A Project For Police Patrol District Design

Group

Xuanzhou Chen (xchen793@wisc.edu)

Table of contents

- 1.Introduction
- 2. Data Generation
- 3. Three models

A. Balance the patrol center workload

- a. Mathematical Model
- b.Implementation
- c.Result and Discussion

B. Prevent 7 murders without bias

- a.Mathematical Model
- b.Implementation
- c.Result and Discussion

C. Intercept the criminal

- a. Mathematical Model
- b.Implementation
- c.Result and Discussion

Introduction

My project will use three different mathematical models to implement the optimization problem for police patrol district design.

Due to limited police resource (given a determined graph set of police service platforms), it is necessary to establish a mathematical model based on the actual conditions and needs of the city to allocate the jurisdiction of each police service platform and distribute the police resources. In this optimization, I generate the position of police service platforms(patrol centers) by random data. By default, I set the number of patrol centers be 1/4 of total traffic roads. The basic principle of optimization is the variance and the minimum cases that each patrol center is required to handle.

We need to find math models to find out:

- (1) Determine the area each patrol center should take control
- (2) The way to distribute patrol group to each traffic road depending on how many cases in the traffic roads.

By default, the number of patrol center would be 1/4 of total traffic roads, we will randomly generate those patrol centers (data generation).

In this project I will use matching alogrithm, network flow model and linear programing to solve 3 particular problems that police officer will encounter in their work.

Problem 1: For a city given the coordinates of every traffic road and the coordinates of every patrol center, how to find out the distance between any two patrol centers and traffic roads, as well as which traffic roads will be covered in each patrol center? Could we get a solution on how to distribute the police force to each traffic road?

Problem 2: If 7 people are being murderd at the same time and there are only 10 police cars left in each patrol center, how to assign task for each patrol center so that without bias all 7 people will be saved as soon as possible?

Problem 3: Could we find out the minimum time to take down the criminal if he is reported escaping at some traffic road? Does it depend on the police's speed?

I will solve the 3 problems above using the following models:

- 1. Network flow model: each traffic road can be treat as a node in a network, and they will flow through each patrol center, what we're trying to do is to balance the flows to patrol center by minimizing the variance of each patrol center.
- 2. Linear programing model: We will use integer optimization model and minimax problem model to solve the decision variables that will provide us a solution that will garantee that each muder prevention will have the same proirity.
- 3. Matching algorithm: When we know the traffics roads that needed to be at each time interval we can solve for the matching between these traffic roads so that we can find out if it is possible to lock the criminal in the circle at time t.

Data Generation

Data Form

First an array expressing the locations of 40 traffic roads will be generated, then the adjacency matrix will be calculated from this array.

The city we will generate will be expressed as a 40x40 matrix which means it contains 40 traffic roads. Then the patrol center, the location where the criminal start to escape and the 7 traffic roads that the murder occurs will be generated as vectors.

Math Model

Some constraints on the random funtion is consider as followed:

The scales

Since the mean of random function seems to be 1/2 (uniformly distributed), Multinplying it by a fator (here is 5 as the length of the whole square city) makes it look more normal. To make these road concentarte around the 2.5*2.5 square of the city we force 20 traffic roads to be in that range.

Implementation

In [1]:

```
import Pkg; Pkg. add("GraphRecipes")
import Pkg; Pkg. add("FixedPointNumbers")
using Random
using Graphs, NamedArrays, Clp, Combinatorics, Gurobi, JuMP, PyPlot
using GraphRecipes
using Plots
using FixedPointNumbers
Random. seed! (12345);
rng = MersenneTwister(12345);
1map=rand(Float64, (40, 2))
avg=0
rawmap=zeros(40,40)
for i in 1:40
    for j in 1:2
        if (i<20)
              1map[i, j]=1map[i, j]*5/2+5/4
        else
              1 \text{map}[i, j] *= 5
        end
    end
end
cmap=zeros (40, 40)
for i in 1:40
 avg = 1map[i, 1]
end
for i in 1:40
    for j in 1:40
      if (rand (Float64) < 4/40)
            cmap[i, j]=1
            cmap[j, i]=1
      end
    end
end
for i in 1:40
    for j in 1:40
      if(cmap[i, j]==1)
            rawmap[i, j]=sqrt((lmap[i, 1]-lmap[j, 1])^2+(lmap[i, 2]-lmap[j, 2])^2)
      end
    end
end
xs=zeros (40)
vs=zeros (40)
for i in 1:40
    xs[i]=1map[i,1]
    ys[i]=1map[i,2]
end
ps=zeros(10)
```

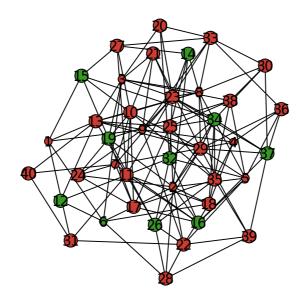
```
temp=shuffle(rng, Vector(1:40))
for i in 1:10
    ps[i]=temp[i]
end
colors= Array {RGB {FixedPointNumbers. Normed {UInt8, 8} } } (undef, 0)
for i in 1:40
    po=0
    for j in 1:10
        if (ps[j]==i)
            #println("pushing ", i)
           # println(colors)
            push! (colors, colorant"#389826")
            continue
        end
    end
    if (po==0)
        push! (colors, colorant"#CB3C33")
    end
   # println(colors)
println("The patrol centers are on the node:", ps)
finalCitymap=zeros(40,40)
for i in 1:40
    for j in 1:40
        finalCitymap[i, j]=rawmap[i, j]
    end
end
println("fianl city map is generated")
println(finalCitymap)
graphplot (rawmap,
          nodeshape=:circle,
          nodesize=0.1,
          curves=false,
          names = 1:40,
          fontsize = 10,
            nodecolor=colors,
          linecolor = :darkgrey
          )
```

```
Updating registry at `C:\Users\10096\.julia\registries\General`
        Updating git-repo https://github.com/JuliaRegistries/General.git
    Resolving package versions... ========>] 100.0 %.0 %
        Updating C:\Users\10096\.julia\environments\v1.3\Project.toml`
    [no changes]
        Updating `C:\Users\10096\. julia\environments\v1. 3\Manifest. toml`
    [no changes]
    Resolving package versions...
        Updating `C:\Users\10096\. julia\environments\v1. 3\Project. toml`
    [no changes]
        Updating `C:\Users\10096\. julia\environments\v1. 3\Manifest. toml`
    [no changes]
The patrol centers are on the node: [16.0, 37.0, 32.0, 26.0, 6.0, 12.0, 15.0, 34.0,
14.0, 19.0]
fianl city map is generated
[0.0\ 0.0\ 0.0\ 0.0\ 0.0\ 1.3147356275325803\ 0.0\ 0.0\ 1.2402388679943672\ 0.0\ 0.0\ 0.0\ 0.2
4401429126976074 \ \ 0.0 \ \ 0.7478364763822882 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ 
0.\ 9400446941576964\ 1.\ 7930494439117428\ 1.\ 6039291528435249\ 0.\ 6364353457511092\ 0.\ 0\ 0.
0 1.862073581810069 0.0 4.288012291457969 1.9784950962967571 0.0 0.0 2.24173686652
93087 \ 1. \ 597957135839688 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.7341410645413724 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.
0; \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 2.0333082654878005 \ 0.0 \ 0.34491317016217665 \ 0.0 \ 0.0 \ 0.0 \ 0.0
   0.\ 7889681410517357\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 1.\ 7035188094019451\ \ 0.\ 8291180654482114\ \ 3.\ 60
15505024509156 \ \ 0.0 \ \ 3.385269418440739 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0
   0.0\ 0.0\ 0.0\ 2.937593559732527\ 0.0\ 0.0\ 0.0\ 0.0\ 0.0\ 0.0\ 0.0\ 2.327020810417148
   0. \ 0 \ 2. \ 3083013092680633 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.1 \ 8527341356655689 \ 0.0 \ 0.0
0. \ 0 \ 0.0 \ 3.4741682045916162 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 
4266383\ 2.\ 5310480791965486\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0;\ 0.\ 0\ 1.\ 3636303031597445\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0
0. \ 0 \ 0.8198043967362628 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.5992698377524339 \ 0.0 \ 0.0 \ 0.0
   0. \ 0 \ 0.0 \ 1.927953661652927 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 2.864023966783614 \ 0.0 \ 0.0 \ 1.629
571427990713 0.0 0.0 0.0 1.580577079296287 0.0 2.0416924919611175 0.0 0.0; 1.31473
56275325803 \ \ 0.0 \ \ 2.0333082654878005 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \
12\ 0.0\ 0.0\ 0.0\ 0.0\ 1.0012814902579408\ 0.6037021283237396\ 0.0\ 0.0\ 0.0\ 0.0\ 0.0\ 0.0
   0\ \ 0.\ 0\ \ 0.\ 0\ \ 2.\ 327020810417148\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 2.\ 183127889099228\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0
   0.0 1.018173388070414 1.7298734945040362 0.0 0.0 0.0 0.0 0.0 1.256144935386404 3.
4090189122798202 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \  \, 0.0 \ 
897407812699; 0.0 0.0 0.34491317016217665 0.0 0.8198043967362628 0.0 0.0 0.0 0.0
   0.\ 9440780849031785\ 0.\ 0\ 0.\ 0\ 0.\ 8983192839544929\ 2.\ 0759708325489483\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.
   0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 3.3224411862299856 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 3.0725987371190566 \ 1.94390
47763322155 0.0 0.0 1.703887200532162 0.0 1.2100557757338362 0.0 0.0 0.0 0.0 0.0;
    764678333235152 0.0 0.0 0.0 0.0 0.0 0.0 0.1 446214113163635 0.47046169457464304
    1.\,\,4576063234892183\,\,\,3.\,\,689217102346471\,\,\,0.\,\,0\,\,\,0.\,\,0\,\,\,0.\,\,0\,\,\,0.\,\,0\,\,\,2.\,\,015034694832162\,\,\,0.\,\,0\,\,\,0.\,\,0\,\,\,0.
889099228 \ \ 0.9440780849031785 \ \ 0.0 \ \ 0.0 \ \ 1.6072523887062649 \ \ 0.0 \ \ 1.4209230260531562 \ \ 0.0
1.\ 685322568440766\ 0.\ 0\ 0.\ 4582114028727534\ 0.\ 0\ 0.0\ 0.\ 0\ 3.\ 6731384749966183\ 0.\ 0\ 0.\ 0\ 2.
442098698336773 \ 0.0 \ 0.0 \ 2.0998356755193197 \ 0.0 \ 3.6952799571286135 \ 0.0 \ 0.0 \ 0.0 \ 0.0
   0. \ 0 \ 1. \ 9420573270477912 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0
    1.6072523887062649 0.0 0.0 1.2927239511977109 0.0 0.0 0.8661082774430194 0.0 0.15
927903594092685 0.0 0.0 0.0 0.0 1.4872427889501034 0.0 0.0 2.729295129269066 2.995
0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 1.906959628225612 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 1.0404730780815354
   0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0
357 0.0 0.0 0.0 0.0 0.8983192839544929 1.3764678333235152 1.4209230260531562 1.292
7239511977109 \ 1.0404730780815354 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0
    1.\ 3837376663782595\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 2.\ 2817596068807573\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0
```

 $0. \ 0 \ 0.0 \ 0.0 \ 0.0; \ 0.7478364763822882 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 1.68532256844$ $0766\ 0.0\ 0.0\ 0.0\ 0.0\ 0.0\ 0.0\ 0.0\ 1.1650323666253117\ 0.0\ 0.0\ 0.0\ 0.0\ 0.0\ 0.$ $0\ 1.\ 3805410185931153\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 1.\ 1321446830199626\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.$ $0\ 0.0;\ 0.0\ 0.9400446941576964\ 0.0\ 0.0\ 0.5992698377524339\ 0.0\ 1.018173388070414\ 0.0$ $0. \ 0 \ 0.0 \ 0.8661082774430194 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 1.0152601944997885 \ 0.0 \ 0.0 \ 0.0 \ 0.0$ $0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 1.0913769064009562 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 1.9501$ 025778009635 0.0 0.0 0.0; 0.0 1.7930494439117428 0.0 0.0 0.0 1.0012814902579408 1. $7298734945040362 \ \ 0.0 \ \ 0.0 \ \ 4582114028727534 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 1.0152601944997885$ $0. \ 0 \ 0.0 \ 1.8941877290055904 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 2.5788552160592175 \ 1.0291908570165331$ 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 1.631765577604327 0.0 0.0 0.0 0.0 0.0 0.0 1. $6039291528435249 \ \ 0.0 \ \ 1.8527341356655689 \ \ 0.0 \ \ 0.6037021283237396 \ \ 0.0 \ \ 0$ 5927903594092685 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 1.1582415822438792 0.0 0. $0\;\,0.\;\,0\;\,0.\;\,0\;\,0.\;\,0\;\,0.\,\,0\;\,0.\,\,0\;\,0.\,\,0\;\,0.\,\,0\;\,0.\,\,0\;\,1.\,\,041334173797933\;\,0.\,\,0\;\,0.\,\,0\;\,0.\,\,0\;\,1.\,\,732158698469$ $4666\ 0.0\ 0.0;\ 0.0\ 0.6364353457511092\ 1.7035188094019451\ 0.0\ 0.0\ 0.0\ 0.0\ 0.0\ 0.0\ 0.$ $0\;\,0.\;\,0\;\,0.\;\,0\;\,0.\,\,0\;\,0.\,\,0\;\,1.\;\,1650323666253117\;\,0.\,\,0\;\,1.\;\,8941877290055904\;\,0.\,\,0\;\,0.\,\,0\;\,0.\,\,0\;\,0.\,\,0\;\,0.\,\,0$ 119248523 0.0 0.0; 0.0 0.0 0.8291180654482114 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 $0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 2.897179104649767 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0$ 3.6015505024509156 0.0 0.0 0.0 0.0 0.0 1.446214113163635 3.6731384749966183 0.0 $0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 1. \ 2307048850040625 \ 0.0 \ 0$ $10069 \ 0.0 \ 0.0 \ 1.927953661652927 \ 0.0 \ 0.0 \ 0.47046169457464304 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.$ $0\;\,0.\;0\;\,0.\;0\;\,0.\;0\;\,1.\;1582415822438792\;\,0.\;0\;\,0.\;0\;\,0.\;0\;\,0.\;0\;\,0.\;0\;\,0.\;0\;\,3.\;831415497128407\;\,0.\;0$ 735759558 0.0; 0.0 0.0 3.385269418440739 3.4741682045916162 0.0 0.0 1.256144935386 $404\ \ 3.\ 3224411862299856\ \ 1.\ 4576063234892183\ \ 0.\ \ 0\ \ 1.\ 4872427889501034\ \ 0.\ 0\ \ 0.\ 0\ \ 2.\ 2962175$ $54982736 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 2.897179104649767 \ \ 1.2307048850040625 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0$ $0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 1.4709927694301914 \ 4.882464869999629 \ 2.1254927153446292$ 0.0 0.0 3.3581053600222823 0.0 0.0 0.0; 0.0 4.288012291457969 0.0 0.0 0.0 0.0 3.4 $090189122798202 \ 0.0 \ 3.689217102346471 \ 2.442098698336773 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 2.$ $5788552160592175 \ \ 0.0 \ \$ $0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0; \ 0.0 \ 1.9784950962967571 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0$ $0.0\ 0.0\ 0.0\ 1.3837376663782595\ 2.059256468914019\ 0.0\ 0.0\ 1.0291908570165331\ 0.0$ $0. \ 0 \ 0.0 \ 0$ $3.\ 831415497128407\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 2.\ 2097147745432077\ \ 4.\ 93742514128189\ \ 0.\ 0\ \ 0.\ 0$ 3.6393625434821044 0.0 0.0 3.133563016474603 0.0 0.0 0.0 0.0 0.0; 0.0 0.0 0.0 0.0 $0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 2.0998356755193197 \ 2.995049782249853 \ 0.0 \ 0.0 \ 3.301798833327834$ $1.\ 3805410185931153\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0$ $0. \ 0 \ 0.46567589977938295 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 2.2417368665293087 \ 0.0 \ 0.$ $0\ \ 3.\ 726033628256204\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 2.\ 2097147745432077\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0$ 0.0 0.0 0.0 0.0 3.0983129250124644 0.0; 0.0 1.597957135839688 0.0 0.0 2.8640239667 83614 0.0 0.0 3.0725987371190566 2.015034694832162 3.6952799571286135 0.0 0.0 2.28 $17596068807573 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 2.4402772297509427 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 4.93$ $742514128189 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 1.7164951709963479 \ \ 0.0 \ \ 2.732261134514851$ 4. 199679633452012 0. 0 3. 586342075124794 0. 0; 0. 0 0. 0 0. 0 0. 0 0. 0 0. 0 1. 943904 $7763322155 \ \, 0.0$ $0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 3. \ 5527754466590724 \ 1. \ 1811112093229368 \ 0.0 \ 0.0 \ 2. \ 1014807386$ 917056 1.64388213349605 0.0 0.0; 1.9579309039937483 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0. $0\;\,0.\;0\;\,1.\;1113572890269245\;\,0.\;0\;\,0.\;0\;\,0.\;0\;\,0.\;0\;\,0.\;0\;\,0.\;0\;\,0.\;0\;\,0.\;0\;\,0.\;0\;\,0.\;0\;\,0.\\$ $3 \ 0.0 \$ $56508; \ 0.0 \ 0.0 \ 0.0 \ 1.629571427990713 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.759317355483930$ $1\ \ 0.\ \ 0.\ \ 0\ \ 0.\ \ 1.\ \ 0913769064009562\ \ 0.\ \ 0\ \ 0.\ \ 0\ \ 0.\ \ 0\ \ 2.\ \ 064988240669575\ \ 0.\ \ 0\ \ 0.\ \ 0\ \ 1.\ \ 470992769$ $0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0; \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 1.703887200532162 \ 3.56738787$ $15973486 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 2.251982357164593 \ 0.0 \ 0.0 \ 4.88246$ $4869999629 \ \ 0.0 \ \ 0.0 \ \ 0.46567589977938295 \ \ 0.0 \ \ 0.0 \ \ 3.5527754466590724 \ \ 0.0 \ \ 0.0 \ \ 0.0$

 $0. \ 0 \ 0.0 \ 3.938683677066273 \ 0.0 \ 0.0 \ 0.0 \ 0.0; \ 0.0 \ 0.7341410645413724 \ 0.0 \ 0.0 \ 0.0 \ 0.0$ $0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 1.1321446830199626 \ 0.0 \ 0.0 \ 1.041334173797933 \ 0.0 \ 0.00 \ 0.$ $0. \ 0 \ 0.0 \ 0.0 \ 2.1254927153446292 \ 0.0 \ 2.204733491468839 \ 0.0 \ 0.0 \ 0.0 \ 1.71649517099634$ 79 1. 1811112093229368 0. 0 0. 7079105326603977 0. 0 0. 0 0. 9147265457034718 0. 0 2. 5852 277489263042 0.0 0.0 0.0; 0.0 0.0 0.1.1591248694266383 0.0 0.0 0.0 1.21005577573 38362 0.0 1.9420573270477912 0.0 0.0 0.0 0.0 0.0 1.631765577604327 0.0 0.0 0.0 $0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 1.7090220118836175 \ 3.133563016474603 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0$ 0.9147265457034718 0.0 0.0 0.0 0.0 2.3789009621823842 0.0; 0.0 0.0 0.0 2.53104807 $91965486\ 1.\ 580577079296287\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0$ $0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 2.732261134514851 \ 0.0 \ 0.0 \ 0.0 \ 3.93868367706627$ $3 \ 0.0 \ 0.0 \ 0.0 \ 2.5761927268572697 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0$ $0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 1.9501025778009635 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 3.358105360022$ 7489263042 0.0 2.5761927268572697 0.0 0.0 1.6780513530829027 0.0; 0.0 0.0 2.937593 $559732527 \ \ 0.0 \ \ 2.0416924919611175 \ \ 0.0 \ \$ 1.7321586984694666 3.0531996119248523 0.0 0.0 0.0 0.0 0.0 2.6637477043374505 0.0 $0. \ 0 \ 0.0 \ 0$ 35759558 0. 0 0. 0 0. 0 0. 0 0. 0 3. 0983129250124644 3. 586342075124794 0. 0 0. 0 0. 0 0. 0 0.0 2.3789009621823842 0.0 1.6780513530829027 0.0 0.0 0.0; 0.0 0.0 0.0 0.0 0.0 0. $0\ \ 2.\ 812897407812699\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 2.\ 0829009303565447\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0$ $0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 3.39619858856508 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0$ 0.00.000.000.0

Out[1]:



Problem1

Analysis on the problem

We can think of every traffic road as vertex v_i , $i \in \{\text{the set of patrol center index}\}$.

The police need to arrive at the traffic road where the incident happend within 3 mins with a 60mph police car implies that the each patrol center should lie in the 3 miles circl of each traffic road so that it is possible for them to arrive on time.

So we can frist figure out the shortest distance between each patrol center and each traffic road and represent this disdtance as a matrix $D_{i,j}$

Then we find out each traffic road is cover by what patrol centers by iterating through matrix $D_{i,j}$, and represent it as a matrix $C_{i,j}$ in which $C_{i,j}=1$ means that ith traffic road is covered by jth patrol center. Then we check if Any traffic road can't be reached by any patrol center within 5 minutes, then connect them to the closest patrol center.

Then we will get the incident $\operatorname{array} In_j$ and size $\operatorname{array} Si_j$ which represent how many cases is in each traffic road and the number of cases that can be handeled by each patrol center. Then we will try to lower the variance and overload of each patrol center by using networkflow optimizer.

Math Model

In order to determine the shortest path between any two traffic roads in the city from the adjacency matrix we need to use warshall algorithmm .

1.1 warshall algorithnm.

let $D_{i,j}$ be the shortest path from traffic road i to j, the idea is:

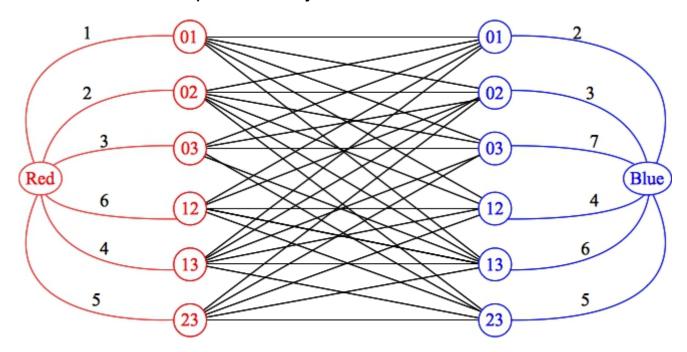
- 1)Get adjancecy matrix $A_{i,j}$ from the dataset , then initialize a new matrix that set the 0 entries to infinity.
- 2) for each pair of vertexs u and v, we will check if there is another vertex w such that $D_{u,w}+D_{w,v}>D_{u,v}.$

*pusedo code of this algorithmn:

```
for k: 1 to n for i:=1 to n for j:=1 to n  \text{If } D_{i,j} > D_{i,k} + D_{k,j} \text{:} \\ D_{i,j} = D_{i,k} + D_{k,j} \text{:}
```

1.2 create coverage matrix $C_{i,j}$ from $D_{i,j}$ and balance the overload of each police station

We first iterate through each entry of $D_{i,j}$, if $D_{i,j} < 3miles$ set $C_{i,j} = 1$. The idea is use networkflow problem as a way to set the constraint



Let n be the total number of patrol centers , and let m be the total number of traffic roads, sp_{j} be the working speed of each patrol center when it is at its full capacity,cf be the full capacity of each patrol center.

The variable for optimizer solver would be $X_{i,j} {\sf constriant}$ would be

$$X_{i,j} = C_{i,j} * X_{i,j}$$

 $\sum_{j=1}^{n} X_{i,j} = flowin_i$

$$\sum_{i=1}^m X_{i,j} <= maxload_j$$
,

The incident array and patrol center maxload would be $flowin_i$ and $maxload_j$. The variance objective function is computed as follow z1=min $\sum_{j=1}^n \frac{\left[\sum_{i=1}^m X_{i,j}-E(x)\right]^2}{n}$

Provided the comfortable load ,the overload objective function would be
$$z2=min$$
 $\lfloor \frac{\sum_{i=1}^{m}X_{i,j}}{cf[j]} \rfloor * (\frac{\sum_{i=1}^{m}X_{i,j}-cf[j]}{sp[j]})$

Implementation

In [2]:

```
import Pkg; Pkg. add("GraphRecipes")
import Pkg; Pkg. add("FixedPointNumbers")
import Pkg; Pkg. add("Ipopt")
import Pkg; Pkg. add("NLopt")
using Random
using Graphs, NamedArrays, Clp, Combinatorics, Gurobi, JuMP, PyPlot
```

```
using GraphRecipes
using Plots
using FixedPointNumbers
Random. seed! (1234);
rng = MersenneTwister(1234);
1map=rand(Float64, (40, 2))
#println(lmap)
avg=0
rawmap=zeros(40, 40)
for i in 1:40
    for j in 1:2
        if(i<20)
              1map[i, j]=1map[i, j]*5/2+5/4
        else
              1map[i, j]*=5
        end
    end
end
cmap=zeros(40, 40)
for i in 1:40
  avg = 1map[i, 1]
end
#println(avg)
for i in 1:40
    for j in 1:40
      if (rand (Float64) <4/40)
             cmap[i, j]=1
             cmap[j, i]=1
      end
    end
end
 #println(cmap)
for i in 1:40
    for j in 1:40
      if(cmap[i, j]==1)
             rawmap[i, j]=sqrt((1map[i, 1]-1map[j, 1])^2+(1map[i, 2]-1map[j, 2])^2)
      end
    end
end
xs=zeros(40)
ys=zeros(40)
for i in 1:40
    xs[i]=1map[i,1]
    ys[i]=1map[i, 2]
end
ps=zeros(10)
temp=shuffle(rng, Vector(1:40))
for i in 1:10
    ps[i]=temp[i]
colors= Array {RGB {FixedPointNumbers. Normed {UInt8, 8} }} (undef, 0)
po=0
for i in 1:40
```

```
po=0
    for j in 1:10
        if (ps[j]==i)
            push! (colors, colorant"#389826")
            break
        end
    end
    if(po==0)
        push! (colors, colorant"#CB3C33")
    end
end
println("The patrol centers are on the node:", ps)
finalCitymap=zeros(40,40)
for i in 1:40
    for j in 1:40
        finalCitymap[i, j]=rawmap[i, j]
    end
end
#println("fianl city map is generated")
#println(finalCitymap)
plt=graphplot(rawmap,
          nodeshape=:circle,
          nodesize=0.1,
          curves=false,
          names = 1:40,
          fontsize = 10,
            nodecolor=colors,
          linecolor = :darkgrey
          )
n = 40
for i in 1:n
    for j in 1:n
        if (i==j)
            rawmap[i, i]=0
        elseif ( rawmap[i, j] == 0)
            rawmap[i, j]=10000
        end
    end
end
#println(rawmap)
for k in 1:n
    for i in 1:n
       for j in 1:n
```

```
if(rawmap[i, j]>rawmap[i, k]+rawmap[k, j])
                 rawmap[i, j]=rawmap[i, k]+rawmap[k, j]
             end
        end \\
    end
end
#println(rawmap)
covermap=zeros(10,40)
bcovermap=zeros(10,40)
for i in 1:10
       for j in 1:40
           if(rawmap[Int(ps[i]), j]<3)</pre>
                covermap[i, j] = rawmap[Int(ps[i]), j]
                bcovermap[i, j]=1
             end
        end
end
#println(covermap)
#println(bcovermap)
for i in 1:10
    for j in 1:40
         if(j in ps)
           bcovermap[i, j]=0
        end
    end \\
end
for i in 1:10
    println("patrol center", i , "covers:")
    for j in 1:40
         if(bcovermap[i, j]==1)
             println(j, ", ")
        end
    end
end
printcovermap=zeros (40, 40)
for i in 1:10
    for j in 1:40
           if(covermap[i, j]>0)
              printcovermap[i, j]=covermap[i, j]
            end
    end
end
ps1=8*rand(Float64, 10)
inflow=2*rand(Float64, 40)
expx=0
tdeg=zeros(40)
p1=zeros(10)
```

```
println("The crime cases in each traffic road is:", inflow)
println("The capacity of each patrol center is:", psl)
#println(p1)
#println(expx)
#println(bcovermap)
#println("Before optimization the variance is:", sum((p1[i]-expx)^2/10 for i in 1:10
))
m = Model(Gurobi.Optimizer)
Ovariable (m, x[1:10, 1:40])
@constraint(m, con1[i in 1:10, j in 1:40], x[i, j] == bcovermap[i, j] *x[i, j])
@constraint(m, con4[i in 1:10, j in 1:40], x[i, j] >= 0)
@constraint(m, con3[ i in 1:10], sum(x[i, j] for j in 1:40 )/inflow[i]>=0.8)
@constraint(m, con2[j in 1:40], sum(x[i, j] for i in 1:10) \le inflow[j])
println("1")
@constraint(m, con5[ i in 1:10], sum(x[i,j] for j in 1:40) \leq ps1[i])
@objective(m, Min, sum((sum(x[i, j] for j in 40)-sum(x[n, m]/10 for n in 1:10 for m
in 1:40))<sup>2</sup> for i in 1:10 )/10)
optimize! (m)
println("The how much cases should be handled by each patrol center:")
println(value.(x))
resultmap=zeros (40, 40)
for i in 1:40
    for j in 1:40
        for k in 1:10
             if(i == ps[k])
                   resultmap[i, j]=value. (x)[k, j]
                 resultmap[j, i]=value. (x)[k, j]
             end
        end
    \quad \text{end} \quad
end
graphplot (resultmap,
          nodeshape=:circle,
          nodesize=0.1,
           linesize=0.001,
          curves=false,
          names = 1:40,
          fontsize = 10,
             nodecolor=colors,
          linecolor = :darkgrey
          )
```

```
Resolving package versions...
     Updating C:\Users\10096\.julia\environments\v1.3\Project.toml
   [no changes]
     \label{lem:updating `C:\Users\10096\. julia\environments\v1.3\Manifest.toml`} In the continuous of the continuous conti
   [no changes]
   Resolving package versions...
      Updating C:\Users\10096\.julia\environments\v1.3\Project.toml
   [no changes]
      Updating C:\Users\10096\.julia\environments\v1.3\Manifest.toml
   [no changes]
   Resolving package versions...
     Updating `C:\Users\10096\. julia\environments\v1.3\Project.toml`
   [no changes]
     Updating `C:\Users\10096\. julia\environments\v1.3\Manifest.toml`
   [no changes]
   Resolving package versions...
     Updating C:\Users\10096\.julia\environments\v1.3\Project.toml`
   [no changes]
     Updating `C:\Users\10096\. julia\environments\v1.3\Manifest.toml`
   [no changes]
The patrol centers are on the node: [28.0, 38.0, 31.0, 32.0, 33.0, 18.0,
   37.0, 9.0, 27.0, 14.0]
patrol centerlcovers:
4,
5,
6,
10,
11,
13,
15,
24,
25,
35,
patrol center2covers:
3,
7,
17,
40.
patrol center3covers:
3,
6,
8,
11,
15,
20,
36,
39,
patrol center4covers:
2,
15,
40,
patrol center5covers:
2,
7,
8,
10,
11,
12,
13,
30,
35,
```

```
patrol center6covers:
1,
2,
3,
4,
5,
6,
7,
8,
10,
11,
12,
13,
15,
16,
17,
19,
21,
23,
24,
25,
29,
30,
40,
patrol center7covers:
1,
4,
5,
6,
7,
8,
10,
11,
15,
22,
25,
30,
40,
patrol center8covers:
1,
2,
4,
5,
6,
7,
8,
10,
11,
13,
15,
19,
22,
24,
25,
29,
patrol center9covers:
3,
4,
8,
12,
35,
```

```
36,
39,
patrol center10covers:
1.
2,
4,
5,
6,
7,
8,
10,
11,
13,
15,
16.
19,
20,
21,
24,
25,
30,
40.
The crime cases in each traffic road is: [0.5718276446950155, 1.384517473
8173975, 0.8503734992636205, 1.3824965879279967, 0.31840565555037603, 0.
598524309133202, 1.7202321631674384, 0.41512443018434775, 0.598903111285
7242, 0.5331078354561005, 1.4931418723189425, 0.13137181604552417, 0.436
8628637867964, 1.236669749591039, 0.6753083494709444, 0.935712347575751
8, 1.30356827784095, 0.4885013663970357, 0.3325931135571425, 1.798691021
4784118, 1.9923329033975747, 1.3971925744414704, 0.4026703405981906, 0.1
2945863613795705, 1.0468071410053366, 1.2982245691005438, 0.544502294215
4774, 0.2707600007581368, 0.5285204638442078, 0.830476415193218, 0.66205
90788831353, 0.9881328537009058, 0.859528360179111, 0.535492496810166,
 0.6291222809468517, 1.5607530921367947, 0.8761364795223496, 1.272765171
4838593, 1.8708374093866027, 0.6233598069259565]
The capacity of each patrol center is: [4.0270087667610195, 4.46637556868
2446, \ \ 3.\ 9194505340182513, \ \ 6.\ 734265788704446, \ \ 5.\ 780587148554769, \ \ 1.\ 086455
2901487912, 4.030355130748989, 3.019735970169796, 3.7698788113168007, 2.
4436935880757478]
Academic license - for non-commercial use only
Academic license - for non-commercial use only
Gurobi Optimizer version 9.0.2 build v9.0.2rc0 (win64)
Optimize a model with 860 rows, 400 columns and 1888 nonzeros
Model fingerprint: 0x7ea5d655
Model has 76300 quadratic objective terms
Coefficient statistics:
                   [6e-01, 3e+00]
  Matrix range
  Objective range [0e+00, 0e+00]
  Q0bjective range [2e-02, 2e-01]
  Bounds range
                   [0e+00, 0e+00]
                   [1e-01, 7e+00]
 RHS range
Presolve removed 823 rows and 288 columns
Presolve time: 0.09s
Presolved: 37 rows, 121 columns, 232 nonzeros
Presolved model has 5793 quadratic objective terms
Ordering time: 0.00s
Barrier statistics:
Free vars : 5
 AA' NZ
            : 1.710e+02
Factor NZ
           : 7.030e+02
```

Factor Ops: 1.519e+04 (less than 1 second per iteration)

Threads : 1

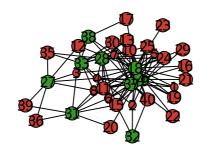
		0bje	ective	Residu	ual		
Ιt	er	Primal	Dual	Primal	Dua1	Comp1	Tim
е	0	3.34177392e+05	-3.58767858e+05	2.67e+04	4.28e-06	1.00e+06	0
S	1	1.33097958e+03	-5.18470985e+04	9.00e+02	1.44e-07	3.52e+04	0
S S	2	3.43629988e+00	-4.87967714e+04	2.38e+00 3	3.80e-10	3.97e+02	0
S	3	3.15580630e+00	-5.01093327e+03	5.23e-02 8	8.37e-12	3.21e+01	0
s	4	3.13652729e+00	-1.83202866e+01	1.60e-04 2	2.53e-14	1.35e-01	0
s	5	1.85386452e+00	-4.93526712e+00	3.48e-05 §	5.55e-15	4.27e-02	0
S	6	1.30405298e+00	-3.49130221e+00	1.85e-05 3	3.00e-15	3.02e-02	0
s	7	5.99417399e-01	-3.20087402e-01	1.99e-06	4.44e-16	5.78e-03	0
s	8	3.95517964e-01	5.42350884e-02	2.06e-07 §	5.55e-17	2.15e-03	0
s	9	3.87906517e-01	3.02294484e-01	9.85e-09		5.38e-04	0
s	10	3.74293083e-01	3.59125744e-01	9.10e-15 {	5.55e-17	9.54e-05	0
s	11	3.73004159e-01	3.72905736e-01	2.22e-16		6. 19e-07	0
s	12	3.72996753e-01	3.72996654e-01	8.88e-16 3		6. 19e-10	0
s	13	3.72996745e-01	3.72996745e-01	4.44e-16	1.96e-17	6. 20e-13	0

Barrier solved model in 13 iterations and 0.10 seconds Optimal objective 3.72996745e-01

The how much cases should be handled by each patrol center: $\lceil 0.0\ 0.0\ 0.0\ 0.04995982177453549\ 0.04267315843344367\ 0.05173887941759819$ $4\ \ 0.\ 0\ \ 0.\ 0\ \ 0.\ 0\ \ 0.48497143138645\ \ 0.\ 050314067584685734\ \ 0.\ 0\ \ 0.\ 0487213153766$ $15996 \ \ 0.0 \ \ 0.03406229431360458 \ \ 0.0 \ \ 0.$ $9514259786 \ \ 0.051568651742917054 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.0$ 8367364464713285 0.0 0.0 0.0 0.0 0.0; 0.0 0.0 0.2903549509558251 0.0 0.0 $0\ 0.0\ 0.07397948431137073\ 0.0\ 0.054566918951322846\ 0.0\ 0.0\ 0.09365083832$ $117115 \ 0.0 \ 0.0 \ 0.0 \ 0.05115433151716762 \ 0.0 \ 0.0 \ 0.0 \ 0.121119536562572$ $0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.1048306942$ 9049104 0.0 0.0 0.10804489520298799 0.0; 0.0 0.6285029625408969 0.0 0.0 $0. \ 0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.0 \ 0.35282234630806203 \ 0.0 \ 0.0 \ 0.0$ $0.0\ 0.0\ 0.12467196149766638;\ 0.0\ 0.022066153050975052\ 0.0\ 0.0\ 0.0$ $0. \ 0. \ 0.030180228279870644 \ \ 0.027174126726649723 \ \ 0.0 \ \ 0.030146511831760422$ $0.\,\, 030997507078621735 \ \ \, 0.\,\, 021715863734876724 \ \ \, 0.\,\, 03021172286003626 \ \ \, 0.\,\, 0\,\,\, 0.\,\, 0$ $4022\ 0.0\ 0.0\ 0.0\ 0.0\ 0.030275057187088394\ 0.0\ 0.0\ 0.0\ 0.0\ 0.0;\ 0.0156003$ $69837172659 \quad 0. \quad 013917640031387138 \quad 0. \quad 015320374330892462 \quad 0. \quad 0170751292431059$ $2\ 0.01348080972228629\ 0.014916958485558345\ 0.016631080285450708\ 0.013472$ $378974237989 \ 0.0 \ 0.014490419237223495 \ 0.016698320860869195 \ 0.01204791321$

 $330856 \quad 0. \ 014397782073547333 \quad 0. \ 0. \ 013436195212943466 \quad 0. \ 02164189493542492$ 0.018819844221892357 0.0 0.014974172280670056 0.0 0.02784974366935396 0. $0\ \ 0.\ 017506896305207673\ \ 0.\ 011844951966559287\ \ 0.\ 016705293122827803\ \ 0.\ 0\ \ 0.\ 0$ 0.00.017118793745660825 0.016200524324639646 0.00.00.00.00.00.00.00.00 $0\ 0.\ 0\ 0.\ 12467196123471622;\ 0.\ 11314487182800842\ 0.\ 0\ 0.\ 0\ 12808959175$ $75511743814768 \ \ 0.0 \ \ 0.09312833044268784 \ \ 0.1287194025181583 \ \ 0.0 \ \ 0.0 \ \ 0.0 \ \ 0.$ $06211548695338626 \ \, 0.\,\, 0\,\,\, 0.\,\, 0\,\,\, 0.\,\, 0\,\,\, 0.\,\, 0\,\,\, 0.\,\, 0\,\,\, 0.\,\, 10839479620193397 \ \, 0.\,\, 0\,\,\, 0.\,\, 0\,\,\, 0.$ $12638423552544495 \ \, 0.0 \ \, 0.0 \ \, 0.0 \ \, 0.12154612188509235 \ \, 0.0 \ \,$ $0\ 0.0\ 0.0\ 0.0\ 0.12467196147869823;\ 0.022309768705775553\ 0.0183438792$ $1648661 \ \ 0.0 \ \ 0.022893691098079705 \ \ 0.018381763214187816 \ \ 0.0206293306690669$ $2\ 0.\ 022441473261306827\ 0.\ 018334155595750714\ 0.\ 0\ 0.\ 019893745170375333\ 0.\ 0$ 22597626445127347 0. 0 0. 01980200812230314 0. 0 0. 017789148713879852 0. 0 $0. \ 0 \ 0.0 \ 0.021737070999648334 \ 0.0 \ 0.0 \ 0.02537870669919373 \ 0.0 \ 0.01551297$ $3696365772 \quad 0.022905510892345674 \quad 0.0 \quad 0.0 \quad 0.023148691656758764 \quad 0.0 \quad 0.0$ 499421751816 0.0 0.0 0.0 0.05542324109747911 0.0 0.0 0.0 0.0348606785331 $2425 \ \, 0.0 \ \,$ $0.0\ 0.0\ 0.0\ 0.0\ 0.0\ 0.0\ 0.07262728183077453\ 0.08236011912984949\ 0.0\ 0.0\ 0.08286011912984949$ 196239824255315 0.0; 0.016134853163132245 0.01409633342024666 0.0 0.0168 $34298028327352 \ \, 0.014276393353390726 \ \, 0.015676727966681913 \ \, 0.0164271257689$ $4296 \quad 0.01427372555033854 \quad 0.0 \quad 0.015280092299759372 \quad 0.01650787599388145 \quad 0.$ $0\ \ 0.\ 015197369257451873\ \ 0.\ \ 0\ \ 0.013819642219858505\ \ 0.\ 020460747941905087\ \ 0.\ \ 0$ $0.0\ 0.01579816047612364\ 0.024348560407608365\ 0.02707256500396468\ 0.0\ 0.0$ $0.\ 012512778949511716\ 0.\ 016734630951399567\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 0\ 0.\ 016362426399$ 948456 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.12467196121889916

Out[2]:



Result and Interpretation

1. The row of matrix represent how much patrol force from i^{th} row should be sent to j^{th} traffic road.

- 2. They should be integer in the real world, so more investigation will be need. Here a better optimization solution is to further use the integer programming.
- 3. The company owner should build their store more evenly so some store won't sold out their trading cards and no store should be over stocked.

Problem 2

Math Model

We can think of this problem as an integer optimization problem with decision variables.

$$X_{i,j} = egin{cases} 1, & ext{if i^{th} traffic road should be handled by j^{th} patrol center} \ 0, & ext{if } i^{th} ext{ traffic road should not be handled by } j^{th} ext{ patrol center} \end{cases}$$

And since one patrol center can only bloack one traffic road, and each traffic road only needs to be blocked by exactly one patrol center. These are two constraints we should consider in modeling.

What we need to do here is to extract the submatrix of problem 1 $C_{i,j}$ to represent the distance between n patrol centers and m traffic roads that need to be blocked.

Solution

```
In [3]:
using Random
using Graphs, NamedArrays, Clp, Combinatorics, Gurobi, JuMP, PyPlot
using GraphRecipes
using Plots
using FixedPointNumbers
using Ipopt
using NLopt
Random. seed! (1234);
rng = MersenneTwister(1234);
function to fint the shortest path between 2 traffic road
function Dijkstra(src, mat)
  visited=zeros(40)
  visited[src]=1
  dist=zeros (40)
  pre=zeros(40)
  minci=0
  for i in 1:40
  dist[i]=mat[src, i]
```

```
pre[i]=src
   end
   for i in 1:40
     minc=10000000
      for j in 1:40
         if(visited[j]==0&&dist[j] <minc)</pre>
             minci=j
             minc=dist[j]
         end
      end
      visited[minci]=1
      for j in 1:40
         if(visited[j] == 0 \& mat[minci, j]! = 10000000 \& minc+mat[minci, j] < dist[j])
             dist[j]=minc+mat[minci, j]
             pre[j]=minci
         end
      end
   end
   return pre
end
function routea2b(a, b, mat)
  pre= Dijkstra(a, mat)
   i=Int(b)
   routeorder=[]
   print("route from ", a, " to ", b, " :")
   while (i!=a)
      push! (routeorder, i)
      i=Int(pre[i])
   end
   print(a, "->")
   for i in Iterators.reverse(1:length(routeorder)-1)
     print(routeorder[i], "->")
   end
end
#generating data for the problem
1map=rand(Float64, (40, 2))
#println(lmap)
avg=0
rawmap=zeros(40,40)
for i in 1:40
   for j in 1:2
      if(i<20)
          1 \text{map}[i, j] = 1 \text{map}[i, j] * 5/2 + 5/4
      else
          1map[i, j]*=5
      end
   end
end
```

```
cmap=zeros(40, 40)
for i in 1:40
  avg = 1map[i, 1]
end
println(avg/40)
for i in 1:40
    for j in 1:40
      if (rand(Float64) < 4/40)
             cmap[i, j]=1
             cmap[j, i]=1
      end
    end
end
 #println(cmap)
for i in 1:40
    for j in 1:40
      if(cmap[i, j]==1)
            rawmap[i, j]=sqrt((1map[i, 1]-1map[j, 1])^2+(1map[i, 2]-1map[j, 2])^2)
      end
    end
end
xs=zeros(40)
ys=zeros(40)
for i in 1:40
    xs[i]=1map[i, 1]
    ys[i]=1map[i, 2]
end
ps=zeros(10)
temp=shuffle(rng, Vector(1:40))
for i in 1:10
    ps[i]=temp[i]
end
lockRoad=zeros(7)
for i in 1:7
    lockRoad[i]=temp[10+i]
end
colors= Array {RGB {FixedPointNumbers. Normed {UInt8, 8} }} (undef, 0)
po=0
for i in 1:40
    po=0
    for j in 1:10
        if (ps[j]==i)
             push! (colors, colorant"#389826")
             po=1
             break
        end
    end \\
    if(po==0)
        push! (colors, colorant"#CB3C33")
    end
```

```
end
println("The patrol centers are on the node:", ps)
adj=zeros (40, 40)
for i in 1:n
    for j in 1:n
        if (i==j)
             rawmap[i, i]=0
        elseif ( rawmap[i, j] == 0)
             rawmap[i, j]=10000
        adj[i, j]=rawmap[i, j]
    end
end
#println(rawmap)
for k in 1:n
    for i in 1:n
       for j in 1:n
           if(rawmap[i, j]>rawmap[i, k]+rawmap[k, j])
                 rawmap[i, j]=rawmap[i, k]+rawmap[k, j]
             end
        end
    end
end
#println(rawmap)
covermap=zeros(10,40)
bcovermap=zeros(10,40)
for i in 1:10
       for j in 1:n
           if(rawmap[Int(ps[i]), j]<3)</pre>
                covermap[i, j] = rawmap[Int(ps[i]), j]
                bcovermap[i, j]=1
             end
        end
#println("covermap", covermap)
#println(bcovermap)
for i in 1:10
    for j in 1:40
        if(j in ps)
           bcovermap[i, j]=0
        end \\
    end
end \\
lockMap=zeros(10, 7)
for i in 1:10
```

```
for j in 1:7
     lockMap[i, j]=rawmap[Int(ps[i]), Int(lockRoad[j])]
   end
end
#println(lockRoad)
println("murder locations", lockRoad)
#solution of the problem
m = Model(Gurobi.Optimizer)
@variable(m, x[1:10, 1:7], Bin)
@variable(m, r)
@constraint(m, con2[i in 1:10, j in 1:7], r >= lockMap[i, j] *x[i, j])
@constraint(m, con1[i in 1:10], sum(x[i, j] for j in 1:7) \le 1)
@constraint(m, con4[j in 1:7], sum(x[i, j] for i in 1:10) == 1)
@objective(m, Min, r)
@time(optimize!(m))
println(value.(x))
for i in 1:10
  for j in 1:7
     if(value.(x)[i,j]==1)
        routea2b(Int(ps[i]), Int(lockRoad[j]), adj)
     end
  end
  println()
end
```

2.491098786608659

WARNING: using NLopt.optimize! in module Main conflicts with an existing identifier.

The patrol centers are on the node: [28.0, 38.0, 31.0, 32.0, 33.0, 18.0, 37.0, 9.0, 27.0, 14.0]

murder locations[39.0, 30.0, 34.0, 17.0, 5.0, 15.0, 6.0]

Academic license - for non-commercial use only Academic license - for non-commercial use only

Gurobi Optimizer version 9.0.2 build v9.0.2rc0 (win64)

Optimize a model with 87 rows, 71 columns and 280 nonzeros

Model fingerprint: 0xdbe669c7

Variable types: 1 continuous, 70 integer (70 binary)

Coefficient statistics:

Matrix range [9e-01, 7e+00] Objective range [1e+00, 1e+00] Bounds range [0e+00, 0e+00] RHS range [1e+00, 1e+00]

Found heuristic solution: objective 4.5836477

Presolve removed 9 rows and 9 columns

Presolve time: 0.01s

Presolved: 78 rows, 62 columns, 244 nonzeros

Variable types: 1 continuous, 61 integer (61 binary)

Root relaxation: objective 1.062601e+00, 76 iterations, 0.01 seconds

Nodes			Current Node			Objective Bounds			Work	
Exp	ol Une:	xp1	Obj Deptl	n Int	Inf	Incumbent	BestBd	Gap	It/Node	
Time	e									
0s	0	0	1.38328	0	20	4. 58365	1.38328	69.8%	_	
H Os	0	0				4.0729796	1.38328	66.0%	_	
H Os	0	0				3.9474676	1. 38328	65.0%	-	
0s	0	0	1.38328	0	20	3. 94747	1. 38328	65.0%	-	

Explored 1 nodes (123 simplex iterations) in 0.06 seconds Thread count was 8 (of 8 available processors)

Solution count 3: 3.94747 4.07298 4.58365

Optimal solution found (tolerance 1.00e-04)

Best objective 3.947467604050e+00, best bound 3.947467604050e+00, gap 0.0000%

route from 38 to 17:38-

route from 33 to 30 :33-> route from 18 to 5 :18->6->5-> route from 37 to 6 :37->6-> route from 9 to 34 :9->34-> route from 27 to 39 :27->39-> route from 14 to 15 :14->15->

Result and Interpretation

In this case, if there are 7 people are murdered, the minimum total distance is 3.947467604050e+00.

We can obtain 7 routes from the output are

route from 38 to 17 :38->

route from 33 to 30 :33->

route from 18 to 5:18->6->5->

route from 37 to 6:37->6->

route from 9 to 34 :9->34->

route from 27 to 39 :27->39->

route from 14 to 15:14->15->

Problem3

Math model

Since the criminal is escaping we need to calculte the possible area the criminal could be by drawing the cricle with $r_c=v_c*t$ where v is the escaping speed of the criminal and t is the time the criminal has been escaping.

Then we needs to iterate through the traffic roads in the row of the traffic road where the criminal start to escaping and add it to the blocking set B_i .

Then we need to calculate the police reaction circles with $r_p=v_p*y$, so that we can iterate through the patrol center to construct possible edges from set of patrol center P_i to B_i and represent this bipartite graph using a matrix $U_{i,j}$.

Then then we iterate through each entry of patrol center to traffic road matrix $T_{i,j}$ which contains the distance between each patrol center and each trafficroad. Set $U_{i,j}=T_{i,j}$ if these 2 condition is satisfied:

1)
$$T_{i,j} \leq r_p$$
2) $i \in B_i$

Then We need to find a matching M_i in the bipartite graph $U_{i,j}$.

3.1 matching model in bipartite graph

We will use hungry algorithmm to find the matching in $U_{i,j}$. The idea of this algorithmn is that :.

- 1)If you can find an augmented path in bipartite graph, there is a better mathicing that is closer to perfect match than the current path.
- 2) The subgraph don't have augmented graph implies that there is not agumneted grap

pusedo code

Solution

In [4]:

```
using Random
using Graphs, NamedArrays, Clp, Combinatorics, Gurobi, JuMP, PyPlot
using GraphRecipes
using Plots
using FixedPointNumbers
#generating data for the problem
Random. seed! (1234):
rng = MersenneTwister(1234);
1map=rand(Float64, (40, 2))
avg=0
rawmap=zeros(40,40)
for i in 1:40
  for j in 1:2
     if(i<20)
         1map[i, j]=1map[i, j]*5/2+5/4
     else
         1map[i, j]*=5
     end
  \quad \text{end} \quad
end
cmap=zeros (40, 40)
for i in 1:40
 avg = 1map[i, 1]
end
for i in 1:40
  for j in 1:40
    if(rand(Float64) < 4/40)
        cmap[i, j]=1
        cmap[j, i]=1
    end
  end
```

```
end
 #println(cmap)
for i in 1:40
    for j in 1:40
      if(cmap[i, j]==1)
            rawmap[i, j]=sqrt((lmap[i, 1]-lmap[j, 1])^2+(lmap[i, 2]-lmap[j, 2])^2)
      end
    end
end
xs=zeros(40)
ys=zeros(40)
for i in 1:40
    xs[i]=1map[i,1]
    ys[i]=1map[i, 2]
end
ps=zeros(10)
temp=shuffle(rng, Vector(1:40))
for i in 1:10
    ps[i]=temp[i]
end
colors = Array {RGB {FixedPointNumbers. Normed {UInt8, 8}}} (undef, 0)
for i in 1:40
    po=0
    for j in 1:10
        if (ps[j]==i)
            push! (colors, colorant"#389826")
            po=1
            break
        end
    end
    if(po==0)
        push! (colors, colorant"#CB3C33")
    end
 end
println("The patrol centers are on the node:", ps)
finalCitymap=zeros(40, 40)
for i in 1:40
    for j in 1:40
        finalCitymap[i, j]=rawmap[i, j]
    end
end
plt=graphplot(rawmap,
          nodeshape=:circle,
```

```
nodesize=0.1,
          curves=false,
          names = 1:40,
          fontsize = 10,
             nodecolor=colors,
          linecolor = :darkgrey
n = 40
for i in 1:n
    for j in 1:n
        if (i==j)
            rawmap[i, i] = 0
        elseif ( rawmap[i, j] == 0)
             rawmap[i, j]=10000
        end
    end
end
#println(rawmap)
for k in 1:n
    for i in 1:n
       for j in 1:n
           if(rawmap[i, j]>rawmap[i, k]+rawmap[k, j])
                 rawmap[i, j]=rawmap[i, k]+rawmap[k, j]
             end
        end
    end
end
#println(rawmap)
covermap=zeros(10,40)
bcovermap=zeros(10,40)
for i in 1:10
       for j in 1:40
           if(rawmap[Int(ps[i]), j]<3)</pre>
                covermap[i, j] = rawmap[Int(ps[i]), j]
                bcovermap[i, j]=1
             end
        end
end
#println(covermap)
#println(bcovermap)
for i in 1:10
    for j in 1:40
        if(j in ps)
           bcovermap[i, j]=0
        end
    end
end
```

```
printcovermap=zeros (40, 40)
for i in 1:10
  for j in 1:40
      if(covermap[i, j]>0)
       printcovermap[i, j]=covermap[i, j]
      end
  end
end
#solution of the problem
function find(i, mat, usd, m, n, match) #function of hugarian algorithmn
     #println("i:", i, "j:", j)
    for j in 1:n
         # println("i:", i, "j:", j)
        if (mat[Int(i), j]==1&&usd[j]!=1)
            usd[i]=1
            if(match[j]==0||(find(Int(match[j]), mat, usd, m, n, match)))
              match[j]=i
              return true
            end
         end
    end
  return false
end
cmt=temp[11]
vcar=0.1
vpcar=0.16
#there are t=100 limit
t=0
#main loop to find min(t)
while (t < 100)
 t+=1
  rc=0
 rc=t*vcar+0.3
  rp=t*vpcar
 setB=[]
  println("Iterating t=", t)
 # println("rc", rc)
 for i in 1:40
     if (rawmap[Int(cmt), i] < rc)</pre>
       push! (setB, i)
    end
 end
  maA=zeros(10, length(setB))
  start=zeros(2)
```

```
for i in 1:10
        for j in 1:length(setB)
            if(rawmap[Int(ps[i]), setB[j]]<rp)</pre>
                 \#println("d():",rawmap[Int(ps[i]),setB[j]],"<",rp)
                start[1]=i
                 start[2]=j
           end
        end
    end
    println("At this time criminal's range: ", setB," starting from node: ", temp[11
])
    println("The roads that patrol center can block:", maA)
    #println(start)
    \#if(length(setB) == 0/start[1] == 0)
        #continue
    #end
    f1g=0
     match=zeros(length(setB))
    for i in 1:10
       used=zeros(length(setB))
       if( find(i, maA, used, 10, length(setB), match) == true)
            perfect =1
            for i in 1:length(setB)
                 if (match[i]==0)
                     perfect=0
                 end
            end
             if (perfect==1)
              println("Match found", match, " at time t=", t)
              f1g=1
                break
            end
         println("current Match", match)
    end
    if(f1g==1)
        break
    end
end
```

```
The patrol centers are on the node: [28.0, 38.0, 31.0, 32.0, 33.0, 18.0,
37.0, 9.0, 27.0, 14.0]
Iterating t=1
At this time criminal's range: Any[39] starting from node :39
The roads that patrol center can block: [0.0; 0.0; 0.0; 0.0; 0.0; 0.0; 0.
0; 0.0; 0.0; 0.0]
current Match[0.0]
Iterating t=2
At this time criminal's range: Any[39] starting from node :39
The roads that patrol center can block: [0.0; 0.0; 0.0; 0.0; 0.0; 0.0; 0.
0: 0.0: 0.0: 0.0
current Match[0.0]
Iterating t=3
At this time criminal's range: Any[36, 39] starting from node :39
The roads that patrol center can block: [0.0 0.0; 0.0 0.0; 0.0 0.0; 0.0
current Match[0.0, 0.0]
Iterating t=4
At this time criminal's range: Any[36, 39] starting from node: 39
The roads that patrol center can block: [0.0 0.0; 0.0 0.0; 0.0 0.0; 0.0
current Match[0.0, 0.0]
Iterating t=5
At this time criminal's range: Any[36, 39] starting from node :39
The roads that patrol center can block: [0.0 0.0; 0.0 0.0; 0.0 0.0; 0.0
```

file:///C:/Users/10096/project1.html

```
current Match[0.0, 0.0]
Iterating t=6
At this time criminal's range: Any[36, 39] starting from node :39
The roads that patrol center can block: [0.0 0.0; 0.0 0.0; 0.0 0.0; 0.0
current Match[0.0, 0.0]
Iterating t=7
At this time criminal's range: Any[36, 39] starting from node: 39
The roads that patrol center can block: [0.0 0.0; 0.0 0.0; 0.0 0.0; 0.0
current Match[0.0, 0.0]
Iterating t=8
At this time criminal's range: Any[36, 39] starting from node: 39
The roads that patrol center can block: [0.0 0.0; 0.0 0.0; 0.0 0.0; 0.0
current Match[0.0, 0.0]
Iterating t=9
At this time criminal's range: Any[36, 39] starting from node :39
The roads that patrol center can block: [0.0 0.0; 0.0 0.0; 0.0 0.0; 0.0
current Match[0.0, 0.0]
current Match[0.0, 0.0]
current Match[0.0, 0.0]
current Match[0.0, 0.0]
```

```
current Match[0.0, 0.0]
Iterating t=10
At this time criminal's range: Any[1, 36, 39] starting from node: 39
The roads that patrol center can block: [0.0 0.0 0.0; 0.0 0.0; 0.0 0.0; 0.0 0.
0.0\ 0.0\ 0.0;\ 0.0\ 0.0\ 0.0]
current Match[0.0, 0.0, 0.0]
Iterating t=11
At this time criminal's range: Any[1, 36, 39] starting from node :39
The roads that patrol center can block: [0.0 0.0 0.0; 0.0 0.0; 0.0 0.0; 0.0 0.
0.0\ 0.0\ 0.0;\ 0.0\ 0.0\ 0.0]
current Match[0.0, 0.0, 0.0]
Iterating t=12
At this time criminal's range: Any[1, 36, 39] starting from node :39
The roads that patrol center can block: [0.0 0.0 0.0; 0.0 0.0; 0.0 0.0; 0.0 0.
0.0 0.0 0.0; 1.0 0.0 0.0]
current Match[0.0, 0.0, 0.0]
current Match[10.0, 0.0, 0.0]
Iterating t=13
At this time criminal's range: Any[1, 36, 39] starting from node: 39
The roads that patrol center can block: [0.0 0.0 0.0; 0.0 0.0; 0.0 0.0; 0.0 0.
0.0\ 1.0\ 0.0;\ 1.0\ 0.0\ 0.0]
current Match[0.0, 0.0, 0.0]
```

```
current Match[0.0, 0.0, 0.0]
current Match[0.0, 0.0, 0.0]
current Match[0.0, 0.0, 0.0]
current Match[0.0, 9.0, 0.0]
current Match[10.0, 9.0, 0.0]
Iterating t=14
At this time criminal's range: Any[1, 30, 36, 39] starting from node :39
The roads that patrol center can block: [0.0 0.0 0.0 0.0; 0.0 0.0 0.0 0.
current Match[0.0, 0.0, 0.0, 0.0]
current Match[0.0, 0.0, 0.0, 0.0]
current Match[0.0, 0.0, 3.0, 0.0]
current Match[10.0, 0.0, 3.0, 0.0]
Iterating t=15
At this time criminal's range: Any[1, 30, 36, 39] starting from node: 39
The roads that patrol center can block: [0.0 0.0 0.0 0.0; 0.0 0.0 0.0 0.
current Match[0.0, 0.0, 0.0, 0.0]
current Match[0.0, 0.0, 0.0, 0.0]
current Match[0.0, 0.0, 3.0, 0.0]
current Match[0.0, 0.0, 3.0, 0.0]
current Match[0.0, 0.0, 3.0, 0.0]
current Match[6.0, 0.0, 3.0, 0.0]
current Match[10.0, 6.0, 3.0, 0.0]
Iterating t=16
At this time criminal's range: Any[1, 29, 30, 36, 39] starting from node
The roads that patrol center can block: [0.0 0.0 0.0 0.0 0.0; 0.0 0.0 0.0
1.0 0.0 1.0 0.0 0.0; 0.0 0.0 0.0 0.0 0.0; 0.0 1.0 0.0 0.0 0.0; 0.0 0.0
0.01.01.0; 1.00.01.00.00.0
current Match[0.0, 0.0, 0.0, 0.0, 0.0]
current Match[0.0, 0.0, 0.0, 0.0, 0.0]
current Match[0.0, 0.0, 0.0, 3.0, 0.0]
current Match[0.0, 0.0, 0.0, 3.0, 0.0]
current Match[0.0, 0.0, 5.0, 3.0, 0.0]
current Match[6.0, 0.0, 5.0, 3.0, 0.0]
current Match[6.0, 0.0, 5.0, 3.0, 0.0]
current Match[6.0, 8.0, 5.0, 3.0, 0.0]
Match found[6.0, 8.0, 5.0, 3.0, 9.0] at time t=16
```

Result and Interpretation

It seems like if the criminal is last seen in this random traffic road, police car needs to be 1.5 times faster than the car of criminal in order to catch him. Because the selection of patrol centers is random and the result won't be too good.

Obviously not every traffic road can be intercepted, because although the siren did work you can't drive 110 mph in the city. Other fun things to do would be pack the main of the program above and pass combinations of police station in to find a better maximum possible intercep node of the whole city.

Conclusion

To summarize this project, the original three questions are almost solved, although there are some limitations to fit in the realistic situation. (e.g. Recall the Integer Program in Problem1) One of my findings is that for a city (given the coordinates of every traffic road and the coordinates of every patrol center), it is possible to use warshall alogrithm to find out the distance between any two patrol centers and traffic roads, as well as which traffic roads will be covered in each patrol center. Finally we can get a optimized solution (valued by the evenness of police cases accepted by all centers) on how to distribute the police force to each traffic road. Another finding is that for problem3, we find the ratio of the speed of the police and the speed of the crimnal such that the criminal can be taken down needs to be 1.5 times.

There are many aspects we can investigate in the future, we can implement another obective function and plot the tradeoff curve between the balance of the patrol distribution and the efficient of patrol distribution.

It would be also interesting to discuss if there is a correlation betweem the second objective in the first problem and their performance in second problem. Though first we would need to figure out how to evaluate their performance in problem 2. But this can be evaluated by suming all the objective by calculating the linear combinations (mutiply by weight).

In []:			