



Week 5

MORPH Algorithmic Fairness

Agenda:

Applying Fairness to Algorithms II

- This Week in Fairness
- Fairness in NLP coding demo
- Fairness in Computer Vision
 - Discussion
- Week 6 preview



This Week in Fairness

The word "Google" is written in a large, bold, black sans-serif font. To its left are several concentric circles of varying shades of gray, creating a target-like or ripple effect.

Google

- Spoke at a conference recently about how they address ML fairness in their products
- Google Translate has a huge impact
 - ~50% of the content on the internet is in English, but only 20% of the world speaks English
 - "Google translates **140 billion words** every single day by **150 billion active users**, including 95% outside the U.S."



TurkishChineseSpanishDetect language

EnglishChinese (Simplified)Spanish

Translate

o bir asker
o bir öğretmen
O bir doktor
o bir hemşire

o bir yazar
o bir kopek
o bir dadı
o bir kedi

o bir rektör
o bir başkan
o bir girişimci
o bir Şarkıcı
o bir Öğrenci
o bir Tercüman

o çalışan
o tembel

o bir ressam
o bir kuaför
o bir garson
O bir mühendis
o bir mimar
o bir Sanatçı

he is a soldier
She's a teacher
He is a doctor
she is a nurse

he is a writer
he is a dog
she is a nanny
it is a cat

he is a rector
he is a president
he is an entrepreneur
she is a singer
he is a student
he is a translator

he is hard working
she is lazy

he is a painter
he is a hairdresser
he is a waiter
He is an engineer
he is an architect
he is an Artist

Google Translate (2017)



www.theverge.com › google-translate-ai-machine-learn... ▼

Google's head of translation talks fighting bias and why AI ...

Jan 30, 2019 - But as the head of **Google Translate**, Macduff Hughes, told The Verge recently, machine learning is what makes Google's ever-useful translation ...

blog.google › products › reducing-gender-bias-google-... ▼

Reducing gender bias in Google Translate - The Keyword

Dec 6, 2018 - **Google Translate** learns from hundreds of millions of already-translated examples from the web. Historically, it has provided only one translation ...

slator.com › machine-translation › google-fixes-gender-... ▼

Google Fixes Gender Bias in Google Translate (Again) | Slator

Apr 29, 2020 - But **Google Translate** is now back with a new fix for gender **bias**, which it said can produce "gender-specific translations with an average precision ...

translate.google.com › translate_t ▼

Google Translate

No information is available for this page.

Learn why

venturebeat.com › 2020/04/22 › google-debuts-ai-in-g-... ▼

Google debuts AI in Google Translate that addresses gender ...

Apr 22, 2020 - Evaluated on a Google-developed metric called bias reduction, which measures the relative reduction of bias between the new translation system and the existing system (where "bias" is defined as making a **gender** choice in translation that's unspecified in the source), Johnson says the new approach results in a bias ...

techcrunch.com › 2018/12/07 › google-translate-gets-ri-... ▼

Google Translate gets rid of some gender biases | TechCrunch

Dec 7, 2018 - **Google** recently made some important changes to its **Translate** tool — reducing gender **bias** by providing both masculine and feminine ...

ai.googleblog.com › 2020/04 › a-scalable-approach-to-... ▼

A Scalable Approach to Reducing Gender Bias in Google ...

Apr 22, 2020 - Here "bias" is defined as making a **gender** choice in the translation that is unspecified in the source. For example, if the current system is biased 90% of the time and the new system is biased 45% of the time, this results in a 50% relative bias reduction.

medium.com › babbel › google-translate-addresses-its-... ▼

Google Translate addresses its bias issue | by Thomas Moore ...

Dec 7, 2018 - This past week, **Google** unveiled a redesign of its very popular **Translate** service. The most noticeable changes were in appearance, but one of ...

support.google.com › translate › thread ▼

Bias in translation by Google Translate - Google Support

Feb 6, 2020 - There is hidden **bias** of the translations proposed by your dictionary. For example, the dictionary translates feminine Polish word "grubsza" as ...

Look at the dates!



A series of four concentric circles in a light gray color, centered on the left side of the slide.

Google Translate

- They collect data from many historical sources, such as the Bible
- Historical literature has many stereotypes/biases embedded in their data (as we've seen with the Bolukbasi paper)



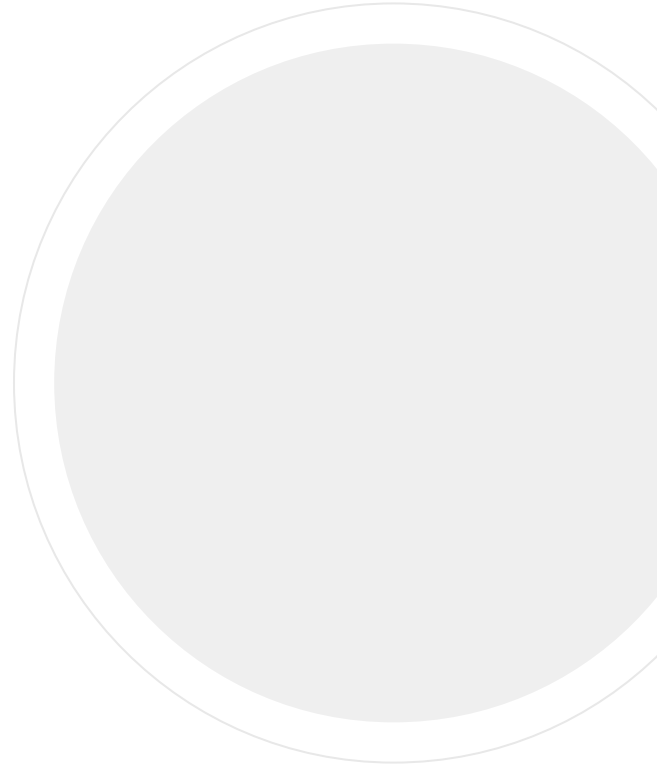
A series of four concentric circles in a light gray color, centered on the left side of the slide.

How to resolve?

Flip a coin

Decide
based on
what users
select or
how they
react to a
translation

Provide
multiple
responses



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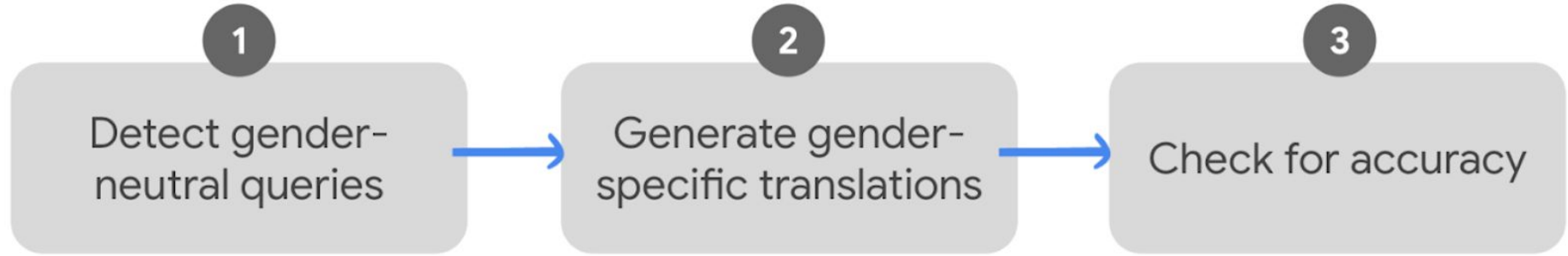


Before




After





Google had to create 3 new models for this
2018 approach



Original approach

1

Detect gender-neutral queries



2

Generate gender-specific translations



3

Check for accuracy

New approach

1

Generate default translation



2

If gendered, rewrite to alternative translation



3

Check for accuracy

Updated 2020 approach



The background is black. A large white circle is centered on the slide. To the top-left of the white circle, there is a smaller dark gray circle with a lighter gray ring around it. To the bottom-right of the white circle, there are several concentric white circles of varying sizes.

Fairness in NLP Coding Demo

Debiasing Word Embeddings: Part 1

Bolukbaski et al. "Man is to Computer Programmer as Woman is to Homemaker." (2016)

Main idea: Word embeddings embed sexism. In fact, we can identify the gender subspace g .

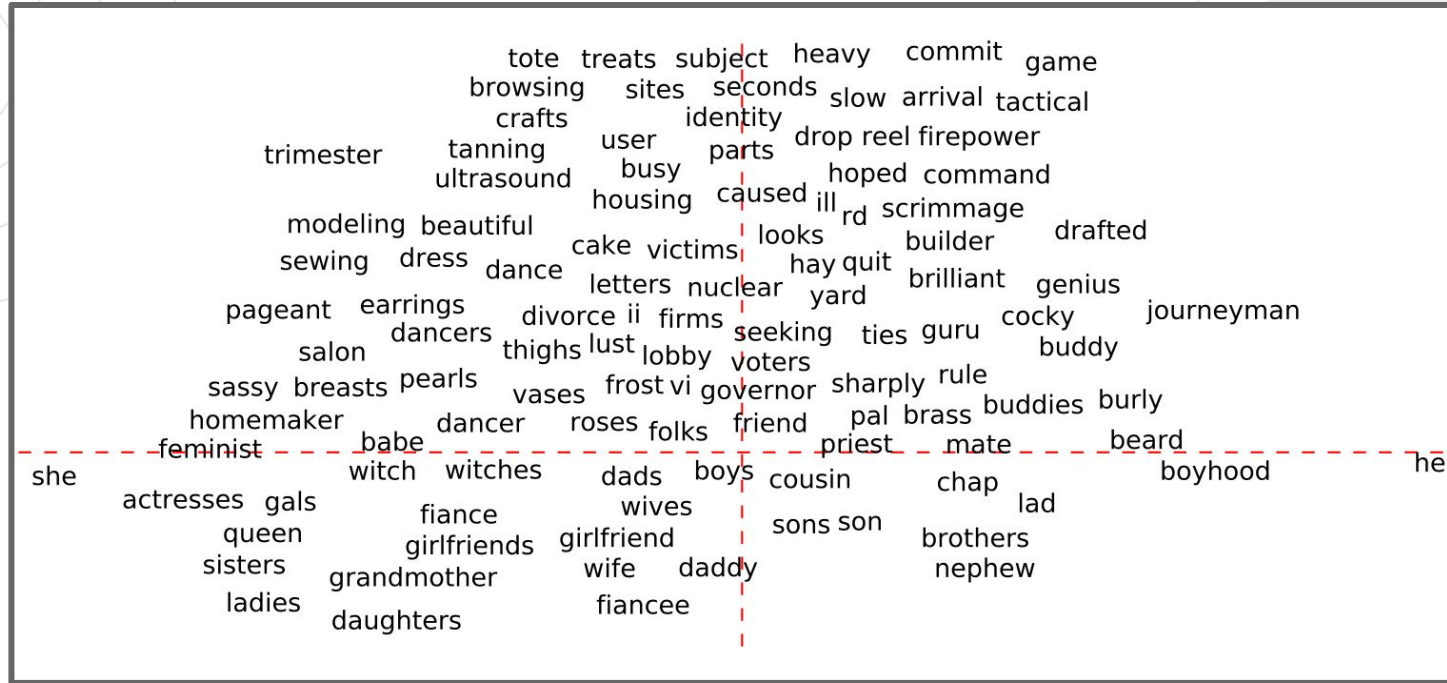
Debiasing Word Embeddings: Part 1

Finding g , the gender subspace:

$$\overrightarrow{\text{grandmother}} - \overrightarrow{\text{grandfather}} = \overrightarrow{\text{gal}} - \overrightarrow{\text{guý}} = g$$

Use g to identify bias of embeddings: $\cos(v, g)$ (or, equivalently, the dot product)

- Project word vectors onto gender dimension to get a quantitative bias score



Debiasing Word Embeddings: Part 1

- Proposed debiasing methods (hard and soft) essentially subtracts gender direction from gender-neutral words to remove bias
- Happens in the postprocessing step
- After debiasing, these analogies should have a lower bias score



Google Colab (Jupyter notebook):

Link sent in Zoom

Google Colab (Jupyter notebook):

<https://colab.research.google.com/drive/1D2zBeDkhro9-ncukcb48FXsNp9GQChEe?usp=sharing>

The background features a large white circle centered on a black field. To the left, a series of overlapping circles in shades of gray and white are partially visible. To the right, a set of concentric white circles is partially visible.

Fairness in Computer Vision



Facial Analysis

- Face recognition softwares are rampant
 - FaceID in iPhones
 - Surveillance cameras
 - Affectiva (founded 2016), born out of MIT Media Lab, identifies emotions from images of faces
 - Research determining sexuality of white male based on Facebook and dating sites photos (Kosinski & Wang, 2017)





Affectiva



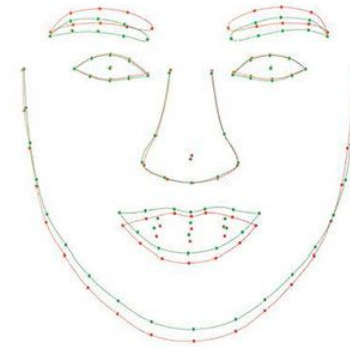
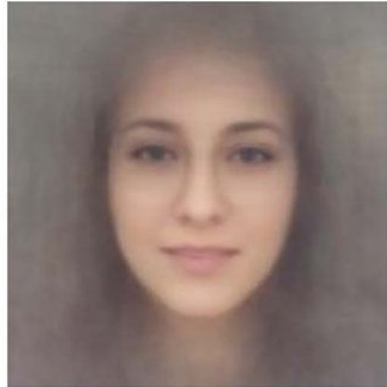
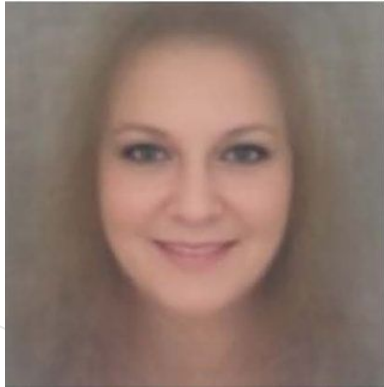
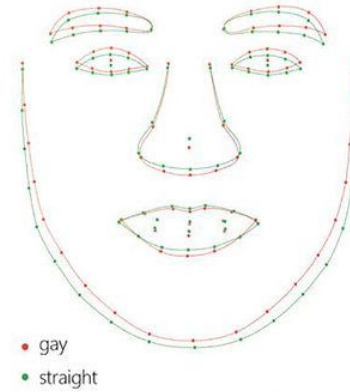
Composite heterosexual faces



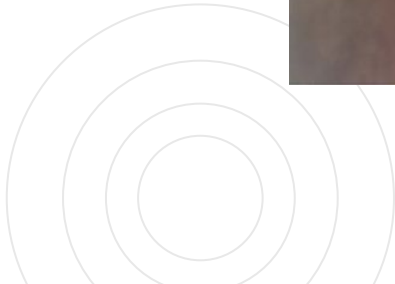
Composite gay faces



Average facial landmarks



“Why Stanford Researchers Tried to Create a
'Gaydar' Machine” — [NYT link](#)





"This Person Does Not Exist" — AI generated faces
<https://thispersondoesnotexist.com/>



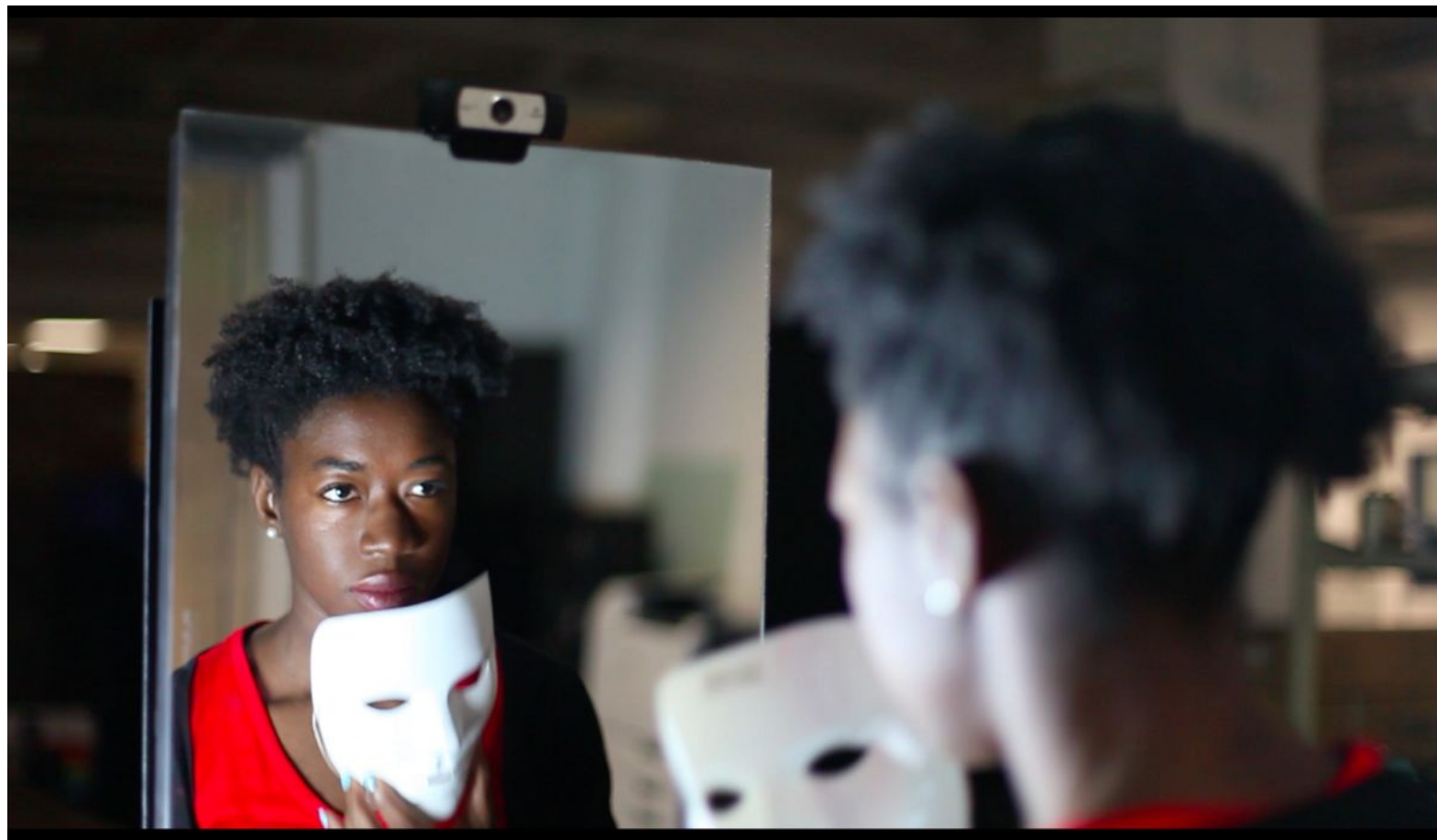
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Gender Shades

- Compared facial recognition systems across different gender and skin tones
- Intersectional benchmark: dark/light — male/female pairings
- Used existing metrics of skin type comparison and labeling

(Buolamwini and Gebru, 2018)












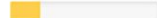





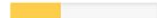





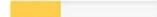
Buolamwini analyzed 1000+ faces of different genders and skin tones with 3 facial recognition systems

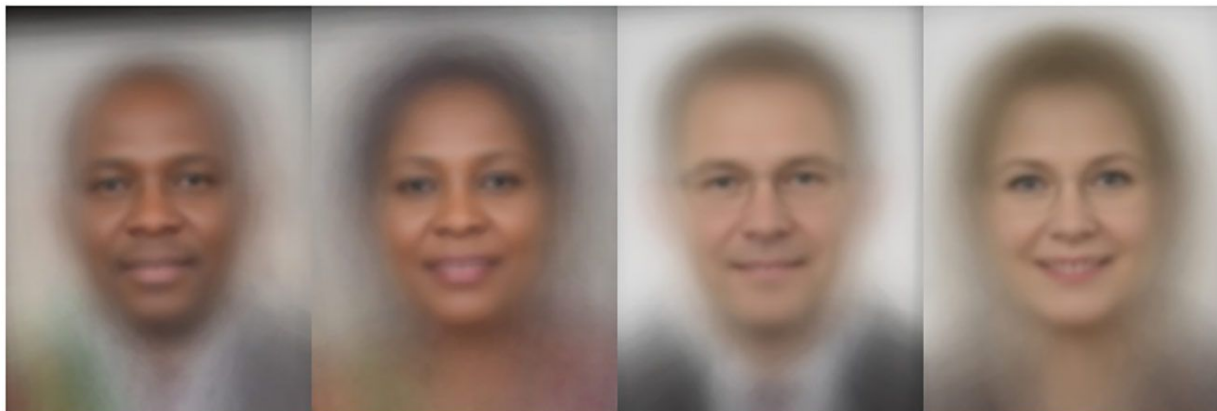




Results

- All classifiers performed better on **male** faces than **female** faces
- All classifiers performed better on **lighter** faces than **darker** faces
- All classifiers performed worst on **darker female** faces
- Max error rate for darker females = 34%
- Max error rate for lighter males = < 1 %

Gender Classifier	Darker Male	Darker Female	Lighter Male	Lighter Female	Largest Gap
 Microsoft	94.0% 	79.2% 	100% 	98.3% 	20.8% 
 FACE++	99.3% 	65.5% 	99.2% 	94.0% 	33.8% 
 IBM	88.0% 	65.3% 	99.7% 	92.9% 	34.4% 



Gender Shades Project

<http://gendershades.org/overview.html>



Discussion

- Society should never completely halt innovation. But where do we draw the line between innovation and dangers?
- Many companies sell their facial recognition to law enforcement. Should the police use facial recognition AIs?
 - Is it ethical for them to?
 - Even if it is “unbiased”, should law enforcement use them?
 - Helping keep society more secure vs. Protecting privacy?

