

Predictive Policing 101

Preventing Crime with Data and Analytics



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Data Management Research Section

ETRI





References

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Depth of Knowledge (DoK) Level of the Seminar

4



DoK1

DoK2

DoK3

DoK4



Outline

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- **Why is the Predictive Policing Important?**
- **Background**
- **Mathematical Frameworks**
- **Data Used in Predictive Policing**
- **Predictive Methodologies for Predictive Policing**
- **Summary**

Why is the Predictive Policing Important?

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Why is the Predictive Policing Important?

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■ What is Crime Analysis?

- **Administrative, Tactical and Strategic Problem Solving**
- **Forecasting and predictive analysis**



August Vollmer
Source: LAPD Web Site

1909

- **Principles of *problem-oriented* policing**
- ***Evidence-based* policing**
- ***Real-time* crime analysis**
- ***Intelligence-led* policing and other proven policing models**



Orlando W. Wilson

1950

ss section
ve
agation

22 Hertz

λ - wavelength 26.71

a - amplitude 80.42

v - velocity 5.02

T - time 1.28

f - frequency 0.22

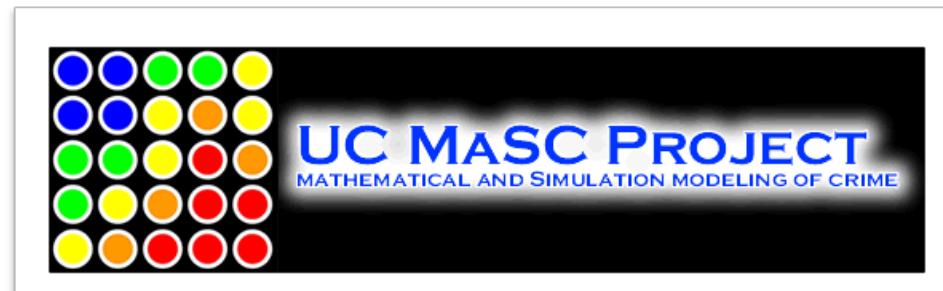
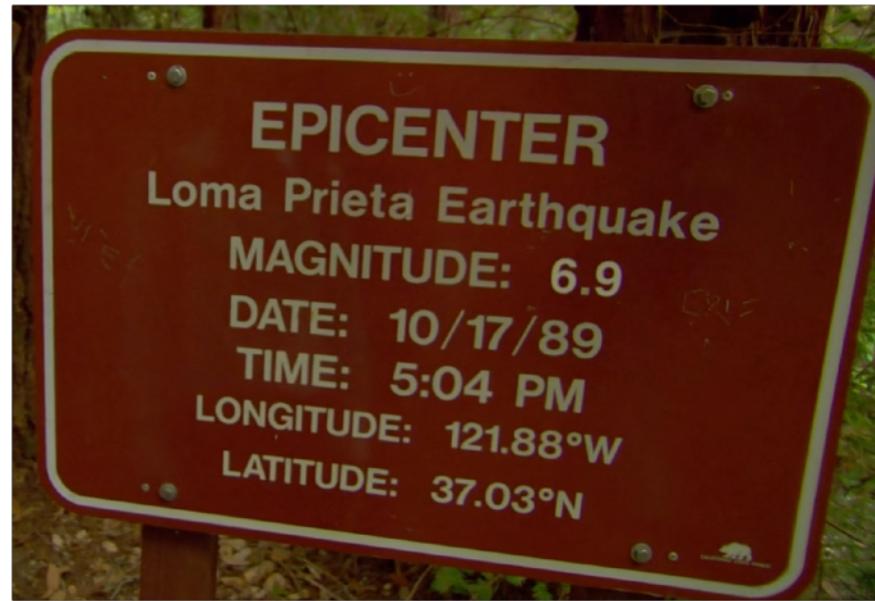
ρ = density
 p = pressure
 μ = viscosity

$$\rho \left(\frac{\partial \mathbf{v}}{\partial t} + (\mathbf{v} \cdot \nabla) \mathbf{v} \right) = -\nabla p + \mu \nabla^2 \mathbf{v} + \mathbf{f}$$
$$\nabla \cdot \mathbf{v} = 0,$$

Clustering Patterns

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- Loma Prieta Earthquake (1989)
 - Earthquake analysis
 - Clustering patterns
- MASC Project
 - Clustering patterns are also seen in *crime data*
 - Future events *nearby in space and time* after shocks of crime
 - Finding and clustering the *patterns* from chaotic human behaviours
 - *Self-exciting point process model similar* to that used in earthquake analysis
 - Pilot Project : Los Angeles
 - Data: 13 million past crimes
 - Duration: past 80 years



Mathematical and Simulation Modeling of Crime

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Jeff Brantingham
UCLA Anthropology



George Mohler
Santa Clara Mathematics



George Tita
**UCI Criminology,
Law and Society**

Mathematical Framework [7]

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■ The Hawkes Process Model (1974)

- any given event, or collection events can be causally linked to a *background Poisson process* and *foreground self-exciting process*.

Self-excitation
the distribution of crimes following an initial event

$$\lambda(t) = \mu + k_0 \sum_{t_k < t} g(t - t_k)$$

The diagram shows the mathematical expression for the Hawkes process rate function $\lambda(t)$. It consists of three main components: a constant background rate μ (highlighted in pink), a sum of past events (highlighted in purple), and a kernel function $g(t - t_k)$ (highlighted in blue). A pink arrow points from the μ term to the text "background rate of events stationary Poisson process". A purple arrow points from the summation term to the text "the density of prior events necessary to trigger excitation". A blue arrow points from the kernel term to the text "how much excitation is generated by a collection of prior events".

*background rate of events
stationary Poisson process*

*the density of prior events
necessary to trigger excitation*

*how much excitation is generated
by a collection of prior events*

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$$\lambda(t) = u + \lambda(t-t_1)$$

17:352

BURGLARY

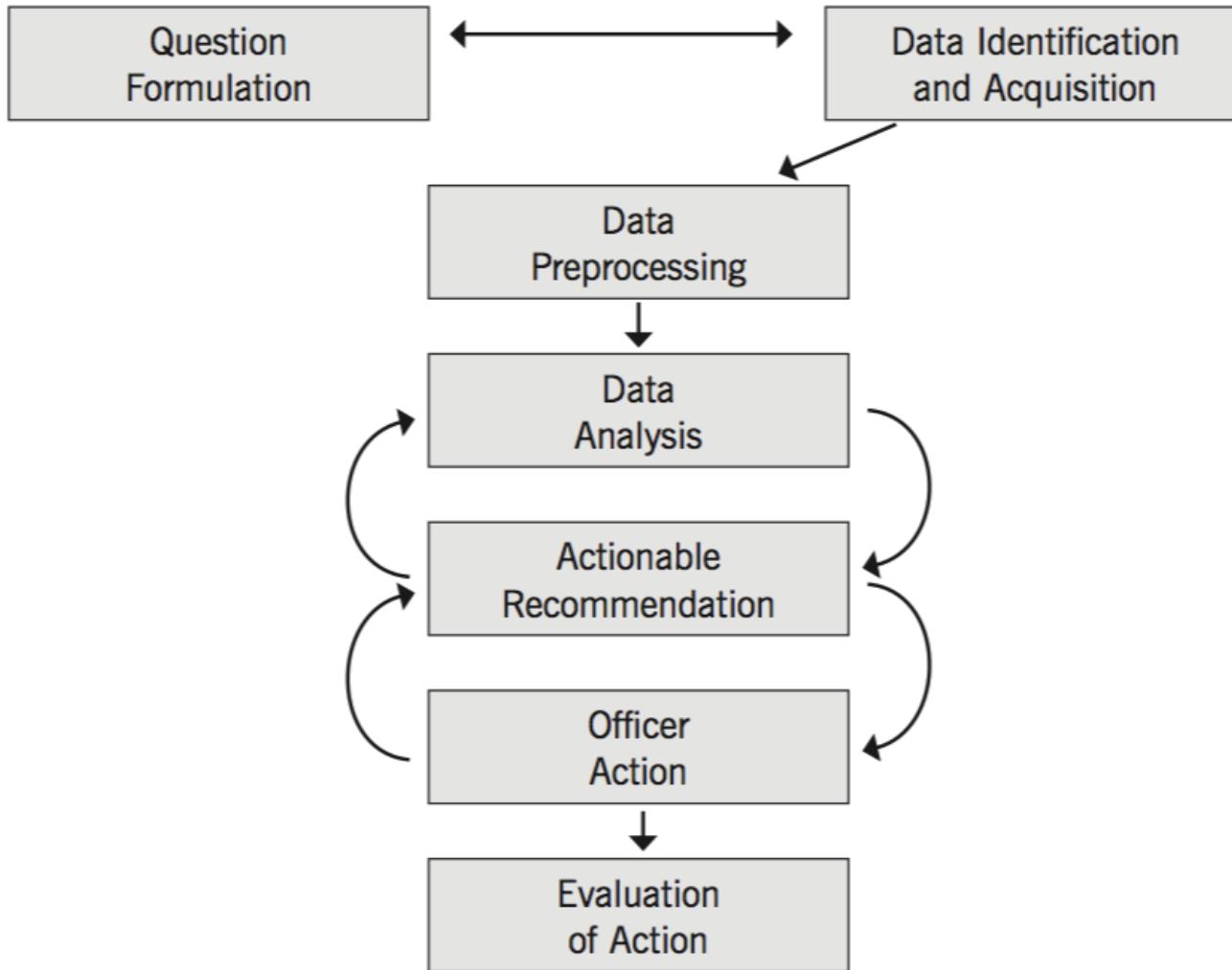
CASES

CASES



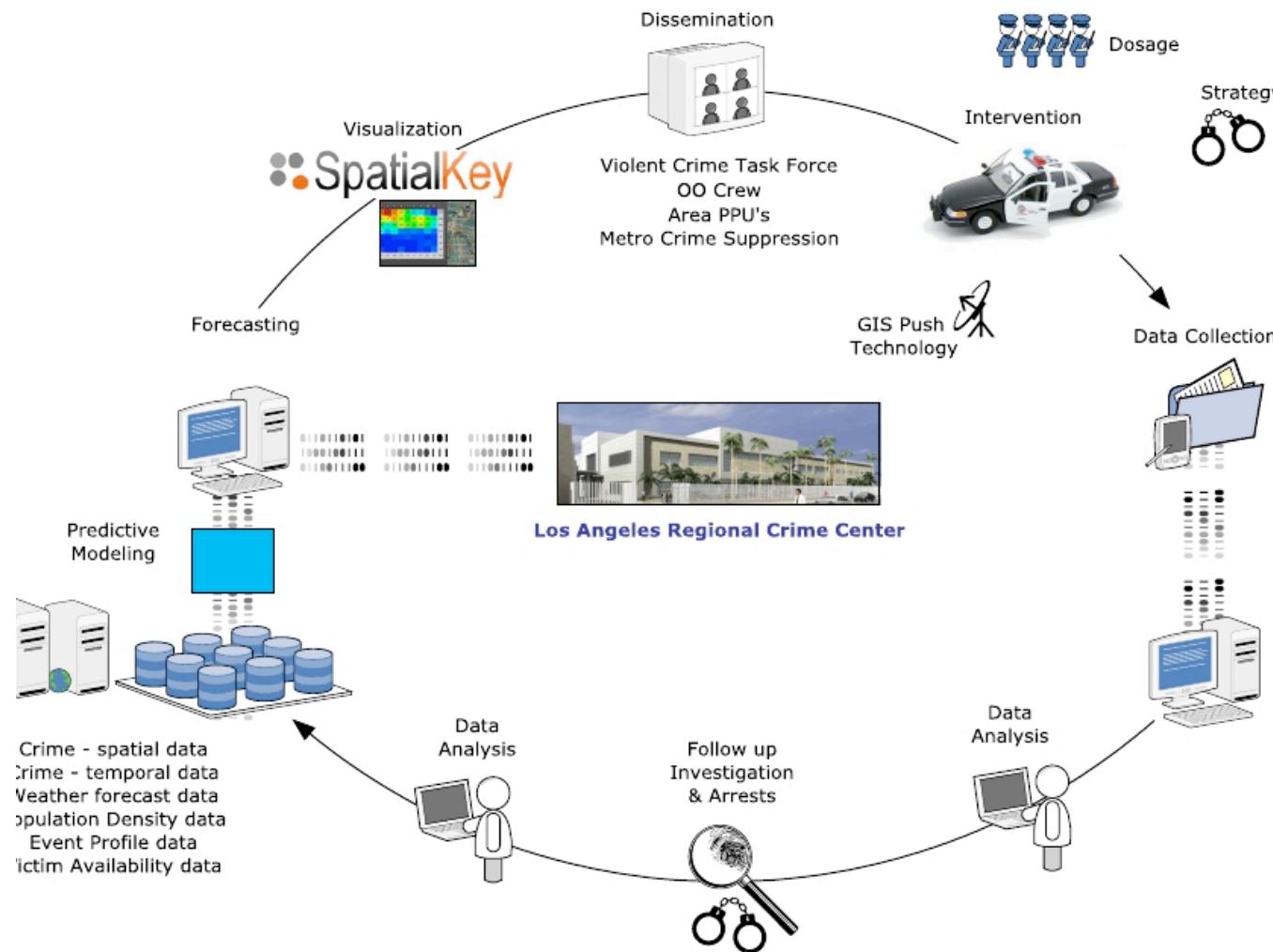
Operational Challenges of Predictive Policing

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Case Study: Los Angeles Regional Crime Center

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Data Used in Predictive Policing

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■ Three Categories

- **Spatial**
- **Temporal**
- **Social Network**

Spatial Variables	Temporal Variables	Social Network Variables
Indicators of Areas with Potential Victims/Targets <ul style="list-style-type: none">• Shopping malls• Property values• Hotels• Area demographics• Population density• Residential instability	<ul style="list-style-type: none">• Payday schedules• Time of day• Weekend vs. weekday• Seasonal weather (e.g., hot versus cold weather)• Weather disasters• Moon phases• Traffic patterns• Sporting and entertainment events	<ul style="list-style-type: none">• Kinship• Friendship• Affiliation with an organization• Financial transaction• Offender/victim
Indicators of Escape Routes <ul style="list-style-type: none">• Highways• Bridges• Tunnels• Public transportation• Railways• Dense foliage		
Indicators of Criminal Residences <ul style="list-style-type: none">• Bars and liquor stores• Adult retail stores• Fast food restaurants• Bus stops• Public health information• Areas with physical decay		

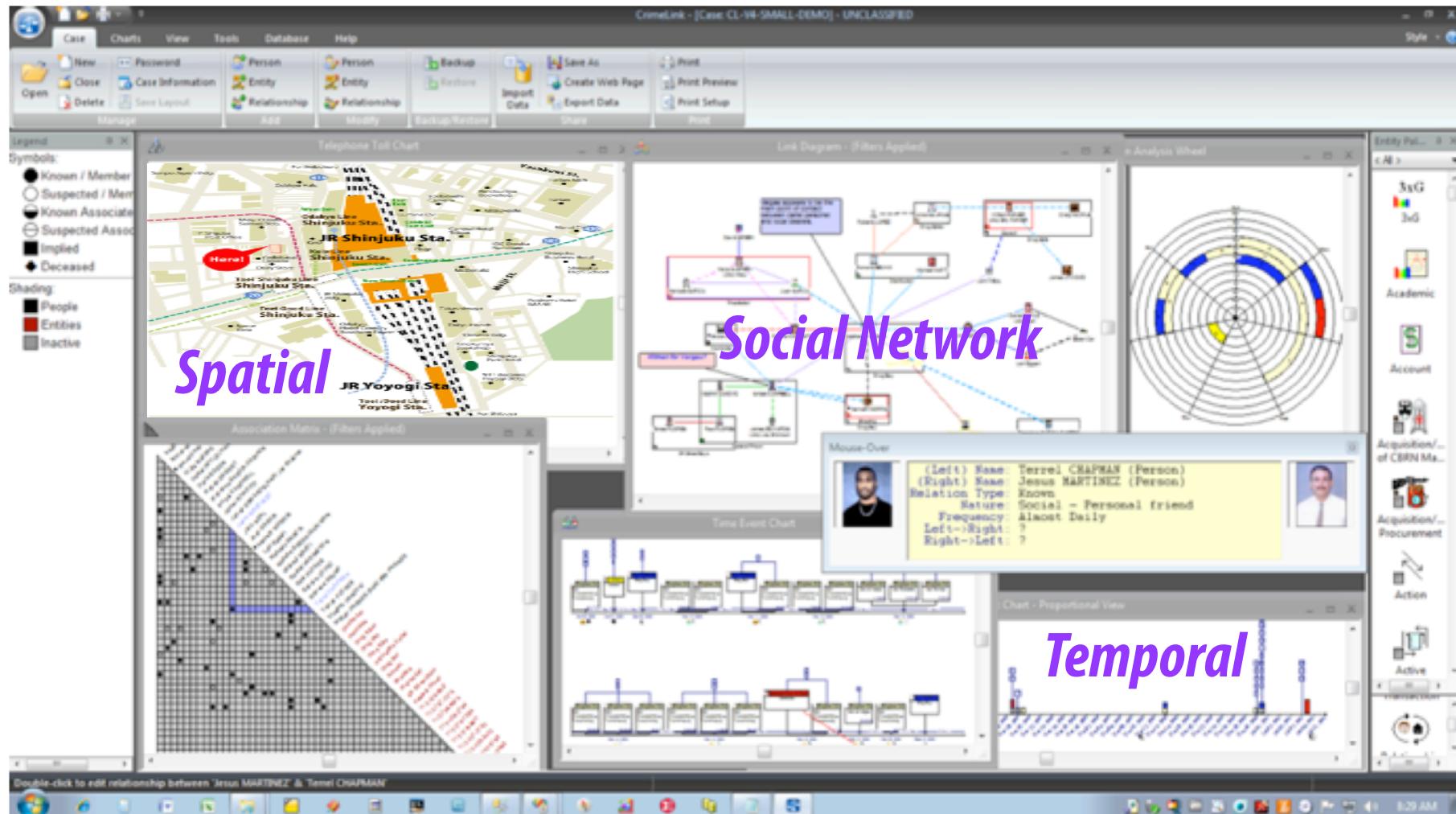
The Transition from Data to Knowledge in a Police Agency

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Data		Information		Knowledge		Result
Analysis	Individual Incident Reports in a records management system	Six of the reports are related in a series of robberies	Communication	Robbery series is prime topic of discussion in next detective's meeting	Strategy & Action	Robbery offender is apprehended
	Statistics showing number of officers per capita throughout the state	Your police department has 20% fewer officers per capita than average		Chief has this information in mind when preparing his budget proposal		Agency is granted additional officers by city
	Crime volume of current year compared to past years; individual records in RMS; jurisdictional information	Auto theft is up 20%, with most of the increase in Police Beat 5 on the midnight shift, probably influenced by new sports arena		Officers internalize this information and consider it when patrolling Beat 5		Auto theft is reduced

Full Scale Analysis

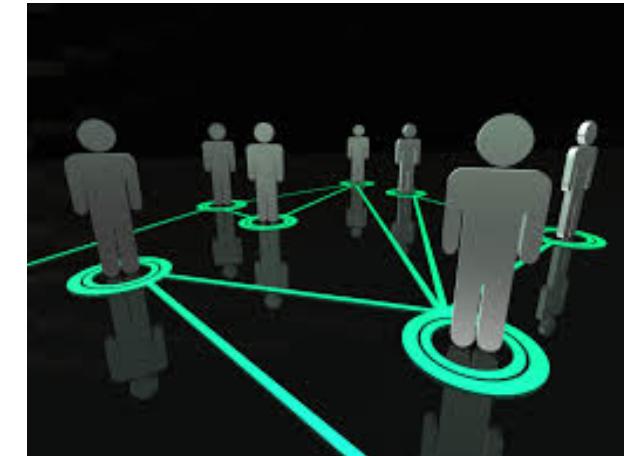
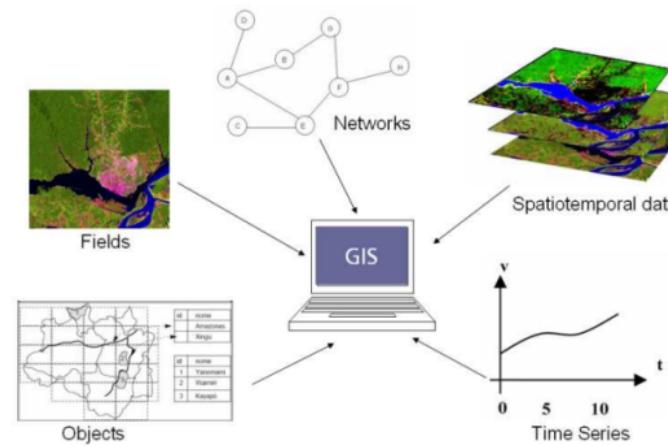
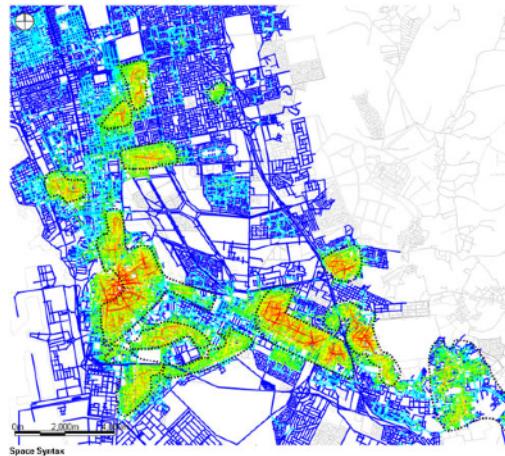
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Predictive Methodologies

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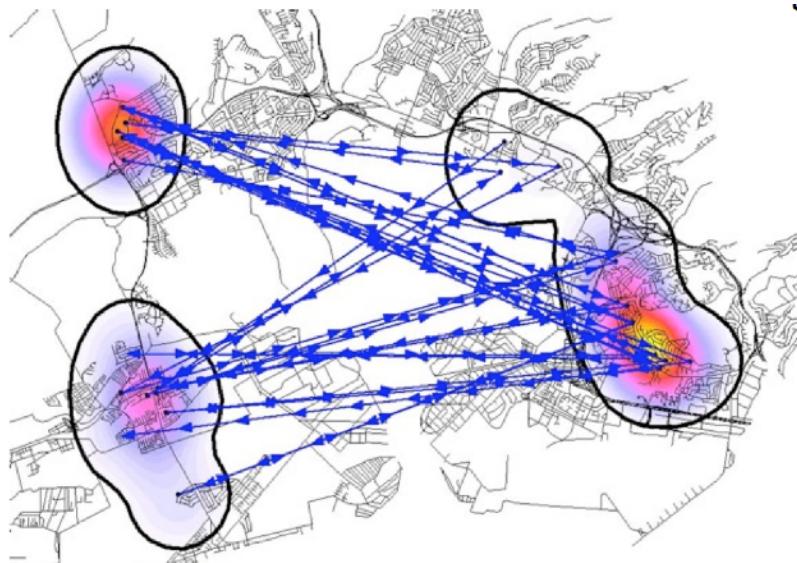
- Analysis of space
- Analysis of time and space
- Analysis of social networks



Predictive Methodology One: Analysis of Space

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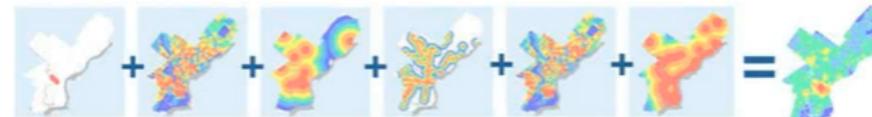
- Point (or offense) locations
- Hierarchical clusters
- Partitioned clusters
- Fuzzy clusters
- Density mapping
- Risk-terrain modeling (RTM) clusters



Predictive Methodology One: Analysis of Space

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- Point (or offense) locations
 - *Theory of repeat victimization, 500x500 feet [PredPol Software]*
- Hierarchical clusters
 - use a *nearest-neighbor technique* [display the clusters: *ellipses, convex hulls*]
- Risk-terrain modeling (RTM) clusters



Gun shootings example



Gun shootings example

Predictive Methodology Two: Analysis of Time and Space

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Predictive Methodology Two: Analysis of Time and Space

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- CrimStat III : a software program (sociologist + National Institute of Justice)
 - spatial-temporal moving average (STMA)
 - the *average time and location* for a subset of incidents
 - correlated walk analysis (CWA) : *temporal and spatial relationships* between incidents
 - computing the *correlation* between *intervals* [time, distance, direction between two events]





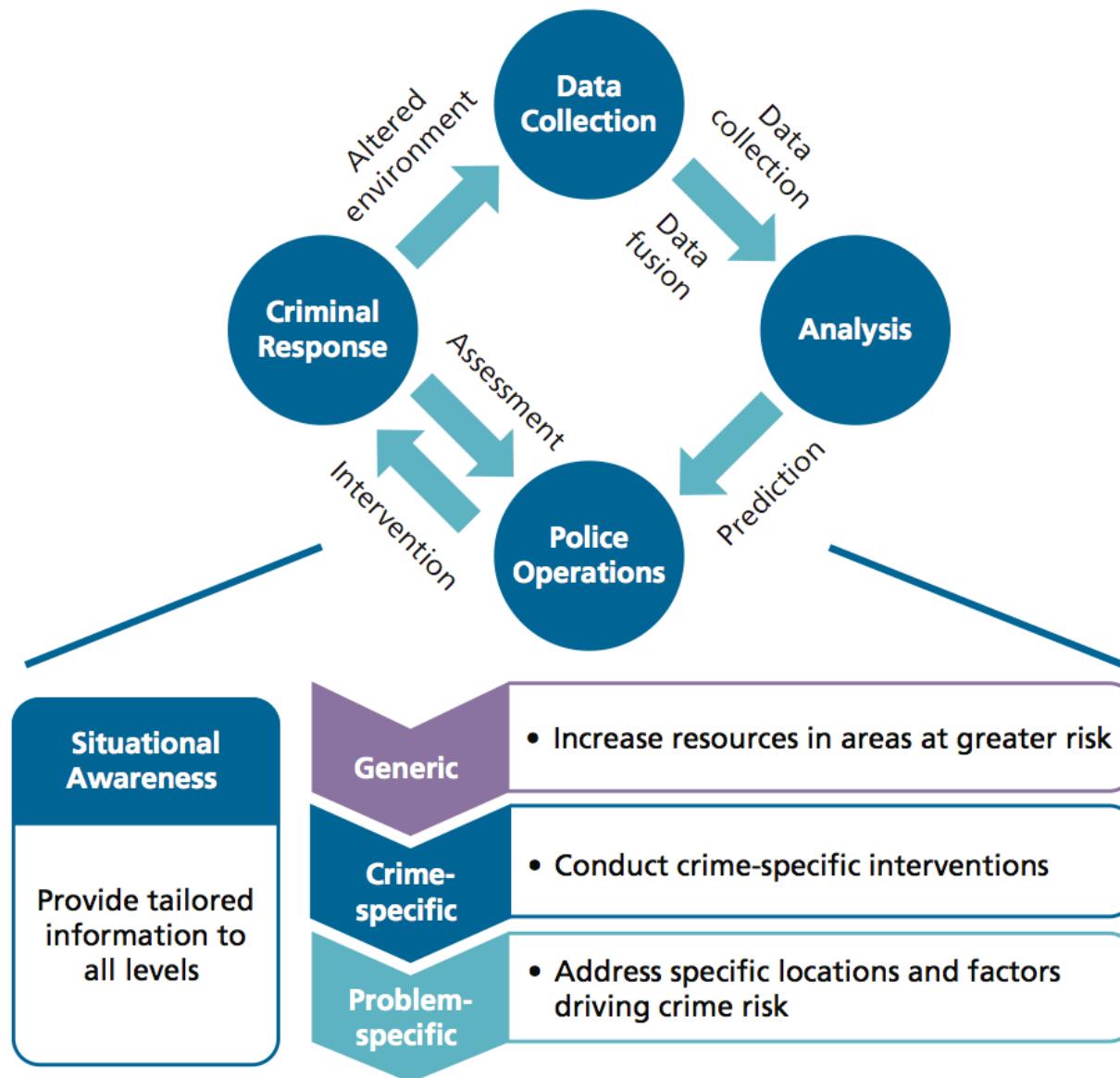
Predictive Methodology Three: Analysis of Social Networks

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- **Social Network Analysis (SNA) : *cutting edge* of crime analysis**
 - detect *persons* of interest, as opposed to *locations* of interest
 - identify individuals that are central to criminal organizations (eg., gangs and drug networks)
- **Building blocks of a social network: *relationships* between two actors**
- **In crime-fighting applications,**
 - SNA is frequently used to identify *central nodes* [*high level of connectivity*]
 - **Measures of *centrality***
 - *degree* : *the number of links* possessed by a node
node's level of connectedness
 - *closeness* : *the total distance* from a node to all other nodes in the network
ease of obtaining information from the network
 - *betweenness* : *the number of instances* a given node appears in the shortest path between other nodes
relevance to the passage of information within the network

Review: Prediction-Led Policing Business Process [2]

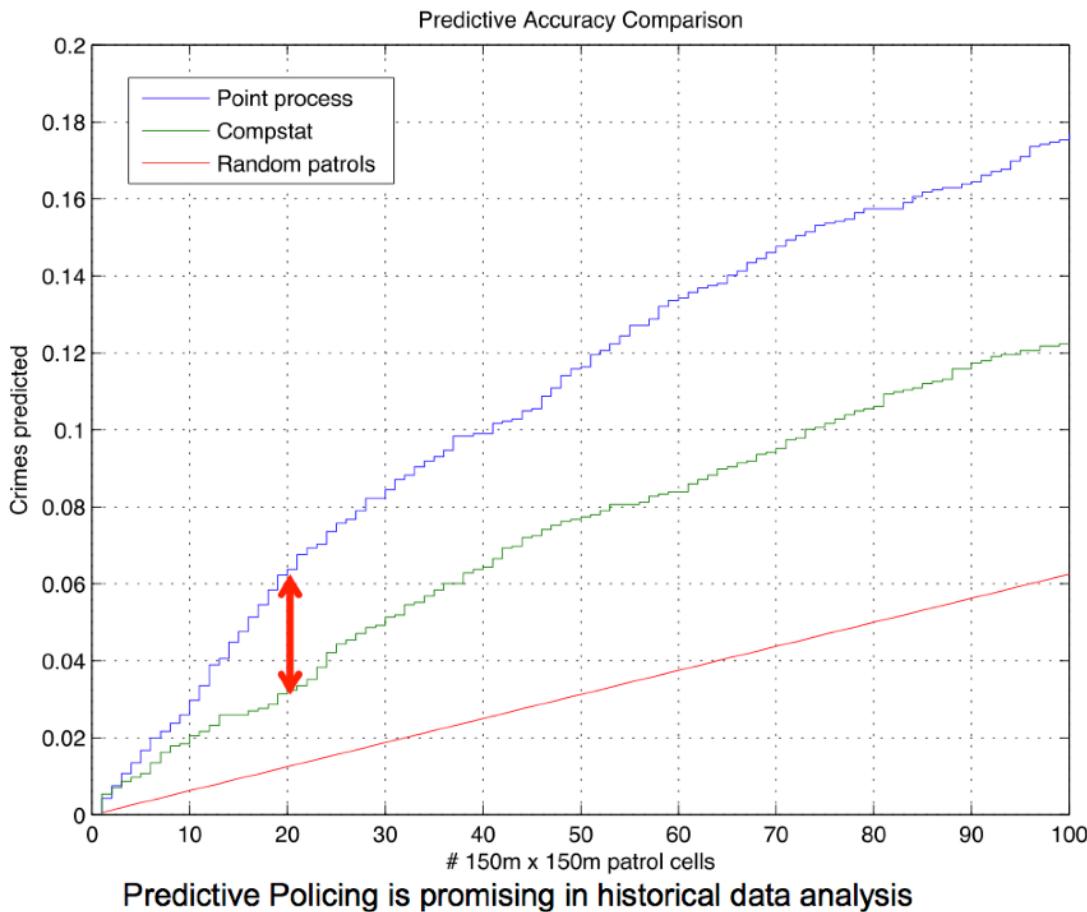
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Places on the Frontier of Predictive Policing in the United States

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- *Los Angeles, California*
- **Santa Cruz, California**
- **Baltimore Country, Maryland**
- **Richmond, Virginia**
- **Memphis, Tennessee**



Summary

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- **Predictive Policing Concept**
- ***Mathematical Frameworks for Predictive Policing***
- ***Data Used in Predictive Policing***
- ***Predictive Methods for Predictive Policing***
- ***Predictive Policing Business Process***

A close-up profile of Steve Jobs' face, showing his beard, glasses, and intense expression.

one more thing

or two...

Future Projects

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- ^ Starting from Fall 2018 ~
- 6 Years : **Korea Predictive Policing Infrastructure [KPPI]**
 - 1st phase (3 years) : Proof-of-concept Stage
 - *human dynamics, criminal analytics frameworks, ...*
 - 2nd phase (3 years) : Experimental Stage
 - *Predictive Policing in Seoul and Daejeon, and more ...*

Candidate Research Collaborators

- Industry [Communications, Systems, Data Scientists]
- Academic [Criminology, Sociology, Anthropology]
- Research Institutes [Nonprofits]



A word cloud centered around the word "thank you" in various languages. The words are colored in a rainbow gradient. Some words have their phonetic pronunciation in parentheses.

- danke (рахмат)
- спасибо (Баярлалаа)
- спасибо (faafetai lava)
- спасибо (kiitos dankie)
- спасибо (dhanyavad)
- спасибо (hvala)
- спасибо (märuvutu)
- спасибо (köszönöm)
- спасибо (hvala)
- спасибо (gręzje)
- спасибо (dziekuje)
- спасибо (obrigado)
- спасибо (sobodi)
- спасибо (dékiji)
- спасибо (mesi)
- спасибо (didî madida)
- спасибо (kam sah hamnida)
- спасибо (তোমাকে ধন্যবাদ)
- спасибо (raaf)
- спасибо (sagolın)
- спасибо (najis tuke)
- спасибо (rahmat)
- спасибо (terima kasih)
- спасибо (감사합니다)
- спасибо (xièxie)
- спасибо (suxiapostú)
- спасибо (dank je)
- спасибо (misaotra)
- спасибо (matondo)
- спасибо (paldies)
- спасибо (grazzi)
- спасибо (djiere dieuf)
- спасибо (sulpáy)
- спасибо (tanemirt)
- спасибо (rahmet)
- спасибо (diolch)
- спасибо (dhanyavadagalu)
- спасибо (shukriya)
- спасибо (mercé)
- спасибо (merci)
- спасибо (teşekkür ederim)
- спасибо (gracias)
- спасибо (təpəðħlejħ)
- спасибо (mochchakkeram)
- спасибо (go raibh maith agat)
- спасибо (arigatō)
- спасибо (dakujem)
- спасибо (trugarez)
- спасибо (mamnun)
- спасибо (хвала)
- спасибо (asante manana)
- спасибо (obrigada)
- спасибо (chokrane murażoze)
- спасибо (tenik)
- спасибо (ngiyabonga)
- спасибо (teşekkür ederim)
- спасибо (gracias)
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- спасибо (mochchakkeram)
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