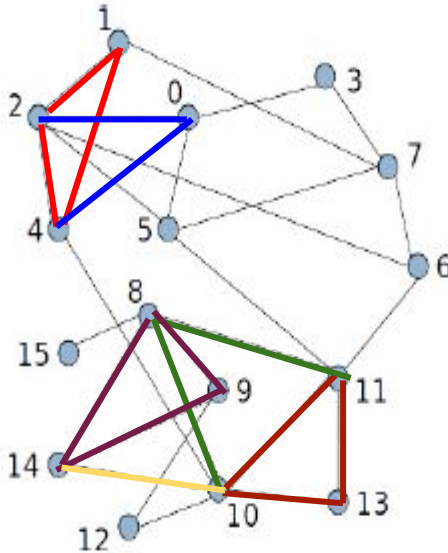
Datasets used - example 2

- $k = 4$: No clique
- $K = 3$

1 - 2 - 4

0 - 2 - 4

8 - 9 - 14



8 - 10 - 11

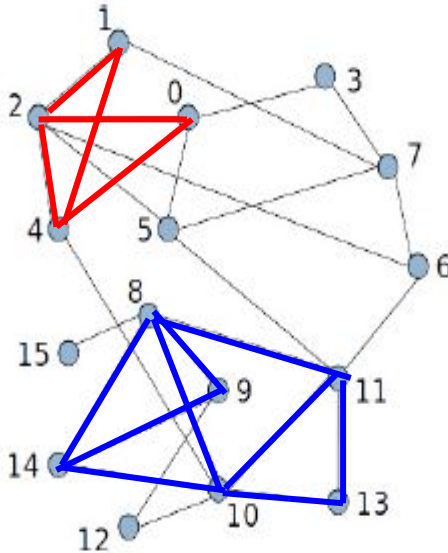
8 - 10 - 14

10 - 11 - 13

Datasets used - example 2

- final result :

2 communities



Optimization score

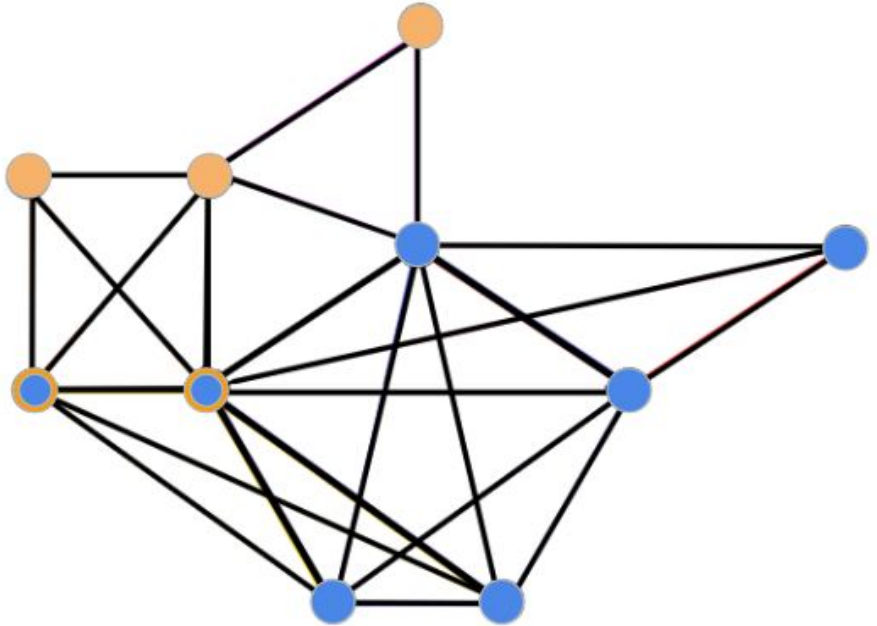
Modularity

$$Q = \frac{1}{m} \sum_{s=1}^K \left(m_s - \frac{d_s^2}{4m} \right)$$

$m = 24$ $k = 2$

$m_1 = 7$ $d_1 = 15$

$m_2 = 17$ $d_2 = 36$



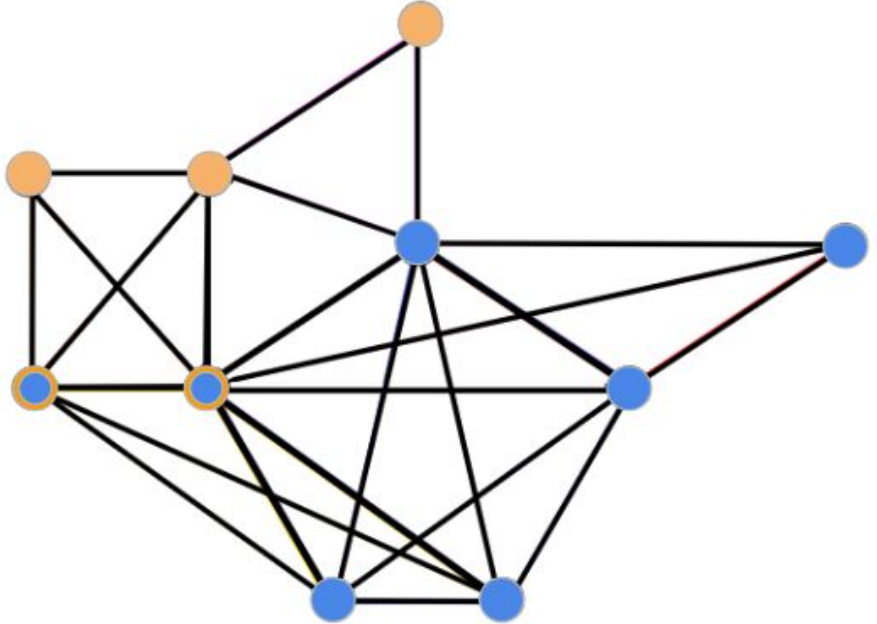
Optimization score

Modularity

$$Q = 0.34$$

The stronger the community, the higher the modularity

Influenced by the number of “single” vertices in our final result



Features of the communities

- The size of the created community is strongly (totally) related to the value of K
- There is no established method to select the best value of K
- The community size depends on which kind of community we want

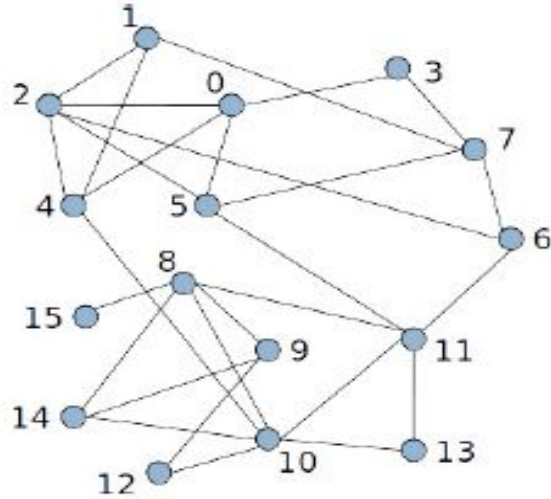
Computation times

After a computation on a very large graph, we remark that:

- the algorithm take a very long time to end
- the longest part of the algorithm is to get all the cliques

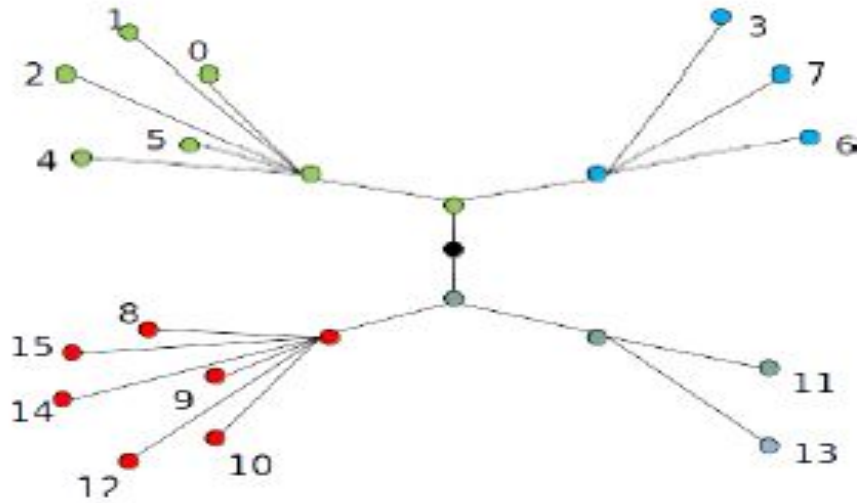
Comparison to Louvain

Let's consider the following graph :



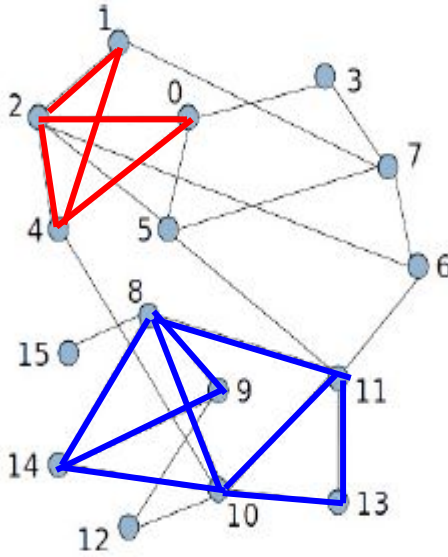
Comparison to Louvain

Community obtained with Louvain algorithm :



Comparison to Louvain

Community obtained with K-cliques algorithm :



Upgrades

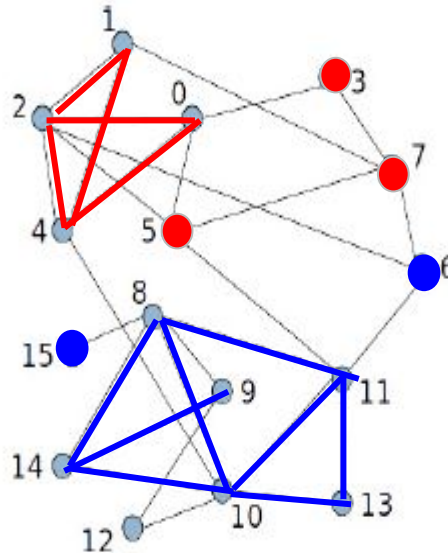
When we check for the cliques:

1, {
 {2,3} <- get_clique
 {3,2} <- get_clique
 {2,4} ...
 {4,2}
 {3,4}
 {4,3}

We could remove every similar subsets, then keep only: →
 {2,3}
 {2,4}
 {3,4}

Upgrades

Add the nodes which are not in any community based on the number of neighbors they have in the communities.



Questions?

