



The Role of Social Network Analysis in Intelligence-Led Policing

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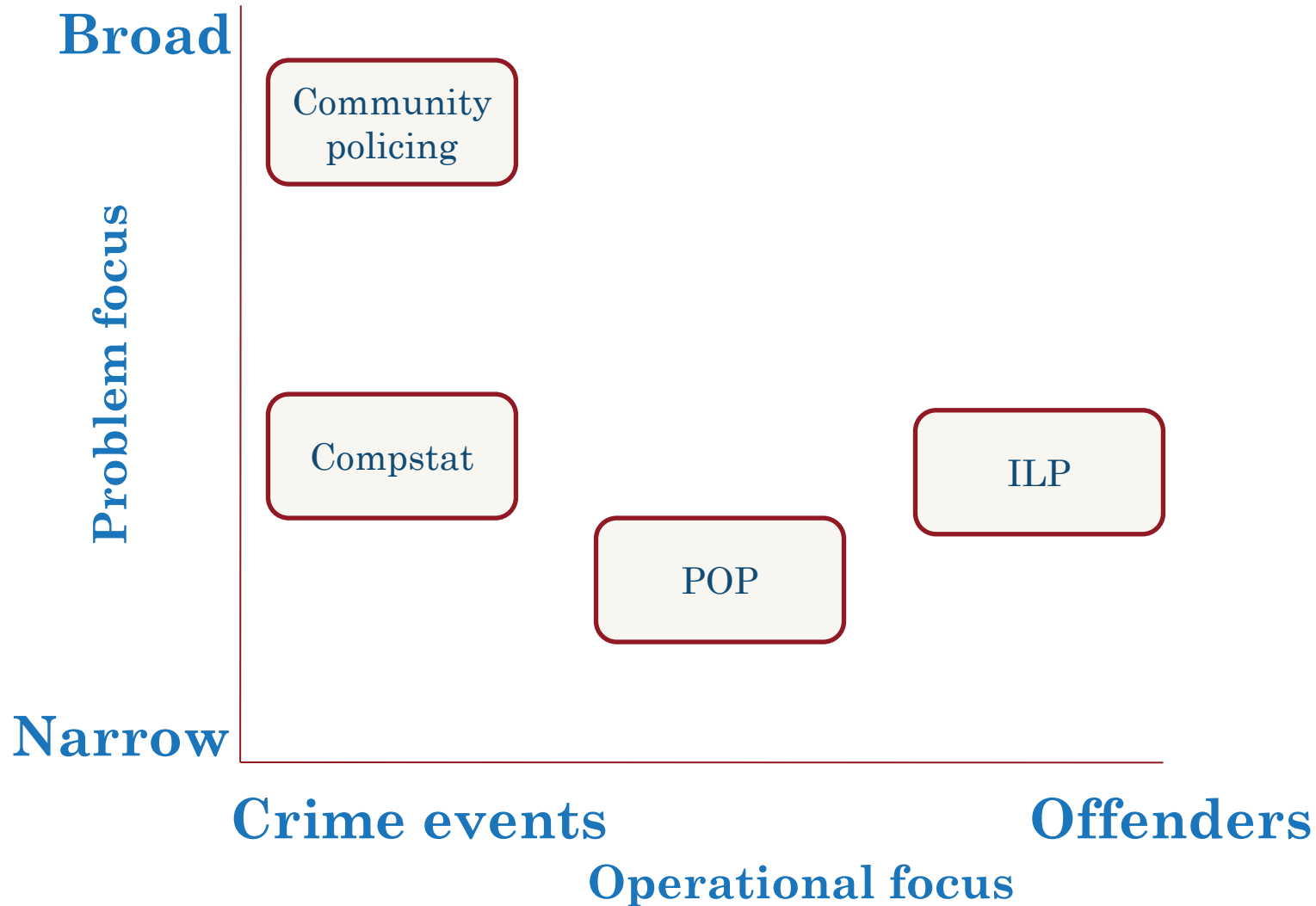
Definition: Intelligence-led policing

“A managerial philosophy where data analysis and crime intelligence are pivotal to an objective, decision-making framework that facilitates crime and problem reduction, disruption and prevention through both strategic management and effective enforcement strategies that target prolific and serious offenders.”

Source: Ratcliffe, J. (2008). Intelligence-Led Policing, Willan Publishing: p. 89.

- Strategic decisions
- Intelligence/data driven
- Focus on problem reduction
- Target prolific offenders

Policing paradigms



The Utility of SNA in ILP

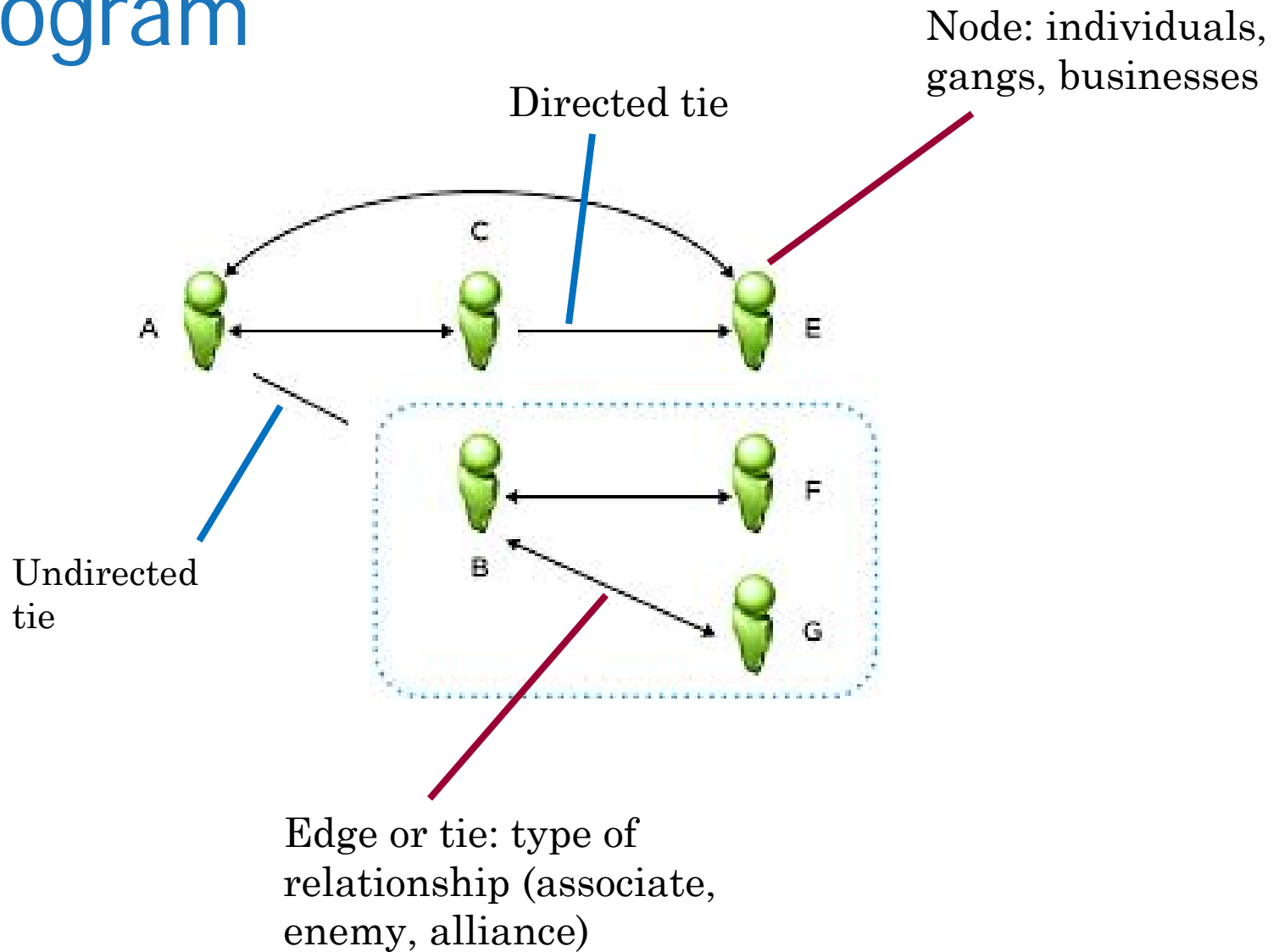
ILP

- Using intelligence to address criminal groups and prolific offenders.
- It often focuses on social relationships

SNA

- Maps social relationships
- Identifies group/organizational structures
- Identifies those at the center of criminal groups

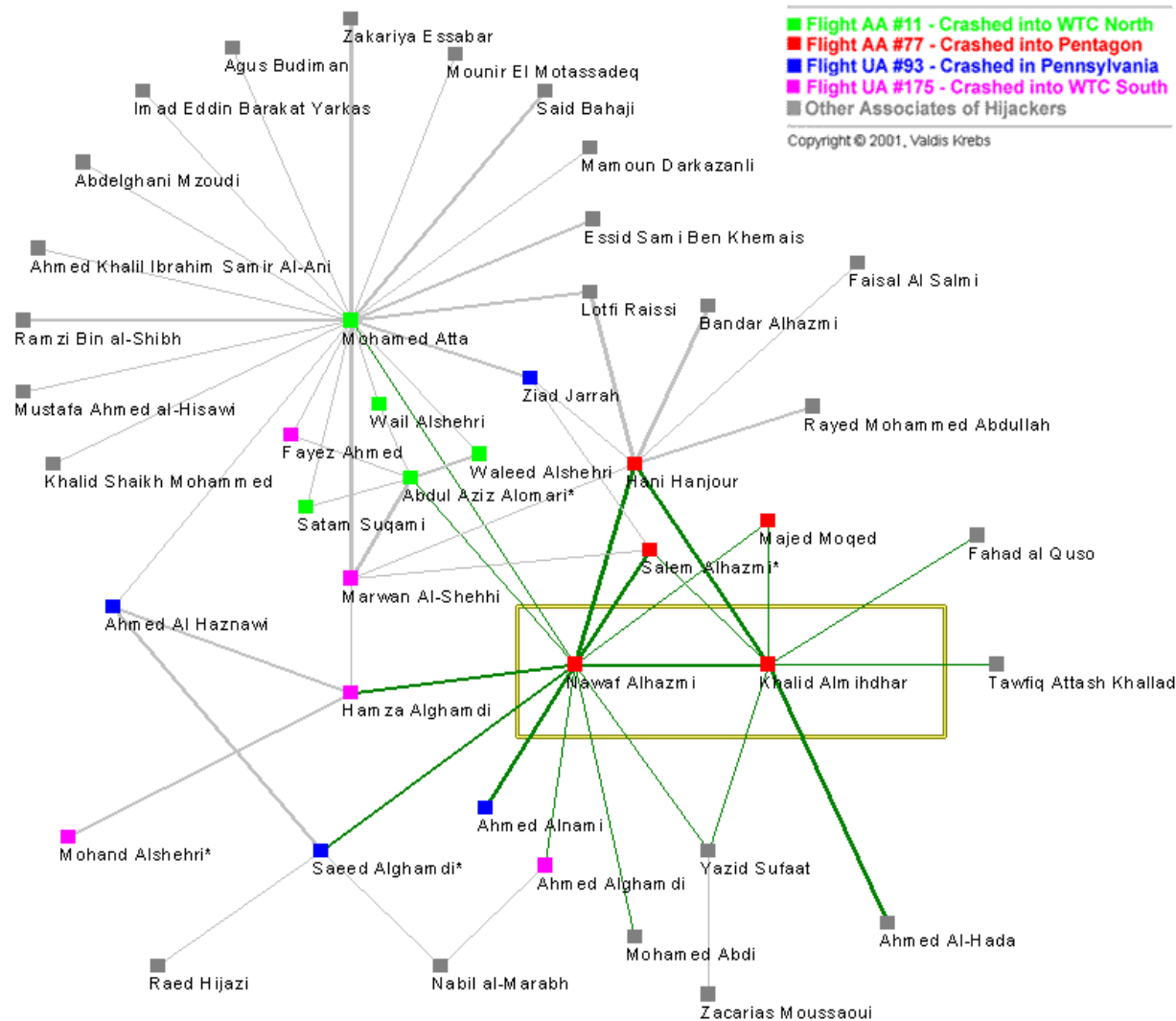
Sociogram



This is not just link analysis!

1. Degree Centrality – Simply the number of ties a node has in the network.
2. Betweenness Centrality – Those who are the intersection on many paths between others.
3. Eigenvector Centrality – Those who are connected to many connected people

Example: The 9-11 Hijacker Network



The 19 terrorists were ALL within two steps of the two original suspects identified in 2000

SOURCE: Valdis Krebs
<http://www.orgnet.com/>

Figure 3 - All 2 step links from two known suspects

Is SNA possible with police data?

Strengths

- Already collected
- Easily accessible
- Structured, relational, and temporal data is plentiful

Weaknesses

- Incompleteness
- Inaccuracies
- Inconsistencies

Advantages of Using SNA

- Layout optimization
 - No lines on top of each other, clear layout
 - Space on the page to equal social distance
- Identifying key players
 - Centrality as a measure of importance
- Free software (Pajek and Excel)

Stop & Think

- What kind of intelligence is available to you?
 - Crime reports
 - Field interview cards
 - GMIC's
 - Court transcripts
 - National Integrated Ballistic Information Network (NIBIN)
 - Telephone records
 - Jail/correctional visits
 - Free talks



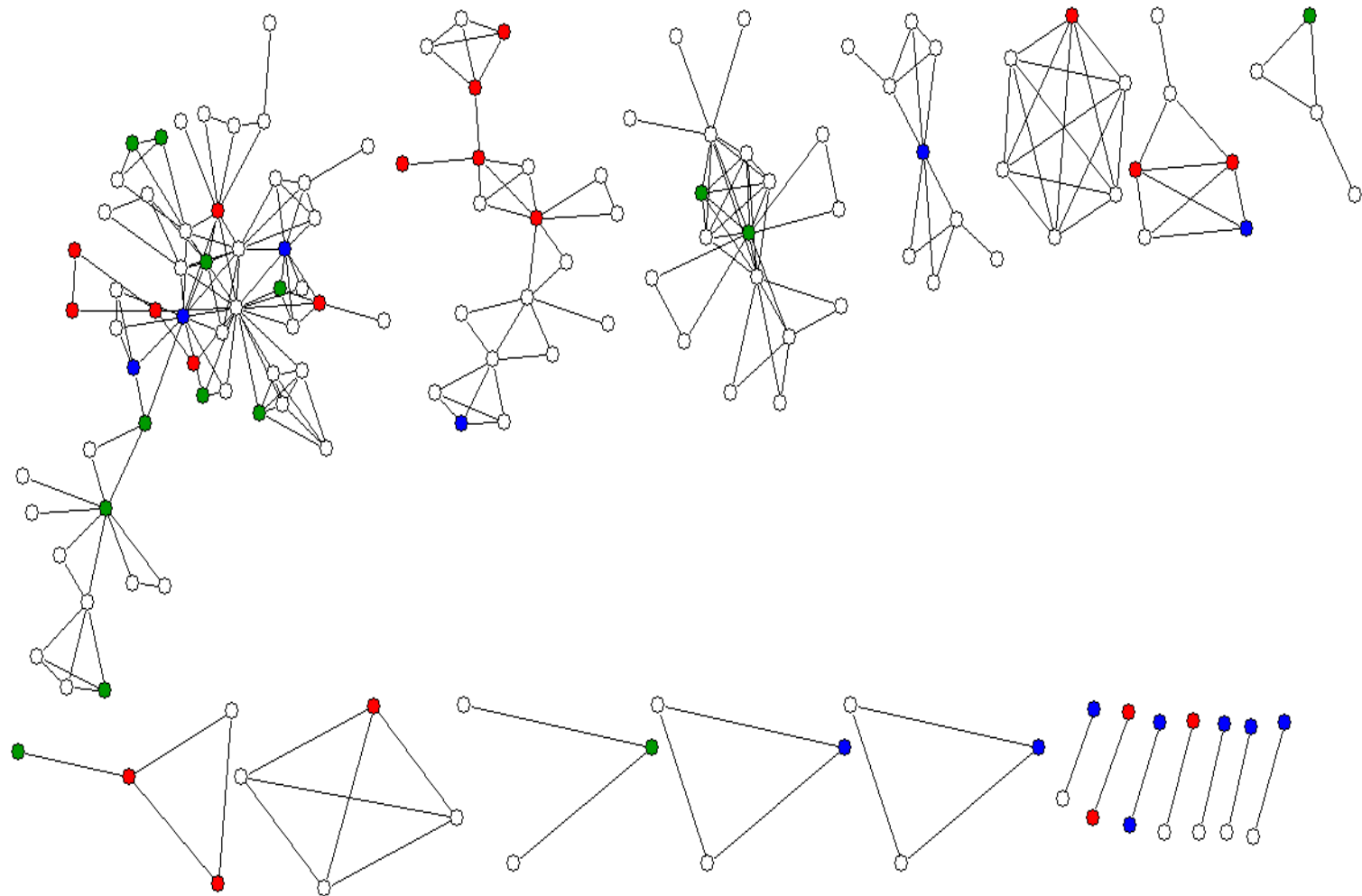
The Glendale Police Department Pilot Study

- ASU collected relational data from 2006-2010
 - GMIC
 - FI Cards
 - Merged with criminal history data
- Major findings
 - Intelligence was fairly reliable
 - Consistently collected
 - Time consuming to pull
 - Data management systems not optimally designed for SNA
 - SNA has strong utility for ILP

Major findings, cont.

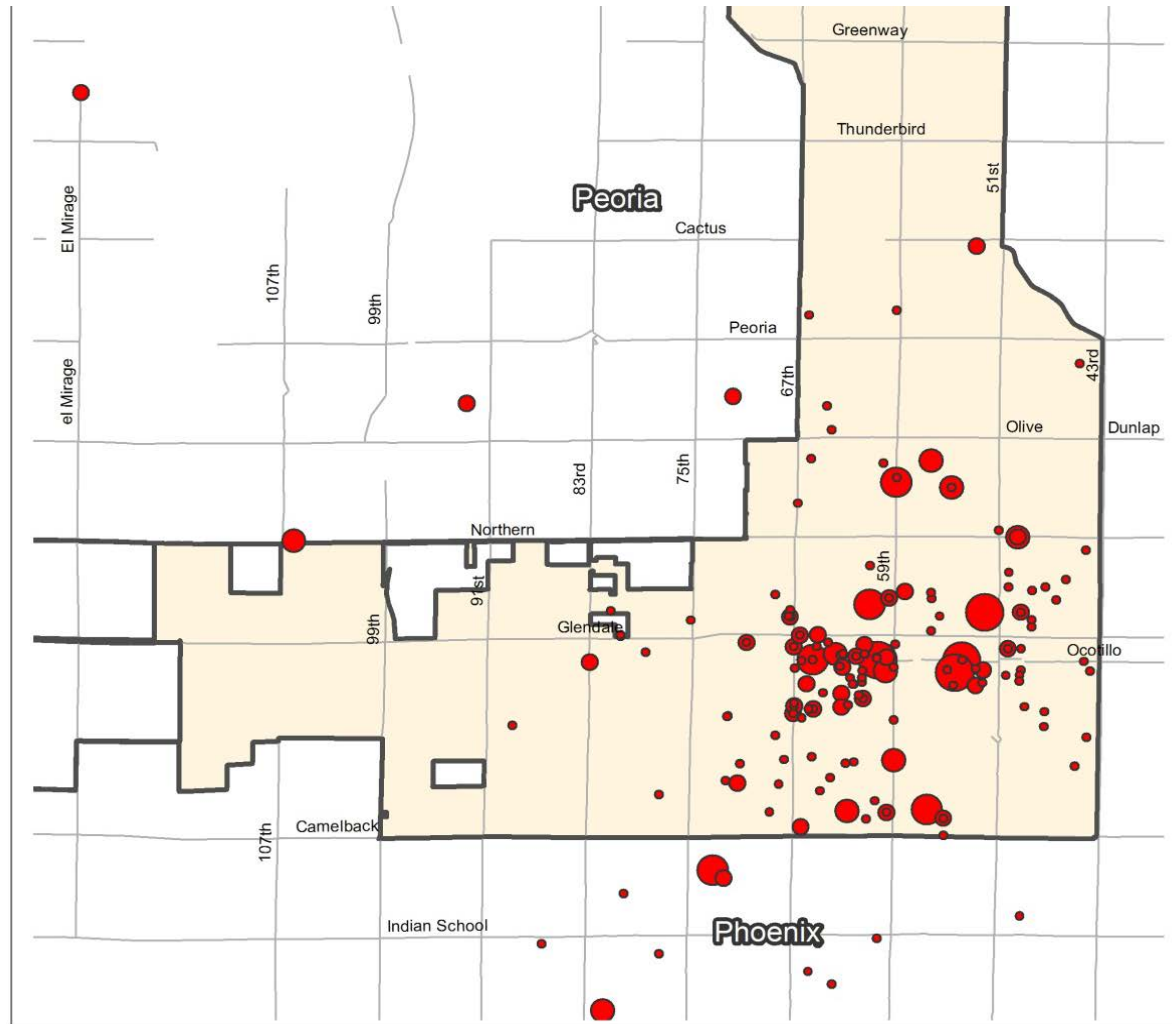
- There was not one large cohesive gang; the network consisted of many smaller connected groups.
- Gang members from different cliques were found to be in the same social network.
- Hybrid gangs were the most criminally involved.
- Betweenness centrality was more important with respect to criminal involvement.
- Gang membership * cohesion = more crime.

Examples of clique affiliations in 2007



Key: Varrio Sixty First = Red; West Side
Grandel = Blue; Varrio Clavalito Park = Green

Betweenness Centrality



Betweenness Centrality

| | Top Betweenness | |
|----------------------------|-----------------|--------|
| | No | Yes |
| Age | 22.86 | 21.89 |
| Gender | | |
| Male | 82.5 | 95.7 |
| Female | 17.5 | 4.3 |
| Race | | |
| Asian | 0.0 | 0.0 |
| Indian | 1.1 | 0.0 |
| Hispanic | 89.8 | 100.0 |
| White | 6.8 | 0.0 |
| Black | 1.7 | 0.0 |
| Other | 0.6 | 0.0 |
| Entered network as: | | |
| Gang member | 41.2 | * 83.3 |
| Gang associate | 51.3 | 16.7 |
| Associate of associate | 7.5 | 0.0 |
| Number of Arrests | | |
| Part 1 Violent | 0.49 | * 0.92 |
| Part 1 Property | 0.96 | 1.17 |
| Part 2 Drug | 0.93 | 1.33 |
| Part 2 Sex | 0.04 | 0.00 |
| Part 2 Misc | 3.57 | * 5.33 |
| Total number of arrests | 5.29 | * 7.79 |

Operation Jenga: Proof of concept for the PPD

What do we need to know at this point?

- How hard is it to get to the data?
- Does it produce ties we otherwise would not know of?
- Do subgroups exist?
- What is the structure of the network?
- What roles do members in the network play?

How we started Operation Jenga

- Step 0, Anna Bella, recommended by a couple of detectives.
 - Suspect of money laundering
 - Has been kidnapped twice (\$1M each)
 - Ex-husband was in a Mexican Cartel
 - She owns several check cashing businesses and a tax service, notary business.
- Data we used:
 - FI's
 - DR's
 - PPD only
- Went back 3 years
- 2 steps

1 lead resulted in 320 people

Person X
5S/2IL/1W

Ms. A.
(Start of network), 1V/1W,
Kidnapped 2x, \$1M ransom

Mr. P, 1S/2IL

Mr. F
(unknown), 4A/3V

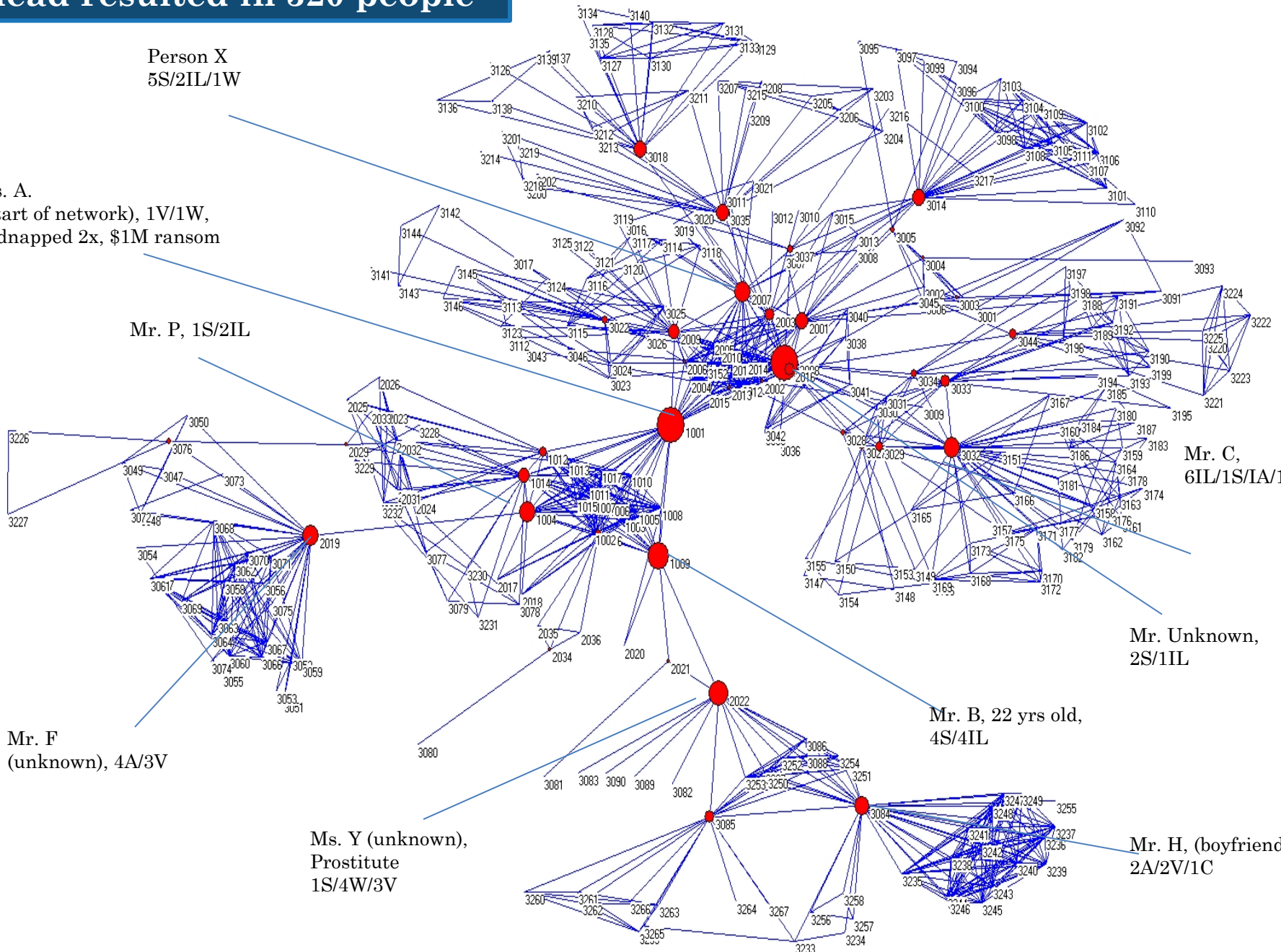
Ms. Y (unknown),
Prostitute
1S/4W/3V

Mr. C,
6IL/1S/1A/1C

Mr. Unknown,
2S/1IL

Mr. B, 22 yrs old,
4S/4IL

Mr. H, (boyfriend),
2A/2V/1C



What else do we now know?

- SNA can work with PPD data
- Labor intensive data collection
- One lead resulted in 320 relationships after 2 steps
- 50% of network reachable through 2 people.
- Key players are not necessarily the most criminally involved

Challenges in using police data

- Labor-intensive
 - Manual look-up and verification of individuals
 - Manual build of edge-lists
- Quality concerns
 - CAD/RMS systems without a reliable unique identifier (with look-up capability) for every individual in the system contributes to errors in both inclusion and exclusion of individuals in the network
- Timeliness
 - Manual processes reduce tactical utility
 - CAD/RMS able to automate edge-list builds would provide near real-time analyses.

So what? What are we going to do with this stuff?

- Degree centrality- number of ties a node has in the network
 - Not the most strategic targets
 - Could be important in collecting information on a network (e.g., informants, free talks, etc.)
- Betweenness centrality - intersection of many paths between others.
 - Strategic targets for disrupting a network
 - Ideal contagion agents for a deterrence message
 - Call-ins would be best directed at these individuals
- Eigenvector centrality - connected to many connected people
 - Collective accountability
 - Pulling levers strategy

Doe, Jane

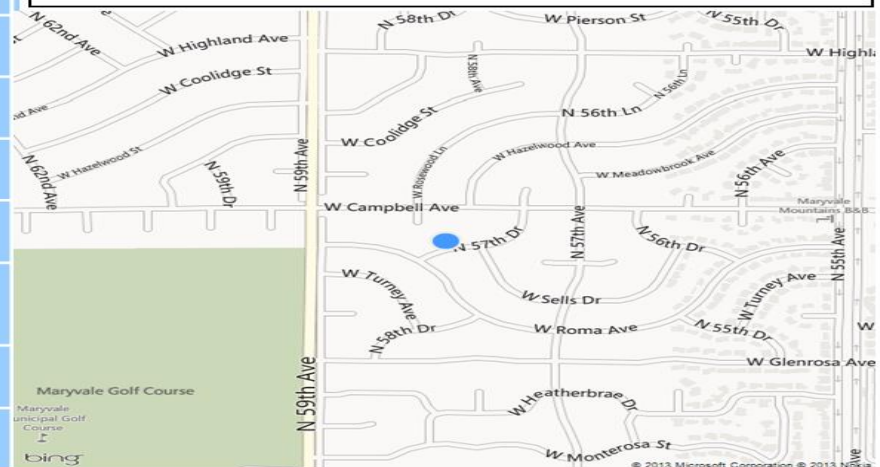
Network ID#: 2034**GLENDALE SPI—VIOLENT NETWORK INTEL SHEET**

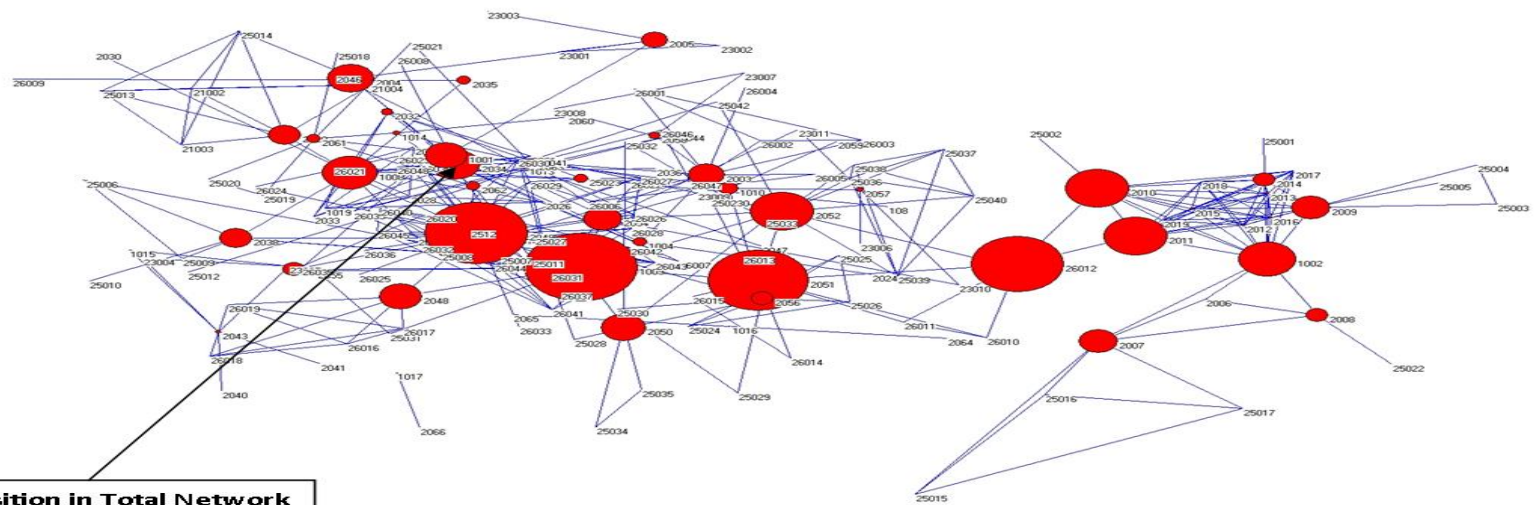
| | |
|---------------------|---|
| AGE: | 29 |
| DOB: | 00/00/1900 |
| SSN: | 999-99-9999 |
| SEX: | FEMALE |
| RACE: | Hispanic/Latina |
| HEIGHT: | 506 |
| WEIGHT: | 120 |
| ADDRESS: | 4412 N ?? ave; Phoenix, AZ 85099 [STABLE—PARENT] |
| EMPLOYER: | Unemployed |
| EMPLOYER'S ADDRESS: | NA |
| GANG: | NA |
| CUSTODY STATUS: | NONE—Released ADC 99/99/2099 |
| PROBATION STATUS: | NONE |
| PROBATION OFFICER | |
| PO CONTACT | |
| RECENT DR#s | |

| | |
|-----------|---|
| 11-119999 | INVOLVE TYPE: S (BO) DESCRIPTION: Misd Warrant |
| 11-9999 | INVOLVE TYPE: S (BO) DESCRIPTION: Misd Warrant |
| 10-9999 | INVOLVE TYPE: S (BO) DESCRIPTION: DV-Assault |
| 10-9999 | INVOLVE TYPE: S (BO) DESCRIPTION: Prohibitive Camping |
| 10-999 | INVOLVE TYPE: FI DESCRIPTION: Suspected selling of drugs |
| 10-99999 | INVOLVE TYPE: S DESCRIPTION: Poss. of Marijuana |
| 10-9999 | INVOLVE TYPE: FI DESCRIPTION: Loitering |

**Other information:**

FBI # : 8abcsde-999999





| ASSOCIATE | DR# | INVOLVEMENT TYPE | DR# | INVOLVEMENT TYPE |
|--------------------|------|---------------------|------|---------------------|
| 1008-MICHAEL | 12-1 | FI | 12-8 | FI |
| | 12-7 | FI | 12-9 | FI |
| | 12-6 | VICTIM | 13-3 | IL |
| | 13-2 | FI | 12-9 | SUSPECT |
| | 12-7 | VICTIM | 12-1 | FI |
| | 12-7 | VICTIM | 12-9 | SUSPECT |
| | 13-5 | FI | 12-3 | FI |
| | 12-1 | FI | 12-4 | SUSPECT |
| | 13-2 | SUSPECT | 10-8 | FI |
| | 13-2 | VICTIM | 12-6 | FI |
| 2004-DANIEL | 12-2 | FI | | |
| 2031- JAMES EDWARD | NONE | | | |
| 2032- NICOLE | NONE | | | |
| 2033- BOB | 13-4 | FI | | |
| 26008- KEN | NONE | | | |

Stop & Think

- If you found these same findings in your community what would you do?



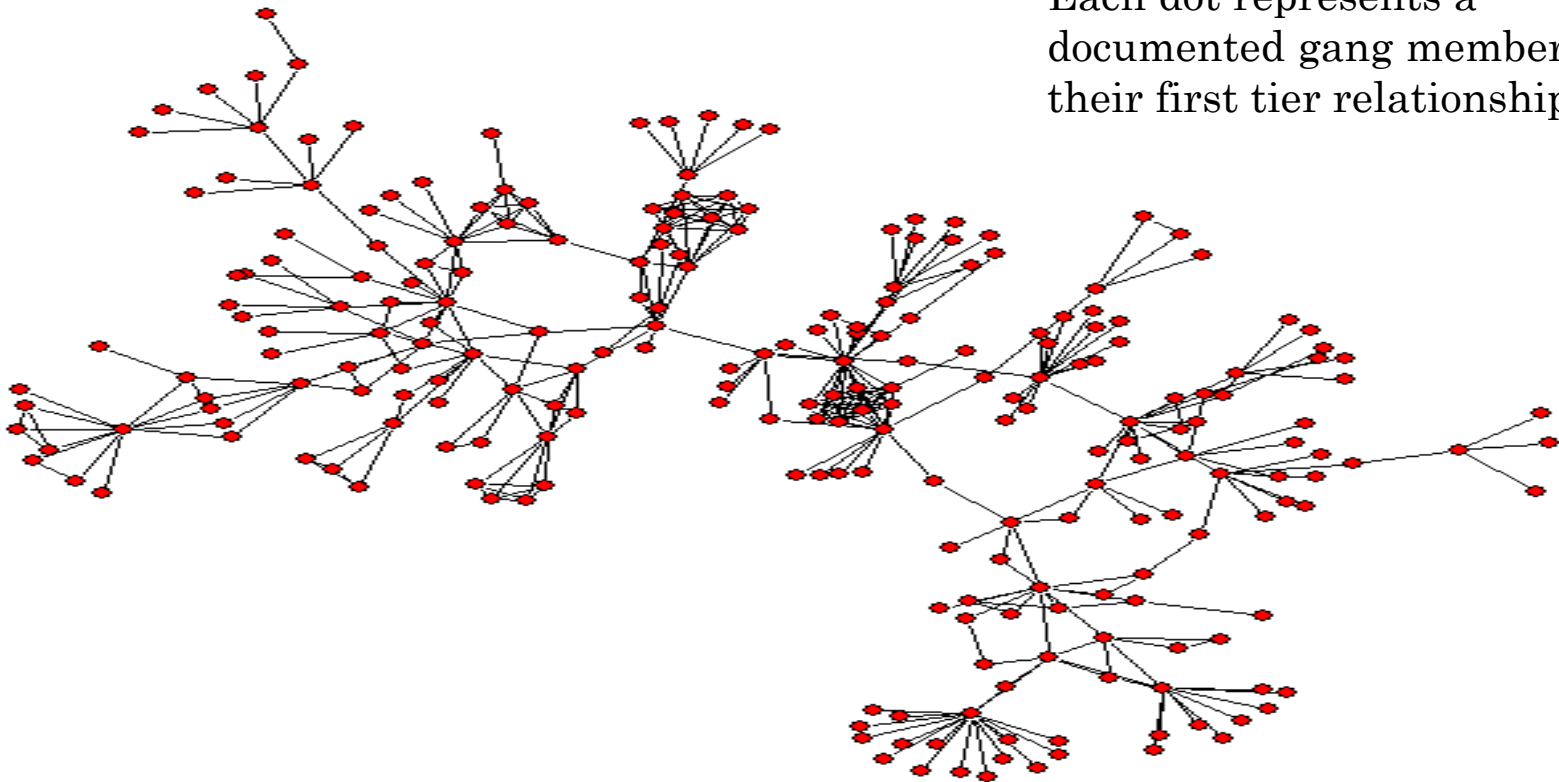


KC NoVA

Kansas City No Violence Alliance

Early Offender Network Model

Each dot represents a documented gang member and their first tier relationships

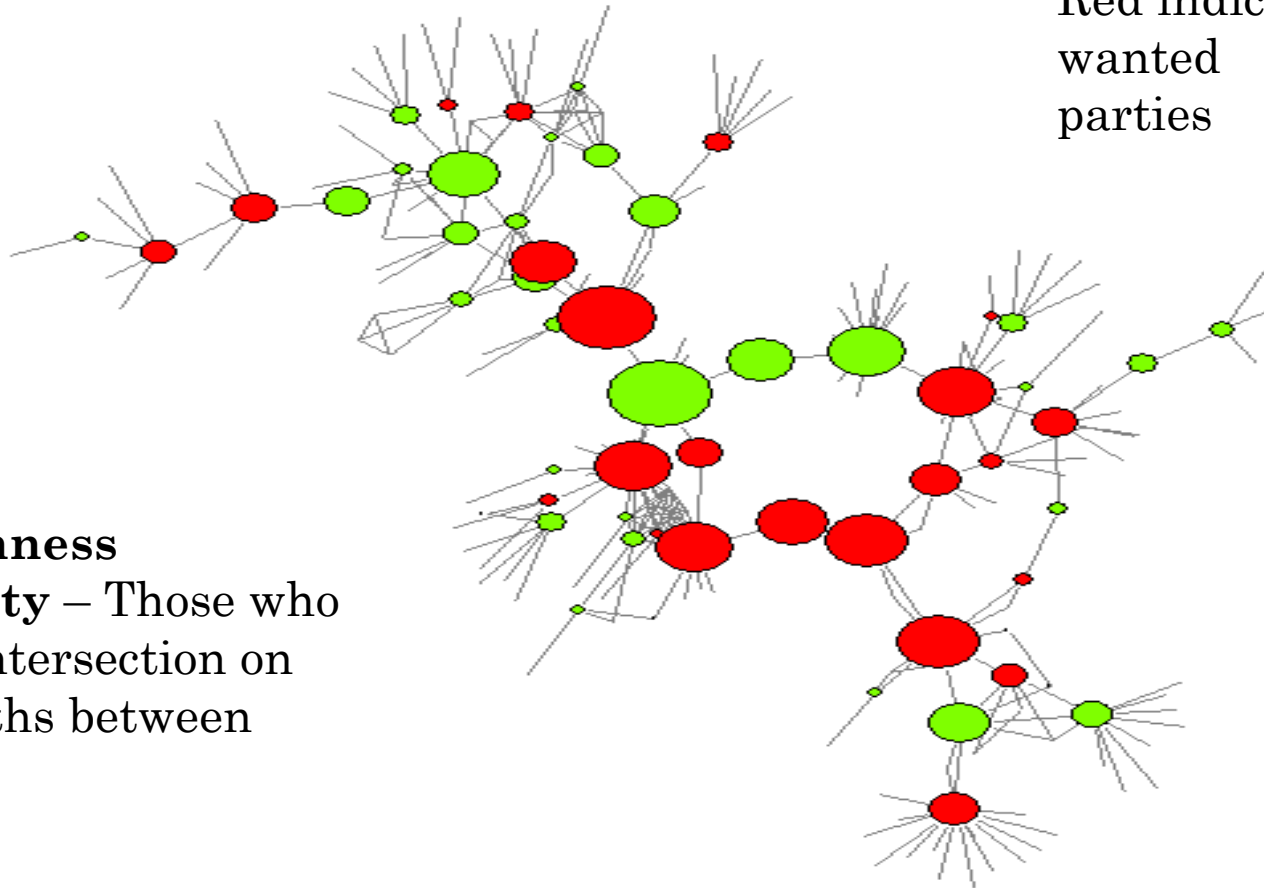


Early offender network

- 360 members in group
- 202 in largest connected group
- 60 currently were on probation / parole
- 32 pending cases were in Jackson County processes
- 126 members had active warrants
- 22 warrants were Felony

Pilot Group High Betweenness and Active Warrants

Red indicates
wanted
parties



**Betweenness
Centrality** – Those who
are the intersection on
many paths between
others

Training

- Finding the right crime analysts
- Giving them time and space to learn
- Need to fully understand PD data systems and how to extract large amounts of data from those systems
- Need to understand the concepts, not just the technique

Analysis

- Software
 - Pajek: Free, Windows-based
 - UCInet: Free, Windows-based
- Resources
 - Wasserman & Faust (1994), *Social Network Analysis: Methods and Applications*
 - Training seminars
 - Local university
 - CNA

Questions, comments?

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