A New Software Prototype for Geographic Profiling

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The Geographic Profiling Problem

- The geographic profiling problem is to estimate the location of the home base of a serial criminal from the known locations of the elements of the offender's crimes.
 - The home base is also called the anchor point of the offender. It may be the offenders home, the home of a relative, a place of work, or even a favorite bar.
- We have developed a new tool for the geographic profiling problem.
 - It is free for download and use, and is entirely open source.
 - http://pages.towson.edu/moleary/Profiler.html
 - The tool is still under very active development, and should be considered a prototype.
 - Effectiveness and accuracy testing has only just begun.

How Does it Work: The Mathematics

- We start by constructing a mathematical model for offender behavior; this gives $P(\mathbf{x}|\mathbf{z})$, the probability that an offender with anchor point \mathbf{z} offends at location \mathbf{x} .
- We assume that the probability that an offender commits a crime is:
 - Proportional to a distance decay function of the distance between the anchor point and the crime location, and
 - Proportional to the number of past crimes that have occurred at or near that location.
- We assume that crime sites are selected independently of one another
- We suppose that different offenders have different average distances they travel to offend.

Distance Decay

- We assume that the distance decay can be modeled by a two-dimensional normal distribution, whose corresponding one-dimensional model is Rayleigh.
- We assume that each offender has an average offense distance α, but that it varies between offenders.
- There is evidence to support this model.
 - Under these assumptions, the ratio of the distance to the crime site divided by the offender's average offense distance should follow a Rayleigh distribution with shape parameter equal to 1.
 - We have calculated this ratio for residential burglaries in Baltimore County. Once commuters are excluded, the Rayleigh distribution is a close match for the observed data.

Distance Decay



How Does it Work: The Mathematics

- $\bullet~$ Our assumptions mean that $\mathsf{P}(\mathbf{x}|\mathbf{z}) \propto \mathsf{D}(d(\mathbf{x},\mathbf{z}),\alpha)\mathsf{G}(\mathbf{x}).$
- Bayes' Theorem then let's us calculate the probability $P(\mathbf{z}|\mathbf{x})$, the probability that an offender who has committed a crime at \mathbf{x} has anchor point at \mathbf{z} .
 - To do so, we need a prior estimate for the location of the offender's anchor point, before information from the crime location is used.
 - For this purpose, reasonable choices include the local population density or the distribution of offenders from other crimes.
 - We also need to estimate the offender's average offense distance.
 - We again use Bayes' Theorem.
 - For the prior, we use the distribution of average offense distances for offenders, as calculated from historical data.
 - For the posterior, we use the distances between the crime locations.
- Provided we assume that crime sites are selected independently, we can extend this to series, and estimate the probability $P(\mathbf{z}|\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_n)$.

How Does it Work: The GUI

• We will use data obtained from Ned Levine's Crime Stat program:



How Does it Work: The GUI

• Let's try the tool...

Results: Population Density



Results: Target Density



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Software Prototype

Results: Coarse Results



Results: Fine Results



Results: Offender Density



Results: Offender Density



Results: Fine Results



How Does it Work: The Code

- There are two programs:
 - The graphical user interface
 - Written in C++ using the Qt Toolkit (4.7) and MinGW GCC (4.5.2)
 - Requires the Shapefile C Library (1.2) http://shapelib.maptools.org/
 - The source code is available
 - It is commented, but the documentation is not yet complete.
 - The analysis engine
 - Written in C++ with the MinGW GCC (4.5.2)
 - Requires the Shapefile C Library (1.2) http://shapelib.maptools.org/
 - Requires the Lapack++ Library (2.5.1)
 - http://lapackpp.sourceforge.net/
 - The source code is available
 - It is fully commented, and the documentation is nearly complete
 - Is scriptable, and can be called from programs other than the graphical user interface

How Does it Work: Scripting

🔐 \\homeshare\moleary\$\Documents\Geographic Profiling\Software\Effectiveness Testing, Summer 2011\Scripts\RunProfiler.py - Notepad++			
File Edit Search View Encoding Language Settings Macro Run TextFX Plugins Window ? X			
2011-0-10-10 Courses.ntml E hunProfiler.py			
1	imp	port multiprocessing	<u>^</u>
2	imp	port os	
3	imp	port subprocess	
4	-		_
	Eder	st ([])	
7		print instance (:03) with pid () (:10 mac(1,05.gecpid())	
8		<pre>budget_budget = billeria/mike/Programs/Edimse/Profiler/Release/Profiler.exe'</pre>	
9			
10		parameter = output path + "/{:03}/data/Parameters.txt".format(i)	
11		output file name = output path + "/{:03}/data/output.txt".format(i)	
12		<pre>output_file = open(output_file_name,'w')</pre>	
13			
14		<pre>p = subprocess.Popen([program,parameter],stdout=output_file)</pre>	- E
15		p.wait()	
16	L	output_file.close()	
17			
18	Ŧ W	windows parallel processing protection	
19	HIL		
20		i We want to use all cores but 1 so that the system will remain responsive	
22		course = multiprocessing.cours()	
23	La la	if(cpu count > 1):	
24	Ŧ	cores = cpu count - 1	
25	¢.	else:	
2.6	-	cores = 1	
27			120
28		# Build the pool	
29		<pre>pool = multiprocessing.Fool(cores)</pre>	
30	1		
31	ť	for 1 in range (1,233):	
32		result = poor.appry_async(r,(r,))	
Python f	ile	length:955 lines:38 Ln:5 Col:10 Sel:0 Dos\Windows ANSI	INS
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Strengths

- Strengths of this approach include:
 - The ability to incorporate geographic distribution of crimes
 - The ability to account for variations in offender distance decay patterns
 - The use of use variations in population density or locations of prior offenders
 - A completely open mathematical model, with all assumptions about offender behavior made explicit
 - A completely open source code base, available for others to examine and modify

Weaknesses

- Weaknesses of this approach include:
 - Effectiveness testing is not yet complete.
 - It is unclear what are the best choices for various parameters and model components
 - The preliminary results do not show significant improvements over existing methods
 - Computation times can be excessive
 - Population data currently only uses 2000 US Census data- though the modifications to allow the use of 2010 data have begun