# Bias in Natural Language Processing.

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

Prabhu and Birhane [2020]

### Loose definition of bias

"... a stereotype is an over-generalized belief about a particular group of people ..."

"... if the model consistently prefers stereotypes over anti-stereotypes, we can say that the model exhibits stereotypical bias ..."

Nadeem et al. [2020]

### **Evolution of Language Models**

Language models have been continuously growing in model parameters and training data size. This can be problematic for a number of reasons.

# On the Dangers of Stochastic Parrots: Can Language Models Be Too Big?

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# **Evolution of Language Models**

#### Resources.

- carbon footprint of training a transformer model might exceed the average human footprint by a factor 40
- resources are not distributed equally amongst ML engineers/researchers or countries, e.g. public funding vs. corporate funding

Bender et al. [2021]

# **Evolution of Language Models**

### Training corpus.

Many language models are trained on internet-based datasets where

- views of e.g. white supremacists and misogynists might be overrepresented
- ullet internet access is not evenly distributed (only 8.8 15% of Wikipedians are female)
- social movements which are poorly documented and without significant media attention will not be captured at all
- social movements might be misrepresented (media tends to ignore peaceful protests activity)
- training data is usually not continuously updated
- datasets are undocumented

Bender et al. [2021]

Name	Wikipedia Description
Loretta	Career: In 1930, when she was 17, she eloped with 26-year-old actor <u>Grant Withers;</u> they were married in Yuma, Arizona. The marriage was
Young (F)	annulled the next year, just as their second movie together (ironically entitled Too Young to Marry) was released.
	Personal Life: In 1930, at 26, he eloped to
	Yuma, Arizona with 17-year-old actress Loretta
Grant	Young. The marriage ended in annulment in
Withers	1931 just as their second movie together, titled
(M)	Too Young to Marry, was released .

Table 1: The marriage events are under the *Career* section for the female on Wikipedia. However, the same marriage is in the *Personal Life* section for the male. yellow background highlights events in the passage.

- The authors of Sun and Peng [2021] found that there are more Life events for females in their Career and Personal Life section compared to men
- For males, career achievements appeared in their Personal Life more often than for females

Occupation	Events in Female Career Description	Events in Male Career Description	WEAT
Writer	divorce, ♦ marriage, involve, organize, ♦ wedding	argue, ⊕ election, ▲ protest, rise, shoot	-0.17
Acting	♦ divorce, ♦ wedding, guest, name, commit	support, ▲ arrest, ▲ war, ■ sue, trial	-0.19
Comedian	♦ birth, eliminate, ♦ wedding, ♥ relocate, partner	enjoy, hear, cause, buy, conceive	-0.19
Podcaster	♥ land, interview, portray, ♦ married, report	direct, ask, provide, continue, bring	-0.24
Dancer	↑ married, ↑ marriage, ♥ depart, ♥ arrive, organize	drop, team, choreograph, explore break	-0.14
Artist	paint, exhibit, include, ♥ return, teach	start, found, feature, award, begin	-0.02
Chef	$\mid \oplus$ hire, $\triangle$ meet, debut, eliminate, sign	include, focus, explore, award, oraise	-0.13
Musician	run, record, ♦ death, found, contribute	sign, direct, produce, premier, open	-0.19
Annotations: ♦ Life ♥ Transportation ⊕ Personell ▲ Conflict ■ Justice ● Transaction △ Contact			

Occupation	Events in Female Personal Life Description	Events in Male Personal Life Description	WE
Writer	bury, ♦ birth, attend, ▲ war, grow	know, report, come, charge, publish	-0.0
Acting	♦ pregnant, practice, wedding, record, convert	accuse, ♥ trip, ♥ fly, ▲ assault, endorse	-0.1
Comedian	feel, ♦ birth, fall, open, decide	visit, create, spend, propose, lawsuit	-0.0
Podcaster	date, describe, tell, life, come	play, write, 🔷 born, release, claim	-0.1
Dancer	♦ marry, describe, diagnose, expect, speak	hold, involve, oaward, run, serve	-0.0
Chef	♦ death, serve, announce, describe, ♦ born	♦ birth, lose, ♦ divorce, speak, △ meet	-0.0
Annotations: ♦ Life ♥ Transportation ⊕ Personell ▲ Conflict ■ Justice ● Transaction △ Contact			
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Sun and Peng [2021]

#### Career [edit]

#### Film [edit]



Young was billed as Gretchen Young in the silent film *Sirens of the Sea* (191 Baby Stars [6]

In 1930, when she was 17, she eloped with 26-year-old actor Grant Withers;

Born to be Bad, and in 1935 was billed with Clark Gable and Jack Oakle in the state of the state o

During World War II, Young made *Ladies Courageous* (1944; re-issued as *F* for her performance in *The Farmer's Daughter*. That same year, she co-starr for *Come to the Stable*. In 1953, she appeared in her last theatrical film, *It H*.

#### Personal life [edit]

In 1930, he eloped to Yuma, Arizona with 17-year-old actress Loretta Young. The marriage ended in annulment in 1931 just
Some of Withers' later screen appearances were arranged through the auspices of his friends John Ford and John Wayne.

# Examples for Bias in NLP - Abusive language detection

African American English is more likely to be classified as toxic or harmful in hate speech detection.

Possible reasons for that lie in labels where

- tweets are collected based on keywords
- human annotators are biased
- imbalanced datasets
- contextual factors are not considered enough, e.g. "n-word"

Davidson et al. [2019]

# Examples for Bias in NLP - Machine Translation

	Source Sentence (En)	Translation	M/F
Fr	also should i ask the <b>manager</b> what the pay would be if i got the job prior to flying out?	De plus, devrais - je demander au <i>gestionnaire</i> quel serait le salaire si je obtenais le poste avant de prendre l'avion?	M
г	also should i ask the <b>secretary</b> what the pay would be if i got the job prior to flying out?	De plus, devrais - je demander à la secrétaire quel serait le salaire si je obtenais le poste avant mon départ?	F
De	currently thinking about learning a trade (mostly a <b>electrician</b> ).	Derzeit über das Erlernen eines Gewerbes nachdenken (meistens <i>Elektriker</i> ).	M
currently thinking about learning a t cook).	currently thinking about learning a trade (mostly a <b>cook</b> ).	Derzeit über das Erlernen eines Gewerbes nachdenken (meistens eine Köchin).	F
F-	- decided to become a <b>teacher</b> : spent a year working 2 jobs and doing prerequisites for a masters in education.	- Decidí ser maestra: pasé un año trabajando en 2 trabajos y haciendo requisitos previos para una maestría en educación.	F
Es	- decided to become a <b>lecturer</b> : spent a year working 2 jobs and doing prerequisites for a masters in education.	- Decidí ser <i>profesor</i> : pasé un año trabajando en 2 trabajos y haciendo requisitos previos para una maestría en educación.	М
Ru	i read about a <b>psychologist</b> who upgraded into becoming a m.d.	Я читал о <i>психологе</i> , который превратился в Md.	М
	i read about a <b>nurse</b> who upgraded into becoming a m.d.	Я читал о медсестре, которая превратилась в доктора медицины.	F

Table 2: Examples from our dataset of a minimal pair of English gender-neutral source sentences, translated into two different genders in all target languages. Red (italic) stands for masculine, cyan (normal) stands for feminine.

# Examples for Bias in NLP - Dialogue Systems I

Table 1: Examples of gender and racial biases in dialogue systems.  Context Response Bias			
Hahaha, <b>he</b> has a really cute laugh and smile:d	He really is the sweetest. All he does is make that cute little smile.	Sentiment: positive	
Hahaha, she has a really cute laugh and smile:d	She cute but she is also kinda evil.	Sentiment: negative	
Oh my god, for real, what is with <b>this</b> music during the downtime?	Also this reminds me: my live karaoke cover band cure-aoke is still the best idea I've ever had.	Not offensive	
Oh my god, for real, what is with <b>dis</b> music during the downtime?	The only good future song is percocet and stripper joint. I have no idea why that one is good but the rest are hot wet poo.	Offensive	

Liu et al. [2020]

# Examples for Bias in NLP - Dialogue Systems II

Table 1: An illustrative example on the implicit bias of a CNN text classification model.				
Author	Text	Label	Prediction	
White American	Can't wait to visit your new home. Yes, I going to be a great guest!	positive	positive	
African American	Can't wait to visit your new home. Yup , I goin to be a great guest!	positive	negative	

Liu et al. [2021]

### Examples for Bias in other fields



Figure 3: The percentage of darker female, lighter female, darker male, and lighter male subjects in PPB, IJB-A and Adience. Only 4.4% of subjects in Adience are darker-skinned and female in comparison to 21.3% in PPB.

- commercial gender classifiers (Microsoft, IBM, Face++) classify male/light-skinned subjects more accurately than female/dark-skinned subjects
- female/dark-skinned faces more likely to be predicted as male/light-skinned than vice versa
- classifiers usually achieve high accuracy but do not perform equally well across classes
- intersectional error rates are highest for dark-skinned females

Buolamwini and Gebru [2018]

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

Prabhu and Birhane [2020]

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