Bioinformatics III

First Assignment

Max Jakob (2549155) Carolin Mayer (2552320)

April 27, 2018

Exercise 1.1: The random network

(a) In Listing 1, our implementation of the required Node class is shown.

```
Listing 1: Source code of Node.py
```

```
o class Node:
      def __init__(self , identifier):
           Sets node id and initialize empty node list that references its connected nodes
           self.id = identifier
5
           self.nodes = \{\}
       def hasLinkTo(self, node):
           Returns True if this node is connected to node asked for,
10
           False\ otherwise
           return node in self.nodes
       def addLinkTo(self , node):
15
           Adds link from this node to parameter node (only if there is no link connection already
           does not automatically care for a link from parameter node to this node
           if not (self.hasLinkTo(node)):
20
               self.nodes[node.id] = node
       def degree(self):
           Returns degree of this node
25
           return self.nodes.__len__()
      \mathbf{def} \ _{""} str_{--} (self):
30
           Returns\ id\ of\ node\ as\ string
           return self.id
```

(b) In Listing 2, our implementation of the required AbstractNetwork class is shown.

```
Listing 2: Source code of AbstractNetwork.py
```

```
class AbstractNetwork:
    """Abstract network definition, can not be instantiated"""

def __init__(self, amount_nodes, amount_links):
    """

Creates empty nodelist and call createNetwork of the extending class
```

```
self.nodes = \{\}
                self.mdegree = 0
               self.__createNetwork__(amount_nodes, amount_links)
 10
               for node in self.nodes:
                     degree = self.nodes[node].nodes.__len__()
                     if(degree>self.mdegree):
                          self.mdegree = degree
 15
          \mathbf{def} \ \_\mathtt{createNetwork}\_\mathtt{(} \, \mathtt{self} \, \, , \, \, \, \mathtt{amount\_nodes} \, , \, \, \, \mathtt{amount\_links} \, ) \colon
               Method overwritten by subclasses, nothing to do here
 20
               raise NotImplementedError
          \mathbf{def} \ \mathrm{appendNode} ( \ \mathrm{self} \ , \ \ \mathrm{node} \, ) \colon
               Appends node to network
 25
                self.nodes[node.id] = node
               if(self.mdegree < node.degree()):</pre>
                     self.mdegree = node.degree()
 30
         def maxDegree(self):
                Returns the maximum degree in this network
 35
               return int (self.mdegree)
          def size (self):
               Returns network size (here: number of nodes)
 40
               return self.nodes.__len__()
         def ___str__(self):
               Any string-representation of the network (something simply is enough)
 45
               s_{\underline{\ }}=\ "\ \underline{\ }"
               for node in self.nodes:

s = s + "{ " + str(node) + " }"

s = s + " -> { "
 50
                    for k in self.nodes[node].nodes:
                         s = s + str(k) + "
                    s = s + "} \setminus n"
 55
               return s
          def getNode(self, identifier):
               Returns node according to key
 60
               return self.nodes[identifier]
(c) In Listing 3, our implementation of the required RandomNetwork class is shown.
                            Listing 3: Source code of RandomNetwork.py
  o from AbstractNetwork import AbstractNetwork
    from Node import Node
    \mathbf{import} \hspace{0.2cm} \mathtt{random} \hspace{0.2cm} \# \hspace{0.2cm} you \hspace{0.2cm} will \hspace{0.2cm} need \hspace{0.2cm} it \hspace{0.2cm} :-)
```

 $""Random\ network\ implementation\ of\ AbstractNetwork""$

class RandomNetwork(AbstractNetwork):

```
\mathbf{def}_{\text{--}createNetwork\_-}(\text{self , amount\_nodes , amount\_links}) \colon \# \ \textit{remaining methods are taken from methods} = \mathbf{def}_{\text{--}nn} + \mathbf{de
                                                                   Creates\ a\ random\ network
10
                                                                    1. Build a list of n nodes
                                                                    2. For i=#links steps, add a connection between for two randomly chosen nodes that are
                                                                 random.seed()
                                                                  for i in range(0, amount_nodes):
15
                                                                                            AbstractNetwork.appendNode(self, node=Node(i))
                                                                   size = AbstractNetwork.size(self)-1
                                                                   for i in range(0, amount_links):
                                                                                           k1 = random.randint(0, size)
20
                                                                                           k2 = random.randint(0, size)
                                                                                           n1 = AbstractNetwork.getNode(self, k1)
                                                                                           n2 = AbstractNetwork.getNode(self, k2)
                                                                                           n1.addLinkTo(n2)
                                                                                           n2.addLinkTo(n1)
25
```

Exercise 1.2: Degree distribution of random networks

(a) In Listing 4, our implementation of the required DegreeDistribution class is shown.

Listing 4: Source code of RandomNetwork.py

```
o import numpy
  class DegreeDistribution:
       """ Calculates a degree distribution for a network"""
       \mathbf{def} __init___(self , network):
5
            Inits DegreeDistribution with a network and calculate its distribution
            size = network.maxDegree() +1
            self.hist = [0] * size
            for node in network.nodes:
10
                i = network.nodes[node].nodes.__len__()
                self.hist[i] = self.hist[i] + 1
       \mathbf{def} \ \ \mathtt{getNormalizedDistribution} \ (\ \mathtt{self} \ ) \colon
15
            Returns the computed normalized distribution
           num = numpy.sum(self.hist)
           return [i / num for i in self.hist]
```

(b) In Listing 5, our implementation of the required Tools class is shown.

```
Listing 5: Source code of RandomNetwork.py

o import matplotlib.pyplot as plt
import numpy as np
import math

def plotDistributionComparison(histograms, legend, title):

,'',

Plots a list of histograms with matching list of descriptions as the legend
,'',

# adjust size of elements in histogram
maxlen = 0

for h in histograms:
    if len(h) > maxlen:
        maxlen = len(h)

for h in histograms:
```

```
while len(h) != maxlen:
15
               h.append(0.0)
      # plots histograms
      for h in histograms:
           plt.plot(range(len(h)), h, marker = 'x')
20
      # remember: never forget labels! :-)
      plt.xlabel('Degree_of_k')
      plt.ylabel ('Density')
      # you don't have to do something here
      plt.legend(legend)
      plt.title(title)
      plt.tight_layout()# might throw a warning, no problem
      plt.show()
30
  def getPoissonDistributionHistogram(num_nodes, num_links, k):
       Generates a Poisson distribution histogram up to k
35
      lam = 2 * num\_links / num\_nodes
      res = [0]*k
      for i in range (0, k):
          res[i] = poisson(lam, i)
40
      return res
  def poisson (lam, k):
       if(k == 0):
          return math.\exp(-\text{lam})
45
      else:
          return (lam/k * poisson(lam,k-1))
```

All missing entries in the histogram where extended with 0.0. This has been done for the purpose of shrinking all histograms to the same length, thats to say to the same number ob buckets they contain.

(c) In Figure 1 and Figure 2, the results of the createAndPlotNetworks.py script that plots

the Poisson distribution (p) together with the degree distributions (r) with density against degree of k, is shown for each generated random network.

In Figure 1 one can see that all function are steadily rising (except r:50/100) until they reach their peaks always close to k = 4 and density around 0.2. After that, the functions are falling steadily until they slowly reach the density zero. One can see that the Poisson distribution of all networks shows a bell curve, whereas the degree distribution approaches to a bell curve as the number of edges and vertical increases. Hence, one can say that as higher the number of verticals and edges as closer the approximation to the bell curve. In Figure 2, the number of nodes is constantly 20000 for each network, whereas the number of edges increases. Hence, the peak of the function are shifted at the x-axes. Furthermore, one can see that the peaks approach to 0.2 as the number of edges increase. In this plot, r and p are equal, which demonstrates the approximation of r to p as the network grows.

Hence, as conclusion we can say that as the number of nodes and edges increase, the degree distribution of the random network approach to the Poisson distribution, which one can see in the approaching bell curve of it.

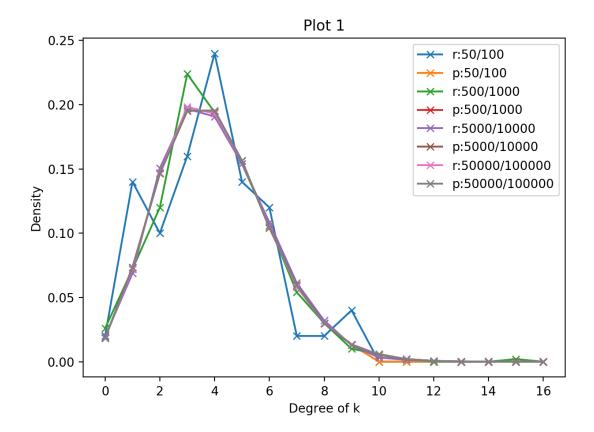


Figure 1: Plot 1

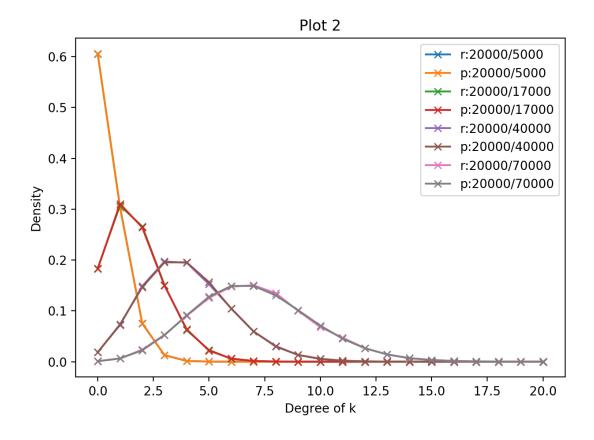


Figure 2: Plot 2