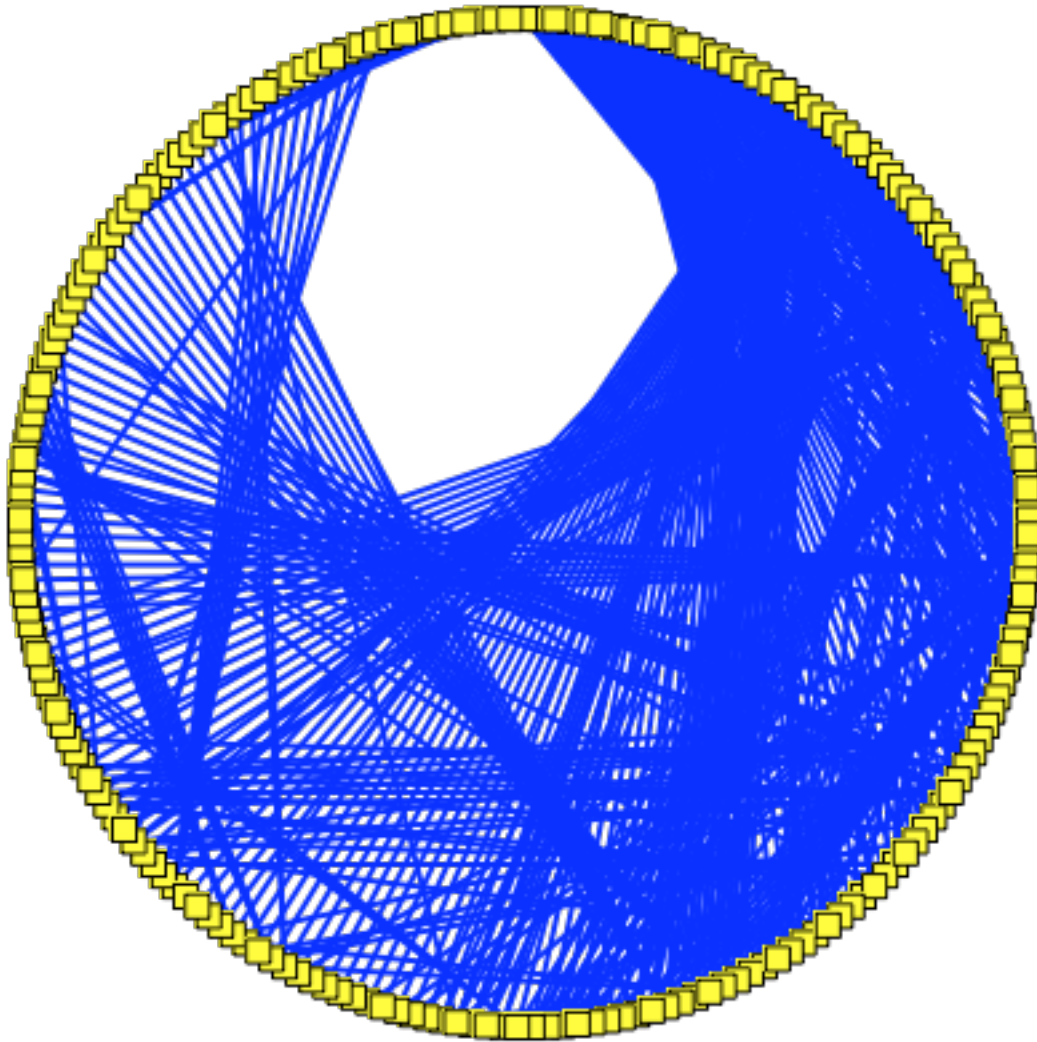
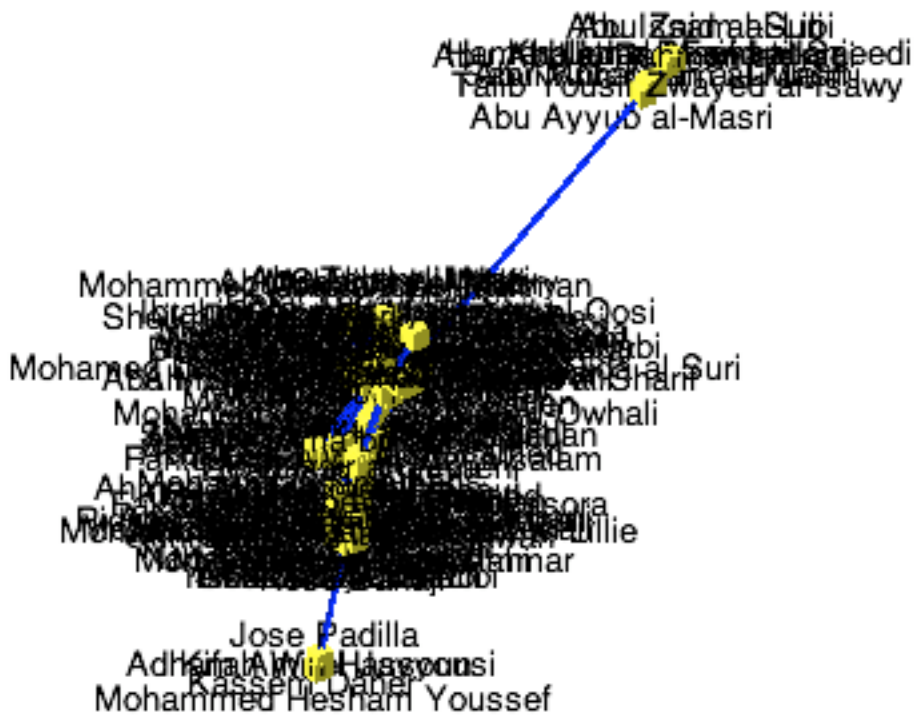


```
Data := ImportMatrix("/Users/sidafa/Documents/Adelaide/Thesis/Matrix.csv", source = csv) :  
convert(Data, 'matrix') :  
Names := ImportMatrix("/Users/sidafa/Documents/Adelaide/Thesis/Names.csv", source = csv) :  
V := convert(Names, 'list') :  
with(GraphTheory) :  
with(networks) :  
with(Statistics) :  
G := Graph(undirected, V, Data) :  
n := nops(Vertices(G)) :  
m := nops(Edges(G)) :  
DrawGraph(G)
```



```
DrawGraph(G, style = spring, dimension = 3)
```

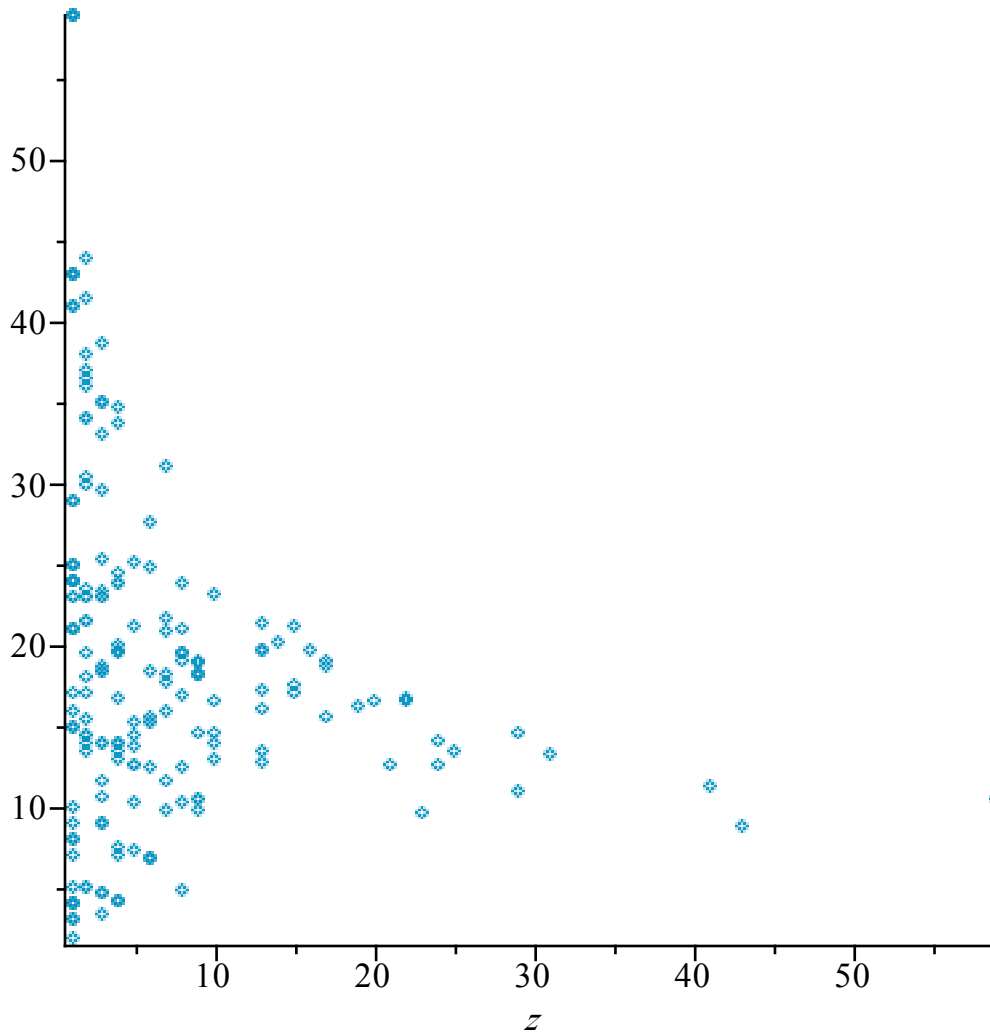


DrawGraph(MinimalSpanningTree(G), showlabels = true)


```

for  $k$  from 1 to  $n$  do
   $counter := 0$ ;
   $x := Neighbors(G, V[k])$ ;
   $y := nops(x)$ ;
   $T[k] := y$ ;
  for  $j$  from 1 to  $y$  do
     $counter := counter + Degree(G, x[j])$ ;
  end do ;
   $Q[k] := evalf\left(\frac{counter}{y}\right)$ ;
end do:
ScatterPlot( $T, Q$ )

```



```
LinearFit([1, t], T, Q, t)
```

21.2980942847207970 – 0.325001260806376369 t

(4)

```

AveragePath := proc( $G, V$ );
local  $j, k, sum, counter$ ;
 $counter := 0$ ;
 $sum := 0$ ;
for  $j$  from 1 to 209 do

```

```

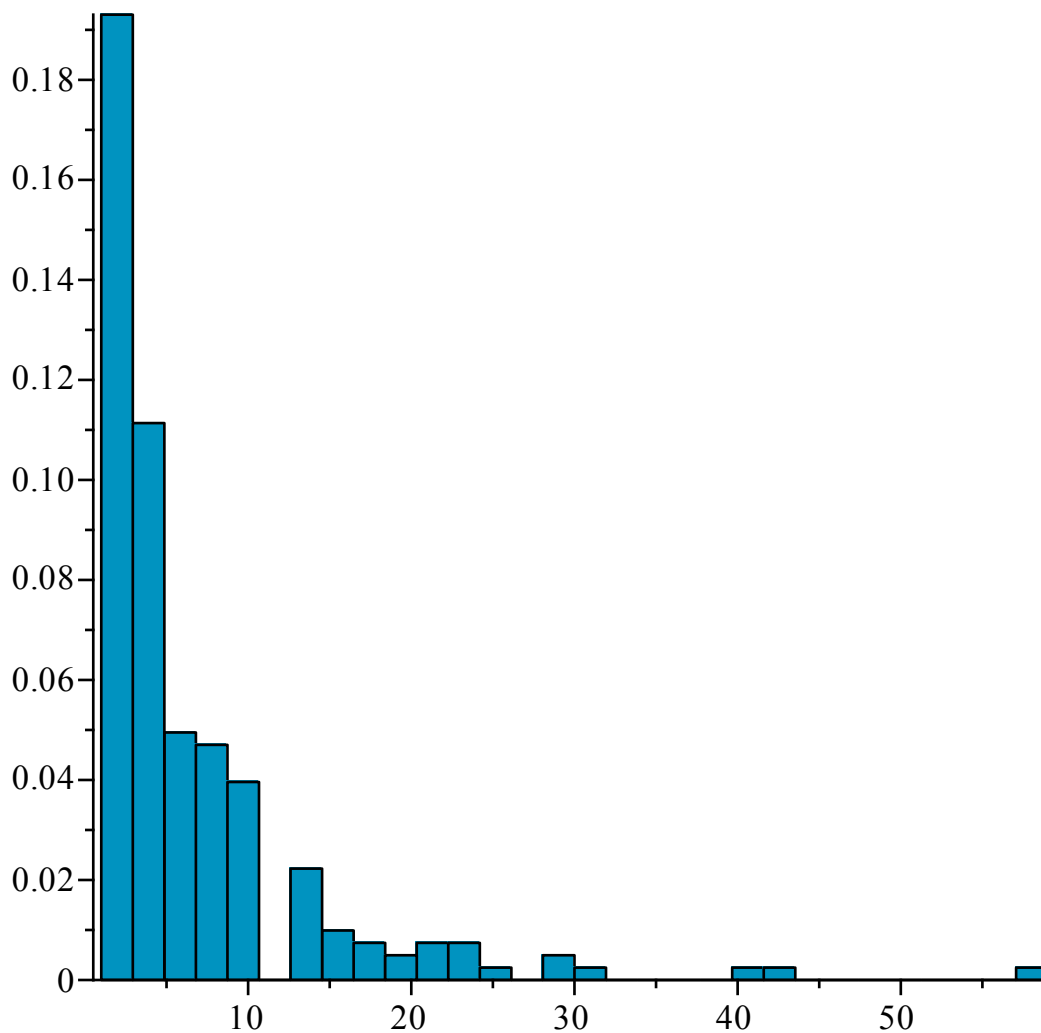
for  $k$  from 1 to 209 do
  if  $j \neq k$  then
     $sum := sum + Distance(G, V[j], V[k]);$ 
     $counter := counter + 1;$ 
  end if
end do
end do;
 $print \left( sum, counter, evalf \left( \frac{sum}{counter} \right) \right);$ 
end proc:
AveragePath( $G, V$ )

```

139790, 43472, 3.215633051

(5)

Histogram(Deg)



```

 $X := Vector([seq(1..MaxDeg)]); Y := Vector(1..MaxDeg, 0);$ 
for  $deg$  from 1 to MaxDeg do
   $freq[deg] := 0;$ 
  for  $vertex$  from 1 to 209 do
    if Degree( $G, V[vertex]$ ) =  $deg$  then  $freq[deg] := freq[deg] + 1$  end if
  end do
end do

```

end do

end do:

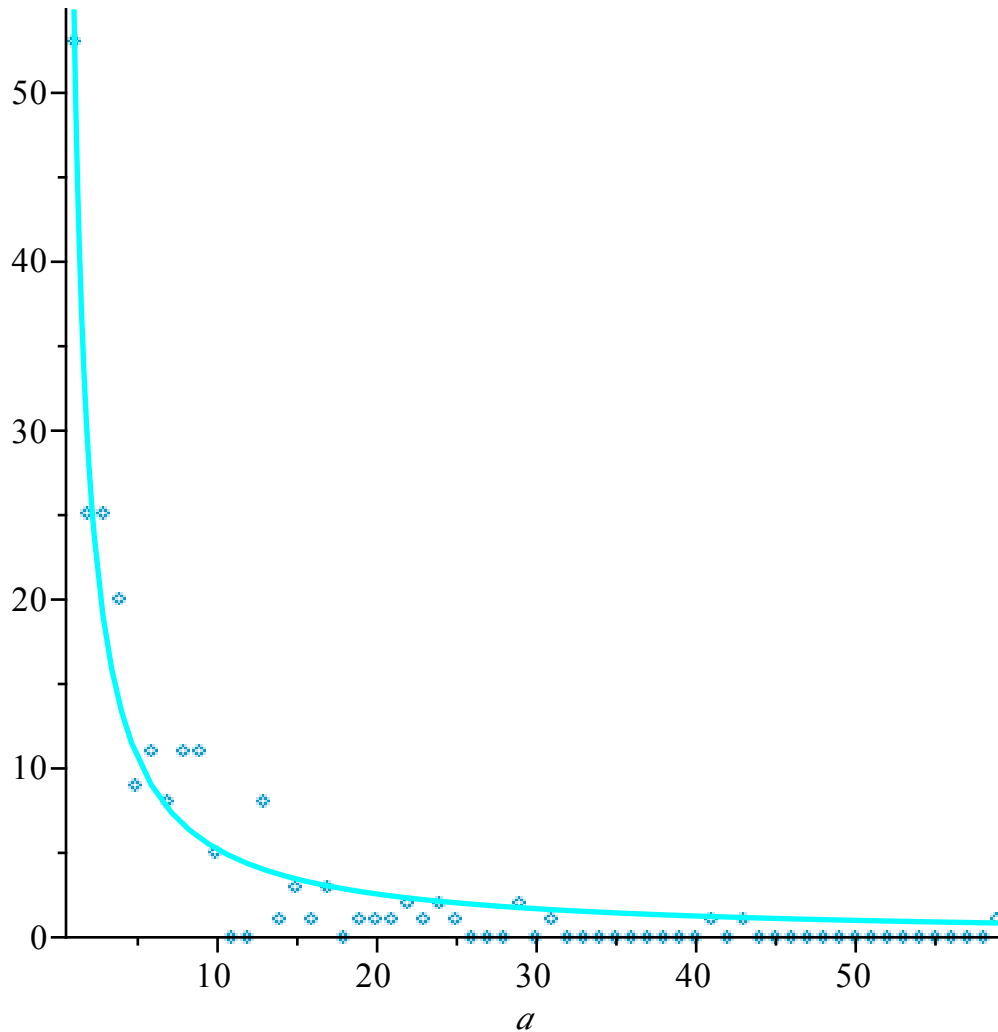
$Y := \text{Vector}([\text{seq}(\text{freq}[i], i = 1 \dots \text{MaxDeg})]) :$

$\text{NonlinearFit}(A \cdot a^b, X, Y, a)$

$$\frac{54.9468818815118212}{a^{1.02308154647380301}}$$

(6)

$\text{ScatterPlot}(X, Y, \text{fit} = [A \cdot a^b, a], \text{thickness} = 2)$



$\text{count} := 0 :$

$\ln X := \text{Vector}(\text{MaxDeg}) ;$

$\ln Y := \text{Vector}(\text{MaxDeg}) ;$

for deg from 1 to MaxDeg do

if $Y[\text{deg}] > 0$ then

$\text{count} := \text{count} + 1 ;$

$\ln X[\text{count}] := \ln(X[\text{deg}]) :$

$\ln Y[\text{count}] := \ln(Y[\text{deg}])$

end if

end do:

$\ln XX := \text{Vector}(\text{count}) ;$

```

lnYY := Vector(count);
for i from 1 to count do
lnXX[i] := lnX[i];
lnYY[i] := lnY[i];
end do:

```

```

[ 1 .. 59 Vectorcolumn
Data Type: anything
Storage: rectangular
Order: Fortran_order ]

```

```

[ 1 .. 59 Vectorcolumn
Data Type: anything
Storage: rectangular
Order: Fortran_order ]

```

```

[ 1 .. 27 Vectorcolumn
Data Type: anything
Storage: rectangular
Order: Fortran_order ]

```

```

[ 1 .. 27 Vectorcolumn
Data Type: anything
Storage: rectangular
Order: Fortran_order ]

```

(7)

```

for i from 1 to count do
print( evalf( lnYY[i] ) )
end do

```

```

3.970291914
3.218875824
3.218875824
2.995732274
2.197224578
2.397895273
2.079441542
2.397895273
2.397895273
1.609437912

```

2.079441542

0.

1.098612289

0.

1.098612289

0.

0.

0.

0.6931471806

0.

0.6931471806

0.

0.6931471806

0.

0.

(8)

Plot1 := ScatterPlot(lnXX, lnYY)

PLOT (...)

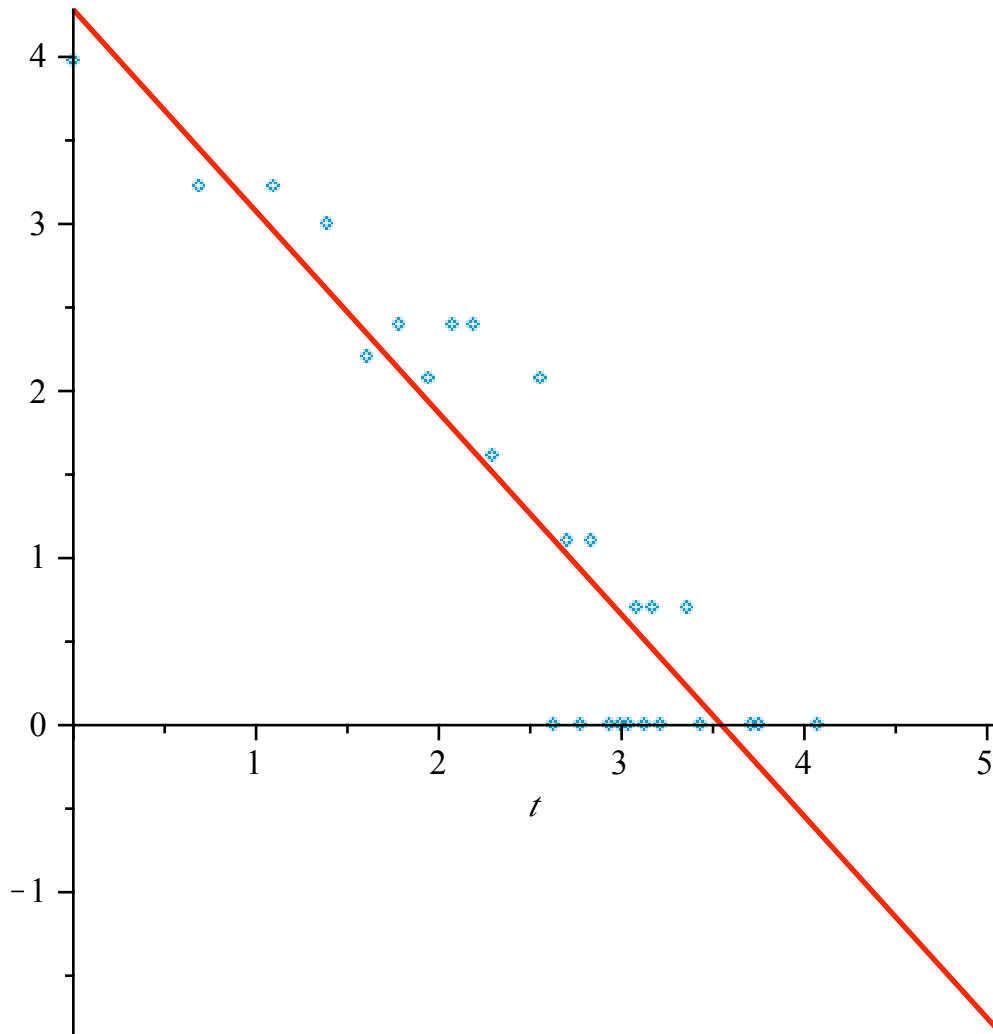
(9)

Plot2 := plot(LinearFit([1, t], lnXX, lnYY, t), t=0..(lnXX[count]+1))

PLOT (...)

(10)

with(plots) : display([Plot1, Plot2], thickness = 2)



Correlation (*lnXX*, *lnYY*)

-0.9142475390

(11)

LinearFit ([1, *t*], *lnXX*, *lnYY*, *t*)

4.28282405433401792 - 1.20722831711125123 *t*

(12)

```

clusteringcoeff := proc (G, V);
  local k, j, Cv, nv, numn, maxedges, m, y, counter;
  Cv := 0;
  counter := 0;
  for k from 1 to n by 1 do nv := Neighbors (G, V[k]);
  y := 0;
  numn := nops (nv);
  if numn ≠ 1 then
    maxedges := binomial (numn, 2);
    for j from 1 to numn - 1 do
      for m from j + 1 to numn do
        if HasEdge (G, {nv[j], nv[m]} ) then y := y + 1;
      end if ;
    end do ;
  end if ;
end do ;

```

```

Cv :=  $\frac{y}{maxedges} + Cv$ ;
counter := counter + 1;
end if
end do;
evalf  $\left( \frac{Cv}{counter} \right)$ ;
end proc:
clusteringcoeff ( G, V )

```

0.7186465269 **(13)**

```

with ( LinearAlgebra ) :
EVmatrix := Eigenvalues ( evalf ( Data ) ) :
EV := convert ( EVmatrix, 'list' ) :
EVsorted := sort ( EV ) :
EVsorted [209]

```

15.0730503143760703 + 0. I **(14)**