

```
clear
cd '~/Documents/Spatial/Project/'
```

```
data = readtable('Berkeley.csv');
```

Part 1 - Preliminary Analysis (Some parts are in Python)

```
%Showing the kinds of crimes in the dataset that were given the 'violent' tag
data.OFFENSE(data.VIOLENT_FLG==1)
```

```
ans =
' GUN/WEAPON '
' GUN/WEAPON '
' ARSON '
' GUN/WEAPON '
' GUN/WEAPON '
' GUN/WEAPON '
' GUN/WEAPON '
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' GUN/WEAPON '
' ARSON '
' ARSON '
' HOMICIDE '
' GUN/WEAPON '
' GUN/WEAPON '
' GUN/WEAPON '
' HOMICIDE '
' GUN/WEAPON '
' GUN/WEAPON '
' GUN/WEAPON '
' ARSON '
' ARSON '
' GUN/WEAPON '
' GUN/WEAPON '
' GUN/WEAPON '
' GUN/WEAPON '
' ARSON '
' GUN/WEAPON '
```

```
%Reading in the shapefile of Berkeley
```

```
[S, A] = shaperead('Census Block Polygons 2010/geo_export_9702b0bc-dfe0-42aa-bab9-e9ca83abd50f')
```

```
%Plotting the Berkeley shapefile
```

```
clf;  
figure  
hold on  
for x=1:size(S)  
    plot(S(x).Lon, S(x).Lat, 'Color',[0.8,0.8,0.8]);  
end  
axis([-122.35 -122.23 37.84 37.91])  
saveas(gcf, 'Berkeley.png')
```

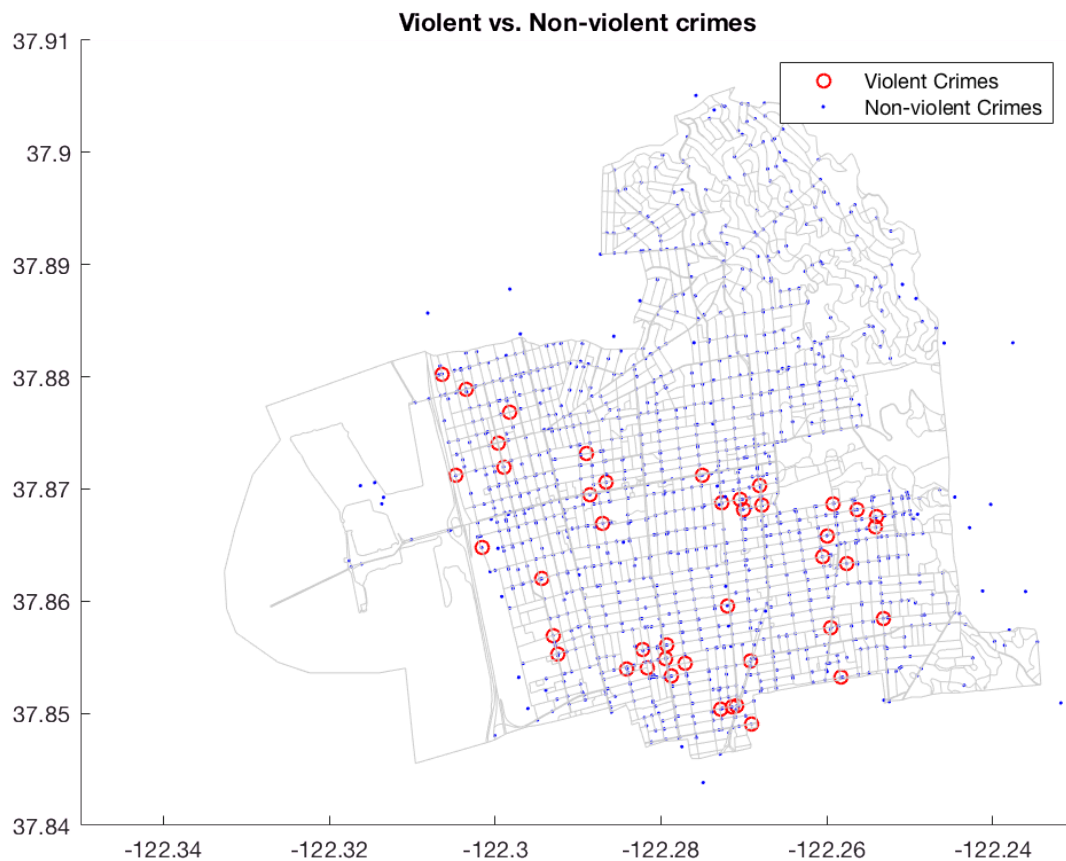


```
%Superimposing the Berkeley crimes onto the Berkeley shapefile
```

```
plot(data.Longitude, data.Latitude, '.', 'Color',[0.5,0.5,0.5], 'MarkerSize',10)  
saveas(gcf, 'Berkeley crimes.png')
```

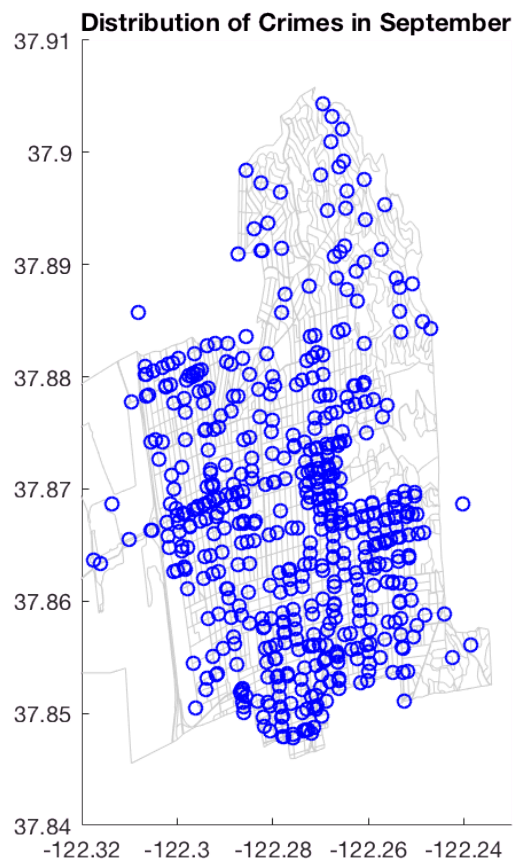
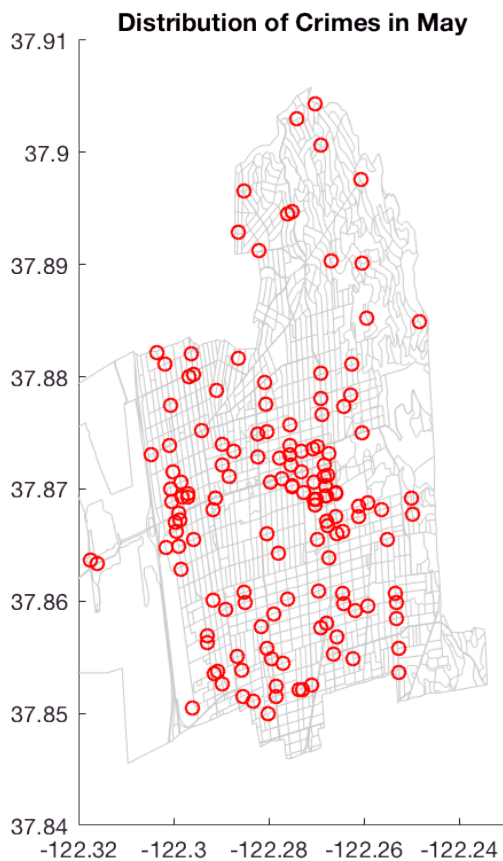


```
%Plotting the violent crimes versus the non-violent crimes
clf;
figure
hold on
plot(data.Longitude(data.VIOLENT_FLG == 1), data.Latitude(data.VIOLENT_FLG == 1), 'or',data.Latitude(data.VIOLENT_FLG == 0))
title('Violent vs. Non-violent crimes')
legend('Violent Crimes','Non-violent Crimes')
saveas(gcf,'Violent crimes.png')
for x=1:size(S)
    plot(S(x).Lon, S(x).Lat, 'Color',[0.8,0.8,0.8]);
end
axis([-122.35 -122.23 37.84 37.91])
```



%Comparing the distributions of Berkeley crimes between two arbitrary months (May and September)

```
clf;
figure
subplot(1,2,1)
hold on
for x=1:size(S)
    plot(S(x).Lon, S(x).Lat, 'Color',[0.8,0.8,0.8]);
end
plot(data.Longitude(data.EVENT_MONTH == 5), data.Latitude(data.EVENT_MONTH == 5), 'or')
axis([-122.32 -122.23 37.84 37.91])
title('Distribution of Crimes in May')
subplot(1,2,2)
hold on
for x=1:size(S)
    plot(S(x).Lon, S(x).Lat, 'Color',[0.8,0.8,0.8]);
end
plot(data.Longitude(data.EVENT_MONTH == 9), data.Latitude(data.EVENT_MONTH == 9), 'ob')
axis([-122.32 -122.23 37.84 37.91])
title('Distribution of Crimes in September')
```



```
length(data.Latitude(data.EVENT_MONTH == 5))
```

```
sprintf('Number of crimes in May: %d', length(data.Latitude(data.EVENT_MONTH == 5)))
```

```
ans = Number of crimes in May: 167
```

```
sprintf('Number of violent crimes in May: %d', length(data.Latitude(data.EVENT_MONTH == 5 & da
```

```
ans = Number of violent crimes in May: 2
```

```
sprintf('Number of crimes in September: %d', length(data.Latitude(data.EVENT_MONTH == 9)))
```

```
ans = Number of crimes in September: 907
```

```
sprintf('Number of violent crimes in September: %d', length(data.Latitude(data.EVENT_MONTH ==
```

```
ans = Number of violent crimes in September: 7
```

This shows that September had a larger number of crimes in general as well as violent crimes than May. This analysis can be replicated for different months.

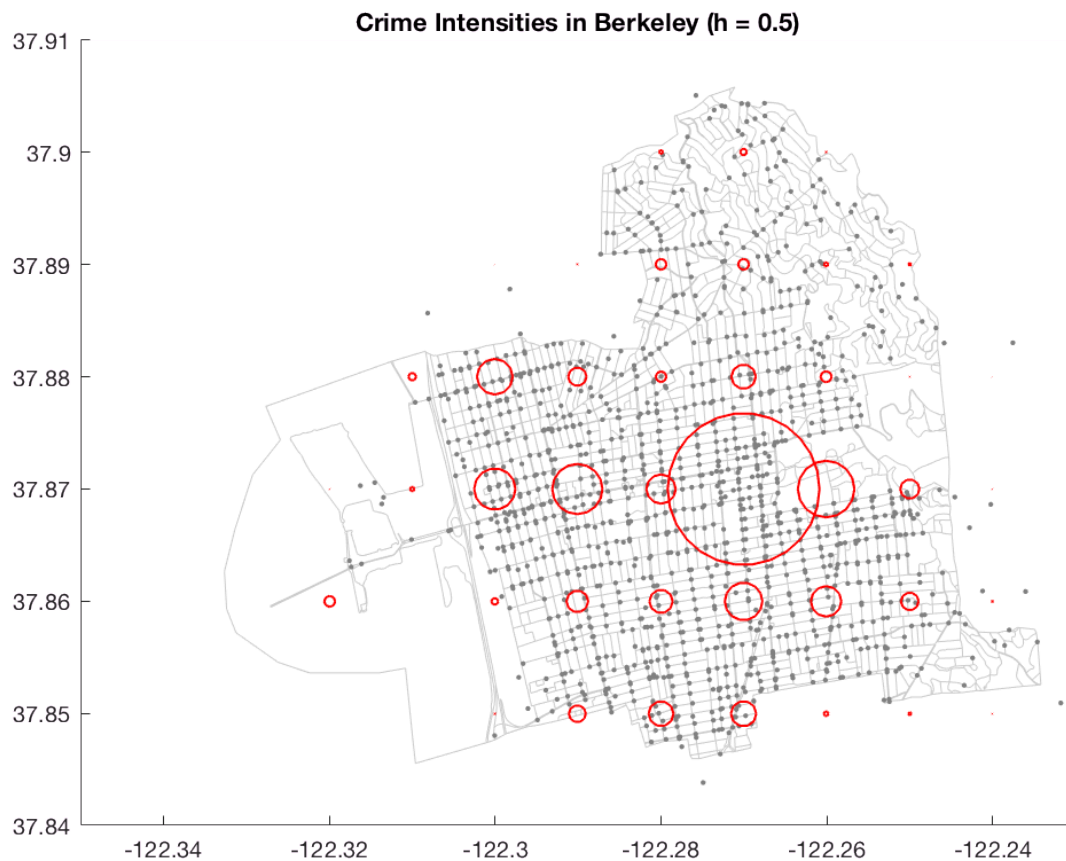
Part 2 - Main Analysis

```
%Creating Intensity Maps based on number of crimes in Berkeley
h = 0.5; %Radius of observation
clf;
figure
hold on

for x=1:size(S)
    plot(S(x).Lon, S(x).Lat, 'Color',[0.8,0.8,0.8]);
end
axis([-122.35 -122.23 37.84 37.91])

%Superimposing the Berkeley crimes onto the Berkeley shapefile
plot(data.Longitude, data.Latitude, '.', 'Color',[0.5,0.5,0.5], 'MarkerSize',10)

for x = -122.32:0.01:-122.23
    for y = 37.84:0.01:37.91
        dist_to_event_j = 111.1*sqrt((x - data.Longitude).^2 + (y - data.Latitude).^2);
        lambda = length(dist_to_event_j(dist_to_event_j < h));
        if lambda > 0
            plot(x,y, 'or', 'MarkerSize', 0.1*lambda);
        end
    end
end
title('Crime Intensities in Berkeley (h = 0.5)')
saveas(gcf, 'Crime Intensities.png')
```



```
%Analysing patterns in distribution of berkeley crimes around the three BART Stations
berkeley_barts = readtable('Bart Stations.csv');
berkeley_barts
```

```
berkeley_barts =
      Name          Lat      Lon
  -----
  'North Berkeley'  37.874  -122.28
  'Downtown Berkeley' 37.87   -122.27
  'Ashby'          37.853   -122.27
```

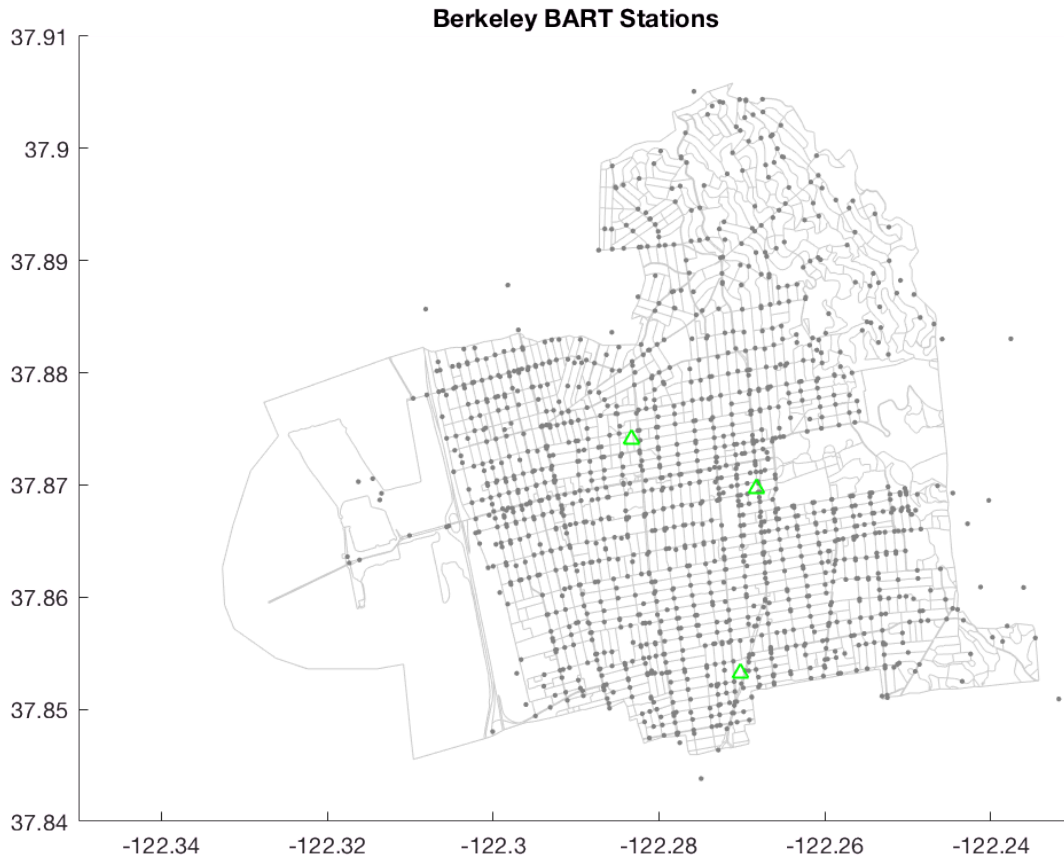
```
clf;
figure
hold on

%Plotting the Berkeley shape file
for x=1:size(S)
    plot(S(x).Lon, S(x).Lat, 'Color', [0.8,0.8,0.8]);
end
axis([-122.35 -122.23 37.84 37.91])

%Superimposing the Berkeley crimes onto the Berkeley shapefile
plot(data.Longitude, data.Latitude, '.', 'Color', [0.5,0.5,0.5], 'MarkerSize', 10)

plot(berkeley_barts.Lon, berkeley_barts.Lat, '^g')
```

```
title('Berkeley BART Stations')
```



```
%Calculating crime intensity around BART Stations by creating buffers
for i=1:3
    [buffer_lat, buffer_lon] = bufferm(berkeley_barts.Lat(i), berkeley_barts.Lon(i), 0.005);
    island_buffer_struct = struct('lat', buffer_lat, 'lon', buffer_lon);
    plot(island_buffer_struct.lon, island_buffer_struct.lat)

    data_lon = data.Longitude;
    data_lat = data.Latitude;
    inside = inpolygon(data_lon,data_lat,buffer_lon,buffer_lat);
    data_lon(inside == false) = NaN;
    data_lat(inside == false) = NaN;

    length(inside(inside==1))
    crime_intensity = length(inside(inside==1))/(pi*0.3^2)

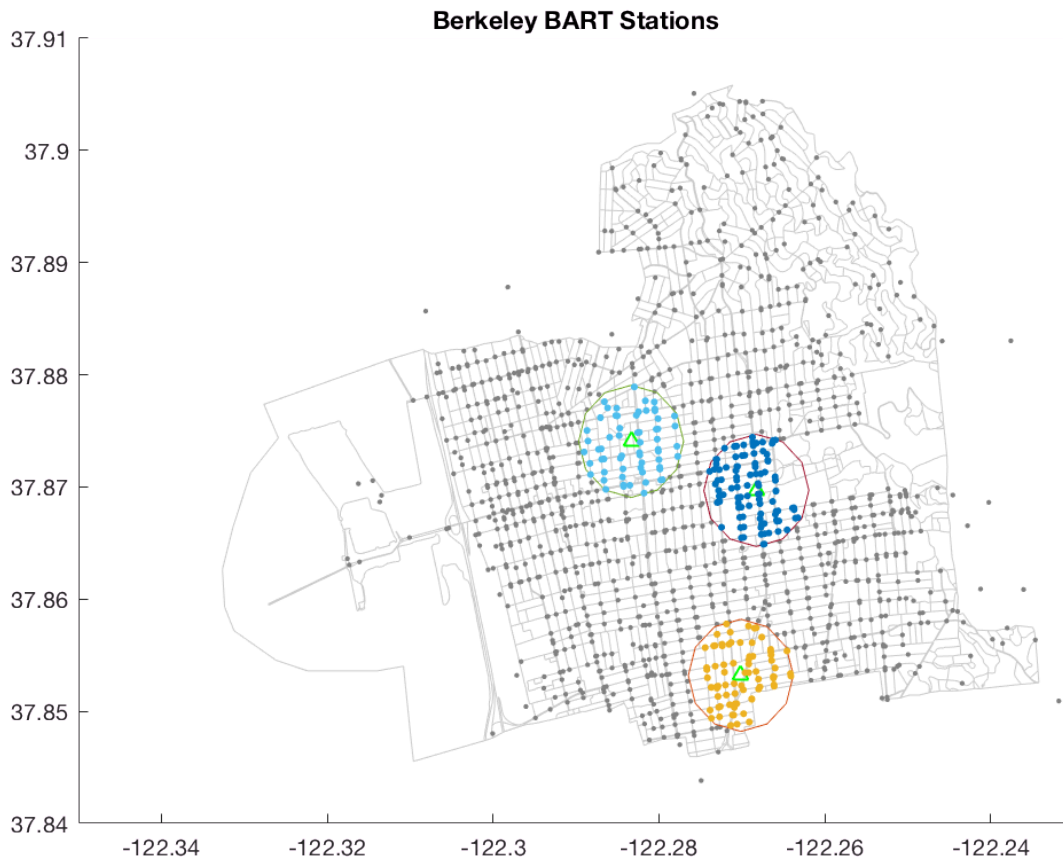
    data_lon(isnan(data_lon)) = [];
    data_lat(isnan(data_lat)) = [];
    simulated_events = struct('lon',data_lon,'lat',data_lat);
    plot(simulated_events.lon, simulated_events.lat, '.', 'MarkerSize',15)
end
```

```
ans = 156
crime_intensity = 551.7371
ans = 766
crime_intensity = 2.7092e+03
```



```
ans = 250  
crime_intensity = 884.1941
```

```
saveas(gcf, 'bart with buffers.png')
```



Part 3 - Recommendation

```
%Calculating the optimal location of a police station in Berkeley using distance optimality  
min_distance = inf;
```

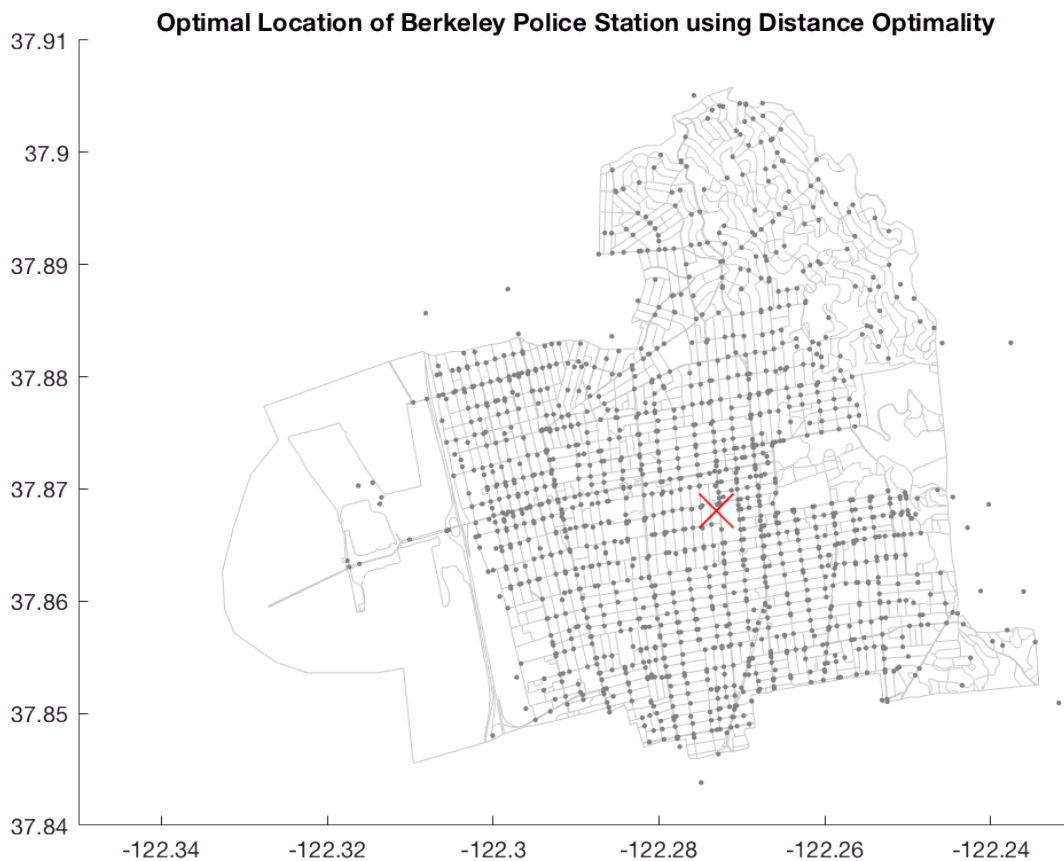
```
for x = -122.32:0.001:-122.23  
    for y = 37.84:0.001:37.91  
        dist_to_event_j = 111.1*sqrt((x - data.Longitude).^2 + (y - data.Latitude).^2);  
        total_dist = sum(dist_to_event_j);  
        %Getting the grid point with the minimum average distance to all crimes  
        if total_dist < min_distance  
            min_distance = total_dist;  
            best_lon = x;  
            best_lat = y;  
        end  
    end  
end  
best_lon
```

```
best_lon = -122.2730
```

```
best_lat
```

```
best_lat = 37.8680
```

```
clf;  
figure  
hold on  
  
%Plotting the Berkeley shape file  
for x=1:size(S)  
    plot(S(x).Lon, S(x).Lat, 'Color',[0.8,0.8,0.8]);  
end  
axis([-122.35 -122.23 37.84 37.91])  
  
%Superimposing the Berkeley crimes onto the Berkeley shapefile  
plot(data.Longitude, data.Latitude, '.', 'Color',[0.5,0.5,0.5], 'MarkerSize',10)  
  
plot(best_lon,best_lat,'xr','MarkerSize',20)  
title('Optimal Location of Berkeley Police Station using Distance Optimality')  
saveas(gcf, 'Optimal Police Station.png')
```

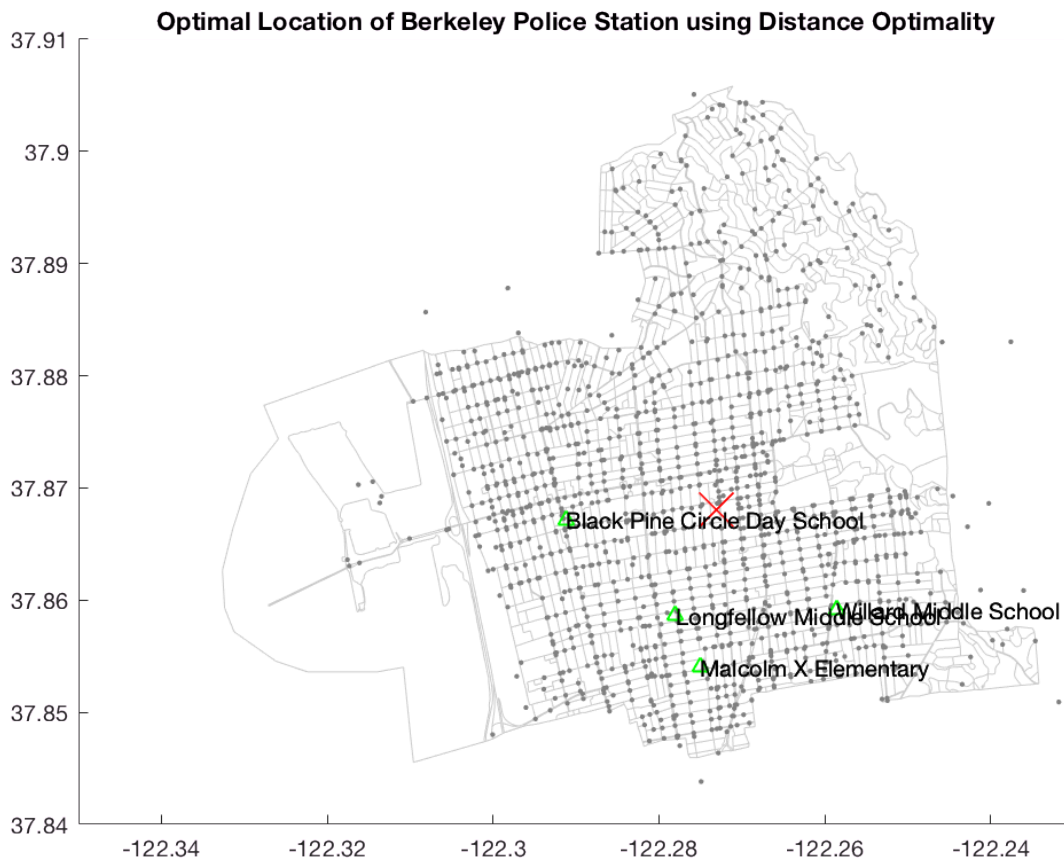


```
%Accounting for the safety of school children while calculating the optimal Berkeley Police St  
berkeley_schools = readtable('Berkeley_Schools - Sheet1 copy.csv'); %Manually chosen the four  
berkeley_schools
```

```
berkeley_schools =
```

Name	Lat	Lng
'Willard Middle School'	37.859	-122.26
'Black Pine Circle Day School'	37.867	-122.29
'Malcolm X Elementary'	37.854	-122.28
'Longfellow Middle School'	37.859	-122.28

```
plot(berkeley_schools.Lng,berkeley_schools.Lat,'^g')
text(berkeley_schools.Lng,berkeley_schools.Lat,berkeley_schools.Name)
```



```
clf;
figure
hold on

%Plotting the Berkeley shape file
for x=1:size(S)
    plot(S(x).Lon, S(x).Lat, 'Color',[0.8,0.8,0.8]);
end
axis([-122.35 -122.23 37.84 37.91])

%Superimposing the Berkeley crimes onto the Berkeley shapefile
plot(data.Longitude, data.Latitude, '.', 'Color',[0.5,0.5,0.5], 'MarkerSize',10)
```

```

plot(best_lon,best_lat,'xr','MarkerSize',20)

school_crimes_lon = [];
school_crimes_lat = [];
for i=1:4
    [buffer_lat, buffer_lon] = bufferm(berkeley_schools.Lat(i), berkeley_schools.Lng(i), 0.005);
    island_buffer_struct = struct('lat', buffer_lat, 'lon', buffer_lon);
    plot(island_buffer_struct.lon, island_buffer_struct.lat)

    data_lon = data.Longitude;
    data_lat = data.Latitude;
    inside = inpolygon(data_lon,data_lat,buffer_lon,buffer_lat);
    data_lon(inside == false) = NaN;
    data_lat(inside == false) = NaN;

    length(inside(inside==1))
    crime_intensity = length(inside(inside==1))/(pi*0.3^2)

    data_lon(isnan(data_lon)) = [];
    data_lat(isnan(data_lat)) = [];
    simulated_events = struct('lon',data_lon,'lat',data_lat);
    school_crimes_lon=[school_crimes_lon;simulated_events.lon];
    school_crimes_lat=[school_crimes_lat;simulated_events.lat];
    plot(simulated_events.lon, simulated_events.lat, '.', 'MarkerSize',15)
end

```

```

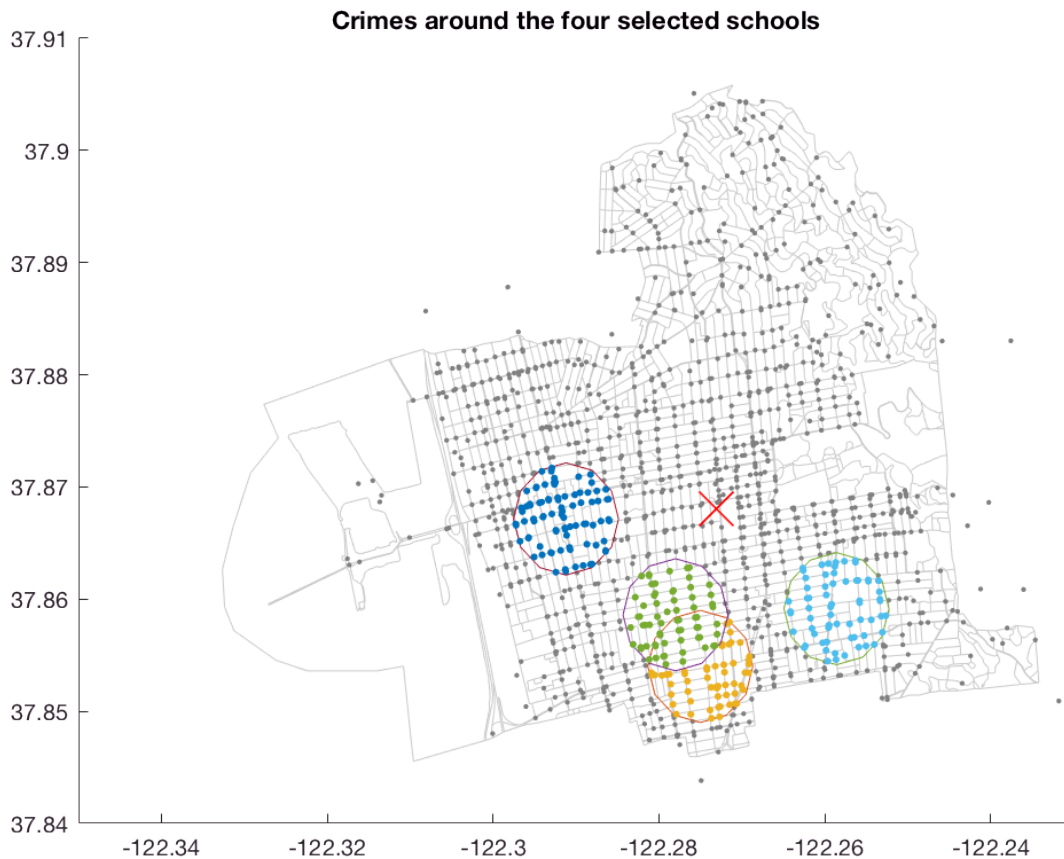
ans = 226
crime_intensity = 799.3115
ans = 285
crime_intensity = 1.0080e+03
ans = 255
crime_intensity = 901.8780
ans = 177
crime_intensity = 626.0094

```

```

title('Crimes around the four selected schools')
saveas(gcf,'schools with buffers.png')

```



`%Prioritizing crimes around the four schools by assigning them larger weight than other crimes`

```

min_distance = inf;
min_school_distance = inf;
for x = -122.32:0.001:-122.23
    for y = 37.84:0.001:37.91
        dist_to_event_j = 111.1*sqrt((x - data.Longitude).^2 + (y - data.Latitude).^2);
        dist_to_school_crime = 111.1*sqrt((x - school_crimes_lon).^2 + (y - school_crimes_lat).^2);
        %Appending the distance to school crimes array 5 times more than the distance to other crimes
        dist_to_event_j = vertcat(dist_to_event_j,dist_to_school_crime,dist_to_school_crime,dist_to_school_crime,dist_to_school_crime);

        total_dist = sum(dist_to_event_j);

        if total_dist < min_distance
            min_distance = total_dist;
            best_lon = x;
            best_lat = y;
        end
    end
end

best_lat

```

`best_lat = 37.8620`

```
best_lon
```

```
best_lon = -122.2760
```

```
plot(best_lon,best_lat, 'xb','MarkerSize',30)  
title('Optimal Location of Berkeley Police Department after prioritizing school crimes')  
saveas(gcf,'final.png')
```

