Determinants of 'Innocence' Statistics on Innocence Project Data JSM August 2011

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The Sine Qua Non of Wrongful Conviction

Timothy Brian Cole: 7.1.1960 - 12.2.1999



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▶ Texas Tech Student Convicted of Rape in 1985

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 Jerry Johnson had confessed to rape in 1995; Court exoneration (posthumous) 2007-2007

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GOAL: Assist (in particular) the Innocence Network's Exoneration Work



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▶ Since 1992-present 273 exonerees



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- Solicit and review prisoner requests for post-conviction review



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Primary but not the sole actors for post-conviction relief [8]

Exoneration

Gross et al ([7]) define exoneration more broadly as:

An official act declaring a defendant not guilty of a crime for which he or she had been previously been convicted.

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Using this definition Gross et al count 340 total exonerated men and women between 1989 and 2003; 80% of whom had been imprisoned for five or more years; 73% of whom were exonerated on the basis of DNA evidence.

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The Innocence Project has focused on cases where exoneration = DNA exculpation.

DNA Evidence

Kaye points out the recent recasting of DNA evidence, [9]:

It is important to note that DNA evidence has assumed an exculpatory role relatively recently...DNA testing for identification in criminal forensics was initially critiqued as too error prone to meet a legal evidentiary standard

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From the early to the late 1990s, the debates about DNA testing standards yielded to near-universal acceptance — partially due to technological advancement — of DNA testing as *the* definitive criminal identification tool, [10], [12] or [2]

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From the early to the late 1990s, the debates about DNA testing standards yielded to near-universal acceptance — partially due to technological advancement — of DNA testing as *the* definitive criminal identification tool, [10], [12] or [2] While DNA is vital to redress a wrongful conviction, its absence weakens cases — the vast majority of exoneration requests — where there simply is no DNA evidence available [13]

Wrongful Conviction

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Wrongful Conviction

Most cases lack DNA evidence.

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Same fundamental errors?

Wrongful Conviction

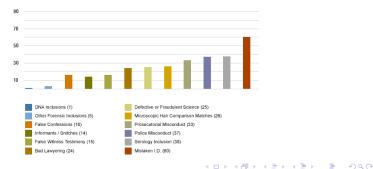
- Most cases lack DNA evidence.
- Same fundamental errors?
- Eyewitness Misidentifications, False Confessions, Jailhouse Snitches, and Flawed Forensics, [4]

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Wrongful Conviction

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Factors Leading to Wrongful Convictions (first 74 exonerations) An initial study of the first 74 DNA exonerations used a different set of categories. This study is from Actual Innocence, by Barry Scheck, Peter Neufeld and Jim Dwyer (Doubleday / 2000).



Factors in Wrongful Conviction Current Work

 Examined through the framework of DNA testing: exclusion and non-identification, [6]

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- Just a small fraction of the entreaties the IPs receive will ever have DNA evidence available...

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- ...Just a fraction of the potentially large numbers of wrongfully convicted
 [5]

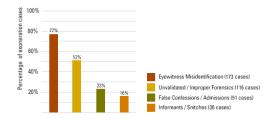
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Contributing Causes of Wrongful Convictions (first 225 DNA exonerations)

Total is more than 100% because wrongful convictions can have more than one cause.



 Endogenous difference: Different projects have different 'rules' and procedures

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- Exogenous differences: Mechanisms that wrongfully convict may differ state by state

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- Endogenous difference: Different projects have different 'rules' and procedures
- Exogenous differences: Mechanisms that wrongfully convict may differ state by state
- Recordkeeping/Coding Data: Inconsistent state by state. National level 'clearing-house' (NY IP) vs. State-level records.

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Georgia Innocence Project

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North	Carolina	Center	for	Actual
Innoce	ence			

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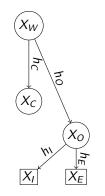


Figure: Multistate Hazard Model for Exoneration Data: X_W - Letter received; X_C - Case Closed; X_I - Case Inculpated; X_E - Case Exonerated.

The Data

	No. Ever	Entries to State					
State	in State	X_W	X_C	Xo	X_{I}	X_E	
X_W	3717		2491	558	-	-	
X_C	2490	-	-	-	-	-	
Xo	558	-	-	-	95	7	
X_{I}	95	-	-	-	-	-	
X_E	7	-	-	-	-	-	

Figure: The GIP data

The Data

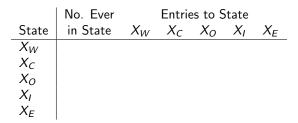


Figure: The NCAIC data...still processing

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Exploit length of time in 'state'...

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Proxy for the IP's 'prior' or ad hoc model for likeliness of exoneration

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'Inflate' data via survival curve

Exploit length of time in 'state'...

Proxy for the IP's 'prior' or ad hoc model for likeliness of exoneration

'Inflate' data via survival curve

... in the presence of covariates

False Confession?

Exploit length of time in 'state'...

Proxy for the IP's 'prior' or ad hoc model for likeliness of exoneration

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'Inflate' data via survival curve

... in the presence of covariates

- False Confession?
- Snitch?

Exploit length of time in 'state'...

Proxy for the IP's 'prior' or ad hoc model for likeliness of exoneration

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'Inflate' data via survival curve

... in the presence of covariates

- False Confession?
- Snitch?
- Race Black?

Exploit length of time in 'state'...

- Proxy for the IP's 'prior' or ad hoc model for likeliness of exoneration
- 'Inflate' data via survival curve

... in the presence of covariates

- False Confession?
- Snitch?
- Race Black?
- Victim White?

Following [14] ([11]) approximate this with 'conditional' proportional hazard curves, on 'left-truncated' data

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Cox proportional hazards



Cox proportional hazards

$$h_j(t) = h_0^j exp\{\beta^T \mathbf{Z}^j\}$$

(1)

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...in the presence of covariates ${\bf Z}$

•
$$Z_1^j = 1$$
 False Confession? Yes.

Cox proportional hazards

$$h_j(t) = h_0^j exp\{eta^T \mathbf{Z}^j\}$$

(1)

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 $\ldots in$ the presence of covariates ${\bf Z}$

- $Z_1^j = 1$ False Confession? Yes.
- $Z_2^j = 1$ Snitch? Yes.

Cox proportional hazards

$$h_j(t) = h_0^j exp\{eta^{\mathsf{T}} \mathbf{Z}^j\}$$

(1)

 $\ldots in$ the presence of covariates ${\boldsymbol Z}$

- $Z_1^j = 1$ False Confession? Yes.
- $Z_2^j = 1$ Snitch? Yes.
- $Z_3^j = 1$ Race Black? Yes.

Cox proportional hazards

$$h_j(t) = h_0^j exp\{\beta^T \mathbf{Z}^j\}$$

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 $\ldots in$ the presence of covariates ${\bf Z}$

- $Z_1^j = 1$ False Confession? Yes.
- $Z_2^j = 1$ Snitch? Yes.
- $Z_3^j = 1$ Race Black? Yes.
- $Z_4^j = 1$ Victim White? Yes

Cox proportional hazards

$$h_j(t) = h_0^j exp\{\beta^T \mathbf{Z}^j\}$$

 $\ldots in$ the presence of covariates ${\boldsymbol Z}$

- $Z_1^j = 1$ False Confession? Yes.
- $Z_2^j = 1$ Snitch? Yes.
- $Z_3^j = 1$ Race Black? Yes.
- $Z_4^j = 1$ Victim White? Yes
- Z_5^j = Duration in previous state

Only Z_5^j is really 'time-varying'

$$h_0^j \equiv h_0$$

$h_0^j \equiv h_0$			
Z	coef	exp(coef)	sig?
Confess?	0.36	1.03	*
Snitch?	59	.55	
Black?	093	.91	
Victim White?	16	.85	**
Duration in Prev. State	1.02	2.76	

$h_0^j \equiv h_0$			
Z	coef	exp(coef)	sig?
Confess?	0.36	1.03	*
Snitch?	59	.55	
Black?	093	.91	
Victim White?	16	.85	**
Duration in Prev. State	1.02	2.76	

Interpretation? Initial review process? Unclear interpretation since 'hazard' (death) means something different in between different states.

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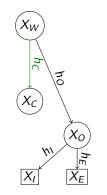


Figure: Multistate Hazard Model for Exoneration Data: X_W - Letter received; X_C - Case Closed; X_I - Case Inculpated; X_E - Case Exonerated.

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$$h_0^j, j = C$$

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$h_0^j, j = C$			
Z	coef	exp(coef)	sig?
Confess?	-0.49	0.61	**
Snitch?	0.0053	1.005	
Black?	081	.92	*
Victim White?	003	.99	
Duration in Prev. State	0.609	1.83	

$h_0^j, j = C$			
Z	coef	exp(coef)	sig?
Confess?	-0.49	0.61	**
Snitch?	0.0053	1.005	
Black?	081	.92	*
Victim White?	003	.99	
Duration in Prev. State	0.609	1.83	

Interpretation? Cases selected because of 'false confession' claim in intake are more quickly dispensed of. Some cases may 'linger' but then closed anyway.

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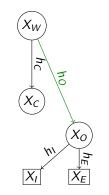


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$$h_0^j, j = O$$

$h_0^j, \ j=O$			
Z	coef	exp(coef)	sig?
Confess?	0.0181	1.02	
Snitch?	0.418	1.51	
Black?	1809	.83	
Victim White?	0.06	1.06	
Duration in Prev. State	0.522	1.685	***

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$h_0^j,j=O$								
Z	coef	exp(coef)	sig?					
Confess?	0.0181	1.02						
Snitch?	0.418	1.51						
Black?	1809	.83						
Victim White?	0.06	1.06						
Duration in Prev. State	0.522	1.685	***					
Interpretation? Cases actually worked. Duration in $X_W = letter received$								
significant implies cases wa	significant implies cases wait awhile?							

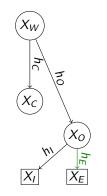


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$$h_0^j, j = E$$

$h_0^j, j = E$			
Z	coef	exp(coef)	sig?
Confess?	0.917	1.02	*
Snitch?	-0.037	0.963	
Black?	-0.326	0.722	***
Victim White?	0.053	1.065	
Duration in Prev. State	0.00323	1.003	

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$h_0^j, j = E$			
Z	coef	exp(coef)	sig?
Confess?	0.917	1.02	*
Snitch?	-0.037	0.963	
Black?	-0.326	0.722	***
Victim White?	0.053	1.065	
Duration in Prev. State	0.00323	1.003	
Interpretation? All the GIF	⁵ exonerees	are black t	thus far.

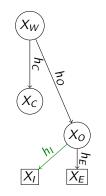


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$$h_0^j, j = I$$

$h_0^j, j = I$			
Z	coef	exp(coef)	sig?
Confess?	0.024	1.024	
Snitch?	-0.066	1.069	
Black?	-0.0571	1.058	
Victim White?	0.0573	1.059	
Duration in Prev. State	0.973	2.646	**

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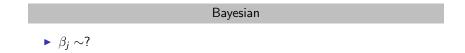
$h_0^j, j = I$			
Z	coef	exp(coef)	sig?
Confess?	0.024	1.024	
Snitch?	-0.066	1.069	
Black?	-0.0571	1.058	
Victim White?	0.0573	1.059	
Duration in Prev. State	0.973	2.646	**

Interpretation? The longer cases waited in the previous state, the longer it took to inculpate. Problems processing cases efficiently?

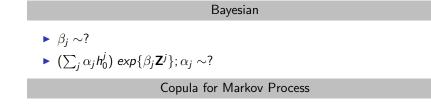
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Bayesian

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$$\beta_j \sim?$$

$$(\sum_j \alpha_j h_0^j) \exp\{\beta_j \mathbf{Z}^j\}; \alpha_j \sim?$$

Copula for Markov Process

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• Following [3] and [1] $(A * B)(x_1, x_2) = \int_0^1 \frac{\partial A(x_1, t)}{\partial x_2} \cdot \frac{\partial B(t, x_2)}{\partial x_1} dt$

Bayesian

$$\beta_j \sim? (\sum_j \alpha_j h_0^j) \exp\{\beta_j \mathbf{Z}^j\}; \alpha_j \sim?$$

Copula for Markov Process

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- ▶ then $C_{X_1X_3} = C_{X_1X_2} * C_{X_2X_3}$ equivalent to CK equations

Other

Bayesian

$$\beta_j \sim? (\sum_j \alpha_j h_0^j) \exp\{\beta_j \mathbf{Z}^j\}; \alpha_j \sim?$$

Copula for Markov Process

- Following [3] and [1] $(A * B)(x_1, x_2) = \int_0^1 \frac{\partial A(x_1, t)}{\partial x_2} \cdot \frac{\partial B(t, x_2)}{\partial x_1} dt$
- then $C_{X_1X_3} = C_{X_1X_2} * C_{X_2X_3}$ equivalent to CK equations

Other

- Processing NC(CAIC) data
- Really useful $\pi(s|H(t)) = \mathbb{P}(X = X_E \text{ in } s > t|H(t)),$ $H(t) = (H_j(t) = \{Z^j; x_1, ..., x_t\}$

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