

# **CS 6385.011 Algorithmic Aspects of Telecommunication Networks**

## **Project 2 “Effect of Individual Link Reliabilities on Network Reliability”**

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## **Project Description :**

The theme of this project is to study experimentally how the network reliability depends on the individual link reliabilities. In order to compute the network reliability as a numerical measure, the method of exhaustive enumeration is used.

## **Network Description :**

### **Network topology:**

A complete undirected graph on  $n = 5$  nodes. This means, every node is connected with every other one (parallel edges and self-loops are excluded in this graph). As a result, this graph has  $m = 10$  edges, representing the links of the network.

### **Components that may fail:**

The links of the network may fail, the nodes are always up. The reliability of each link is  $p$ , the same for every link. The parameter  $p$  will take different values in the experiments.

### **Reliability configuration:**

The system is considered operational, if the network topology is connected.

## **Inputs and Outputs :**

Input is a complete graph with 5 vertices which is auto generated by the program. There will be 10 edges in the graph. All edges will be assigned a minimum weight(1.0) and have the same reliability measure of ' $p$ ' which ranges from [0-1].

Output is always a reliability value of the entire network calculated using exhaustive enumeration.

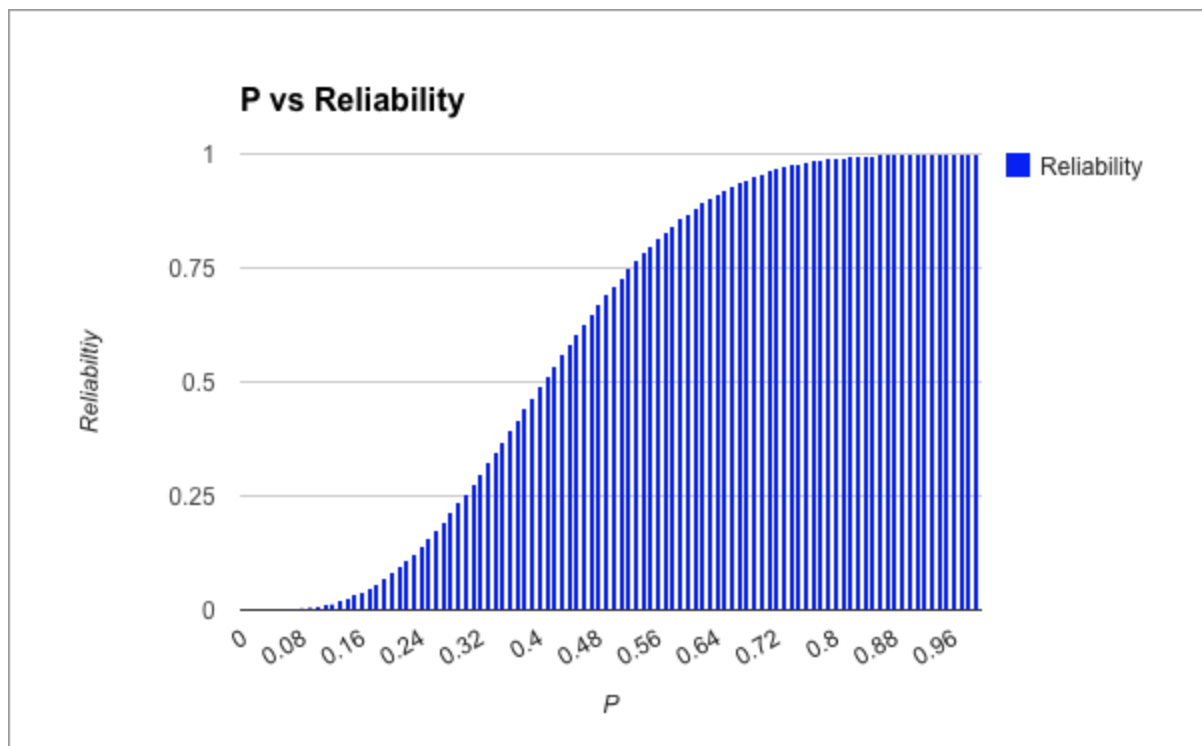
## **Scenarios :**

1. The first scenario is to run the program for different values of ' $p$ '. The parameter ' $p$ ' runs over the  $[0,1]$  interval in steps of 0.01.
2. The second scenario is to fix the parameter ' $p$ ' at 0.95 and flip the system states of ' $k$ ' random combinations; thereby calculating the reliability of the network. The value of ' $k$ ' varies from 0 to 99. To reduce the effect of randomness, 100 runs are made for each ' $k$ ' value and averaged.

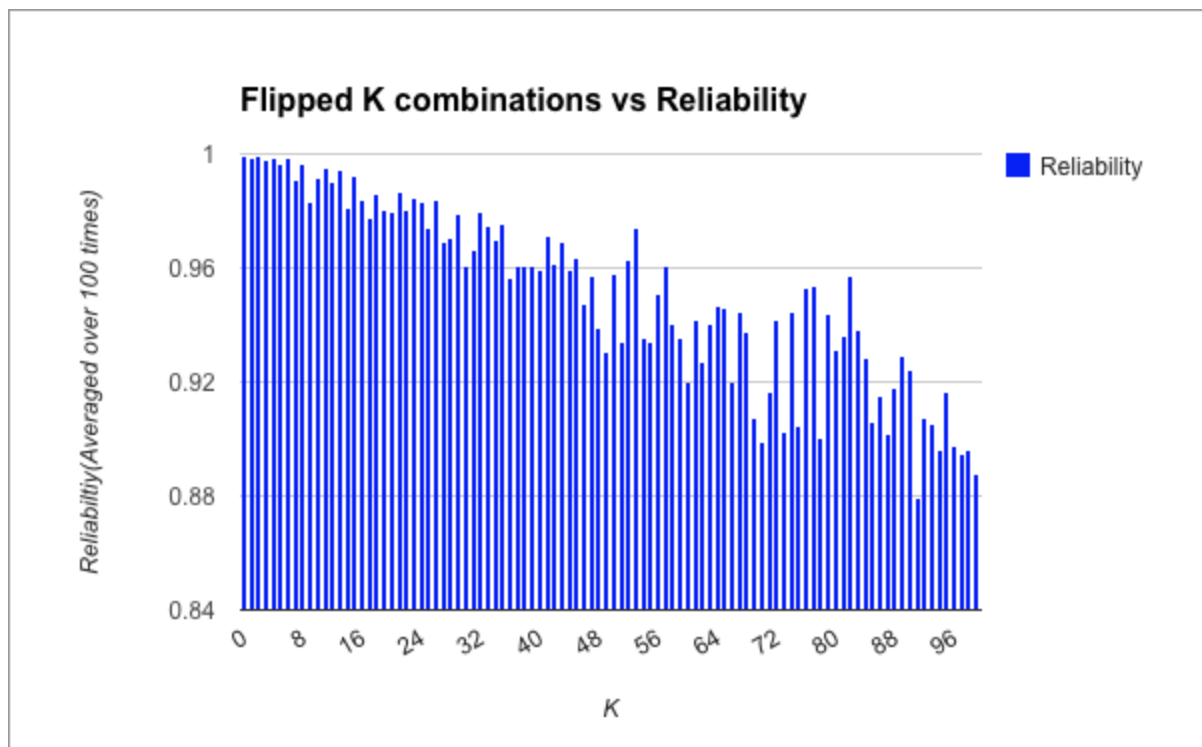
## Pseudocode :

1. Generate a complete graph with 5 vertices and assign proper edge weights and link reliabilities 'p'.
2. Calculate all the possible combinations of link failures.
3. For each of the combinations, check the connectedness of the graph using Dijkstra's algorithm[1]. If the graph is connected for a particular combination of link failures, then the system state is assigned to true. Otherwise, assign it to false meaning disconnected.
4. Using the method of exhaustive enumeration, calculate the network reliability for the combinations and the corresponding system states.
5. For scenario 1, repeat steps 1-4 for all values of 'p' in the [0,1] interval in steps of 0.01.
6. For scenario 2, do steps 1-3 once, assign  $k=0$  and goto step 7.
7. Flip 'k' randomly chosen system states and calculate the network reliability for the combinations and the corresponding system states.
8. Flip back the same 'k' system states to original state.
9. Repeat steps 7-8 for 100 times for each value of 'k' from 0-99 and average it.
10. Plot graphs for scenario 1 and scenario 2.

## Graphs :



From the graph shown above, it can be clearly observed that the network reliability increases with increase in 'p' and stabilises towards the end. Intuitively, the probability of the network state to be ON is higher when there are lesser number of link failures. According to the method of exhaustive enumeration, if there are lesser number of failed links in a particulate combination, then there are lesser number of (1-p) terms and more of 'p' terms. So higher values of 'p' directly increases the value of network reliability. Also from the graph, there is a slower increase in reliability for lesser values of 'p' from 0.00 to 0.2. After that, there is a steep increase till 0.7 and then it almost flattens.



From the graph above, it can be seen that increase in the value of 'k' decreases the reliability of the network. This is because, for a complete graph network of 5 vertices, there can be 1024 possible link failure combinations, out of which more than 70% have the system state as true. Therefore, flipping a few of the 1024 states may not affect the reliability much, but flipping 100 of the 1024 will definitely reduce the network reliability because there will be more number of TRUE states that will become FALSE now. And since majority of the states were TRUE and now many of them have become FALSE, the network reliability drops.

## Results:

### Scenario 1:

Reliability when  $p = 0.0 : 0.0$

Reliability when  $p = 0.01 : 0.0$

Reliability when  $p = 0.02 : 0.0$

Reliability when  $p = 0.03 : 0.0$

Reliability when  $p = 0.04 : 0.0$

Reliability when  $p = 0.05 : 0.001$

Reliability when  $p = 0.06 : 0.001$

Reliability when  $p = 0.07 : 0.002$

Reliability when  $p = 0.08 : 0.004$

Reliability when  $p = 0.09 : 0.006$

Reliability when  $p = 0.1 : 0.008$

Reliability when  $p = 0.11 : 0.011$

Reliability when  $p = 0.12 : 0.015$

Reliability when  $p = 0.13 : 0.02$

Reliability when  $p = 0.14 : 0.026$

Reliability when  $p = 0.15 : 0.033$

Reliability when  $p = 0.16 : 0.04$

Reliability when  $p = 0.17 : 0.049$

Reliability when  $p = 0.18 : 0.059$

Reliability when  $p = 0.19 : 0.07$

Reliability when  $p = 0.2 : 0.082$

Reliability when  $p = 0.21 : 0.095$

Reliability when  $p = 0.22 : 0.109$

Reliability when  $p = 0.23 : 0.124$

Reliability when  $p = 0.24 : 0.141$

Reliability when  $p = 0.25 : 0.158$

Reliability when  $p = 0.26 : 0.176$

Reliability when  $p = 0.27 : 0.195$

Reliability when  $p = 0.28 : 0.215$

Reliability when  $p = 0.29 : 0.235$

Reliability when  $p = 0.3 : 0.256$

Reliability when  $p = 0.31 : 0.278$

Reliability when  $p = 0.32 : 0.3$

Reliability when  $p = 0.33 : 0.323$

Reliability when  $p = 0.34 : 0.347$

Reliability when  $p = 0.35 : 0.37$

Reliability when  $p = 0.36 : 0.394$

Reliability when  $p = 0.37 : 0.418$   
Reliability when  $p = 0.38 : 0.442$   
Reliability when  $p = 0.39 : 0.466$   
Reliability when  $p = 0.4 : 0.49$   
Reliability when  $p = 0.41 : 0.513$   
Reliability when  $p = 0.42 : 0.537$   
Reliability when  $p = 0.43 : 0.56$   
Reliability when  $p = 0.44 : 0.583$   
Reliability when  $p = 0.45 : 0.606$   
Reliability when  $p = 0.46 : 0.628$   
Reliability when  $p = 0.47 : 0.65$   
Reliability when  $p = 0.48 : 0.671$   
Reliability when  $p = 0.49 : 0.691$   
Reliability when  $p = 0.5 : 0.711$   
Reliability when  $p = 0.51 : 0.73$   
Reliability when  $p = 0.52 : 0.749$   
Reliability when  $p = 0.53 : 0.766$   
Reliability when  $p = 0.54 : 0.784$   
Reliability when  $p = 0.55 : 0.8$   
Reliability when  $p = 0.56 : 0.815$   
Reliability when  $p = 0.57 : 0.83$   
Reliability when  $p = 0.58 : 0.844$   
Reliability when  $p = 0.59 : 0.858$   
Reliability when  $p = 0.6 : 0.87$   
Reliability when  $p = 0.61 : 0.882$   
Reliability when  $p = 0.62 : 0.893$   
Reliability when  $p = 0.63 : 0.904$   
Reliability when  $p = 0.64 : 0.913$   
Reliability when  $p = 0.65 : 0.922$   
Reliability when  $p = 0.66 : 0.93$   
Reliability when  $p = 0.67 : 0.938$   
Reliability when  $p = 0.68 : 0.945$   
Reliability when  $p = 0.69 : 0.952$   
Reliability when  $p = 0.7 : 0.958$   
Reliability when  $p = 0.71 : 0.963$   
Reliability when  $p = 0.72 : 0.968$   
Reliability when  $p = 0.73 : 0.972$   
Reliability when  $p = 0.74 : 0.976$   
Reliability when  $p = 0.75 : 0.979$

Reliability when  $p = 0.76 : 0.983$   
Reliability when  $p = 0.77 : 0.985$   
Reliability when  $p = 0.78 : 0.988$   
Reliability when  $p = 0.79 : 0.99$   
Reliability when  $p = 0.8 : 0.992$   
Reliability when  $p = 0.81 : 0.993$   
Reliability when  $p = 0.82 : 0.995$   
Reliability when  $p = 0.83 : 0.996$   
Reliability when  $p = 0.84 : 0.997$   
Reliability when  $p = 0.85 : 0.997$   
Reliability when  $p = 0.86 : 0.998$   
Reliability when  $p = 0.87 : 0.999$   
Reliability when  $p = 0.88 : 0.999$   
Reliability when  $p = 0.89 : 0.999$   
Reliability when  $p = 0.9 : 0.999$   
Reliability when  $p = 0.91 : 1.0$   
Reliability when  $p = 0.92 : 1.0$   
Reliability when  $p = 0.93 : 1.0$   
Reliability when  $p = 0.94 : 1.0$   
Reliability when  $p = 0.95 : 1.0$   
Reliability when  $p = 0.96 : 1.0$   
Reliability when  $p = 0.97 : 1.0$   
Reliability when  $p = 0.98 : 1.0$   
Reliability when  $p = 0.99 : 1.0$   
Reliability when  $p = 1 : 1.0$

## **Scenario 2 :**

Reliability when  $p=0.95$  and  $k=0$  is 0.9999  
Reliability when  $p=0.95$  and  $k=1$  is 0.9996  
Reliability when  $p=0.95$  and  $k=2$  is 0.9991  
Reliability when  $p=0.95$  and  $k=3$  is 0.9975  
Reliability when  $p=0.95$  and  $k=4$  is 0.9977  
Reliability when  $p=0.95$  and  $k=5$  is 0.9979  
Reliability when  $p=0.95$  and  $k=6$  is 0.9984  
Reliability when  $p=0.95$  and  $k=7$  is 0.985  
Reliability when  $p=0.95$  and  $k=8$  is 0.9906  
Reliability when  $p=0.95$  and  $k=9$  is 0.9976  
Reliability when  $p=0.95$  and  $k=10$  is 0.995  
Reliability when  $p=0.95$  and  $k=11$  is 0.9782

Reliability when  $p=0.95$  and  $k=12$  is 0.9955  
Reliability when  $p=0.95$  and  $k=13$  is 0.9871  
Reliability when  $p=0.95$  and  $k=14$  is 0.9935  
Reliability when  $p=0.95$  and  $k=15$  is 0.993  
Reliability when  $p=0.95$  and  $k=16$  is 0.9741  
Reliability when  $p=0.95$  and  $k=17$  is 0.986  
Reliability when  $p=0.95$  and  $k=18$  is 0.9854  
Reliability when  $p=0.95$  and  $k=19$  is 0.9932  
Reliability when  $p=0.95$  and  $k=20$  is 0.9775  
Reliability when  $p=0.95$  and  $k=21$  is 0.969  
Reliability when  $p=0.95$  and  $k=22$  is 0.9802  
Reliability when  $p=0.95$  and  $k=23$  is 0.9842  
Reliability when  $p=0.95$  and  $k=24$  is 0.9767  
Reliability when  $p=0.95$  and  $k=25$  is 0.9793  
Reliability when  $p=0.95$  and  $k=26$  is 0.9709  
Reliability when  $p=0.95$  and  $k=27$  is 0.9782  
Reliability when  $p=0.95$  and  $k=28$  is 0.9819  
Reliability when  $p=0.95$  and  $k=29$  is 0.9773  
Reliability when  $p=0.95$  and  $k=30$  is 0.9776  
Reliability when  $p=0.95$  and  $k=31$  is 0.977  
Reliability when  $p=0.95$  and  $k=32$  is 0.9764  
Reliability when  $p=0.95$  and  $k=33$  is 0.975  
Reliability when  $p=0.95$  and  $k=34$  is 0.9625  
Reliability when  $p=0.95$  and  $k=35$  is 0.9725  
Reliability when  $p=0.95$  and  $k=36$  is 0.9675  
Reliability when  $p=0.95$  and  $k=37$  is 0.9786  
Reliability when  $p=0.95$  and  $k=38$  is 0.9522  
Reliability when  $p=0.95$  and  $k=39$  is 0.9617  
Reliability when  $p=0.95$  and  $k=40$  is 0.937  
Reliability when  $p=0.95$  and  $k=41$  is 0.9575  
Reliability when  $p=0.95$  and  $k=42$  is 0.9499  
Reliability when  $p=0.95$  and  $k=43$  is 0.961  
Reliability when  $p=0.95$  and  $k=44$  is 0.9444  
Reliability when  $p=0.95$  and  $k=45$  is 0.9567  
Reliability when  $p=0.95$  and  $k=46$  is 0.9651  
Reliability when  $p=0.95$  and  $k=47$  is 0.9683  
Reliability when  $p=0.95$  and  $k=48$  is 0.9569  
Reliability when  $p=0.95$  and  $k=49$  is 0.9674  
Reliability when  $p=0.95$  and  $k=50$  is 0.9181



Reliability when  $p=0.95$  and  $k=51$  is 0.9602  
Reliability when  $p=0.95$  and  $k=52$  is 0.9478  
Reliability when  $p=0.95$  and  $k=53$  is 0.9491  
Reliability when  $p=0.95$  and  $k=54$  is 0.9596  
Reliability when  $p=0.95$  and  $k=55$  is 0.9517  
Reliability when  $p=0.95$  and  $k=56$  is 0.9473  
Reliability when  $p=0.95$  and  $k=57$  is 0.9192  
Reliability when  $p=0.95$  and  $k=58$  is 0.9666  
Reliability when  $p=0.95$  and  $k=59$  is 0.9424  
Reliability when  $p=0.95$  and  $k=60$  is 0.9392  
Reliability when  $p=0.95$  and  $k=61$  is 0.9229  
Reliability when  $p=0.95$  and  $k=62$  is 0.9311  
Reliability when  $p=0.95$  and  $k=63$  is 0.9553  
Reliability when  $p=0.95$  and  $k=64$  is 0.9317  
Reliability when  $p=0.95$  and  $k=65$  is 0.9443  
Reliability when  $p=0.95$  and  $k=66$  is 0.9251  
Reliability when  $p=0.95$  and  $k=67$  is 0.9367  
Reliability when  $p=0.95$  and  $k=68$  is 0.9329  
Reliability when  $p=0.95$  and  $k=69$  is 0.9074  
Reliability when  $p=0.95$  and  $k=70$  is 0.9091  
Reliability when  $p=0.95$  and  $k=71$  is 0.9196  
Reliability when  $p=0.95$  and  $k=72$  is 0.9529  
Reliability when  $p=0.95$  and  $k=73$  is 0.9598  
Reliability when  $p=0.95$  and  $k=74$  is 0.9368  
Reliability when  $p=0.95$  and  $k=75$  is 0.9163  
Reliability when  $p=0.95$  and  $k=76$  is 0.9524  
Reliability when  $p=0.95$  and  $k=77$  is 0.933  
Reliability when  $p=0.95$  and  $k=78$  is 0.9068  
Reliability when  $p=0.95$  and  $k=79$  is 0.9335  
Reliability when  $p=0.95$  and  $k=80$  is 0.931  
Reliability when  $p=0.95$  and  $k=81$  is 0.9191  
Reliability when  $p=0.95$  and  $k=82$  is 0.9087  
Reliability when  $p=0.95$  and  $k=83$  is 0.922  
Reliability when  $p=0.95$  and  $k=84$  is 0.9456  
Reliability when  $p=0.95$  and  $k=85$  is 0.9337  
Reliability when  $p=0.95$  and  $k=86$  is 0.9054  
Reliability when  $p=0.95$  and  $k=87$  is 0.9137  
Reliability when  $p=0.95$  and  $k=88$  is 0.9127  
Reliability when  $p=0.95$  and  $k=89$  is 0.9259

Reliability when  $p=0.95$  and  $k=90$  is 0.8986  
Reliability when  $p=0.95$  and  $k=91$  is 0.8925  
Reliability when  $p=0.95$  and  $k=92$  is 0.9428  
Reliability when  $p=0.95$  and  $k=93$  is 0.8815  
Reliability when  $p=0.95$  and  $k=94$  is 0.8962  
Reliability when  $p=0.95$  and  $k=95$  is 0.9113  
Reliability when  $p=0.95$  and  $k=96$  is 0.875  
Reliability when  $p=0.95$  and  $k=97$  is 0.9223  
Reliability when  $p=0.95$  and  $k=98$  is 0.8763  
Reliability when  $p=0.95$  and  $k=99$  is 0.9131

## **Conclusion :**

Thus from the experiments, two results can be concluded.

1. Higher the reliability of individual links, higher the network reliability.
2. Flipping more number of system states will decrease the network reliability.

## Appendix:

### Source Code:

#### Graph.java

//Contains all the classes for creating a graph(vertices, edges) and Dijkstra's algorithm

```
import java.util.ArrayList;
import java.util.Collections;
import java.util.Comparator;
import java.util.List;
import java.util.PriorityQueue;
```

```
/**
 *
 * @author rahulrdhanendran
 */
```

//This class creates a complete graph for the given number of vertices

```
public class Graph {
    public ArrayList<Vertex> vertices;
    public ArrayList<Edge> edges;

    public Graph(int n, double p)
    {
        vertices = new ArrayList<>();
        edges = new ArrayList<>();
        for(int i=0; i<n ; i++)
        {
            Vertex v = new Vertex("V"+i);
            v.incidentEdges = new ArrayList<>();
            vertices.add(v);
        }
        for(int i=0; i<n ; i++)
        {
            for(int j=0;j<n;j++)
            {
                if(i==j)
                    continue;
                Edge e=new Edge(vertices.get(i),vertices.get(j),p,1);

                if(!edges.contains(e))
                {
                    edges.add(e);
                }
                vertices.get(i).incidentEdges.add(e);
            }
        }
    }

    public void computePaths(Vertex source)
```

```

{
    source.minDistance = 0;
    PriorityQueue<Vertex> vertexQueue = new PriorityQueue<Vertex>();
    vertexQueue.add(source);

    while (!vertexQueue.isEmpty()) {
        Vertex u = vertexQueue.poll();

        // Visit each edge exiting u
        for (Edge e : u.incidentEdges)
        {
            Vertex v = e.vertexB;
            double weight = e.weight;
            double distanceThroughU = u.minDistance + weight;
            if (distanceThroughU < v.minDistance) {
                vertexQueue.remove(v);
                v.minDistance = distanceThroughU ;
                v.previous = u;
                vertexQueue.add(v);
            }
        }
    }
}

public List<Vertex> getShortestPathTo(Vertex target)
{
    List<Vertex> path = new ArrayList<Vertex>();
    for (Vertex vertex = target; vertex != null; vertex = vertex.previous)
        path.add(vertex);
    Collections.reverse(path);
    return path;
}

}

class Vertex implements Comparable<Vertex>
{
    public final String name;
    public ArrayList<Edge> incidentEdges;
    public double minDistance = Double.POSITIVE_INFINITY;
    public Vertex previous;
    public Vertex(int n)
    {
        name = Integer.toString(n);
    }
    public Vertex(String argName) { name = argName; }
    public String toString() { return name; }
    public int compareTo(Vertex other)
    {

```

```

        return Double.compare(minDistance, other.minDistance);
    }
}

```

class Edge

```

{
    public final Vertex vertexA;
    public final Vertex vertexB;
    public final double weight;
    public final double probability;
    public Edge(Vertex a,Vertex b,double p, double argWeight)
    {
        vertexA = a;
        vertexB = b;
        weight = argWeight;
        probability = p;
    }
}

```

```

@Override
public boolean equals(Object o) {
    Edge o1 = (Edge) o;
    if((o1.vertexA==this.vertexA    &&    o1.vertexB==this.vertexB)    ||    (o1.vertexA==this.vertexB    &&
o1.vertexB==this.vertexA) )
        return true;

    return false;
}

```

```

@Override
public int hashCode(){
    return 1;
}
}

```

## MainClass.java

//responsible for implementing the pseudocode mentioned earlier

```
import java.util.*;
```

```
/**
```

```
 * @author rahulrdhanendran
```

```
 */
```

```
public class MainClass {
```

```
    public HashSet<String> combinations;
```

```
    public ArrayList<TableEntry> table;
```

```
    public MainClass()
```

```
{
```

```

    table = new ArrayList<>();
    combinations = new HashSet<>();
}

public void getAllCombinations(int []array, int length, String curr)
{
    if(curr.length() == length) {
        char[] chars = curr.toCharArray();
        Arrays.sort(chars);
        combinations.add(new String(chars));
    } else {
        for(int i = 0; i < array.length; i++) {
            String oldCurr = curr;
            if(curr.contains(Character.toString(Character.forDigit(array[i],10))))
                continue;
            curr += array[i];
            getAllCombinations(array,length,curr);
            curr = oldCurr;
        }
    }
}

```

```

public double getReliability(double prob, int scenario)
{
    table.clear();
    int a[] = new int[10];

    for(int i=0;i<10;i++)
    {
        a[i]=i;
    }
    for(int i=0;i<10;i++)
    {
        combinations.clear();
        getAllCombinations(a, i , "");
        if(combinations.size()==1)
        {
            table.add(new TableEntry());
        }
        else
        {
            for(String s : combinations)
            {
                Graph g = new Graph(5,prob);
                TableEntry t = new TableEntry();
                char badEdge[] = s.toCharArray();
                for(int j=0;j<s.length();j++)
                {

```

```

        int index = Character.getNumericValue(badEdge[j]);
        t.edgeNumber[index] = 0; //set as bad edge
        Edge e = g.edges.get(index);

g.vertices.get(Character.getNumericValue(e.vertexA.name.charAt(1))).incidentEdges.remove(e);
g.vertices.get(Character.getNumericValue(e.vertexB.name.charAt(1))).incidentEdges.remove(e);
    }
    g.computePaths(g.vertices.get(0));
    for(int k=0;k<g.vertices.size();k++)
    {
        List<Vertex> path = g.getShortestPathTo(g.vertices.get(k));
        if(path.get(path.size()-1).minDistance == Double.POSITIVE_INFINITY)
        {
            t.state = false;
            break;
        }
    }
    table.add(t);
}
}
}
TableEntry last = new TableEntry();
last.state = false;
for(int m=0;m<10;m++)
{
    last.edgeNumber[m]=0;
}
table.add(last);

if(scenario == 2)
{
    flipAndCalculate(table,prob);
}
return calculateReliability(table, prob);
}

public void flipAndCalculate(ArrayList<TableEntry> t, double prob)
{
    for(int k=0;k<100;k++)
    {
        double reliability = 0.0;
        for(int n=0;n<100;n++)
        {
            HashSet<Integer> randIndices = getRandomIndices(k);
            //flip
            for(int index : randIndices)
            {
                t.get(index).state = !t.get(index).state;
            }
        }
    }
}

```

```

    }
    //calculate
    reliability += calculateReliability(t, prob);
    //flip it back to original
    for(int index : randIndices)
    {
        t.get(index).state = !t.get(index).state;
    }
}

//System.out.println("Reliability when p=0.95 and k="+k+" is
"+(double)Math.round(reliability/50.0*10000)/10000);
System.out.println((double)Math.round(reliability/100.0*10000)/10000);
}

System.exit(0);
}

public HashSet getRandomIndices(int i)
{
    HashSet<Integer> hs = new HashSet<Integer>();
    Random rand = new Random();
    do
    {
        int index = rand.nextInt(1024);
        if(!hs.contains(new Integer(index)))
            hs.add(new Integer(index));
    }while(hs.size()<i);
    return hs;
}

public double calculateReliability(ArrayList<TableEntry> t, double prob)
{
    double reliability = 0.0;
    for(TableEntry te : t)
    {
        if(te.state)
        {
            double product=1.0;
            for(int i=0;i<10;i++)
            {
                if(te.edgeNumber[i]==0)
                    product = product*(1-prob);
                else
                    product = product*prob;
            }
            reliability += product;
        }
    }
}

```



```

        return reliability;
    }

    public static void main(String[] args) {
        MainClass m = new MainClass();
        for(double i=0.0;i<=1;i+=0.01)
        {
            //System.out.println("Reliability when p = "+(double)Math.round(i*100)/100+" :
            "+(double)Math.round(m.getReliability((double)Math.round(i*100)/100,1)*1000)/1000);
            System.out.println((double)Math.round(m.getReliability((double)Math.round(i*100)/100,1)*1000)/1000);
        }

        //scenario 2
        m.getReliability(0.95, 2);
    }
}

class TableEntry{
    int edgeNumber[] = new int[10];
    boolean state;
    //default: all edges are good and the system is up.
    public TableEntry()
    {
        state = true;
        for(int i=0;i<10;i++)
        {
            edgeNumber[i]=1;
        }
    }
}

```

## References :

- Dijkstra's Algorithm : [http://www.algolist.com/code/java/Dijkstra's\\_algorithm](http://www.algolist.com/code/java/Dijkstra's_algorithm)