

# Searching for Clique-k

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#### Outline



PROBLEM DESCRIPTION



ALGORITHM DESCRIPTION



ALGORITHM ANALYSIS



EXPERIMENTAL ANALYSIS



**TESTING** 



**CONCLUSION** 

# Problem Description (1)

Clique as a word means a small group of people, with shared interests or other features in common, who spend time together and do not readily allow others to join them.

In the context of computer science, clique is a subset of vertices in an undirected graph that has all the vertices adjacent to each other which means they are all connected by edges to each other.

Clique problem is the computational problem of finding cliques. Based on the need and given parameters, there may be different formulations in the problem.

Common formulations of the clique problem include finding a maximum clique which is a clique with the largest possible number of vertices, finding a maximum weight clique in a weighted graph, or listing all maximal cliques that are cliques that cannot be enlarged.



Clique computation is used in various fields, with bioinformatics being the most common. Other than modeling in bioinformatics, it is also used in electrical engineering and chemistry fields.



In bioinformatics, the problem of clustering gene expression data requires finding the minimum number of changes needed to transform a graph describing the data into a graph formed as the disjoint union of cliques.



In electrical engineering, cliques are used to analyze communications networks, and to design efficient circuits for computing partially specified Boolean functions.



In chemistry, cliques are used to describe chemicals in a chemical database that have a high degree of similarity with a target structure.

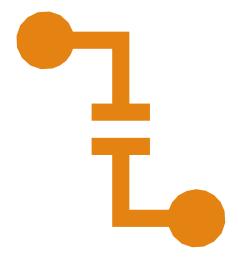
# Problem Description (2)

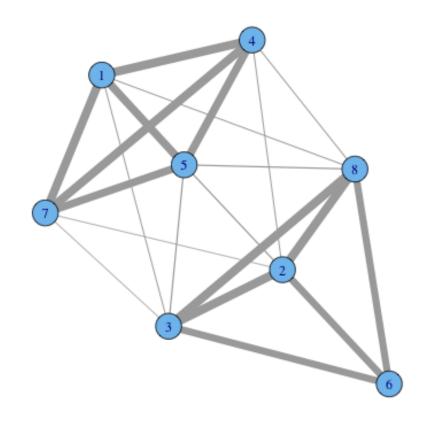
## Problem Description (3)

In this project's context, the problem specified is finding a clique with a requested size. It is described more formally below:

- **Input:** We have an n-node undirected graph G(V, E) with node set V and edge set E; a positive integer k with  $k \le n$ .
- Problem: Does G contain a k-clique, i.e. a subset W of the nodes V such that W

has size k and for each distinct pair of nodes u, v in W,  $\{u, v\}$  is an edge of G?





#### About Problem

 The problem of finding the maximum clique is both fixed-parameter intractable and hard to approximate. And, listing all maximal cliques may require exponential time as there exist graphs with exponentially many maximal cliques.

#### Algorithm Description

```
def p_clique(graph,k):
 if k == 1:
   return True
 clique = []
 vertices = get_keys_array(graph)
 max_ln=0
 max_dict=[]
 for i in range(0,len(graph)):
  clique=[]
  clique.append(vertices[i]) for v in vertices:
     if v in clique:
       continue
     isNext = True
```

```
for u in clique:
        if u in graph[v]:
         continue
        else:
         isNext = False
         break
    if isNext:
      clique.append(v)
if k <= len(clique):</pre>
             return True
 if len(clique)>max_ln:
   max_ln=len(clique)
   max_dict=clique
if k <= Ten(clique)
   return True
else:
   return False
```

```
clique = []
 vertices =
get_keys_array(graph)
 max ln=D
 max_dict=[]
  for i in range(0,len(graph)):
   clique=[]
                                                O(V)
   clique.append(vertices[i])
   for v in vertices:
       if v in clique:
                                                    O(V*V)
           continue
       isNext = True
       for u in clique:
           if u in graph[v]:
             continue
           else:
                                                         O(V*V*k)
             isNext = False
             break
       if isNext:
         clique.append(v)
         if k <= len(clique):
           return True
   if len(clique)>max ln:
     max_ln=len(clique)
     max dict-clique
  if k <= len(clique):
            return True
 else :
            return False
```

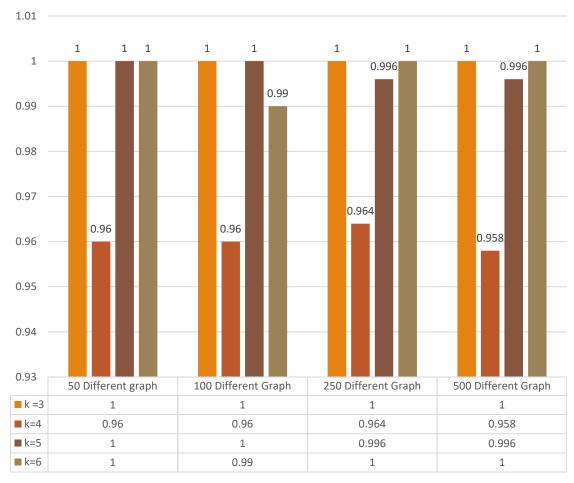
# Algorithm Analysis

Running Time: O( V2 + k)



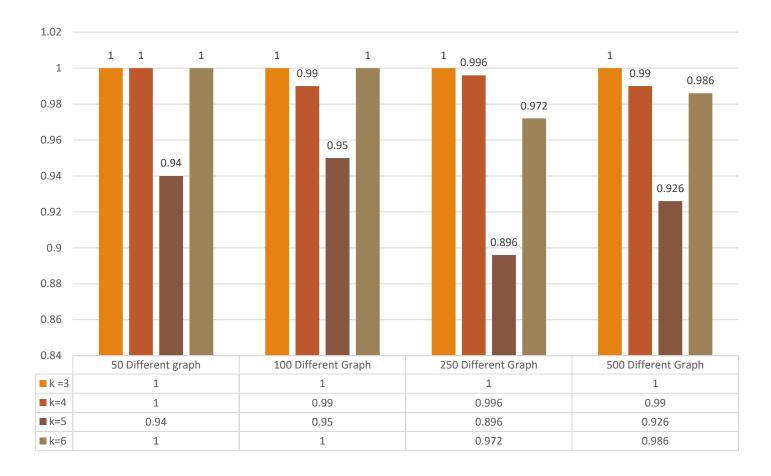
# Experimental Analysis

We tested the algorithm with randomly generated graphs of sizes 10, 15, 20, 30 with k values of 3, 4, 5, 6 for 50, 100, 250, 500 times in sequence.

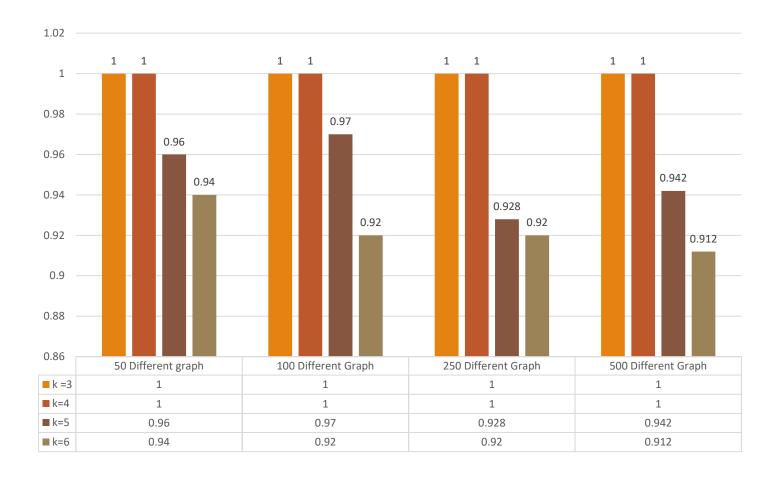


#### Success Rates

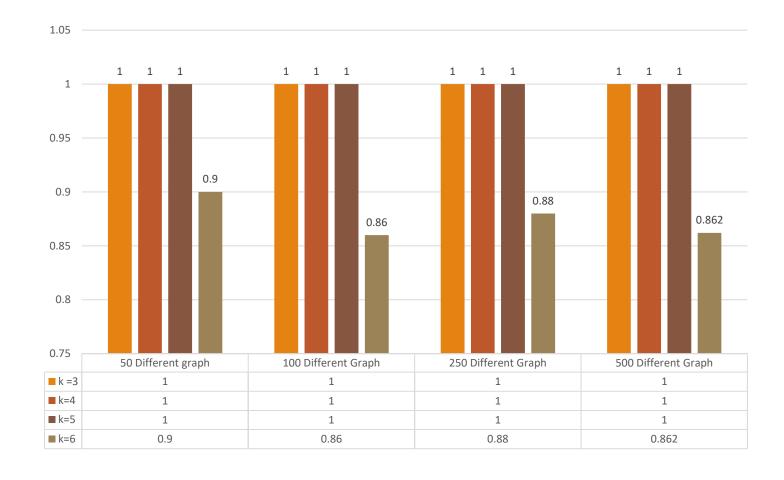
Success Rates are determined as if our algorithm can find k-clique precisely.



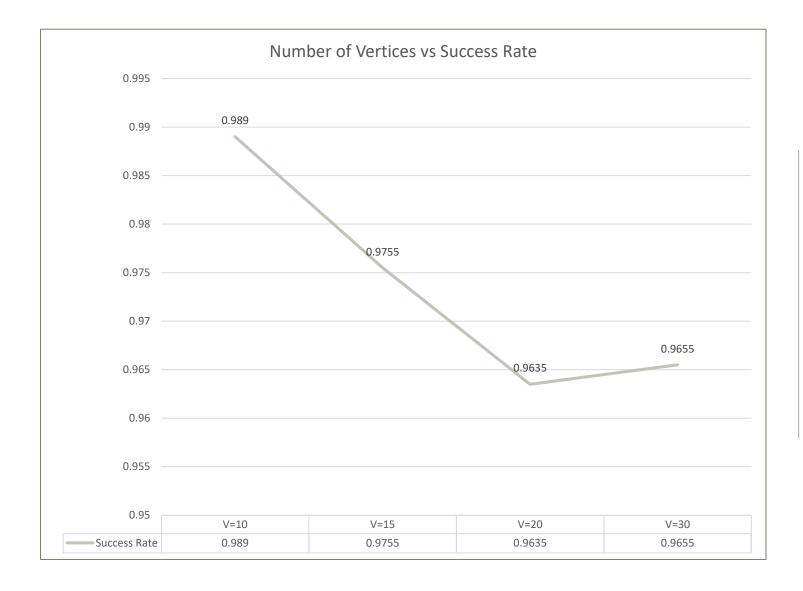
## Success Rates



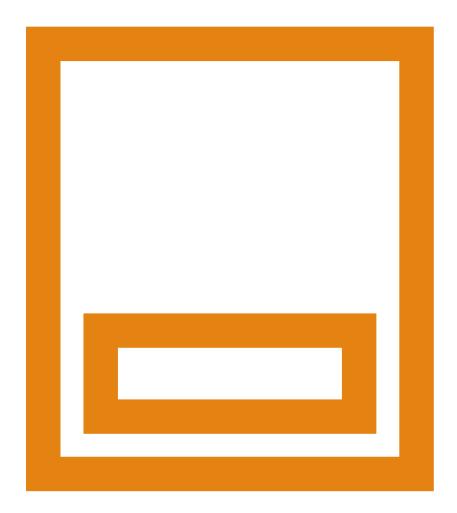
## Success Rates



## Success Rates



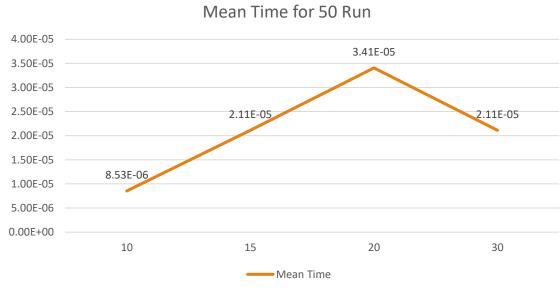
## Number of Vertices vs Success Rate



## Running Time Experimental Measurement

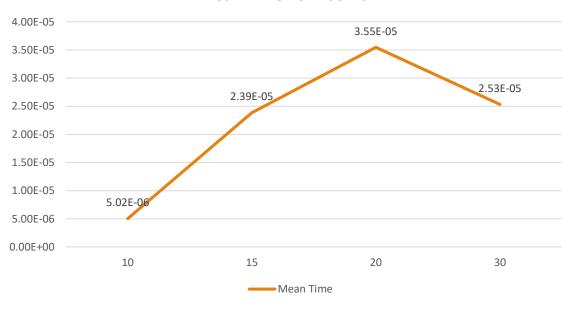
To experimentally analyze running time of our algorithm we created some functions to help us to interpret some statistic concepts using python libraries. We fixed our K value to 5 because we want to find the effect of only the number of vertices to running time.

Size	Mean Time	Standard Deviation	Standard Error	%90-CL	%95-CL
10	8.534e-06	2.75e-05	1.21e-06	2.95e-05 - 2.55e-05	2.99e-05 - 2.51e-05
15	2.110e-05	3.507e-05	3.015e-06	4.003e-05 - 3.011e-05	4.098e-05 - 2.9166e- 05
20	3.407e- 05,	4.050e-05	4.867e-06	4.851e-05 - 3.249e-05	5.004e-05 - 3.096e-05
30	2.111e-05	2.256e-05	3.016e-06	2.752e-05 - 1.760e-05	2.847e-05 - 1.665e-05



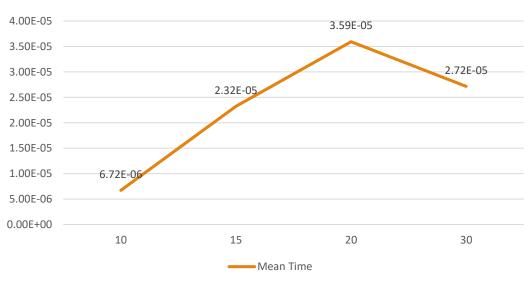
Size	Mean Time	Standard Deviation	Standard Error	%90-CL	%95-CL
10	5.016e-06	2.57e-05	5.04e-07	2.65e-05 - 2.49e-05	2.673e-05 - 2.476e-05
15	2.385e-05	3.368e-05	2.397e-06	3.7625e-05 - 2.9737e-05	3.838e-05 - 2.898e-05
20	3.547e-05	4.064e-05	3.564e-06	4.650e-05 - 3.478e-05	4.763e-05 - 3.365e-05
30	2.533e-05	2.420e-05	2.546e-06	2.839e-05 - 2.001e-05	2.919e-05 - 1.921e-05

#### Mean Time For 100 Run



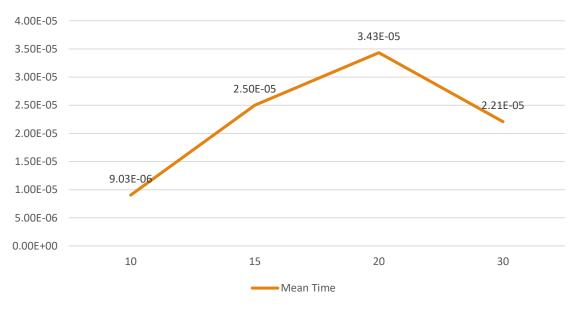
Size	Mean Time	Standard Deviation	Standard Error	%90-CL	%95-CL
10	6.718e-06	2.765e-05	4.257e-07	2.835e-05 - 2.695e-05	2.848e-05 - 2.682e-05
15	2.323e-05	3.928e-05	1.472e-06	4.170e-05 - 3.686e-05	4.216e-05 - 3.639e-05
20	3.593e-05	3.887e-05	2.277e-06	4.262e-05 - 3.513e-05	4.333e-05 - 3.441e-05
30	2.715e-05	2.780e-05	1.720e-06	3.063e-05 - 2.497e-05	3.117e-05 - 2.443e-05

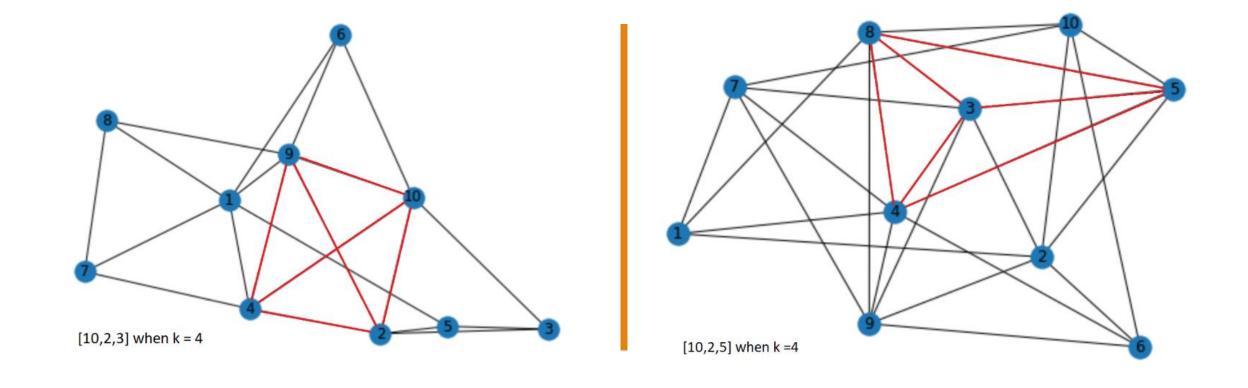
#### Mean Time For 250 Run



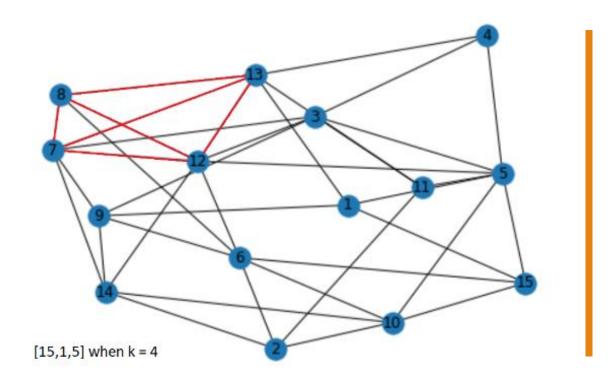
Size	Mean Time	Standard Deviation	Standard Error	%90-CL	%95-CL
10	9.031e-06	2.753e-05	4.042e-07	2.820e-05 - 2.687e-05	2.833e-05 - 2.674e-05
15	2.502e-05	4.352e-05	1.120e-06	4.537e-05 - 4.168e-05	4.572e-05 - 4.133e-05
20	3.434e-05	3.734e-05	1.537e-06	3.987e-05 - 3.481e-05	4.035e-05 - 3.432e-05
30	2.208e-05	2.581e-05	9.885e-07	2.744e-05 - 2.419e-05	2.775e-05 - 2.387e-05

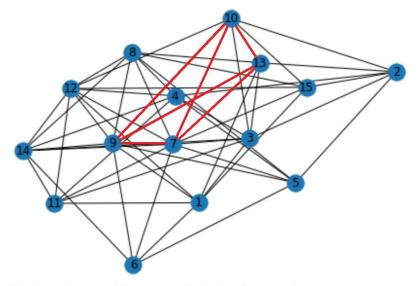
#### Mean Time For 500 Run





## Testing (Failed Conditions)

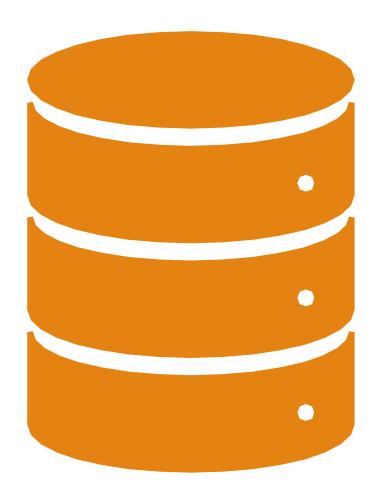




When k = 4 both algorithm returns true(p algo = [7,3,10,13]) When k = 5 p algorithm returns false but np algo finds =[7,9,8,12,14]

## Failed Case

#### Successful Case



## Conclusion (1)

To sum up, we have seen that finding k-clique problem is NP-Complete problem that is reduced from 3-SAT. Also, there is not such an exact efficient algorithm that solves that problem. There are just heuristic algorithms such as greedy max clique algorithms. We changed this algorithm to find k-clique. We examined and analyzed this algorithm in this report and have seen that this algorithm does not always work.

#### Conclusion (2)

After experimental analysis of this algorithm, we deduced that it has approximately 99% chance to find answer for decision problem.

If we consider the measurement of running time of the algorithm in the experimental analysis part shows a consistent trend with our result. However, when the graph size is 30, the running time decreased because when the graph size increases, number of connections are also increased so finding 5-clique is become easier and fast. Then it is normal to algorithm become faster.