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Locating spatial analyses of crime: the crime analysis framework

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8th International Investigative Psychology Conference, The Keyworth Centre, London, 15th-16th December 2005 Locating Spatial Analyses of Crime: The Crime Analysis Framework

Professor Alex Hirschfield, HonMFPH
Professor of Criminology and Director
International Centre for Applied Criminology
ICAC

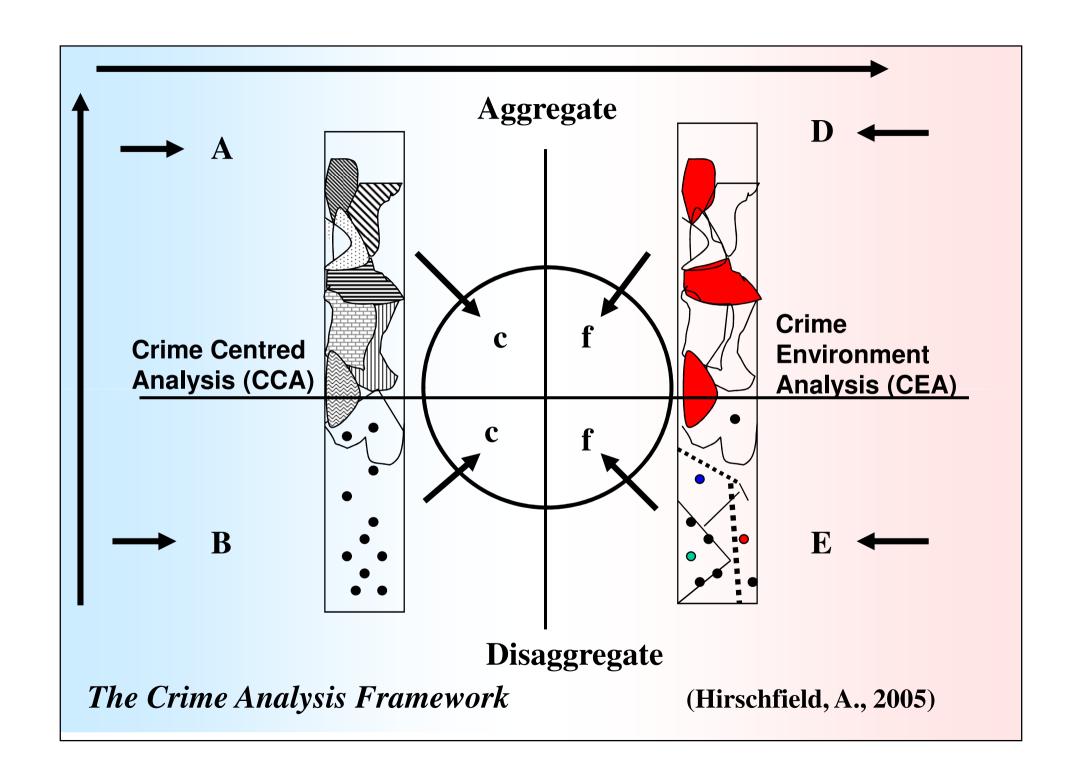
University of Huddersfield, Floor 14, CSB, Queensgate, Huddersfield, UK HD1 3DH

E-mail: a.hirschfield@hud.ac.uk

Why Map & Analyse Crime Data?

- to identify the scale and distribution of crime and disorder
- to explore relationships between crime and the environment (physical & social)
- to target resources for crime prevention
- to evaluate the impact of crime prevention

- to inform police operations
- to apprehend offenders
- to profile the spatial behaviour of offenders.
- to predict the spatial and temporal distribution of offences
- to develop Early Warning Systems of emerging problems
- >to communicate with and to engage communities
- > to support bids for extra resources from government



Crime Centred Analysis I

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Where do crimes occur?
When do crimes occur?
When crimes occur, where do they occur?
Where crimes occur, when do they occur?
How do crimes occur (MO analysis)
Do areas with one crime problem have other crime
  problems?
Where are these areas?
Which and how many crimes do they have?
How much of the population is affected (prevalence)?
How concentrated is crime (socially, temporally, over
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space)?

Crime Centred Analysis II

To what extent are there repeat crimes?

What is the time interval between repeats?

Where are repeat crimes concentrated?

Who are the victims? Who are the offenders?

Do offenders live in the areas with the highest crime rates?

Do offence locations relate to those of previous offences?

Is the volume of crime decreasing or increasing?

Are crimes affecting the same areas or new areas?

Are crimes diffusing or concentrating?

Is there evidence of displacement or crime switch?

Crime Environment Analysis I



Physical & Built Environment

Land use, Terrain, Urban Design, Communications

Social Environment

Migration, ethnicity, deprivation, social cohesion

Policy Environment

Target Hardening, CCTV,

Alley-gates, Street Wardens, Home watch, other ABIs

Crime Environment Analysis II

What types of area have high crime?

Are they student areas or deprived estates?

Do they have particular types of housing /built environment?

Are they Policy Priority Areas?

What types of transport and communications do they have?

Are they accessible to offenders physically/ socially?

Do they have poor natural surveillance?

Do they have a large number of potential crime attractors?

Do they have crime prevention measures already?

Are they deployed in the right places at the right times?

How does the crime prevention relate to crime change?

Crime Centred Analysis(CCA)

Techniques for Aggregate CCAs

- Tabulation of crime counts and derivation of crime rates
- Identification of areas with significantly high and significantly low crime
- Calculation of the concentration of crime at area level
- Identification of crime mix and its variation across areas

Distinguishing High and Low Crime Rates

Ward	Households		Theft of	Theft From
			Vehicle	Vehicle
1	2,752	35.6	10.9	37.0
2	1,459	21.2	12.3	30.1
2	2,366	29,5	15.2	44.8
4	2,394	12.9	6.2	30.4
5	2,284	52.1	10.5	30.6
6	2,149	14.4	19.5	101.4
7	2,839	24.6	13.7	31.7
8	2,509	25.1	19.5	56.6
9	2,876	47.2	19.8	70.3
Mean		25.1	12.0	40.4
SD		12.2	5.2	20.0

Malicious Ignition Dwelling Fires 1998/99 Resource Targeting Table (RRT)

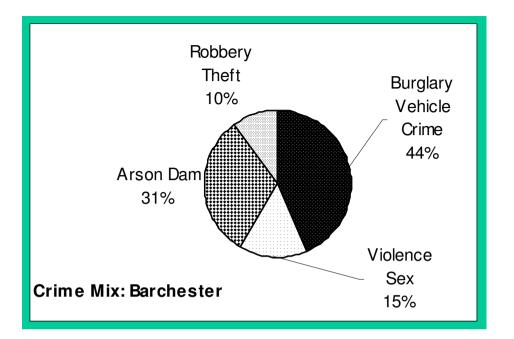
	Ward Code & Name	Cum %	Cum %	No. of	Cum %
		Pop	Hhlds	Incidents98/99	Incidents98/99
1	BNFG Bradford	0.46	0.49	47	4.00
2	BNFE Beswick and Clayton	0.94	0.98	39	7.33
3	BNFD Benchill	1.45	1.46	37	10.48
4	BNFM Cheetham	2.01	2.00	37	13.63
5	BNFY Lightbowne	2.54	2.55	36	16.70
6	BNFZ Longsight	3.14	3.11	34	19.59
7	BRFJ Langworthy	3.55	3.57	32	22.32
8	BRFC Broughton	3.96	4.00	31	24.96
9	BNFK Central	4.32	4.43	30	27.51
10	BNFF Blackley	4.80	4.94	25	29.64
11	BNFU Harpurhey	5.27	5.45	25	31.77
12	BPFW Werneth	5.73	5.86	23	33.73
13	BRFK Little Hulton	6.19	6.34	23	35.69
14	BNFB Baguley	6.68	6.84	20	37.39
15	BQFD Central and Falinge	7.11	7.28	20	39.10
16	BRFL Ordsall	7.39	7.60	20	40.80
17	BPFR St.Marys	7.88	8.05	18	42.33
18	BQFP Middleton West	8.16	8.33	18	43.87
19	BNGA Moss Side	8.68	8.88	17	45.32
20	BPFJ Hollinwood	9.08	9.28	17	46.76
21	BNFA Ardwick	9.47	9.69	15	48.04
22	BNGC Newton Heath	10.00	10.23	15	49.32

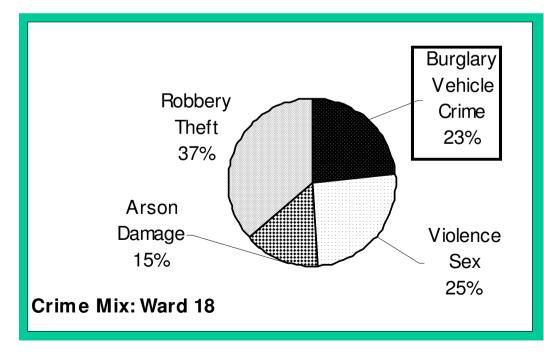
 \leftarrow 25% of Incidents

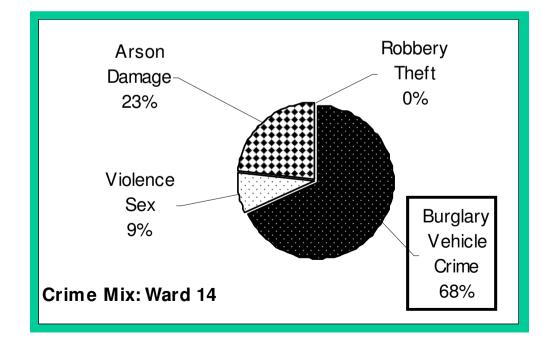
 \leftarrow 50% of Incidents

CCA:

Crime Mix





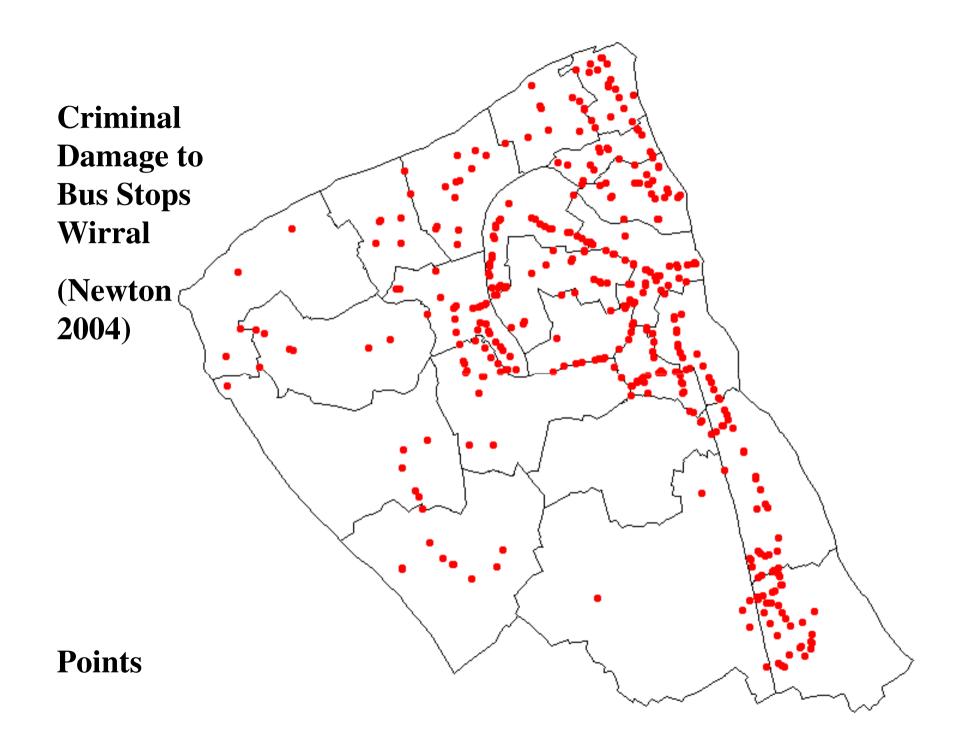


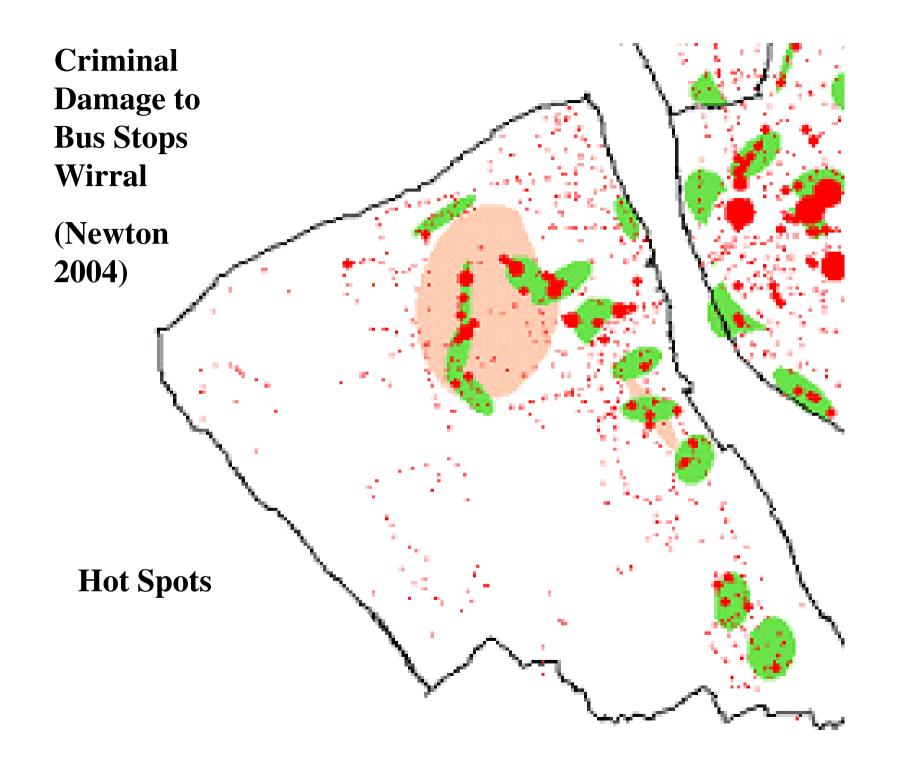
Ward	Burglary		Burglary		Burglary	
	Prevalence		Concentration		Prominence	
<i>17</i>	52.1	(1)	8.8	(2)	20.3	(1)
21	47.2	(2)	10.0	(1)	11.6	(14)
1	42.2	(3)	7.3	(4)	12.0	(12)
8	35.6	(4)	6.5	(7)	13.6	(8)
13	35.6	(5)	7.2	(5)	14.4	(7)
9	34.1	(6)	8.5	(3)	18.9	(3)
7	33.1	(7)	6.6	(6)	12.2	(11)
15	26.5	(8)	5.2	(9)	10.7	(17)
2	25.9	(9)	4.2	(11)	16.3	(5)
20	25.1	(10)	4.6	(10)	7.6	(20)
19	24.6	(11)	5.2	(9)	11.3	(16)
14	21.2	(12)	2.3	(16)	19.0	(2)

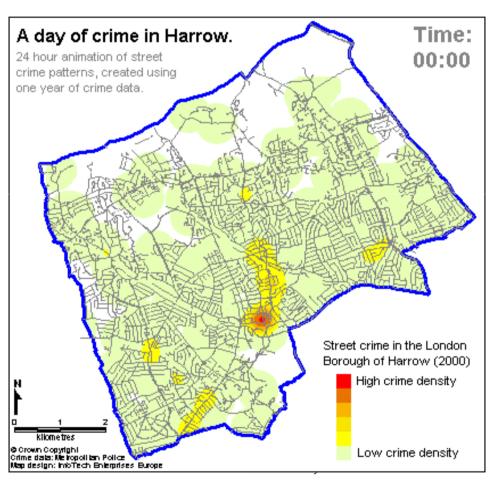
CCA Mapping Techniques

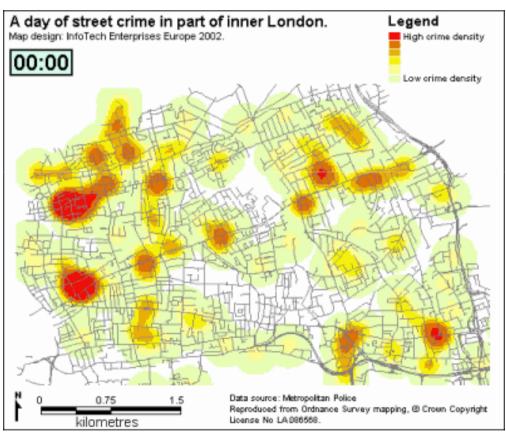
Disaggregate Data Analyses

- Mapping the distribution of individual incidents (offence, victim, offender locations);
- Mapping the distribution of repeat incidents (multiple incidents, repeat victims, prolific offenders)
- Identifying clusters / hot spots' from points
- Exploring space-time clustering









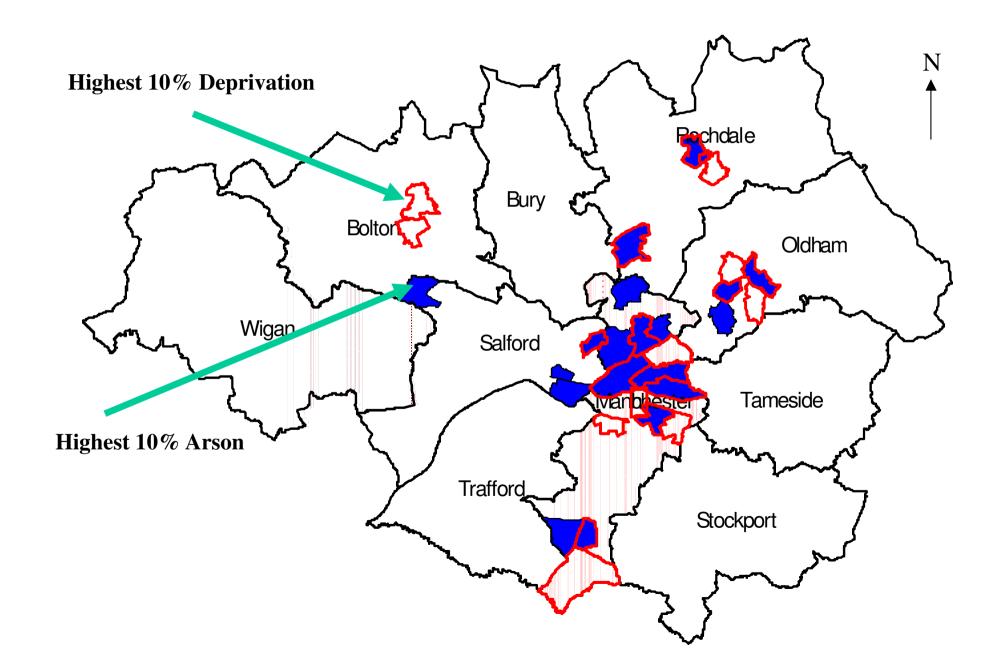
(Chainey, 2001)

Crime Environment Analysis(CEA)

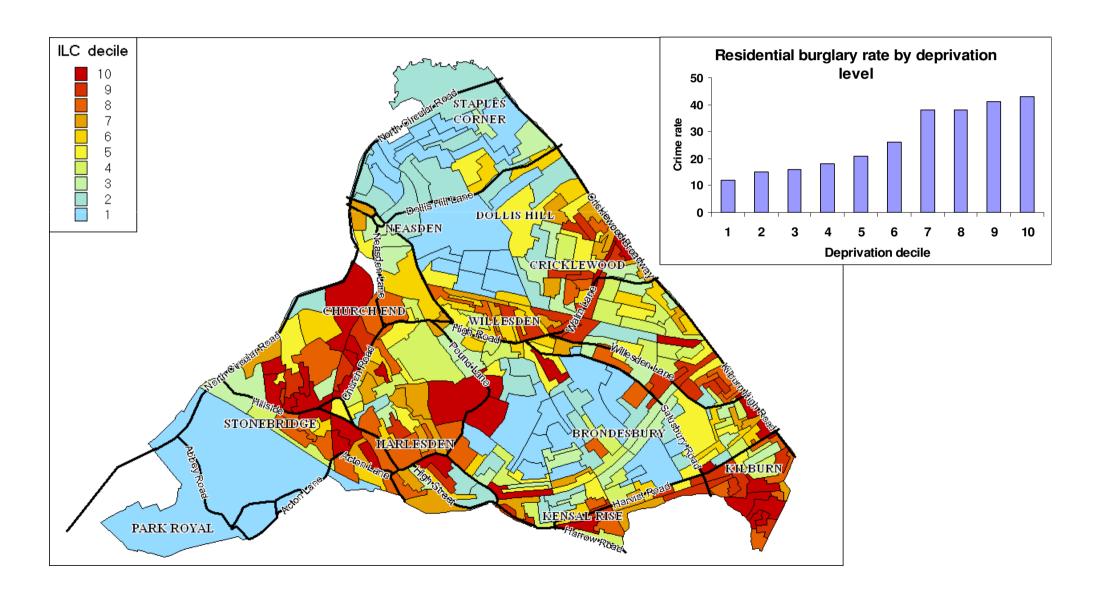
Techniques for Aggregate CEAs

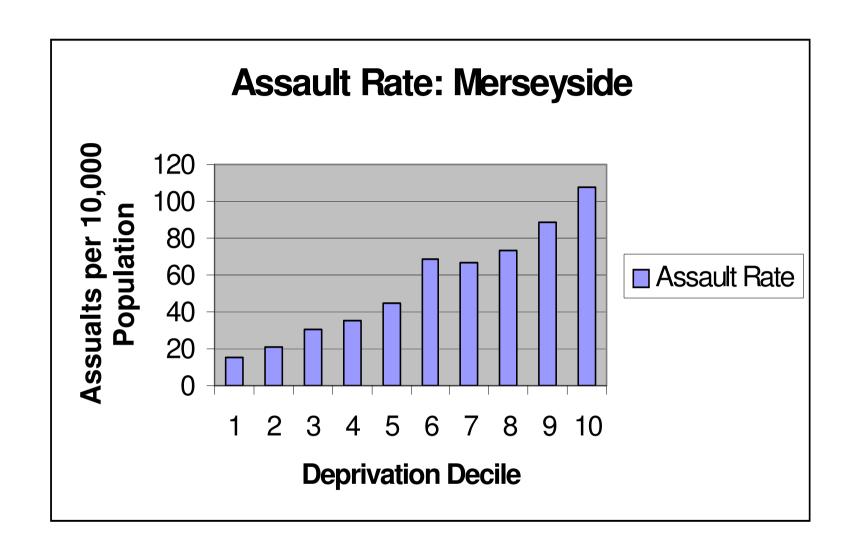
- Derivation of crime rates for areas ranked by deprivation level
- Derivation of crime rates for different types of residential neighbourhood
- Identification of overlap between high crime and high values on other social indicators (e.g. unemployment)
- Calculation of the concentration of crime by area type, social indicator

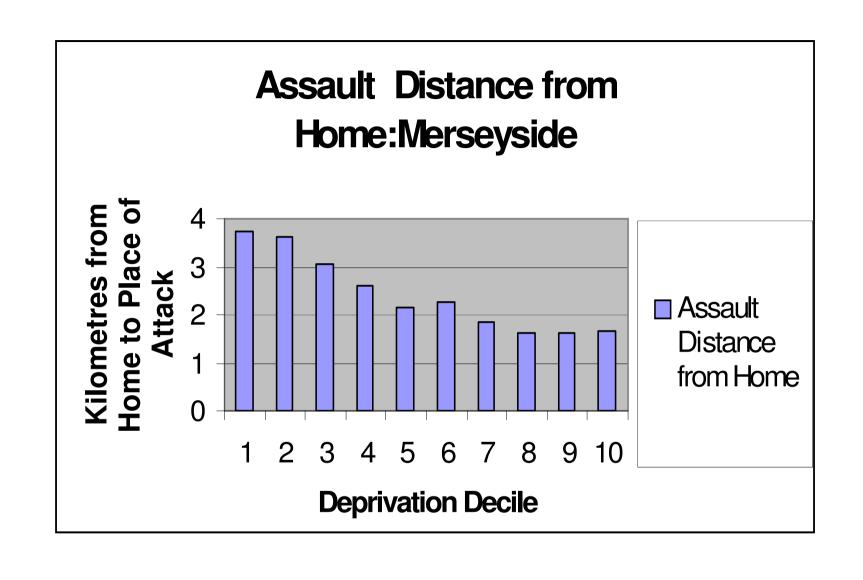
HIGHEST ARSON & HIGHEST DEPRIVATION



Mapping crime with deprivation







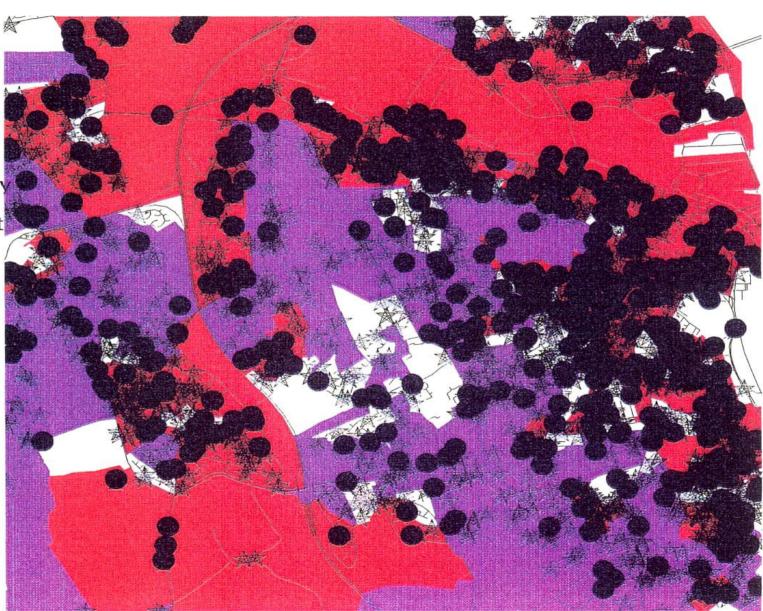
CEA Mapping Techniques

Disaggregate Data Analyses

- Mapping incidents on contextual backcloths (Geodemographics, land use maps, digital aerial photos)
- Mapping hot spots and spatial-temporal clusters in relation to the environment
- Identifying 'hot spot' demographics & land use
- Conducting specific site and RADIAL analyses

Offences in Oxton

Wmars
Mdist
Recorded Burglary
Recorded Assault
Affluent
Disadvantaged



Offenders in Oxton

Wm

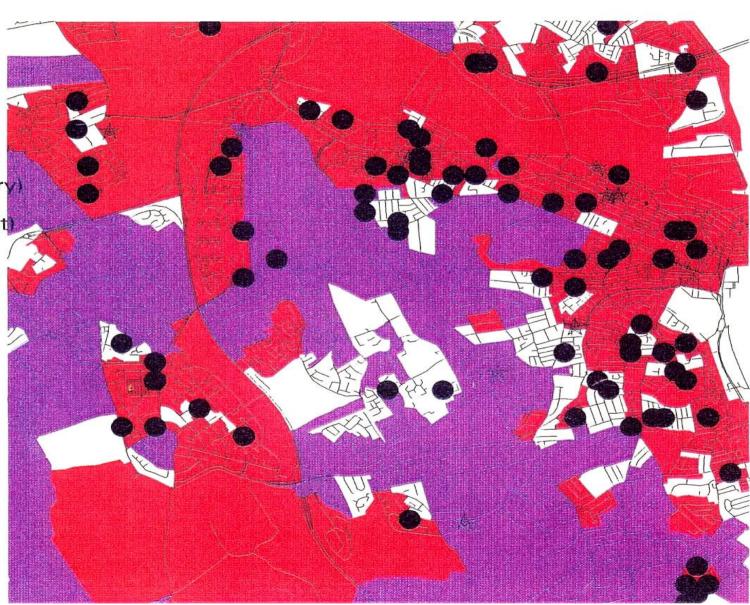
Mdi: t

Offende Buig arv

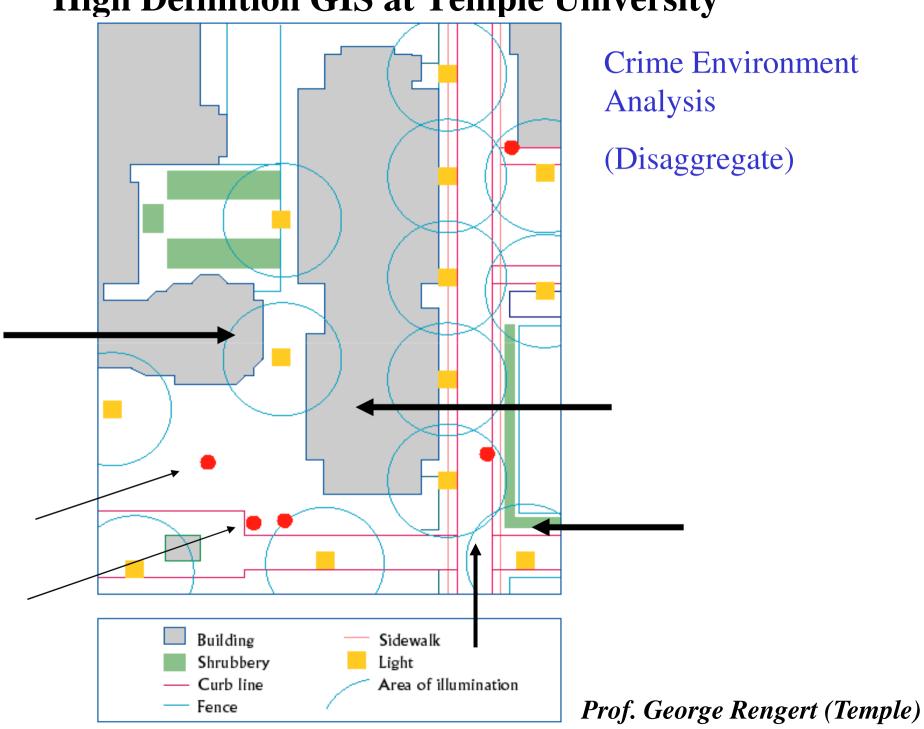
Offende Assault

A:fluent

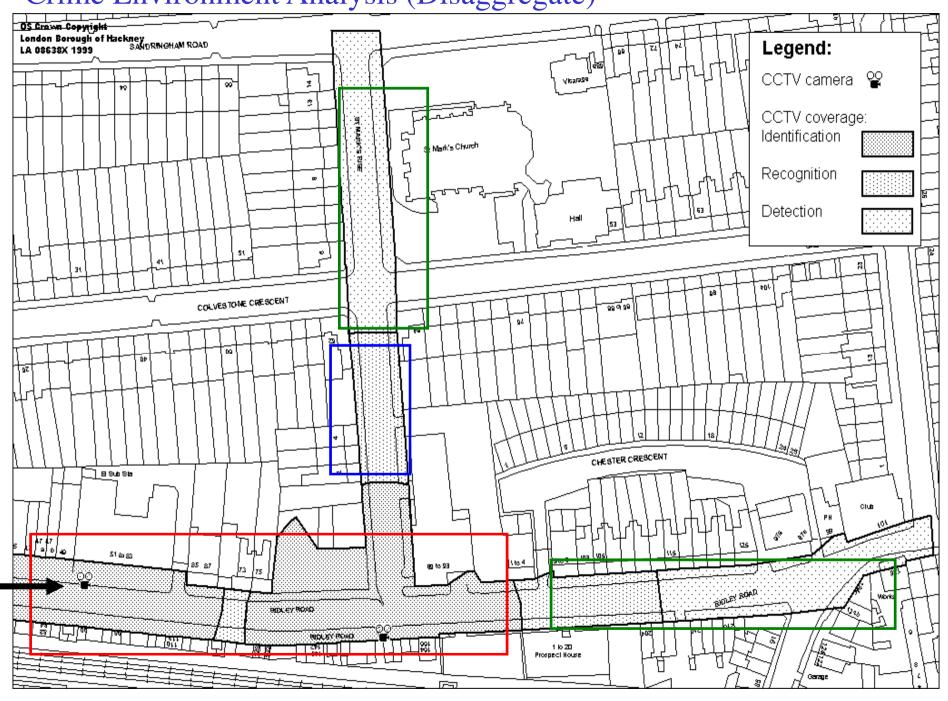
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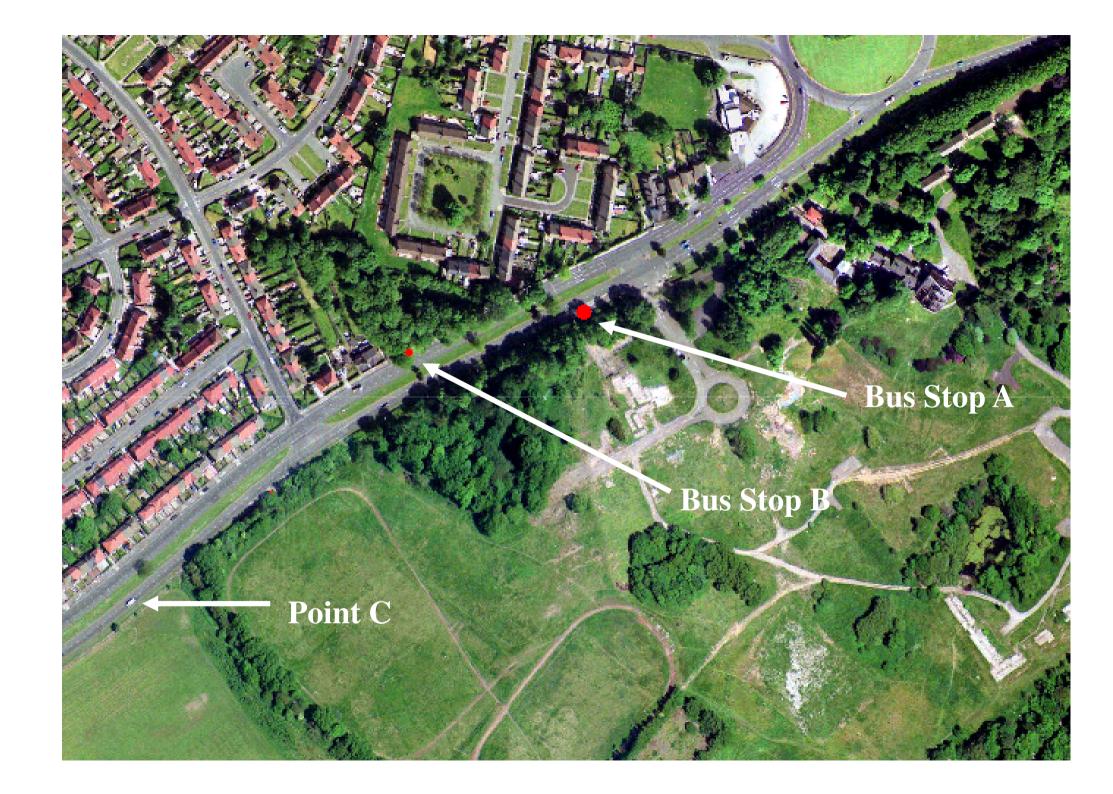
High Definition GIS at Temple University

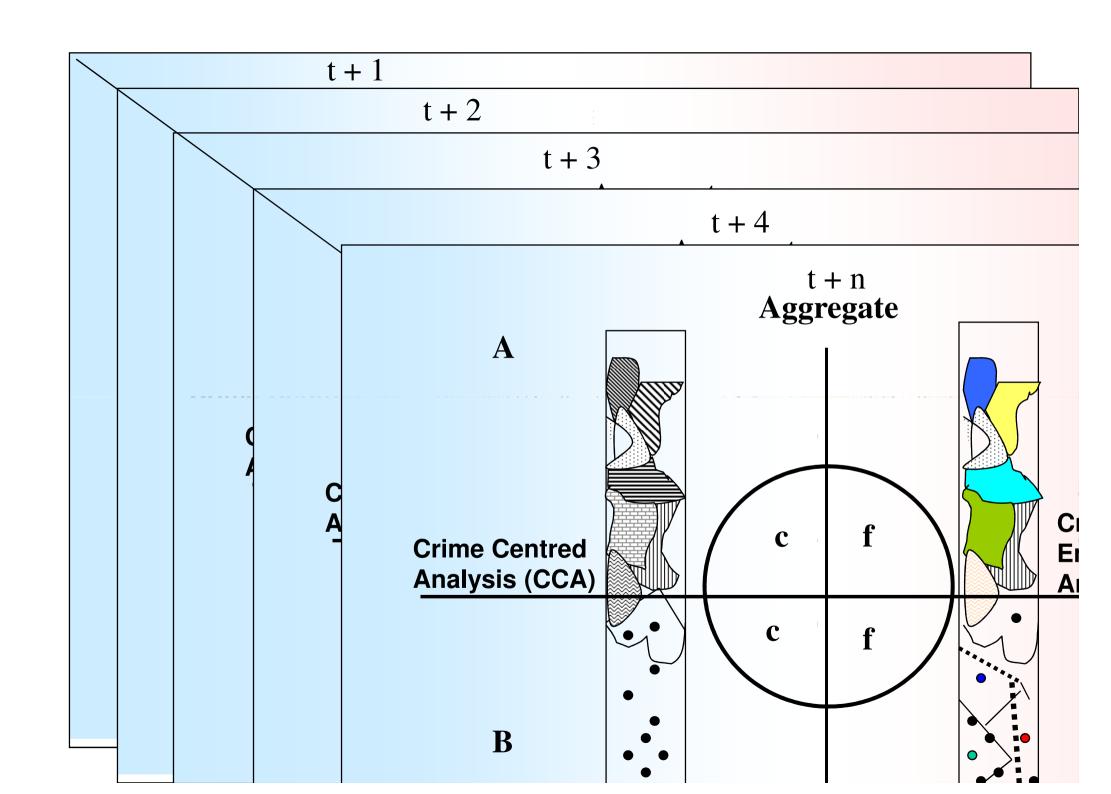


Crime Environment Analysis (Disaggregate)



Spencer Chainey (Jill Dando Institute, UCL, London)





Conclusion

- Much can be gained solely through CCAs
- **CEAs** add further insights by identifying factors that facilitate/inhibit crime (e.g. low/ high social cohesion, good/poor natural surveillance)

Both CCA and CEA require:

- Awareness of sources of data on crime, disorder, land use and socio-demographic conditions
- Expertise in data manipulation and processing
- Basic skills in data analysis
- Competence in the use of GIS
- An ability to interpret the results from analysis