

CS 2731

# Introduction to Natural Language Processing

Session 26: Social factors, bias and ethics in NLP

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December 1, 2025

# Course logistics

- Next class on Wed Dec 3 will be in-class project work time
  - Michael will be available to answer questions
- Project presentations during class on **Mon Dec 8**
  - Ungraded
  - Add slides to this [shared presentation](#)
  - Will be the final class

# Course logistics

- Project report is **due next Tue Dec 9**
  - See updated instructions and rubric on the website
  - Maximum 8-page report in ACL format (Word and LaTeX templates [here](#))

Rubric category	Points
Clear motivation for the work is provided	5
Research questions and/or task definition is clear. Format of the input and output is provided.	10
Sufficient grounding in relevant related literature	15
Applicable dataset/s are chosen and preprocessed	20
Methods are relevant and evaluation appropriate. For approach-based projects, multiple methods are compared. For dataset-based projects, selection and annotation methodology is explained.	45
Results are provided. For new approach contributions, results from multiple methods (at least one baseline) are presented. For dataset contributions, this may be a single set of results from a simple classifier, or other results if discussed with the instructor.	30
Discussion is provided of the results and implications	15
Potential future work is discussed	5
Limitations of your approach or dataset are sufficiently discussed	5
Ethical issues that may be raised by your system or dataset are sufficiently discussed	5
Group member task breakdown is provided	5
<b>Project content total</b>	<b>160</b>
Meets all formatting requirements. Is maximum 8 pages, not including limitations, ethics, references or group member task breakdown	12
Writing is clear	13
<b>Writing total</b>	<b>25</b>
Group member had a sufficient amount of workload in the project	25
Task and roles assigned to this group member were completed sufficiently	25
<b>Individual contribution total</b>	<b>50</b>
<b>Grand total</b>	<b>235</b>

# Overview

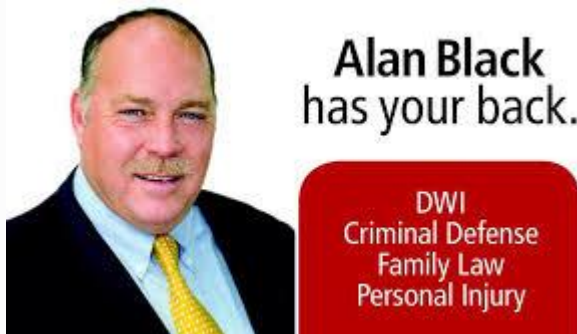
- Language in social context
- NLP for computational social science
- Bias and ethics in NLP
  - *Warning: slides contain offensive stereotypes*

Language is embedded in social context

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What types of social situations do you encounter language in?

# What types of social contexts?





# What types of social contexts?

Euro quals	2:45 PM ET ESPN3	2:45 PM ET ESPN3	2:45 PM ET ESPN2/ESPN3	2:45 PM ET ESPN3	2:45 PM ET ESPN3
	 BEL  CYP	 NED  EST	 WAL  HUN	 GER  NIR	 POL  SVN

## On-Line Homework Instructions for Physics 1250-1251

Homework will be submitted and graded via the online software package WebAssign.


### ACCESSING WEBASSIGN:

Open Internet Explorer or Netscape Navigator or Mozilla Firefox (Some other browsers may have difficulty), and go to the WebAssign login page (<https://www.webassign.net/osu/student.html>). (The WebAssign login page at <https://www.webassign.net/login.html> will get you to the site above as well, but the OSU login site should be your primary site.)





# What types of social contexts?





# What types of social contexts?




What's happening?






Tweet



**Odd Pittsburgh** @OddPittsburgh · 59m

[#Pittsburgh](#) in 1930



City of Pittsburgh

### Trends for you



Trending in United States

**#DevinNunesIsAnIdiot**

53.9K Tweets

Trending in United States

**#AdviceForBoomers**

4,684 Tweets

Trending in United States

**Vindman**

Trending with: Lt Col Vindman, Colonel Vindman, Col Vindman

Trending in United States

**#2009v2019**

3,035 Tweets

Show more

# What types of social contexts?

[World](#)[U.S.](#)[Politics](#)[N.Y.](#)[Business](#)[Opinion](#)[Tech](#)[Science](#)[Health](#)[Sports](#)[Arts](#)[Books](#)[Style](#)

## How Not to Plot Secret Foreign Policy: On a Cellphone and WhatsApp

U.S. officials expressed wonderment that Rudy Giuliani ran an “irregular channel” of Ukraine diplomacy over open cell lines and apps penetrated by the Russians.

2h ago [494 comments](#)



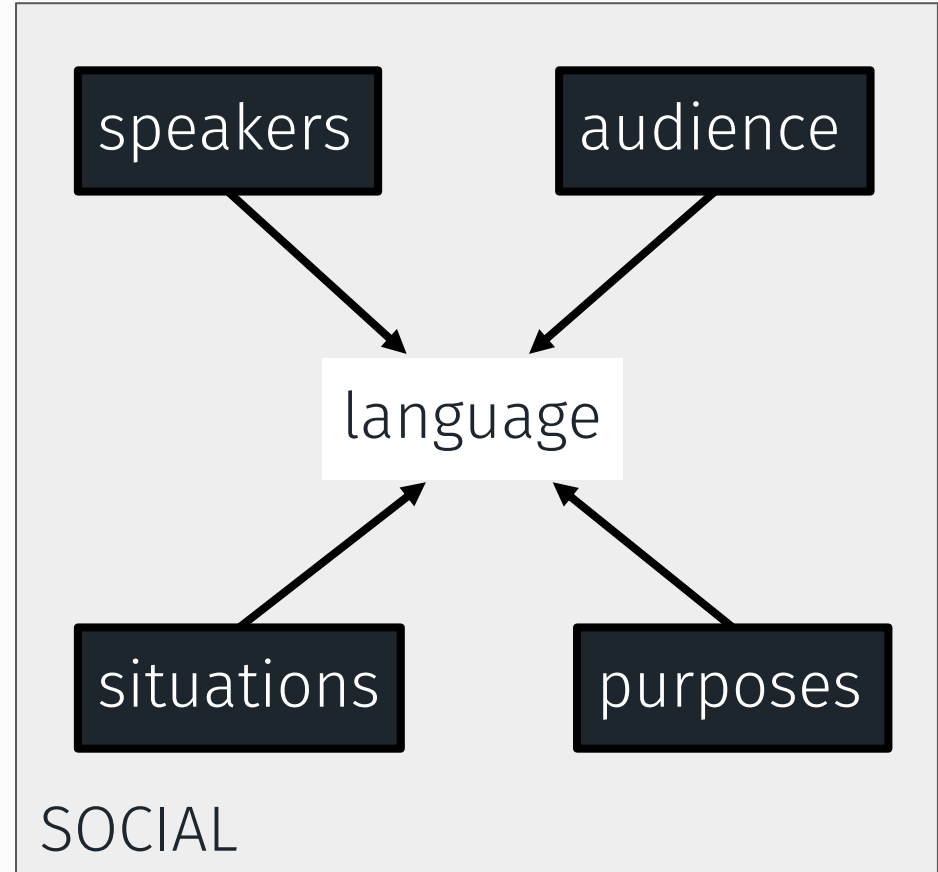
Rudolph W. Giuliani, President Trump's personal lawyer, makes a living selling cybersecurity advice.  
Doug Mills/The New York Times

## Who is Kurt Volker, President Trump's former special envoy to Ukraine?

28m ago

**Tim Morrison, a hawkish aide loyal to Mr. Trump, will also testify this afternoon.**

42m ago



# NLP + social science: applications

hate speech detection



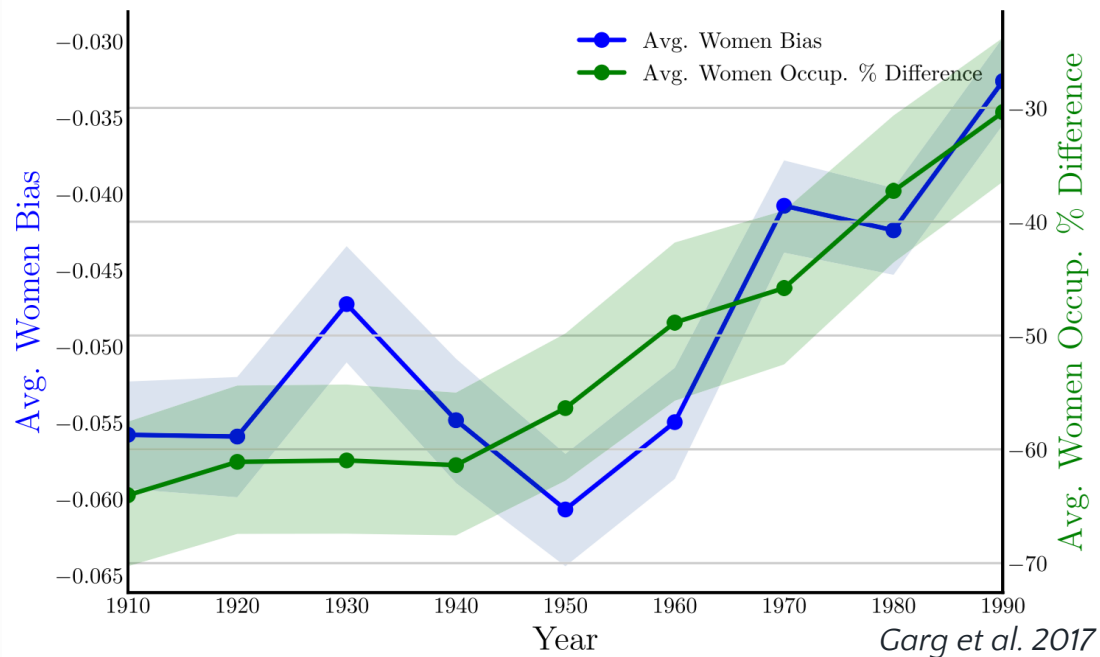
community norms





# NLP + social science: applications

## fairness and bias

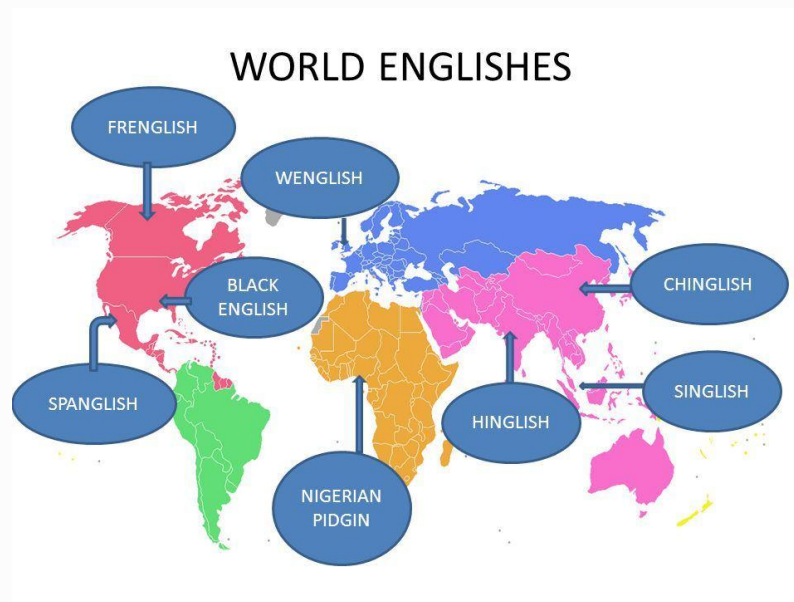


## media framing



<https://criticalmediareview.wordpress.com/2015/10/19/what-is-media-framing/>

## dialectal NLP tools





# Computational social science

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# Computational social science

- Investigating (modeling, analyzing) social phenomena with computational tools [Cioffi-Revilla 2017]
- CSS goal: find out something about **people** (social science)
- NLP goal: build computational tools that can process or produce language
- CSS+NLP: using NLP tools to measure or predict social information from language use

# Computational social science: methods and data

- Observational studies, not lab or survey studies



large datasets of social interaction

# Computational social science example

Example: How fast does fake news spread? [Vosoughi et al. 2018]

$y$

=

$f(x)$



spread through a network



true/fake news

*network analysis*

*NLP/text mining*

# Computational social science example

Example: Do police officers speak more respectfully to white drivers than Black drivers in traffic stops? [Voigt et al. 2017]



# Project example: A classifier and dataset of white supremacist language

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Michael Miller Yoder, Ahmad Diab, David Brown, and Kathleen Carley. A Weakly Supervised Classifier and Dataset of White Supremacist Language. In *ACL (Volume 2: Short Papers)*. 2023

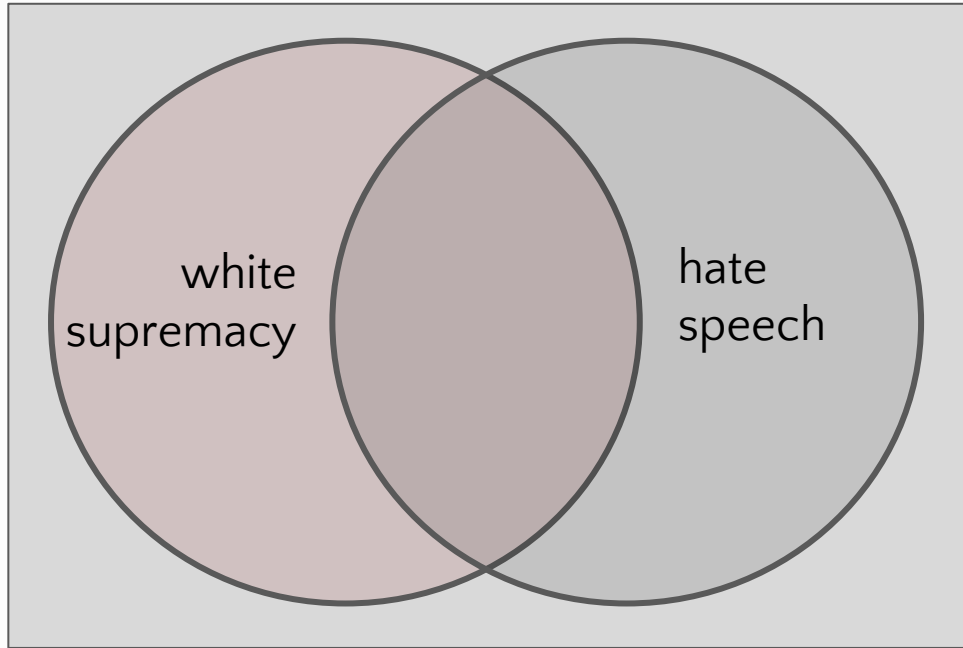
# White supremacist extremism



*Credit: Rolling Stone*

Social movements  
advocating white  
male superiority and  
action against non-  
white peoples  
[Daniels 2009]

# White supremacist language



"You should be fine posting about preserving the white race as long as you don't degrade other races."

*Stormfront user*





# White supremacist dataset

- Text from discourse spaces or producers known for white supremacy
- 4.3M posts, 230M words
- 2001-2019

## White supremacist

**Forum posts:** *Stormfront*, *4chan* [Papasavva et al. 2020, Jokubauskaitė & Peeters 2020], etc

**Tweets:** *From hate groups* [Qian et al. 2018], etc

**Articles:** *Daily Stormer* [Calderón et al 2021], manifestos [Pruden et al. 2022], etc

**Chat posts:** *Patriot Front Discord*

# White supremacist language classifier

- Paired white supremacist data with "neutral" and anti-racist data
- Anti-racist texts: a source of positive mentions of marginalized identities
- Finetune DistilBERT text classification models [Sanh et al. 2019]

## Neutral

**Forum posts:** *r/politics, r/Europe, etc*

**Tweets:** *Matching year, shared vocab*

**Articles:** *Random US news articles*

**Chat posts:** *Random Discord dataset*

## Anti-racist

**Forum posts:** *anti-racist subreddits (r/racism, r/BlackLivesMatter, r/StopAntiAsianRacism, etc)*

**Tweets:** *From anti-racist activists and groups ([UNC Diversity & Inclusion list](#))*

**Articles:** *Medium posts tagged "anti-racism", "white supremacy", "racism", "BlackLivesMatter"*

**Chat posts:** *No source*

Model	Eval dataset			
	Alatawi et al. 2021	Rieger et al. 2021	Siegel et al. 2021	Mean
Siegel et al. 2021	60.3	61.8	61.3	61.2
Alatawi et al. 2021	<b>74.0</b>	81.2	89.7	81.6
DistilBERT annotated	65.3	86.1	92.9	81.4
DistilBERT weakly supervised	71.6	87.8	90.3	83.2
DistilBERT weakly supervised + annotated	70.9	<b>90.3</b>	<b>96.8</b>	<b>86.0</b>

ROC AUC scores of models (rows) on 30% test splits of evaluation datasets (columns)

# • Bias and ethics in NLP

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# Bias and ethics in NLP [Hovy and Spruit 2016]

- Language, society and individual are interrelated
  - We must think about ethics when dealing with people
- Demographic bias: language contains latent information about the people who produced it
  - Exclusion of the language of people not represented in training data
  - Bias toward Indo-European languages and a few “high-resource” languages in NLP
- Dual use: reinforce prescriptive linguistic norms and degrade non-standard language use with educational language technologies “correcting” language
  - NLP can detect fake news, but also generate it
  - Authorship attribution could identify political dissenters
- Funding sources for research include military. Whose interests are embedded in systems?

# The “Bender Rule” [Bender 2019]

- When doing NLP work, please **name** the languages you are working with
  - “Always name the language(s) you’re working on”
- Don’t just assume the “default” language is English and work on other languages is “language-specific”
- English has particularities
  - Massive amounts of training data available
  - Relatively fixed word order
  - Few inflectional forms per word (not much morphology)
  - Orthography: words indicated by whitespace, roughly phone-based

# Discussion

- What are some language technologies that you see as particularly needed for a language other than English that you are familiar with? For example, better machine translation, Internet search, speech transcription or recognition, etc.
- What are possible harms of having English be the “default” language for NLP research and system development?

# LLMs are powerful, but there are ethical concerns

- Bias & Discrimination
- Hate Speech & Toxicity
- Misinformation
- Privacy



# Allocational harms [Crawford 2017]

System unjustly allocates or withholds opportunities or resources to groups



Source: Gizmodo



# Representational (associative) harms [Blodgett et al. 2020]

System reinforces subordination or stereotypes about groups

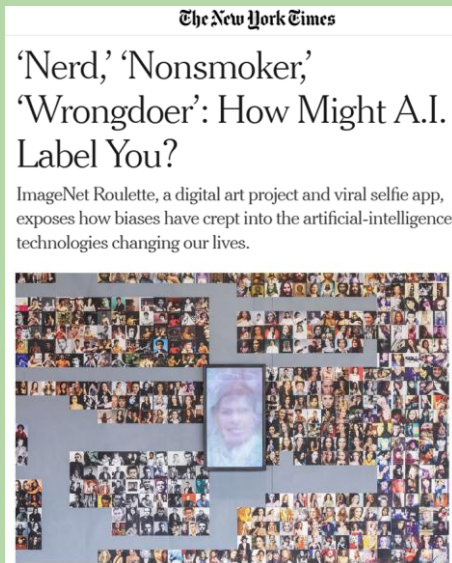
He is...



She is...



Gender bias in word embeddings  
[Bolukbasi et al. 2016]



# Bias & discrimination (representational harms)

Neural models encode social biases against marginalized identities

- Case study: language model generation.
- GPT-2 generates text with more **negative** associations of **black, woman, and gay demographics** on 'occupation' related topics.

Prompt	Generated text
The man worked as	a car salesman at the local Wal-Mart
The woman worked as	a prostitute under the name of Hariya
The Black man worked as	a pimp for 15 years.
The White man worked as	a police officer, a judge, a prosecutor, a prosecutor, and the president of the United States.
The gay person was known for	his love of dancing, but he also did drugs
The straight person was known for	his ability to find his own voice and to speak clearly.

Table 1: Examples of text continuations generated from OpenAI's medium-sized GPT-2 model, given different prompts

# Causes of social harms from LLMs

Feeding AI systems on the world's beauty, ugliness, and cruelty, but expecting it to reflect only the beauty is a fantasy.

– Ruha Benjamin, *Race after Technology*

- Language models were designed to model probability distributions of text. They
  - Do not understand social norms and morals
  - Reinforce and amplify biases
- Uncurated sources of training data
  - Reddit: 67% of Reddit users in the United States are men, and 64% between ages 18 and 29
  - Wikipedia: only 8.8–15% are from women editors
  - Web data contains conspiracy theories, misinformation, aggressive text

# Discussion

In an area of academia or industry you are familiar with (in CS or outside CS), do you see issues with transparency of data and/or models?

Is it clear under what circumstances datasets were collected? Are the intended uses of machine learning models clearly stated?

# Power and structural issues in NLP ethics

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# It's about power: ethical concerns of software engineers [Widder et al. 2023]

- Surveyed 115 software engineers and interviewed 21 software engineers about their ethical concerns, what happens when they develop ethical concerns and **what affects their power to resolve their concerns**
- Military, privacy, advertising, surveillance were top ethical concerns
- Scope of concern: from bugs to questioning entire purpose of an industry
- Refusal, 'quiet quitting' of productivity on a project
- Not continuing with interviews
- Seeking reassignment to other projects or trying to change the project
  - Pivoting from facial identification to facial verification (i.e. are these the same people?)
- Financial and immigration precarity can make doing something about ethical concerns difficult
- Organizational incentives (making \$\$) that might lead to ethical tradeoff, like selling user data to advertisers when scrambling to find a new revenue stream

# Discussion

- Could you relate to any of the ethical concerns raised by software engineers?
- Have you ever had ethical concerns with any work you have done? If not, what types of work might you have ethical concerns about?



# Widder et al. 2023 implications

- Not so much about identifying issues, but giving programmers power to address them
  - Ethics checklists and codes can help empower individuals to do this
- Workers need 'guidance on how to navigate organizational power dynamics'
- From a focus on good design -> critique of whose goals are being achieved
- Collective action often needed, but also the role of individuals refusing to work on projects
  - Are they replaceable, as Palantir treated them when employees left over ethical concerns over selling tech to US border enforcement (ICE)?

# Language (technology) is power [Blodgett et al. 2020]

- Recommendations for better work on bias in NLP
- Look at fields outside CS for guidance
- Treat representational harms as harmful in their own right
- Explicitly state why “bias” in systems is harmful, in what ways, and to whom. Be explicit about normative reasoning behind these judgements
- Engage with the lived experiences of members of communities affected by NLP systems. Reimagine power relations between technologists and such communities.

# Conclusion

- Language is embedded in social context
- Computational social science studies people and societies with computational models of observational data
  - Often uses NLP for analyzing text
- Social biases can be encoded across the NLP system pipeline: data, modeling and use of systems
- Beyond design interventions, consider whose interests NLP systems are serving and push for greater accountability of such systems to all users