

Point Pattern Analysis: Basics

SERGIO REY

GPH 483/598

Geographic Information Analysis

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Spring 2010

1 Point Pattern Analysis Objectives and Examples

- Objectives
- Definitions
- Examples and Terminology

2 Centrography

- Central Tendency
- Dispersion and Orientation
- Geometry

Point Pattern Analysis Objectives

Goals

- Pattern detection
- Assessing the presence of *clustering*
- Identification of individual *clusters*

General Approaches

- Estimate intensity of the process
- Formulating an idealized model and investigating deviations from expectations
- Formulating a stochastic model and fitting it to the data

Point Pattern Analysis Definitions

Spatial Point Pattern

A set of events, irregularly distributed within a region A and presumed to have been generated by some form of stochastic mechanism.

Representation

$\{Y(A), A \subset \mathbb{R}\}$, where $Y(A)$ is the number of events occurring in area A .

Events, points, locations

Event an occurrence of interest

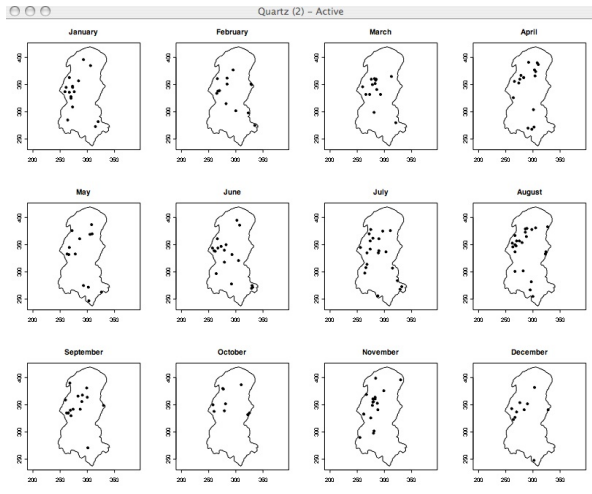
Point any location in study area

Event location a particular point where an event occurs

Region: A

- Most often planar (two-dimensional Euclidean space)
- One dimensional applications also possible
- Three-dimensional increasingly popular (space + time)
- Point processes on networks (non-planar)

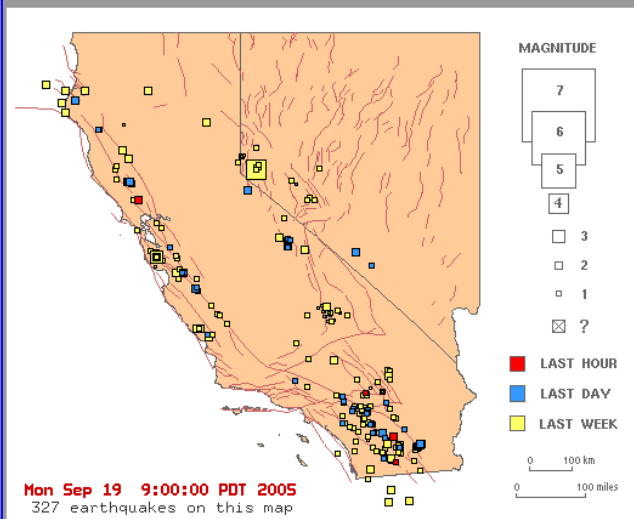
Space-Time Point Patterns



Space-Time Point Patterns

Index Map of Recent Earthquakes in California-Nevada

USGS • UCB • Caltech • UCSD • UNR



Point Patterns on Networks



**Figure 2: Retail stores assigned to the street network in Shibuya, Tokyo
(cells are indicated by different colors)**

Unmarked Point Patterns

- Only location is recorded
- Attribute is binary (presence, absence)

Marked Point Patterns

- Location is recorded
- Non-binary stochastic attribute
- e.g., sales at a retail store, dbh of tree

Mapped Point Patterns

- *All* events are recorded and mapped
- Complete enumeration of events
- Full information on the realization from the process

Sampled Point Patterns

- *Sample* of events are recorded and mapped
- Complete enumeration of events impossible or intractable
- Partial information on the realization from the process
- Presence/“absence” data (ecology, forestry)

Mapped Patterns

not a sample

Selection Bias

- events are mapped
- but non-events are not

Research Questions

Location Only

are points randomly located or patterned

Location and Value

- marked point pattern
- is combination of location and value random or patterned

Both Cases

What is the Underlying Process?

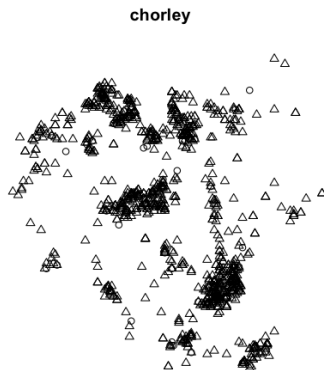
Classic Point Pattern Analysis

- points on an isotropic plane
- no effect of translation and rotation
- classic examples: tree seedlings, rocks, etc

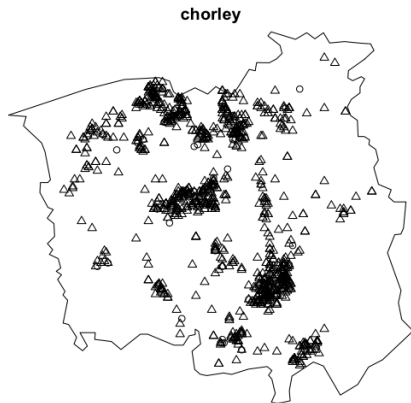
Distance

- no directional effects
- no translational effects
- straight line distance only

Events: Point Map



Points in Context



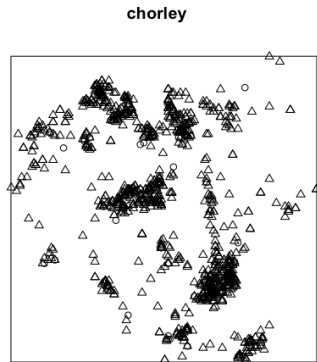
First Moment

- number of points N , area of study $|A|$
- intensity: $\lambda = N/|A|$
- area depends on bounds, often arbitrary

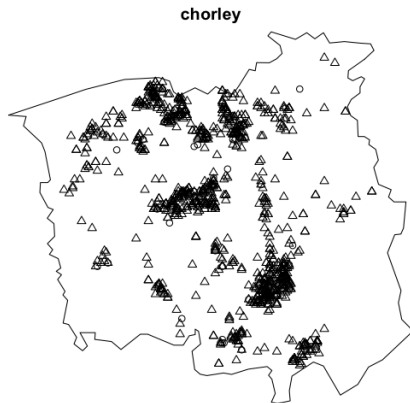
Artificial Boundaries

- bounding box (rectangle, square)
- other (city boundary)

Bounding Box



District Boundary



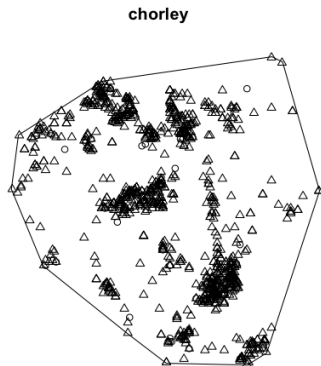
Tightest fit

various algorithms

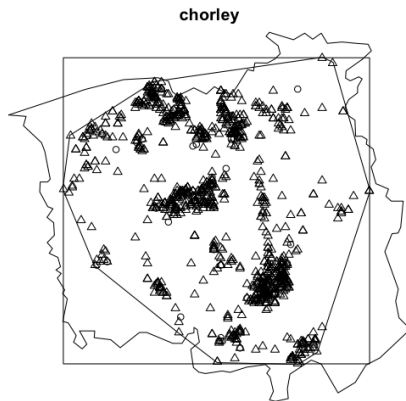
Rescaled Convex Hull (Ripley-Rasson)

- adjust to properly reflect spatial domain of point process
- use centroid of convex hull
- rescale by $1/[\sqrt{(1 - m/N)}]$
- m : number of vertices of convex hull

Convex Hull



Multiple Boundaries



Intensity Estimates

	Area km^2	Intensity $cases/km^2$
District Boundary	315.155	3.29
Bounding Box	310.951	3.33
Convex Hull	229.421	4.52

N=1036

Points on a Network

Realistic Location

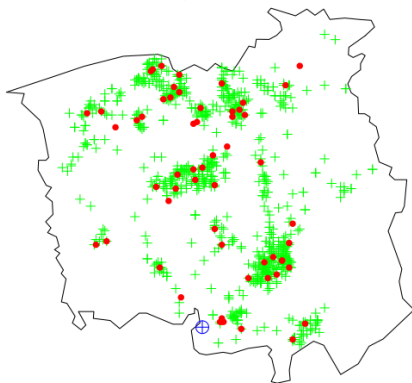
- addresses
- remove impossible locations (lakes)

Network Distance

- shortest path along network
- Manhattan block distance
- distance vs. travel time or cost

Case-Control Design: Lancashire Cancer

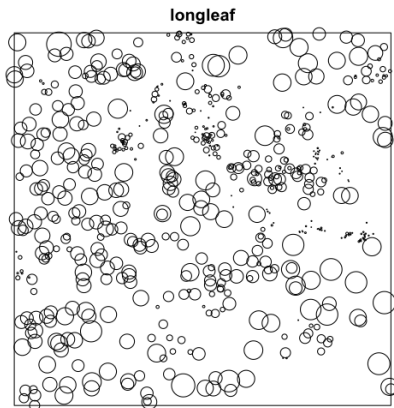
Chorley-Ribble Data



Both Location and Value

- Patterns in the Location
- Value Associated with Location

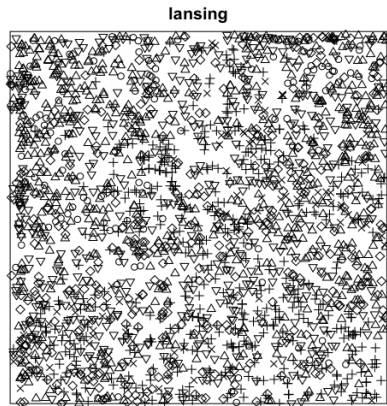
Marked Point Pattern: Longleaf Pine



Multiple Categories

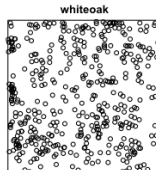
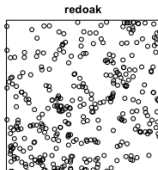
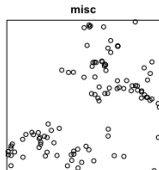
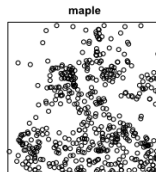
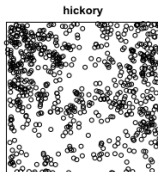
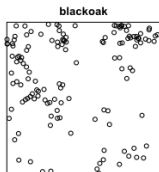
- Patterns in Single Category
- Association between Patterns in Other Categories
- Repulsion or Attraction

Multi-Type Pattern: Lansing Woods



Multi-Type Pattern: Lansing Woods

split(lansing)



Areal Aggregation

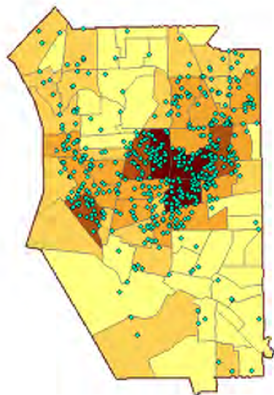
Event Counts

- points aggregated by areal unit
- spatially extensive variable

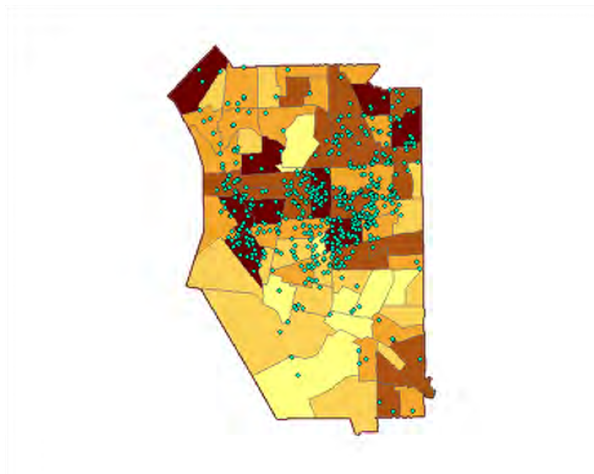
Rates

- events / population at risk
- non-homogeneous population at risk
- risk = probability of an event
- rate is an estimate of underlying risk

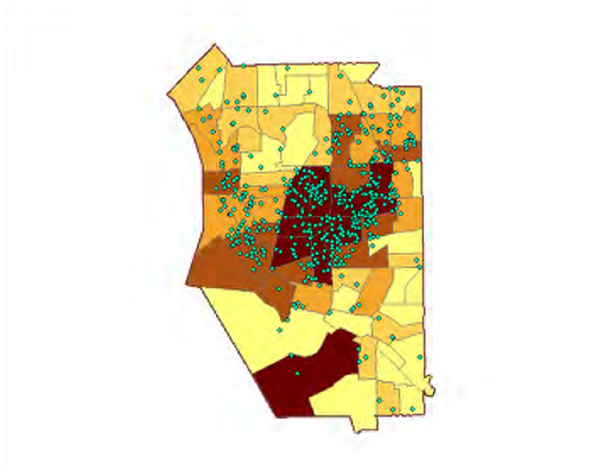
Homicide Counts by Census Tracts



Population Count by Census Tracts



Homicide Rate by Census Tracts



Central Tendency

Purpose

- Provide a “center point”
- Similar to first moment of a distribution
- “Representative point”

Measures

- Mean Center
- Weighted Mean Center
- Median Center
- Center of Minimum Distance

Example Data

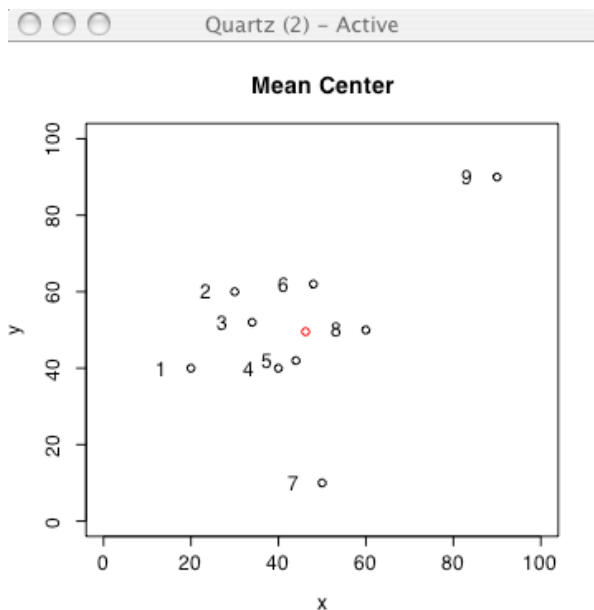
i	x_i	y_i	w_i
1	20	40	10
2	30	60	20
3	34	52	10
4	40	40	20
5	44	42	10
6	48	62	80
7	50	10	10
8	60	50	90
9	90	90	100

(x_m, y_m)

$$x_m = 1/n \sum_{i=1}^n x_i \quad (1)$$

$$y_m = 1/n \sum_{i=1}^n y_i \quad (2)$$

Mean Center



Weighted Mean center

(x_m, y_m)

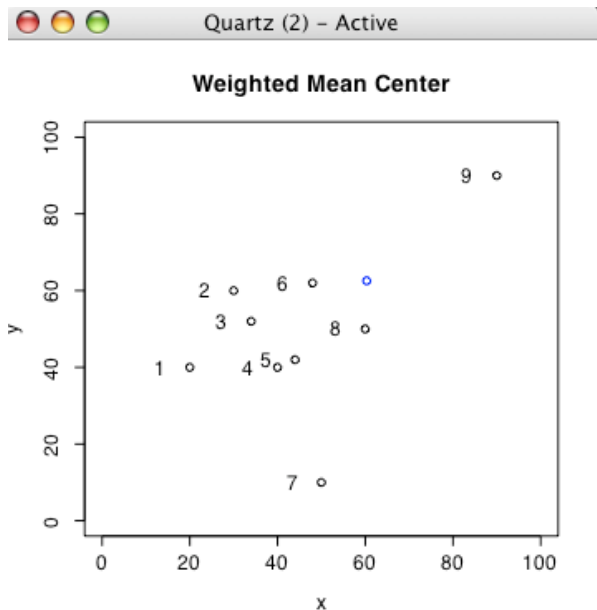
$$x_m = 1/n \sum_{i=1}^n x_i \frac{w_i}{\sum_{i=1}^n w_i} \quad (3)$$

$$y_m = 1/n \sum_{i=1}^n y_i \frac{w_i}{\sum_{i=1}^n w_i} \quad (4)$$

w_i weight

- Marked point patterns
- Continuous mark
- Not categorical mark

Weighted Mean Center



Definition(s)

English Statistics The intersection of two orthogonal axes, each which has an equal number of points on either side.

American The center of minimum travel.

Manhattan Median

$$\text{Min } f(x_m, y_m) = \sum_{i=1}^n |x_i - x_m| + |y_i - y_m| \quad (5)$$

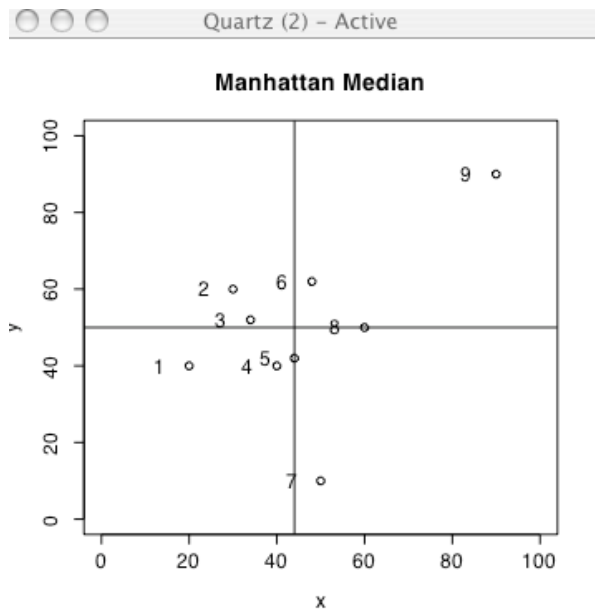
Advantages

- Can be found very quickly
- No calculations are typically required (other than intersection)

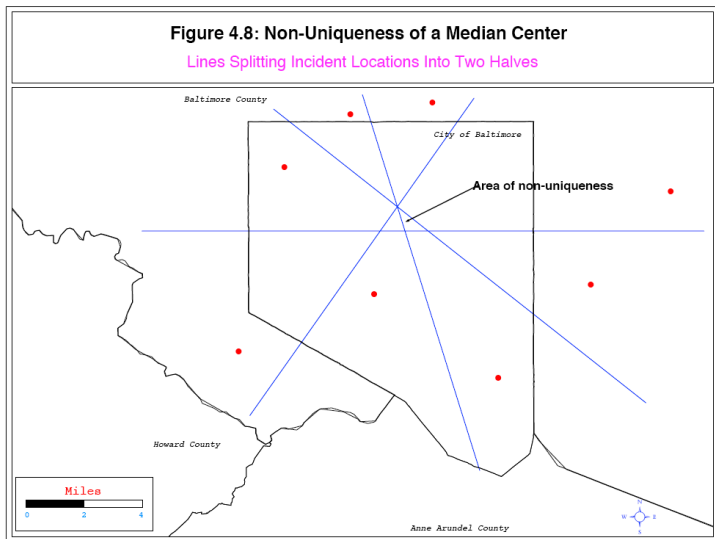
Disadvantage

- Never unique with even n
- Always unique with odd n
- Not unique under axis rotation

Manhattan Median



Non-Uniqueness



Center of Minimum Travel

Euclidean Median

The location from which the sum of the Euclidean distances to all points in a distribution is a minimum.

Euclidean Median

$$\text{Min } f(x_m, y_m) = \sum_{i=1}^n \sqrt{(x_i - x_m)^2 + (y_i - y_m)^2} \quad (6)$$

Weighted Euclidean Median

$$\text{Min } f(x_m, y_m) = \sum_{i=1}^n \frac{w_i}{\sum_{i=1}^n w_i} \sqrt{(x_i - x_m)^2 + (y_i - y_m)^2} \quad (7)$$

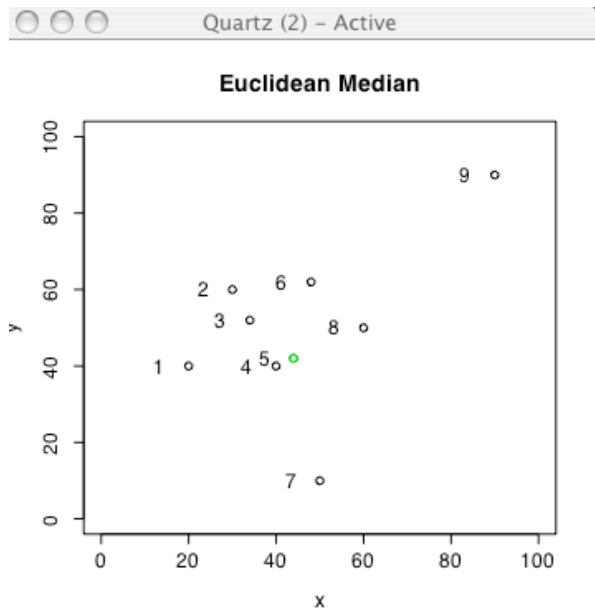
Weber Problem

Find the optimal location for a factory: one that minimizes transport costs between sources of raw materials and delivery to the market.

Solutions

- No closed form solution
- First iterative algorithm: Kuhn and Kuenne (1962)
- Important for more general location allocation problems

Euclidean Median



Measures

- Standard Distance
- Major/minor axes
- Standard Deviational Ellipse

Standard Distance

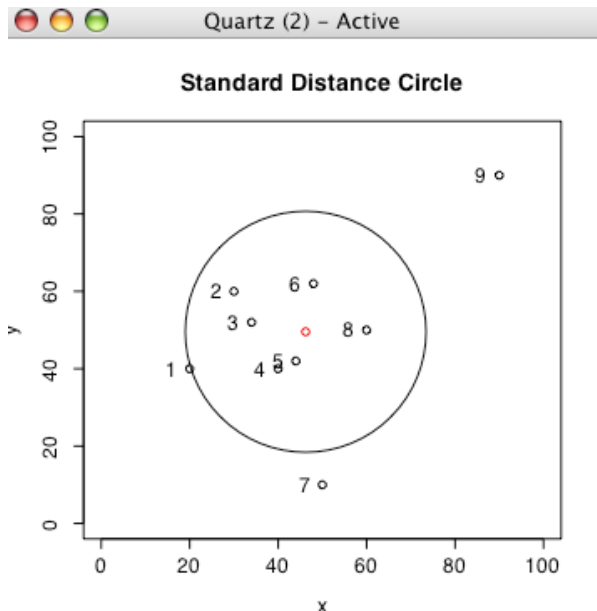
Euclidean Based

$$SD = \sqrt{\frac{\sum_{i=1}^n (x_i - x_m)^2}{n} + \frac{\sum_{i=1}^n (y_i - y_m)^2}{n}} \quad (8)$$

Uses

- Similar to standard deviation
- Combine with Mean Center for “outlier detection”
- Sensitive to extreme values

Standard Distance Circle



Standard Deviational Ellipse

Relative to Standard Distance

- Measures dispersion
- Sensitive to *shape* of distribution
- Measures dispersion in two dimensions

Components

- Angle of rotation
- Dispersion along major axis
- Dispersion along minor axis

Standard Deviational Ellipse

Major, minor axes

- Major axis defines the direction of maximum spread in the distribution
- Minor axis is orthogonal to major axis
- Minor axis defines the direction of minimum spread

Steps

- 1 Determine rotation angle of Y -axis
- 2 Calculate standard deviations for transposed axes
- 3 Determine length of axes
- 4 Determine area of the ellipse

Rotation Angle Θ

$$\Theta = \text{ARCTAN} \left\{ \left(\sum_i (x_i - \bar{x})^2 - \sum_i (y_i - \bar{y})^2 \right) + \left[\left(\sum_i (x_i - \bar{x})^2 - \sum_i (y_i - \bar{y})^2 \right)^2 + 4 \left(\sum_i (x_i - \bar{x})(y_i - \bar{y}) \right)^2 \right]^{1/2} \right\} / 2 \sum_i (x_i - \bar{x})(y_i - \bar{y})$$

Standard Deviations On Transposed Axes

S_x

$$S_x = \sqrt{2 \frac{(\sum_{i=1}^n (x_i - \bar{x}) \cos(\Theta) - \sum_{i=1}^n (y_i - \bar{y}) \sin(\Theta))^2}{n - 2}} \quad (9)$$

S_y

$$S_y = \sqrt{2 \frac{(\sum_{i=1}^n (x_i - \bar{x}) \sin(\Theta) - \sum_{i=1}^n (y_i - \bar{y}) \cos(\Theta))^2}{n - 2}} \quad (10)$$

Ellipse Axes

Lengths

$$L_x = 2S_x \quad (11)$$

$$L_y = 2S_y \quad (12)$$

Mid Point

Mean Center of Point Pattern (x_m, y_m)

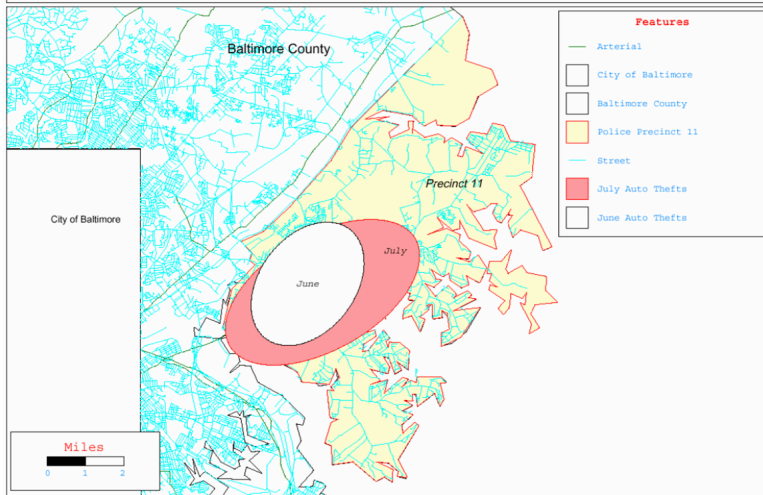
Area

$$A = \pi S_x S_y \quad (13)$$

Standard Deviation Ellipse

Figure 4.19: Auto Theft Change in Precinct 11

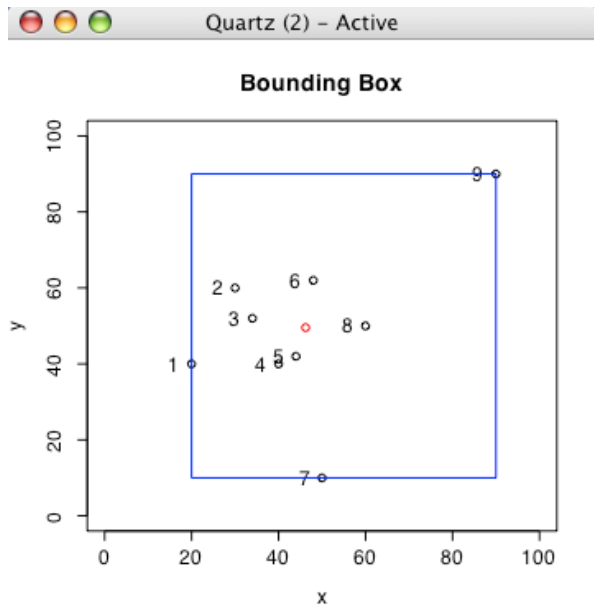
Ellipses of June and July 1996



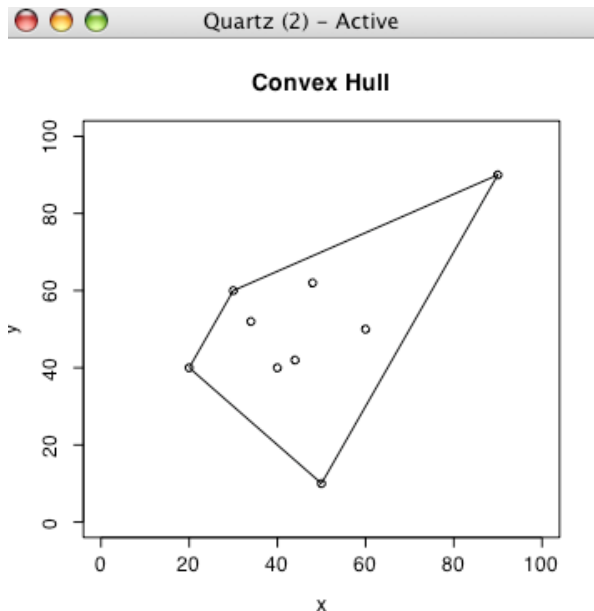
Geometry

- Bounding Box
- Convex Hulls

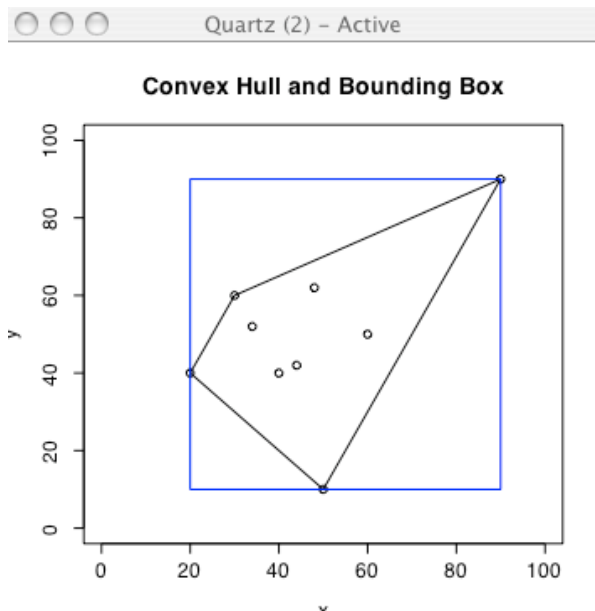
Bounding Box



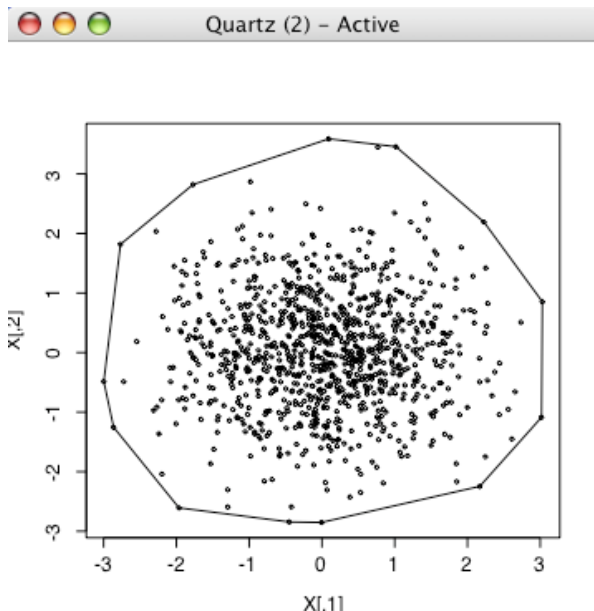
Convex Hull



Convex Hull and Bounding Box



Convex Hull (Large n)



Nested Convex Hulls

