

```
In [1]: using Graphs
using GraphPlot
using DataFrames
using CSV
using LinearAlgebra
```

# Lucas Schmidt Ferreira de Araujo

## Report 01

### Exercise I

- Alice - 1, Bob - 2, Gail - 3, Irene - 4, Carl - 5
- Harry - 6, Jen - 7, David - 8, Ernest - 9, Frank - 10

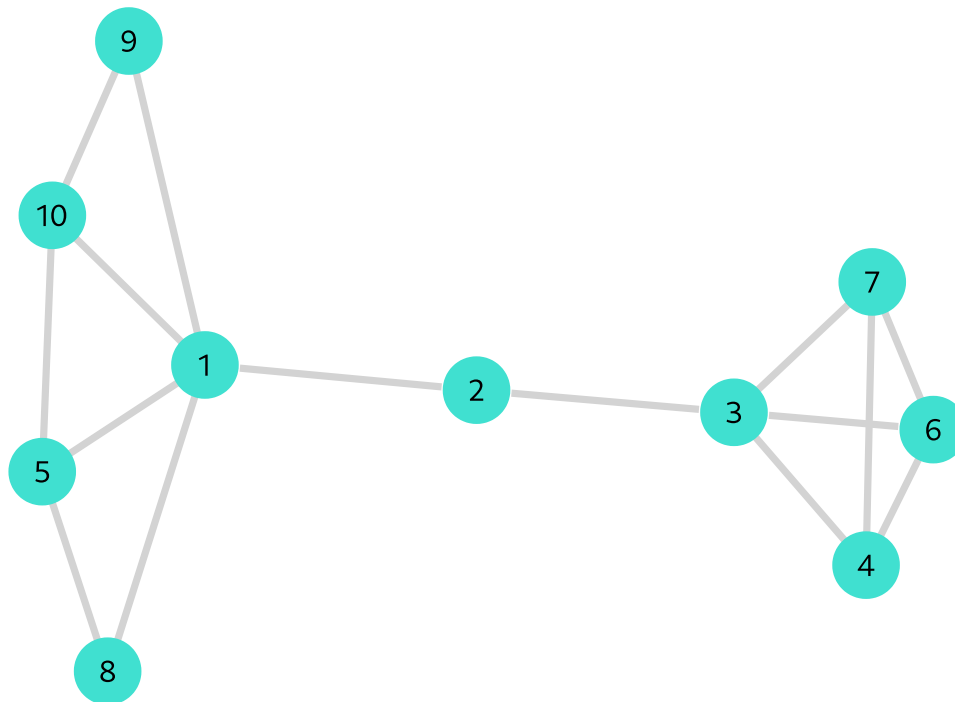
This is a graph with:

- Number of nodes:  $n = 10$
- Number of edges:  $m = 15$

```
In [2]: n = 10
m = 15

G = Graph(n) # graph with 10 vertices
add_edge!(G, 1, 2)
add_edge!(G, 2, 3)
add_edge!(G, 4, 3)
add_edge!(G, 5, 1)
add_edge!(G, 3, 6)
add_edge!(G, 4, 7)
add_edge!(G, 1, 8)
add_edge!(G, 6, 7)
add_edge!(G, 9, 10)
add_edge!(G, 1, 9)
add_edge!(G, 7, 3)
add_edge!(G, 8, 5)
add_edge!(G, 1, 10)
add_edge!(G, 6, 4)
add_edge!(G, 5, 10)

gplot(G, nodelabel=1:10)
```



c) The Density of the network is given by the formula

$$\rho = \frac{m}{\binom{n}{2}} = \frac{2m}{n(n-1)}$$

```
In [3]: ρ = 2*m/(n*(n-1))
        println("The density ρ = $(ρ)")
```

The density  $\rho = 0.3333333333333333$

d) The degree  $k_i$  of the node  $i$  is given by the number of edges connected to it

```
In [4]: degrees = degree(G)
        println("Node i | Degree ki")
        for v in 1:n
            println("i = $v, ki = $(degrees[v])")
        end
```

Node i | Degree ki

i = 1, ki = 5

i = 2, ki = 2

i = 3, ki = 4

i = 4, ki = 3

i = 5, ki = 3

i = 6, ki = 3

i = 7, ki = 3

i = 8, ki = 2

i = 9, ki = 2

i = 10, ki = 3

e) The clustering coefficient  $C_i$  for a node  $i$  is given by

$$C_i = \frac{\text{Number of links between neighbors of } i}{\text{Number of all possible connection between them}}$$

```
In [5]: clust_coeff = local_clustering_coefficient(G)
println("Node i | Coefficient Ci")
for v in 1:n
    println("i = $v, Ci = $(clust_coeff[v])")
end
```

```
Node i | Coefficient Ci
i = 1, Ci = 0.3
i = 2, Ci = 0.0
i = 3, Ci = 0.5
i = 4, Ci = 1.0
i = 5, Ci = 0.6666666666666666
i = 6, Ci = 1.0
i = 7, Ci = 1.0
i = 8, Ci = 1.0
i = 9, Ci = 1.0
i = 10, Ci = 0.6666666666666666
```

f) The Closeness Centrality coefficient for a vertex  $i$  is given by the relation

$$C_B(v_i) = \frac{n-1}{\sum_k d(v_k, v_i)}$$

```
In [6]: cb = closeness centrality(G)
println("Node i | Coefficient Cbi")
for v in 1:n
    println("i = $v, Ci = $(cb[v])")
end
```

```
Node i | Coefficient Cbi
i = 1, Ci = 0.5625
i = 2, Ci = 0.5625
i = 3, Ci = 0.5
i = 4, Ci = 0.375
i = 5, Ci = 0.4090909090909091
i = 6, Ci = 0.375
i = 7, Ci = 0.375
i = 8, Ci = 0.391304347826087
i = 9, Ci = 0.391304347826087
i = 10, Ci = 0.4090909090909091
```

f) The Degree of Centrality is devinde by

$$C_D = \frac{d_v(s, t)}{d(s, t)}$$

Where

- $d_v(s, t)$ — Is the number of shortest path that passe through the vertex  $v$
- $d(s, t)$ — Is the total amount of shortest paths

```
In [7]: cb = degree centrality(G)
println("Node i | Coefficient Cbi")
for v in 1:n
    println("i = $v, Ci = $(cb[v])")
end
```

Node i | Coefficient Cbi  
i = 1, Ci = 0.5555555555555556  
i = 2, Ci = 0.2222222222222222  
i = 3, Ci = 0.4444444444444444  
i = 4, Ci = 0.3333333333333333  
i = 5, Ci = 0.3333333333333333  
i = 6, Ci = 0.3333333333333333  
i = 7, Ci = 0.3333333333333333  
i = 8, Ci = 0.2222222222222222  
i = 9, Ci = 0.2222222222222222  
i = 10, Ci = 0.3333333333333333

## Exercise II

a) Prepare a CSV with the edge list

```
In [8]: df = DataFrame( From = [1,2,4,5,3,4,1,6,9,1,7,8,1,6,5], To = [2,3,3,1,6,7
CSV.write("data.csv",df)
df
```

15x2 DataFrame

Row	From	To
	Int64	Int64
1	1	2
2	2	3
3	4	3
4	5	1
5	3	6
6	4	7
7	1	8
8	6	7
9	9	10
10	1	9
11	7	3
12	8	5
13	1	10
14	6	4
15	5	10

## Exercise III

a) The vector  $k$  of nodes degrees  $k_i, \quad i \in \{1, 2, \dots, N\}$

$$k = Ae$$

```
In [9]: A = adjacency_matrix(G)
e = ones(n)

k = A * e
```

10-element Vector{Float64}:

```
5.0
2.0
4.0
3.0
3.0
3.0
3.0
2.0
2.0
3.0
```

b) The ammount of links  $L$  in the network

$$L = \frac{1}{2}k \cdot e$$

```
In [10]: L = .5 * sum(k.*e)
```

15.0

c) The matrix  $N$  whose elements  $n_{ij}$  is equal to the number of common neighbors of nodes  $i$  and  $j$

$$N = AA^T - D$$

such that  $D$  is the diagonal matrix representing the degrees of the nodes

```
In [11]: N = A * transpose(A) - Diagonal(k)
```

10×10 SparseArrays.SparseMatrixCSC{Float64, Int64} with 48 stored entries:

```
·   ·   1.0   ·   2.0   ·   ·   1.0  1.0  2.0
·   ·   ·   1.0  1.0  1.0  1.0  1.0  1.0  1.0
1.0   ·   ·   2.0   ·   2.0  2.0   ·   ·
·   1.0  2.0   ·   ·   2.0  2.0   ·   ·
2.0  1.0   ·   ·   ·   ·   ·   1.0  2.0  1.0
·   1.0  2.0  2.0   ·   ·   2.0   ·   ·
·   1.0  2.0  2.0   ·   2.0   ·   ·   ·
1.0  1.0   ·   ·   1.0   ·   ·   ·   1.0  2.0
1.0  1.0   ·   ·   2.0   ·   ·   1.0   ·   1.0
2.0  1.0   ·   ·   1.0   ·   ·   2.0  1.0   ·
```

d) The number  $T$  of triangles present in the network

$$T = \frac{1}{6}tr A^3$$

```
In [12]: T = tr(A^3) / 6
```

7.0

e) How would you determine whether the network is connected only by looking at the adjacency matrix?

A Network is connected if all the nodes are connected to at least with 1 other node. Since the element  $a_{ij}$  of the matrix  $A^k$  represents the number of paths of size  $k$  from the node  $i$  to  $j$ , we might say that if there is any 0 entry in the  $A^n$  matrix then there is a point not connected with anybody and therefore the network is disconnected

In [13]:  $A^n$

```
10x10 SparseArrays.SparseMatrixCSC{Int64, Int64} with 100 stored entries:
25032 10328 9291 6376 18504 6376 6376 13858 13858 18504
10328 6849 9870 8577 7762 8577 8577 5975 5975 7762
9291 9870 23289 20648 5561 20648 20648 4232 4232 5561
6376 8577 20648 18670 3774 18669 18669 2959 2959 3774
18504 7762 5561 3774 14179 3774 3774 10654 10709 14090
6376 8577 20648 18669 3774 18670 18669 2959 2959 3774
6376 8577 20648 18669 3774 18669 18670 2959 2959 3774
13858 5975 4232 2959 10654 2959 2959 8116 8082 10709
13858 5975 4232 2959 10709 2959 2959 8082 8116 10654
18504 7762 5561 3774 14090 3774 3774 10709 10654 14179
```