

A New Mathematical Technique for Geographic Profiling

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Geographic Profiling

- **The Question:**

Given a series of linked crimes committed by the same offender, can we make predictions about the anchor point of the offender?

- The anchor point can be a place of residence, a place of work, or some other commonly visited location.

Implementation

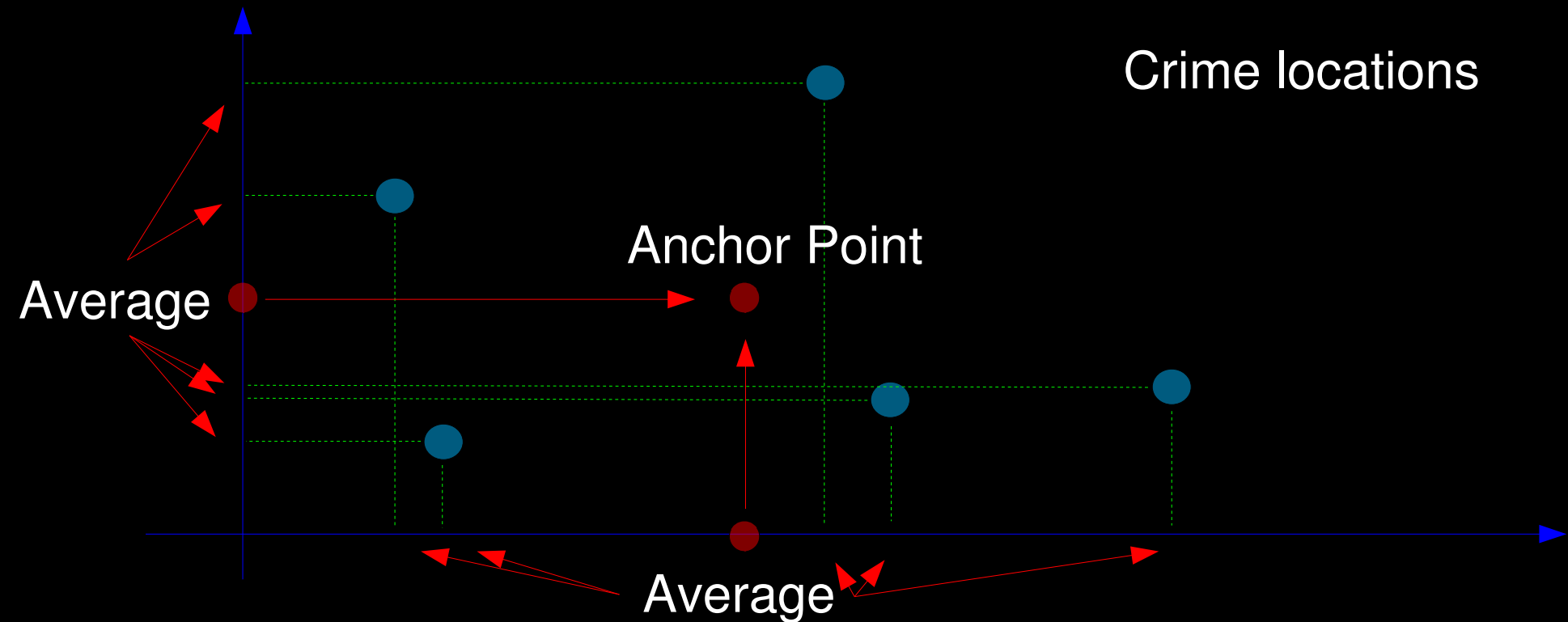
- **CrimeStat**
Ned Levine
- **Dragnet**
David Canter
- **Rigel**
Kim Rossmo
- **Predator**
Maurice Godwin

Current Techniques

- Spatial distribution strategies
- Probability distance strategies

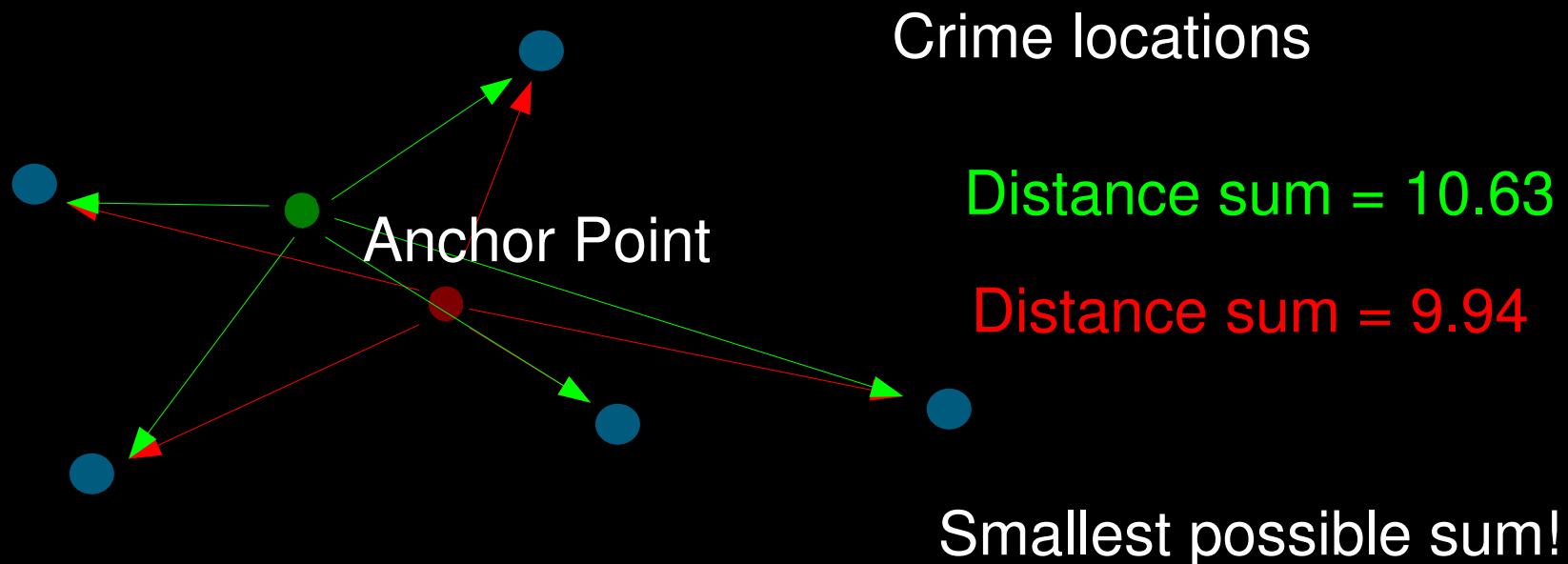
Spatial Distribution Strategies

- Centroid:
 - Use the average value of the crime coordinates



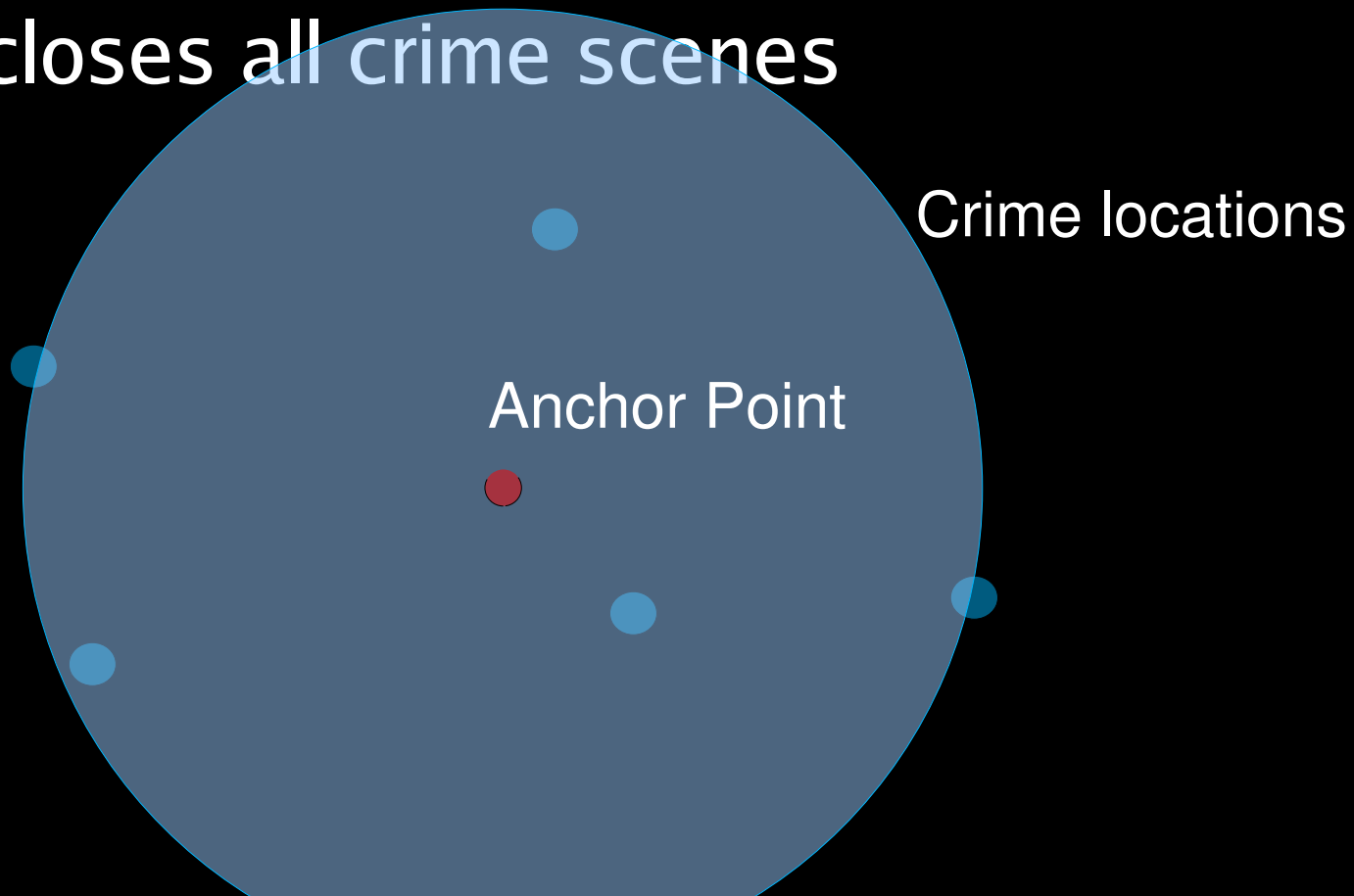
Spatial Distribution Strategies

- Center of minimum distance:
 - Find the point where the sum of the distance to all crime sites is minimized.



Spatial Distribution Strategies

- Circle Method:
 - Use the center of the smallest circle that encloses all crime scenes



Probability Distribution Strategies

- The anchor point is located in a region with a high “hit score”.
- The hit score $H(\mathbf{z})$ has the form

$$\begin{aligned} H(\mathbf{z}) &= \sum_{i=1}^n h(\mathbf{z}, \mathbf{x}_i) \\ &= h(\mathbf{z}, \mathbf{x}_1) + h(\mathbf{z}, \mathbf{x}_2) + \cdots + h(\mathbf{z}, \mathbf{x}_n) \end{aligned}$$

where \mathbf{x}_i are the crime locations and $h(\mathbf{z}, \mathbf{x})$ has a defined form.

Probability Distribution Strategies

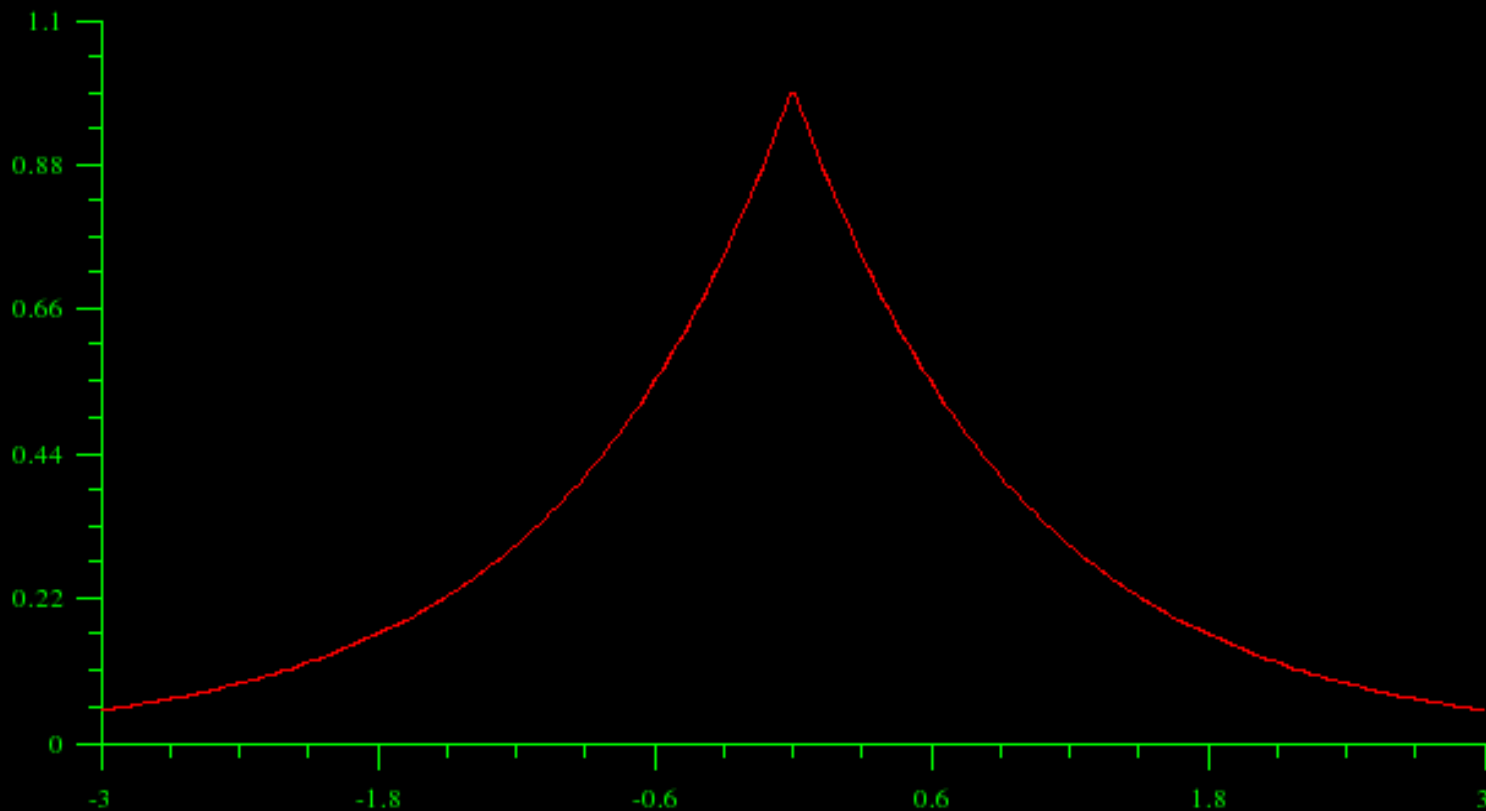
- Linear:

- $h(\mathbf{z}, \mathbf{x}) = a - b|\mathbf{x} - \mathbf{z}|$



Probability Distance Strategies

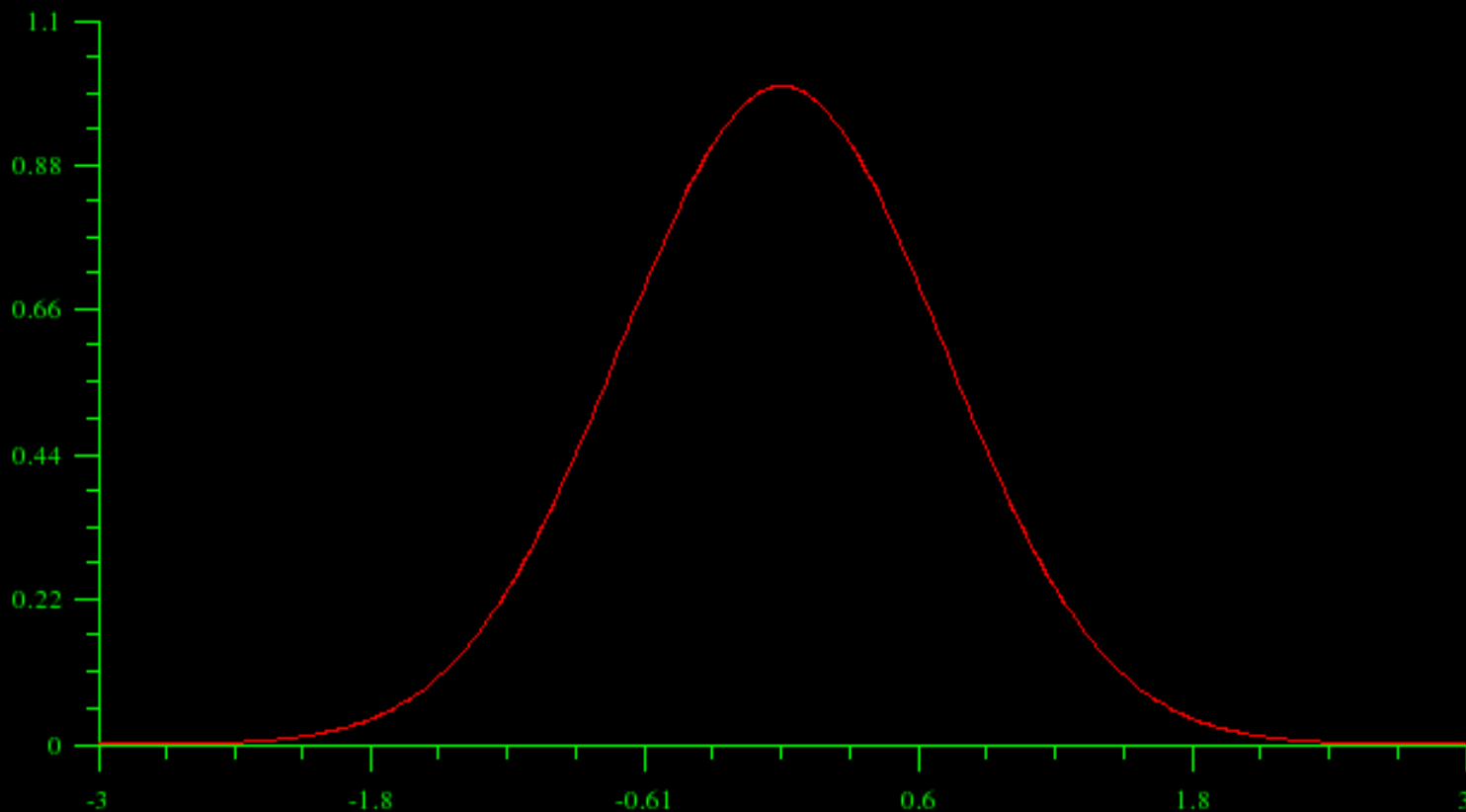
- Negative exponential
 - $h(\mathbf{z}, \mathbf{x}) = A \exp(-B|\mathbf{x} - \mathbf{z}|)$



Probability Distance Strategies

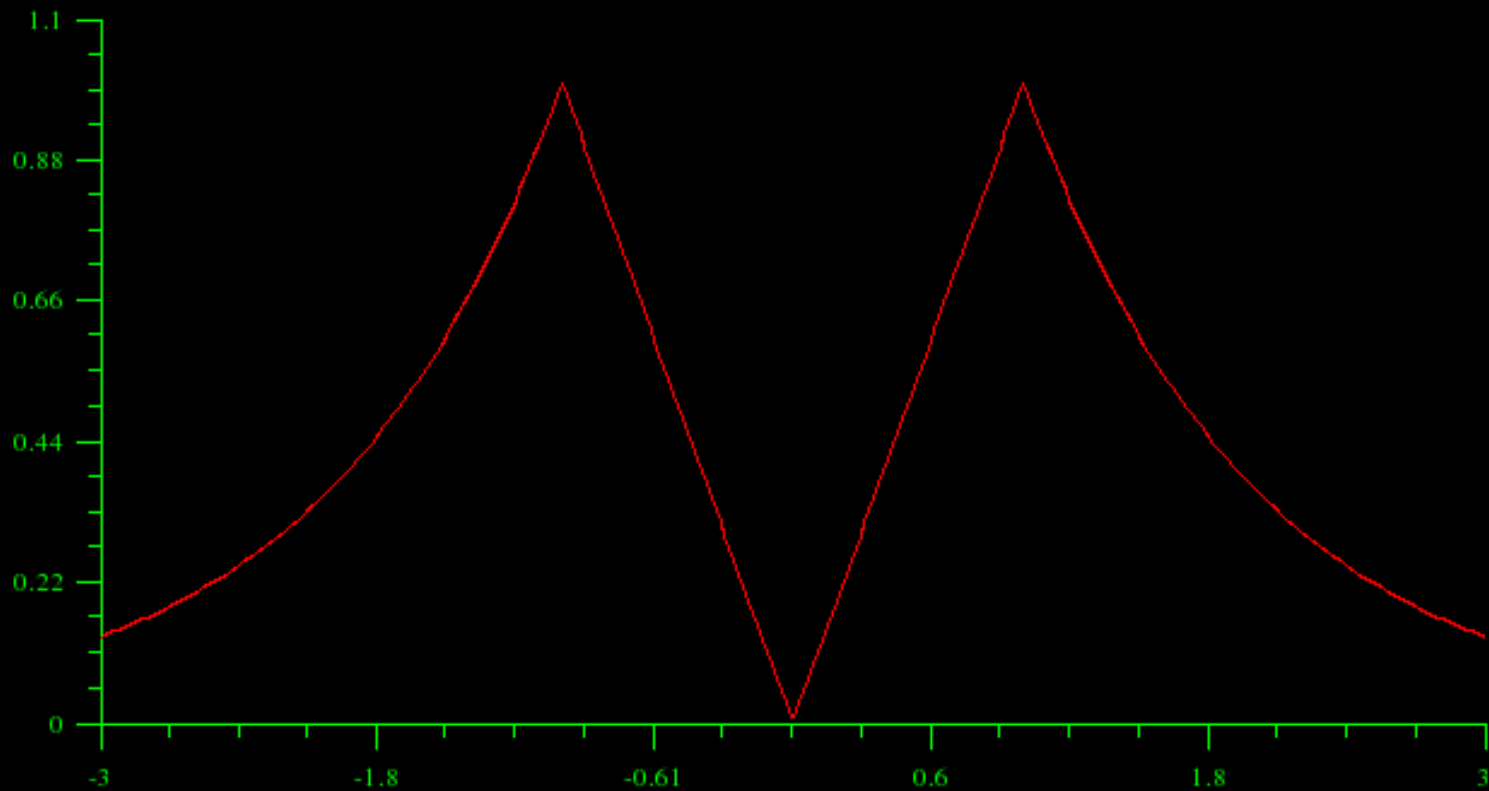
- Normal distribution

- $h(\mathbf{z}, \mathbf{x}) = A \exp(-B |\mathbf{x} - \mathbf{z}|^2)$



Probability Distance Strategies

- Truncated negative exponential:



Shortcomings

- What is the theoretical justification?
 - What assumptions are being made about criminal behavior?
 - What mathematical assumptions are being made?
 - How do you check the assumptions?

Shortcomings

- How do you add in local information?
 - How could you incorporate socio-economic variables into the model?

Snook, *Individual differences in distance travelled by serial burglars*

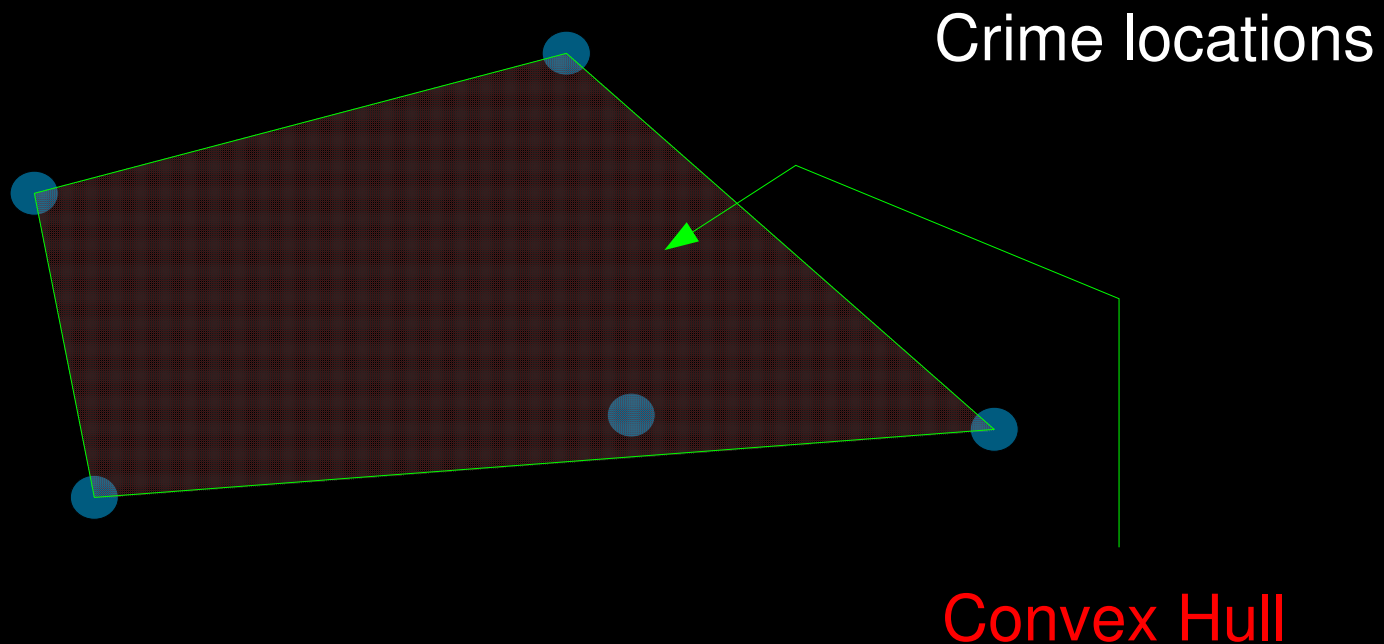
Malczewski, Poetz & Iannuzzi, *Spatial analysis of residential burglaries in London, Ontario*

Bernasco & Nieuwbeerta, *How do residential burglars select target areas?*

Osborn & Tseloni, *The distribution of household property crimes*

Shortcomings

- The convex hull effect:
 - The anchor point always occurs inside the convex hull of the crime locations.



A New Approach

- In previous methods, the unknown quantity was:
 - The anchor point
(spatial distribution strategies)
 - The hit score
(probability distance strategies)
- We use a different unknown quantity.

A New Approach

- Let $P(x; z)$ be the density function for the probability that an offender with anchor point z commits a crime at location x .
 - This distribution is our new unknown.
 - This has criminological significance.
 - In particular, assumptions about the form of $P(x; z)$ are equivalent to assumptions about the offender's behavior.

The Mathematics

- Given crimes located at $\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n$ the *maximum likelihood estimate* for the anchor point \mathbf{z} is the value of \mathbf{z} that maximizes

$$\begin{aligned} L(\mathbf{z}) &= \prod_{i=1}^n P(\mathbf{x}_i, \mathbf{z}) \\ &= P(\mathbf{x}_1, \mathbf{z}) P(\mathbf{x}_2, \mathbf{z}) \cdots P(\mathbf{x}_n, \mathbf{z}) \end{aligned}$$

or equivalently, the value that maximizes

$$\begin{aligned} \lambda(\mathbf{z}) &= \sum_{i=1}^n \ln P(\mathbf{x}_i, \mathbf{z}) \\ &= \ln P(\mathbf{x}_1, \mathbf{z}) + \ln P(\mathbf{x}_2, \mathbf{z}) + \cdots + \ln P(\mathbf{x}_n, \mathbf{z}) \end{aligned}$$

Relation to Spatial Distribution Strategies

- If we make the assumption that offenders choose target locations based only on a distance decay function in normal form, then

$$P(\mathbf{x}; \mathbf{z}) = A \exp(-B|\mathbf{x} - \mathbf{z}|^2)$$

- The maximum likelihood estimate for the anchor point is the centroid.

Relation to Spatial Distribution Strategies

- If we make the assumption that offenders choose target locations based only on a distance decay function in exponentially decaying form, then

$$P(\mathbf{x}; \mathbf{z}) = A \exp(-B|\mathbf{x} - \mathbf{z}|)$$

- The maximum likelihood estimate for the anchor point is the center of minimum distance.

Relation to Probability Distance Strategies

- We can generate a hit score by using either

$$L(\mathbf{z}) = \prod_{i=1}^n P(\mathbf{x}_i, \mathbf{z}) \quad \lambda(\mathbf{z}) = \sum_{i=1}^n \ln P(\mathbf{x}_i, \mathbf{z})$$

- If we multiply rather than add in the usual method of probability distance strategies, we obtain our method.

Advantages

- Our method recaptures existing methods.
- Assumptions about offender behavior can be directly used in the model.
- We can explicitly incorporate information about geography and socio-economic factors into the model.
- We do not suffer from the convex hull problem.

Better Models

- Recall that $P(\mathbf{x}; \mathbf{z})$ is the density function for the probability that an offender with anchor point \mathbf{z} commits a crime at the point \mathbf{x} .
- Suppose that $P(\mathbf{x}; \mathbf{z})$ has the general form

$$P(\mathbf{x}; \mathbf{z}) = K(|\mathbf{x} - \mathbf{z}|) \cdot G(\mathbf{x}) \cdot N(\mathbf{x}; \mathbf{z})$$

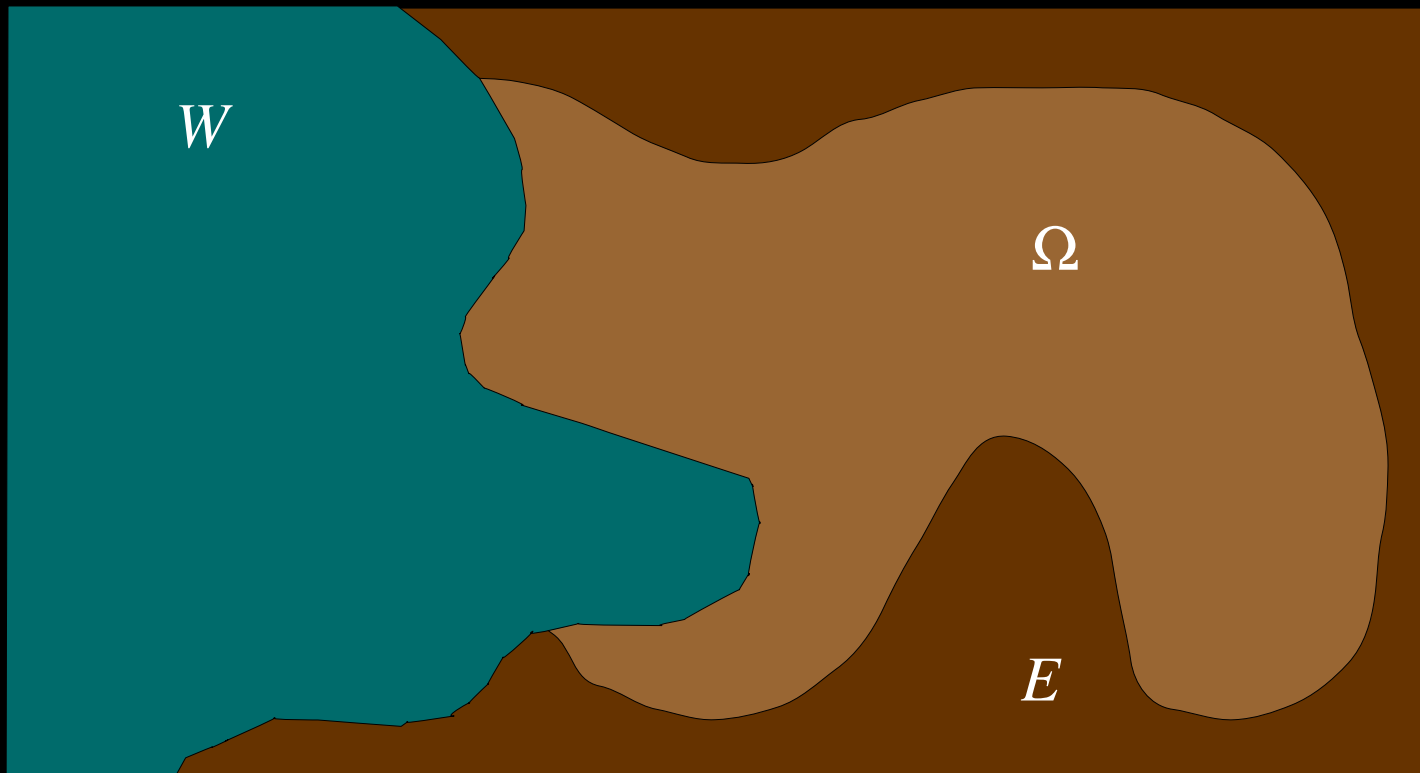
Dispersion
kernel

Geographic
factors

Normalization

The Simplest Case

- We have information about crimes committed by the offender only for a portion of the region.



The Simplest Case

- Regions
 - Ω : Jurisdiction(s). Crimes and anchor points may be located here.
 - E : “elsewhere”. Anchor points may lie here, but we have no data on crimes here.
 - W : “water”. Neither anchor points nor crimes may be located here.
- In all other respects, we assume the geography is *homogeneous*.

The Simplest Case

- We know $z \notin W$ and $P(\mathbf{x}; \mathbf{z}) = 0$ if $\mathbf{x} \notin \Omega$.

- We set

$$G(\mathbf{x}) = \begin{cases} 1 & \mathbf{x} \in \Omega \\ 0 & \mathbf{x} \notin \Omega \end{cases}$$

We choose an appropriate dispersion kernel; say

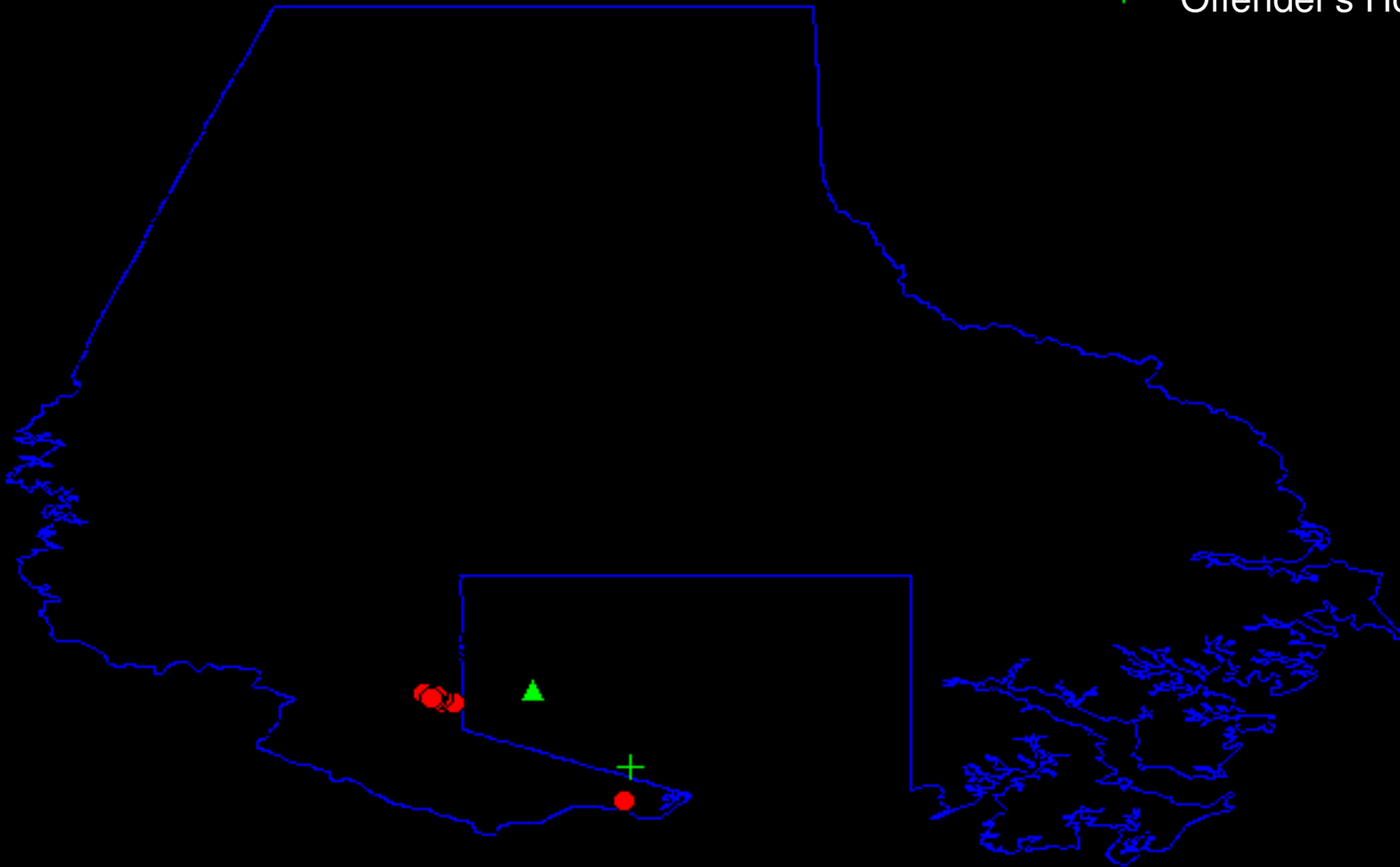
$$K(\mathbf{x}; \mathbf{z}) = \exp(-|\mathbf{x} - \mathbf{z}|^2 / \sigma^2)$$

- The required normalization function is

$$N(\mathbf{x}; \mathbf{z}) = \left[\iint_{\Omega} \exp(-|\mathbf{y} - \mathbf{z}|^2 / \sigma^2) dy_1 dy_2 \right]^{-1}$$

Sample Results

- Baltimore County
- Vehicle Theft
- ▲ Predicted Anchor Point
- + Offender's Home



Sample Results

- Crimes were vehicle thefts in 2003-2004.
 - Data provided by Phil Canter, Baltimore County Police Department.
- Predicted anchor point was not in the convex hull of the crime locations.

Better Models

- Method is just a modification of the centroid method that accounts for possibly missing crimes outside the jurisdiction.
- Clearly, better models are needed.
- This is ongoing work.
- More data!

Questions?

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