

Implicit Attitudes, NLP, and the "Real World"

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Other Work

- Quantifying "modernity" in Chinese poetry
- NAACL Comp Ling for Literature 2013
 Discourse-level effects on reference
- vho "y " n r A 201 bil gu ls INTERSPEECH 2016
- Sociophonetic embodiment: Body movement and head positioning

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'you know how like when

Today - dissertation time!

Implicit Attitudes

"introspectively unidentified (or inaccurately identified) traces of past experience that mediate favorable or unfavorable feeling, thought, or action toward social objects"

Greenwald and Banaji 1995

NLP allows us to analyze linguistic "actions" at a large scale

...while accounting for crucial aspects of the "real world" social context

Project 1

Racial Disparities in Police Officer Respect

with Nick Camp, Camilla Griffiths, Will Hamilton, David Jurgens, Vinod Prabhakaran, Rebecca Hetey, Dan Jurafsky, and Jennifer Eberhardt

Our Question

Do officers treat White community members with a greater degree of respect than they afford to Blacks?

Police-Community Interaction

- Media focus on explosive incidents
- Research focus on outcomes

but:

 one quarter of adults have contact with the police during the course of a year
 majority occurring in traffic stops

Respect is Important

- A person who is treated with respect
 ... has more trust in the individual
 - officer's fairness Tyler and Ho 2001
 - ... and the procedural fairness of the institution Tyler and Sunshine 2003
 - ... and is more willing to support or cooperate with the police

Tyler 1990, Mazerolle et al 2013

Previous work on procedural fairness

• Relies on:

• citizens' recollection of past interactions

Epp et al 2014

researcher observation of officer behavior

Mastrofski et al 2009, Dai et al 2011, Jonathan-Zamir et al 2015

• These are invaluable but indirect

 ... and presence of researcher may influence police behavior

Mastrofski and Parks 1990

Police body camera footage

- Oakland PD has been wearing body cameras since 2010
- Usually used only as evidence
- ... but, a window into everyday behavior!



Our proposal: Footage as Data

- 981 stops by 245 officers in April 2014
 - Drivers: 682 black, 299 white
 - 183 hours of footage
- Professionally transcribed and diarized
- Resulting data set:
 - 36,738 officer utterances, 350k+ words

Sample transcription

0:00:00 0:00:09 OFFICER [to dispatch]: Unknown occupant and it's going to be for registration. It should be code four. 0:00:20 0:00:20 OFFICER: Hi. 0:00:20 0:00:20 FEMALE: Hi. 0:00:21 0:00:23 OFFICER: I pulled you over because your registration is expired by almost a year. 0:00:25 0:00:28 FEMALE: Okay, I have the paperwork for it, a moving permit? 0:00:28 0:00:28 OFFICER: I'm sorry? 0:00:29 0:00:30 FEMALE: I have the paperwork for it.

0:00:30 0:00:31 OFFICER: Okay.

Project 1

Study 1

Perceptions of Officer Treatment from Language

Study 1: Goals

- Can human raters judge respect from officers' language?
- Are there differences in officer respect towards Black versus White community members?

"Thin Slice" Utterance Rating Task

- Participants (N=70) blind to race labeled 414 officer utterances
 - 10 coders per utterance
 - 4-point Likert scales

Respectful, Polite, Friendly, Formal, Impartial (high rater agreement αs=.73-.91)

Utterance Rating Task

Read the following interaction with a police officer: The citizen just said:

It's in my glove compartment.

And then the officer says:

Let me take a look at it. How about insurance?

How *impolite* or *polite* was the officer?

OOVery
ImpoliteSomewhat
ImpoliteSomewhat
PoliteVery
Polite

Utterance Rating Task



The Latent Space of Respect

Two PCs explain 93% of the variance:

	Respect	Formality
variance explained:	71%	22%
Formal	0.27	0.91
Friendly	0.47	-0.39
Polite	0.49	-0.04
Respectful	0.47	0.03
Impartial	0.50	-0.11

The Latent Space of Respect

Race on these dimensions:



Project 1

Study 2

Modeling Respect with Computational Linguistics

Study 2: Goals

- Develop a computational linguistic model capable of estimating Respect
- Use the human labeled data as supervised training data to learn weights on interpretable features

Methodology

- Hand-engineered features
 - Lexicons, gazetteers, regexes, dependencies, joint pattern matching ("bald commands")
 - Drawn primarily from linguistic and computational work on politeness

Goffman 1967, Lakoff 1973, Culpepper 1976, Brown and Levinson 1978 Prabhakaran et al 2012, Danescu-Niculescu-Mizil 2013, Krishnan and Eisenstein 2014

Statistical Model: simple linear regression
 log-tranformed counts of features per utterance

Feature Weights

Apologizing For You Gratitude *** Reassurance ++ Last Names *** Formal Titles ** For Me Safety Give Agency Filled Pauses (Um/Uh) ** -Adverbial Just **Positive Words** Hedges Introductions 0 -1

Perceived as more

Disrespectful

Respect Model Coefficients



Log Odds Ratio by Race



Respectful † is p < 0.1, * is p < 0.05, ** is p < 0.01, *** is p < 0.001

Respect Model Coefficients Log Odds Ratio by Race Questions *** Linguistic Negation *** **Negative Words Feature** Ask for Agency Disfluency ** Weights Informal Titles First Names Hands on the Wheel -0.5 0.0 0.5 0 -1 Perceived as more ... More common in... Disrespectful Respectful Black Stops White Stops † is p < 0.1, * is p < 0.05, ** is p < 0.01, *** is p < 0.001





Results

• *Respect* model is able to perform roughly like an average annotator

Model Adjusted R ²	0.258
Model RMSE	0.840
Average annotator RMSE	0.842 (range from 0.497 - 1.677)

• *Formality* model is worse but still reasonable

Model Adjusted R ²	0.190
Model RMSE	0.882
Average annotator RMSE	0.764 (range from .517 - 1.703)



Study 3

Racial Disparity Across the Entire Dataset

Study 3: Goals

- Do the results from Study 1 hold across an entire month of traffic stops?
- ... even controlling for contextual factors?

Study 3: Results

		Respect			Formality		
	β	\mathbf{CI}	р	eta	CI	р	
Arrest Occurred	-0.00	-0.03 - 0.03	.933	0.01	-0.02 - 0.04	.528	
Citation Issued	0.04	0.02 - 0.06	<.001	0.01	-0.01 - 0.03	.209	
Search Conducted	-0.08	-0.110.05	<.001	-0.00	-0.03 - 0.02	.848	
Age	0.07	0.05 - 0.09	<.001	0.05	0.03 - 0.07	<.001	
Gender (F)	0.02	-0.00 - 0.04	.062	0.02	0.00 - 0.04	.025	
Race (W)	0.05	0.03 - 0.08	<.001	-0.01	-0.04 - 0.01	.236	
Officer Race (B)	0.00	-0.03 - 0.04	.884	0.00	-0.03 - 0.03	.987	
Officer Race (O)	-0.00	-0.04 - 0.03	.809	-0.00	-0.03 - 0.02	.783	
Officer Race (B) : Race (W)	-0.01	-0.03 - 0.02	.583	0.01	-0.01 - 0.03	.188	
Officer Race (O) : Race (W)	-0.01	-0.03 - 0.02	.486	-0.00	-0.02 - 0.02	.928	

Interpretation

White community members are 57% more likely to hear an officer say one of the **top 10% most respectful** utterances in our dataset

Black community members are 61% more likely to hear an officer say one of the **top 10% least respectful** utterances in our dataset

Controls

- Holds even considering:
 - Only "everyday" interactions (no arrest, no search)
 - Crime rate in the area
 - Density of businesses in the area
 - Whether driver race was known before the stop
 - Officer years of experience

Controls - Severity

- We asked OPD officers to rate the stops for severity
 - **1 Very minor** (expired registration)
 - 4 very severe (speeding)

- Black drivers are stopped for less severe offenses
- ... but no impact on respect



Controls - Officer Race



Surprisingly, not a factor!

Across the Interaction

- Respect rises throughout the interaction
- ... but rises faster for whites



Across the Interaction

- No race effect for Formality
- Officers less formal over the interaction



Conclusions from the first paper

- Confirms community reports: interactions with black community members are more fraught
- Provides concrete strategies for officers
- Cooperation with Oakland to integrate results into procedural justice training
 ... and we can measure impact



Moving Forward

- Tone of Voice:
 - Preliminary results suggest a similar trend
- Community member language:
 - Escalation
 - Compliance, politeness
- Other Departments