



#### Re-Shaping the Police Landscape: The Glacier of Forecasting



By Lawrence Sherman *Cambridge University* 5<sup>th</sup> EBP Conference July, 2012



## 2011 Landscape Changes: Sudden and Gradual

<u>Earthquakes</u>

Record Number: 202 = 6+ Richter <u>Glaciers Melting</u> (faster than expected)

- Japan
- New Zealand
- Myanmar (Burma)
- Turkey
- India-Nepal

- Greenland
- Himalayas
- Antarctica
- Andes
- Patagonia

#### Policing Landscape Changes: Sudden v. Gradual

Earthquakes Police killings  $\rightarrow$  Arab Spring

Elected Police Commissioners—UK

Budget Cuts—UK, US

<u>Glaciers Melting?</u> Slow, steady increase in **Better Evidence:** 

- 1. Forecasting
- 2. Focusing
- 3. Follow-Through

#### **Police Glaciers**

- Growth of Police Knowledge, Research
- Cambridge Police Executive Programme
- Society for Evidence-Based Policing
- More experiments in UK than ever before

# **Campbell Collaboration**

- <u>www.campbellcollaboration.org</u>
- Crime and Justice Coordinating Group
- 29 Completed systematic reviews
- Hosted by Norway
- Swiss and Israeli co-chairs
- New Campbell Centre in China?
- Global basis for peer-reviewed evidence

#### 3 Kinds of Evidence

- 1. Prediction: risk and harm levels
- 2. Prevention: Cost-effectiveness evidence
- 3. Punishment: Managing resources

# Predicting Serious Harm not just minor crime

- From arrest to parole release
- Who is dangerous, who is not
- Restrain loss of liberty for the worst
- Use restoration, rehabilitation for most

#### 3 Ways to Predict Behaviour

- 1. Clinical (untested) assessments
- 2. Checklists validated by testing
- 3. Data mining validated by testing

# Testing Forecasts: What Actually Happens?

	Actually Happens	Actually does <i>NOT</i> Happen
Predicted to Happen	1. True Positive	2. False Positive
Predicted NOT To Happen	3. False Negative	4. True Negative

# From Clinical to Data Mining

#### Predicting Domestic Violence

Cambridge MSt Thesis

**Tested Predictions** 

All murder, attempted cases = false negatives

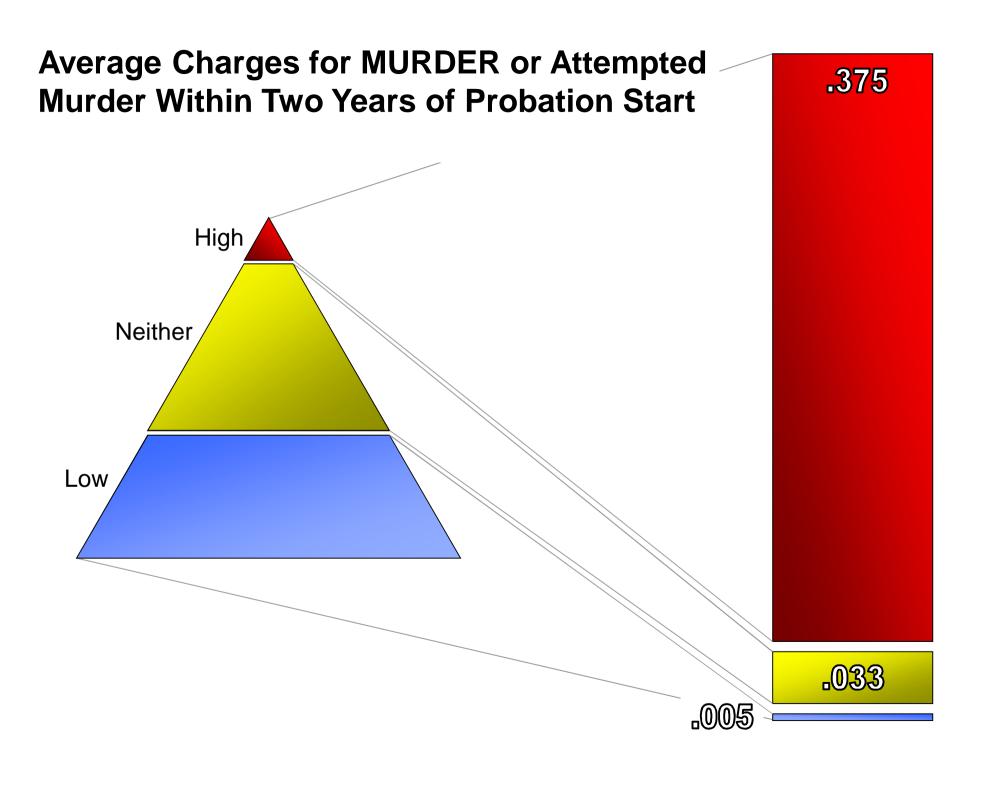
**Predictions missed** 

Chief Sara Thornton, Thames Valley



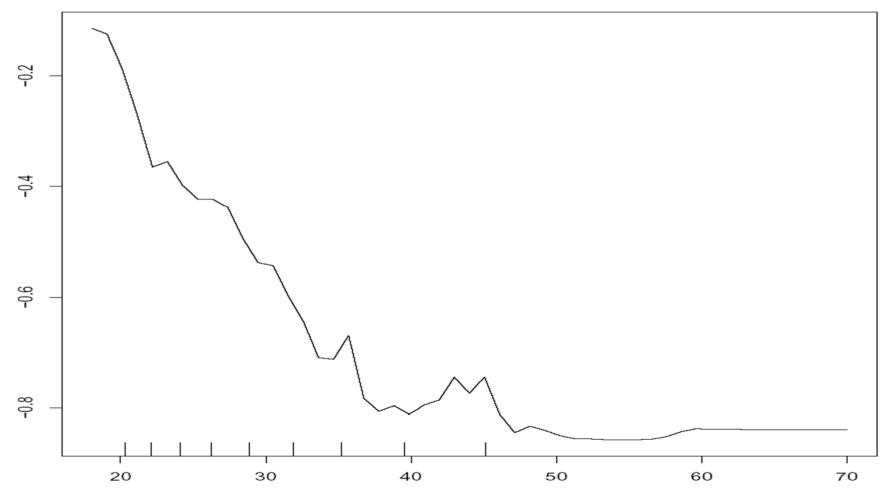
# **Forecasting Murder**

- JOURNAL OF THE ROYAL STATISTICAL SOCIETY
- Berk, Sherman, et al, 2009
- Philadelphia Probation Cases
- 300-400 murders per year
- 1.5 million population
- Rate = 14 X Scotland's



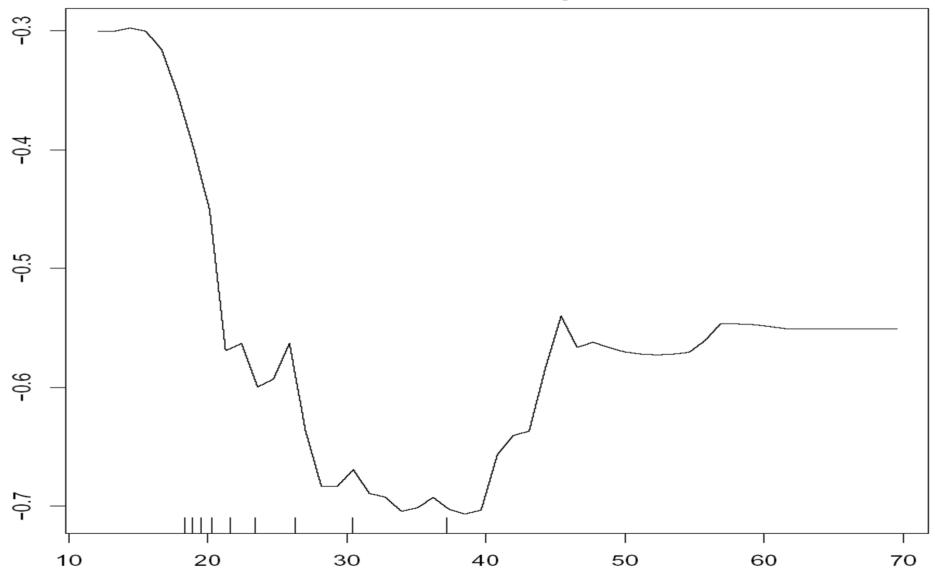
#### Risk of Murder by Age at Time of Crime (Phila)

Partial Dependence on age



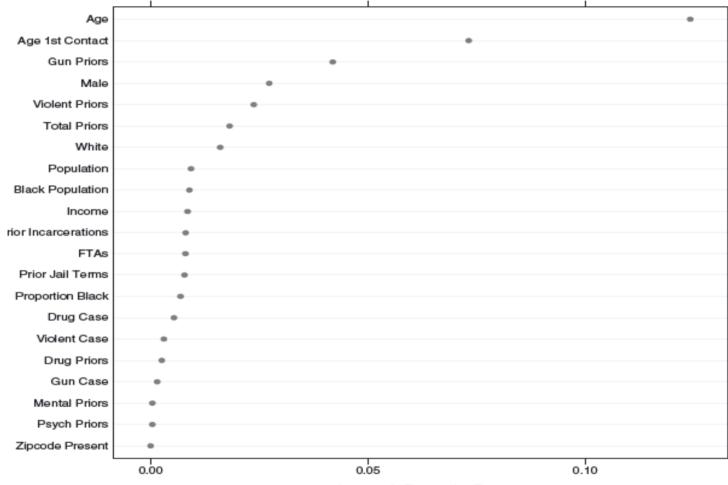
age

#### Risk of Future Murder By Age of First Adult Disposition





## Value of Each Murder Predictor



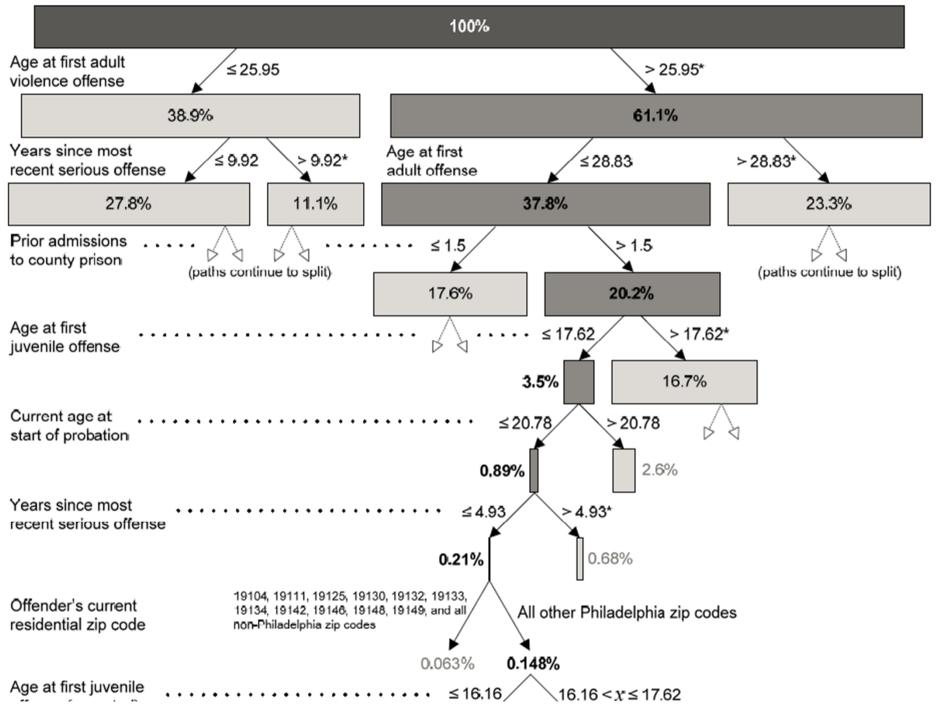


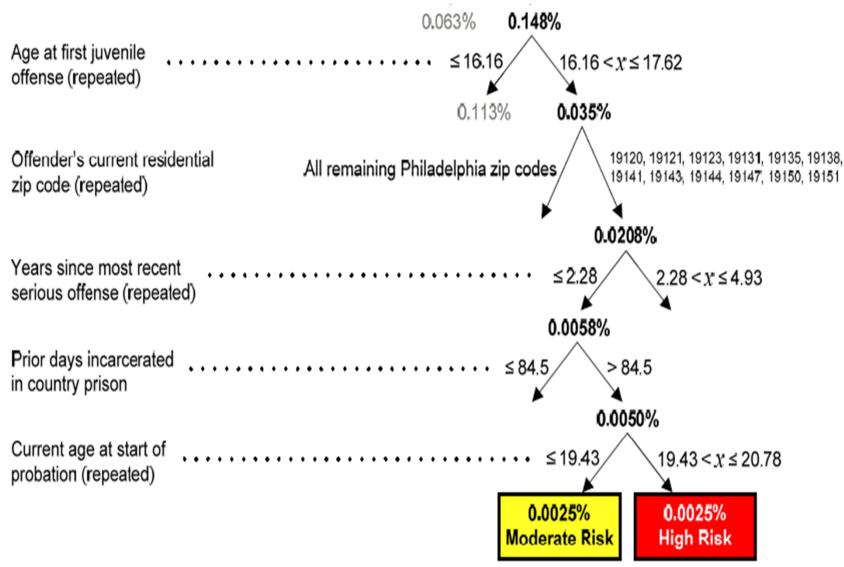
#### How Does This Work? Barnes and Hyatt, 2012

Table 1: Simplified confusion matrix for the most recent Philadelphia forecasting model (i.e., Model C), based on construction sample

		Actual		Actual			
		High	Non-High		Totals	Percent	
Forecast High Risk	A	7,112	B	11,700	18,812	15.7%	
Forecast Non-High	С	4,468	D	96,655	101,123	84.3%	
Totals		11,580		108,355	119,935		
Percent		9.7%		90.3%			

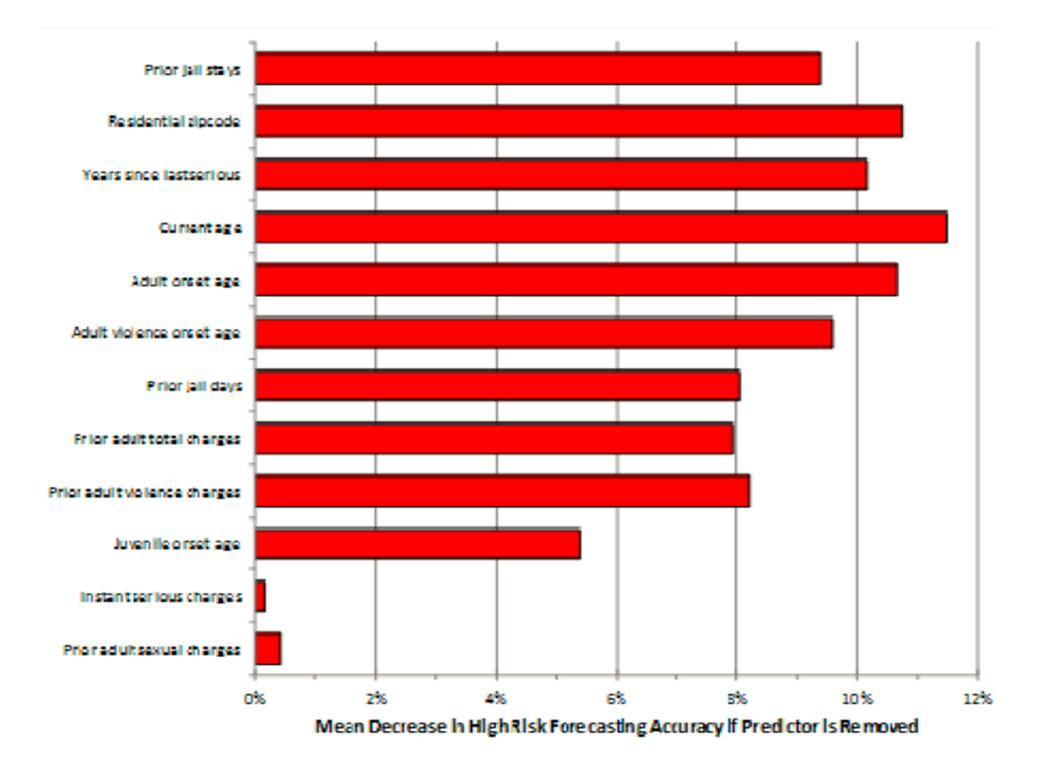
119,988 New Probation Case Starts, 2002-2007





\* including offenders with no prior record within the indicated offense category

Figure 1: An example of one path through one tree in Philadelphia's latest random forest model



# Announcing England & Wales

- 100,000 cases
- Convictions in year 2000
- Ten years of Followup
- Data Mining for Serious Offending
- Also for identifying low HARM cases
- Even with high risk of repeats (LOW harm)

# 2-year Definitions: England and Wales

Low Harm = NO new crimes over 2 years

#### Medium Harm = No SERIOUS crimes 2 yrs

High Harm = Violence, Sex crime, robbery

# In England, Who Is Safe to Caution?

NrPriorConvictions				<u> </u>				····o
Sex							· · · · · · · · · · · · · · · · · · ·	
AgeFirstCharge							· · · · · · · · · · · · · · · · · · ·	
InstantRobberyChargeCount							· · · · · 0 · · · · · ·	
InstantDrugChargeCount		O						
JuvHistory							0	
TotalDaysPriorCustody						• • • •		
AgeFirstViolenceCharge						•••••		
NrPriorCustodialConvictions						····· 0 ···		
AgeFirstCriminalDamageCharge						····· 0 ···		
PriorCriminalDamageChargeCount						·····o		
AgeFirstDrugCharge						••••••		
PriorViolenceChargeCount						0		
PriorDrugChargeCount						•••••		
AgeFirstRobberyCharge					• • • • • • • • • • • • • • • • • • • •			
AgeFirstSexualCharge					• • • • • • • • • • • • • • • • • • • •			
PriorRobberyChargeCount					• • • • • • • • • • • • • • • • • • •			
InstantCriminalDamageChargeCount					0			
PriorSexualChargeCount				• • • • • •				
InstantViolenceChargeCount			•••••••••••••••••••••••••••••••••••••••					
InstantSexualChargeCount	····· 0							
Age	•••• •••••••							
			1	1	1	1	1	1
	-0.	.04	-0.02	0.00	0.02	0.04	0.06	0.08

#### **Error of Low-Harm Forecast**

39,598 forecast to be low harm

799 were actually high-harm = 2% off
98% Accurate
6719 were high or medium = 17%
83% accurate

#### The Moral Issues

- False Positive—money, fairness
- False negative—danger, harm
- Half of all convictions police-finalized
- Prosecutions are very expensive

## Moral Balance

- Fair predictions—not INTUITION (bias)
- Give suspects fair chance at diversion
- Unbiased estimate of their dangerousness
- Reduce costs of prosecution for low-risk
- Save costs for high-harm
- Avoid cautioning high-harm; prosecute
- Overall, fewer prosecuted, or prison

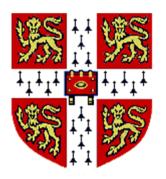
#### Police Landscape

- Emphasis on risk assessment accepted
- Clinical Risk assessment-- low accuracy
- Data Mining—high accuracy
- Now make it widely available
- Test police offender management strategies
- Legitimacy of best evidence for selection





## Thank You



Reshaping the Police Landscape with

Forecasting

Lawrence Sherman *Cambridge University* 

