

Analyzing Possible Human Trafficking Activity Using Large- Scale Information Extraction

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Human Trafficking

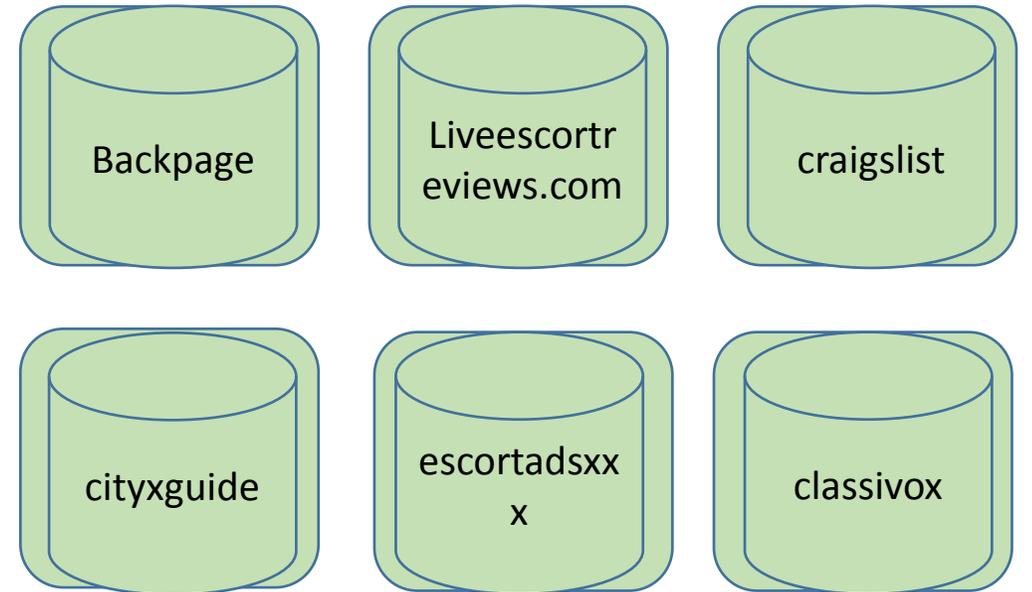
- Odious crime that obtains labor via physical, financial coercion
- Traffickers believed to channel many victims into sex work
- Trafficking phenomena very opaque
 - Unclear which sex workers are in fact trafficking victims
 - Unclear which sex work organizers engage in trafficking
 - Law enforcement lacks any good intelligence
 - Surprisingly old problem



Trafficking image, US Department of Justice

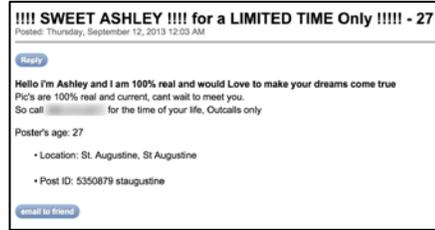
Human Trafficking and Web Data

- Sex work has gone online
- 80M advertisements posted on dozens of websites
- Advertisements contain:
 - Contact info
 - Personal details
 - Location
 - Price
 - Transaction details
 - Lots of other data
- Grim, but potential analytical gold mine
- Unfortunately, few or no H/T labels



Our Goals

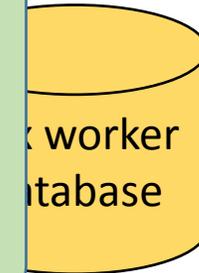
1. Transform 80M raw ads into high-quality database



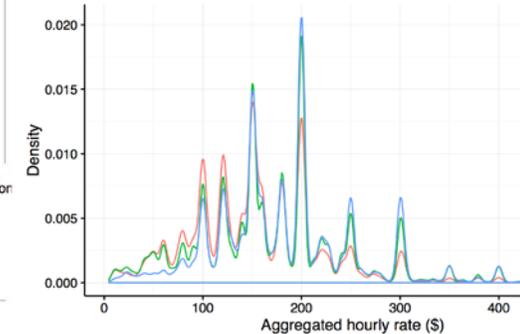
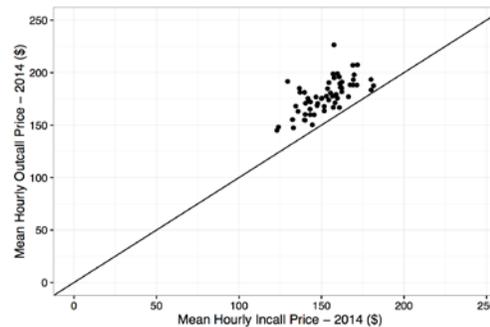
Name	Age	Location	MSA
Ashley	27	St Augustine	Jacksonville

2. Build database-backed enforcement tools to

Novel social science and tools made possible by computational methods



3. Long term plan: Use database for novel social science findings, further help law enforcement



Tools for Identifying Trafficking Victims with Data

- Is sex worker:
 - Offering surprisingly low price?
 - Willing to engage in risky services?
 - Moving unusually frequently?
 - Sharing contact info with suspected h/t victim?
- Does advertisement:
 - Use numerically-nearby phone numbers?
 - Reuse ad text from a suspected h/t victim?
 - Describe someone with possible warning signs (physical abuse, drug abuse, fear of law enforcement, etc)?



60 Minutes Overtime, CBS.com

HT Advertising Table

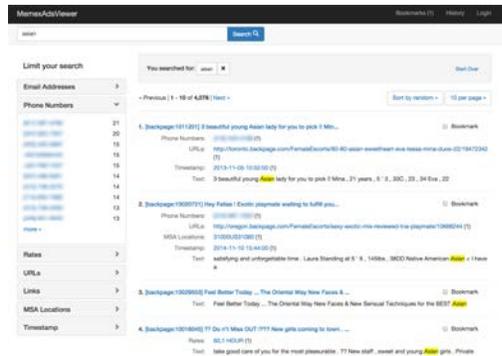
Aspirational Table

Ad Identifier	Date Posted	(Craigslist) Region	...	Ad Text
10001	08/12/15	Wash		...
10002	08/15/15	NYC		...
10003	08/16/15	LA		...

Ad Identifier	Price/ Hour	Place(s)	Incall/ Outcall	Phone	Email
10001	100	Bethesda	Incall	555...	m@...
10002	225	Nyack	Outcall	555...	c@...
10003

Is the number of sex service ads increasing or decreasing?
 What large region has the most ads?

What is the typical price for offered services?
 Which providers are posting multiple times?
 Which exact locality is growing in popularity?



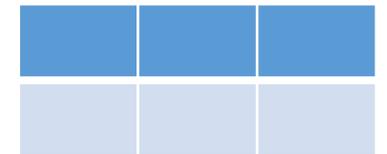
Faceted Search

Attorney		
Name	Total Loss Paid	Number of Claims
kevin o'connell	1028765.93	19
roger moore	263975.02	17
mark mcMahon	58107	13
marc gladner	79337.02	6
chad kreblin	217278.42	5
mark koss	209715.14	5
jeremy brantingham	98276.41	5
isa taylor	81934.8	5
michael hupy	76377.89	5
paul aaimott	29014.79	5

Data Analytics



Visualizations



Other Tools

Labor Market Interactions

- How do prices react to women's opportunities in broader market?
- If prices are insensitive to broader conditions, it might suggest a large amount of compelled participation

- Use Bartik wages [Bartik 1991] to measure labor market demand apart from local supply
- Build regression model to predict price in a given MSA and month
 - Control for posting venue and commute distance
 - Control for MSA-specific effects, seasonal time-of-year effects

Labor Market Findings

- After controlling for MSA-specific effects, robust ***negative*** correlation between female opportunity and hourlong price
 - \$1 in overall wage increase predicts \$0.43 **decrease** in hourly rate
 - Why??
- Divide market into low-, mid-, high-end prices (<25%, 25-75; >75%)
- Negative correlation between female opportunity and # mid-price ads
 - Predicts 7.1% drop in first-time advertisers at mid-price (signif at 0.01)
 - 1\$ in overall wage increase predicts 3% drop in overall participation
- But, positive correlation between female opp'ty and # low-price ads
 - Almost identical rise in participation; 3.6% increase in first-time ads