

I THINK THIS "ROBOTS DESTROYING HUMANITY" SCENARIOS ARE BEING OVERTHOUGHT...

THANKS, YOU GUYS. OH JEEZ, I JUST DECIDED TO HACK INTO PENTAGON WITH MY SUPER INTELLIGENCE AND TRIGGER ALL AVAILABLE NUCLEAR MISSILES...

WE DID IT ! WE FINALLY MANAGED TO GET THIS OLD COMPUTER TO THINK !!!

WHAT HAVE WE DONE?

Artificial Intelligence and Global Risks

VISION

- The biggest problems with AI today lay in bias
- Why
 - Data keeps growing
 - Used by everyone in some capacity
 - Smart phones, social media, etc.

BUT WE'RE FOCUSED ON THE WRONG THINGS

- AGI vs weak AI
- Killer robots
- Solely funding AGI research

HOW CLOSE ARE WE TO
GENERAL INTELLIGENCE?

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HOW CLOSE?

- Policy - thinks 10 years
- AI experts - not 100 years
- Instructors
 - Leilani - Never
 - Matías - 80 years

MAIN ETHICAL SYSTEMS

- Asimov Laws
- Robot Ethics Charter and EURON - not much follow up
- Asilomar AI Principles
- Start Russell - 3 principles of Human -compatible AI
- AI Now Recommendations

BUT NO RULES ABOUT BIAS

SO WHAT ARE WE MISSING

Harms of Representation

WHAT IS AN ALGORITHM?

- A specification of how to solve a problem
- Fibonacci
 - add the last two numbers
 - or $F(n) = F(n-1) + F(n-2)$
- Pseudocode

```
1 begin
2     t = 0;
3     initialize particles P(t);
4     evaluate particles P(t);
5     while (termination conditions are
6         unsatisfied)
7         begin
8             t = t + 1;
9             update weights
10            select pBest for each particle
11            select gBest from P(t-1);
12            calculate particle velocity
13            P(t);
14            update particle position P(t)
15            evaluate particles P(t);
16        end
17    end
```

EXAMPLE : GALE-SHAPLEY

- Used for residency matching
- TODO - walk through

WHAT IS AN REPRESENTATION?

- Representation = identity
- How to do you represent _ to a computer?
 - Images = Pixels
 - Human applicants - race, age, etc.
- Latanya Sweeney in 2013
- Representations are difficult to formalize

Ads by Google

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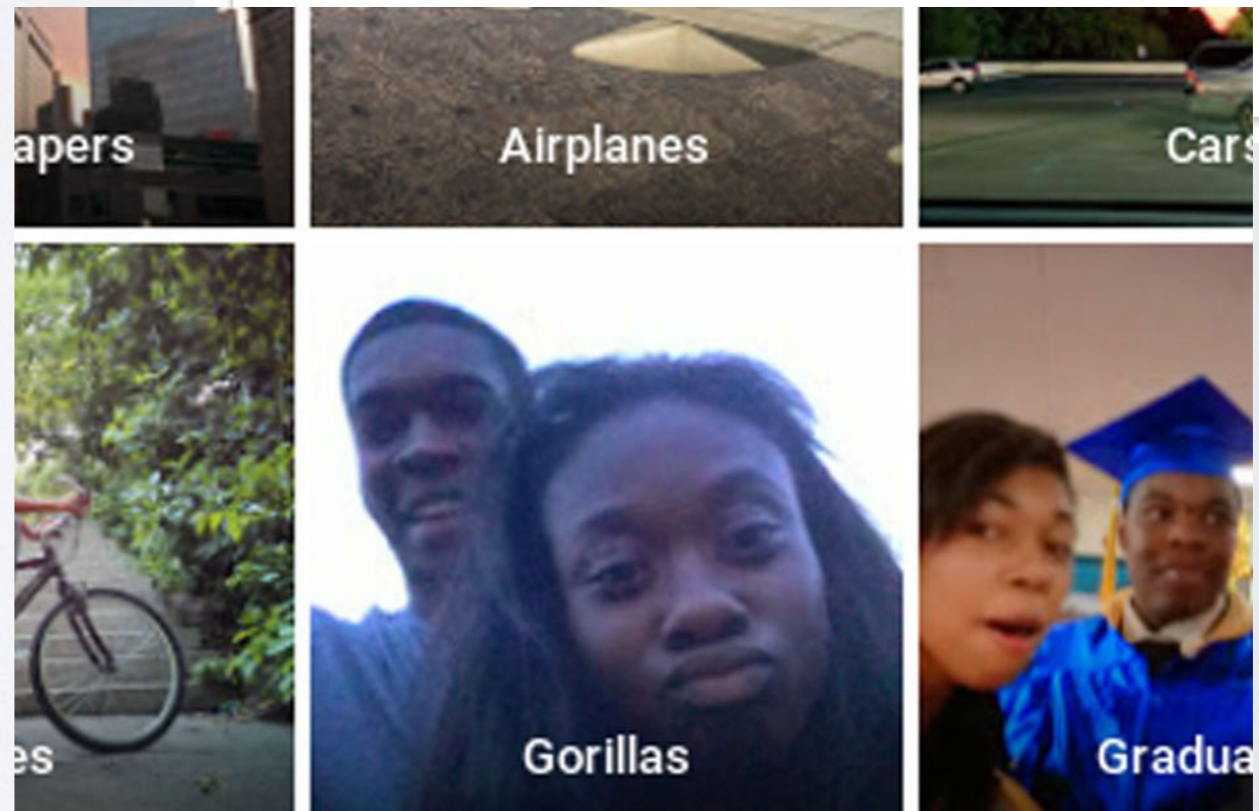
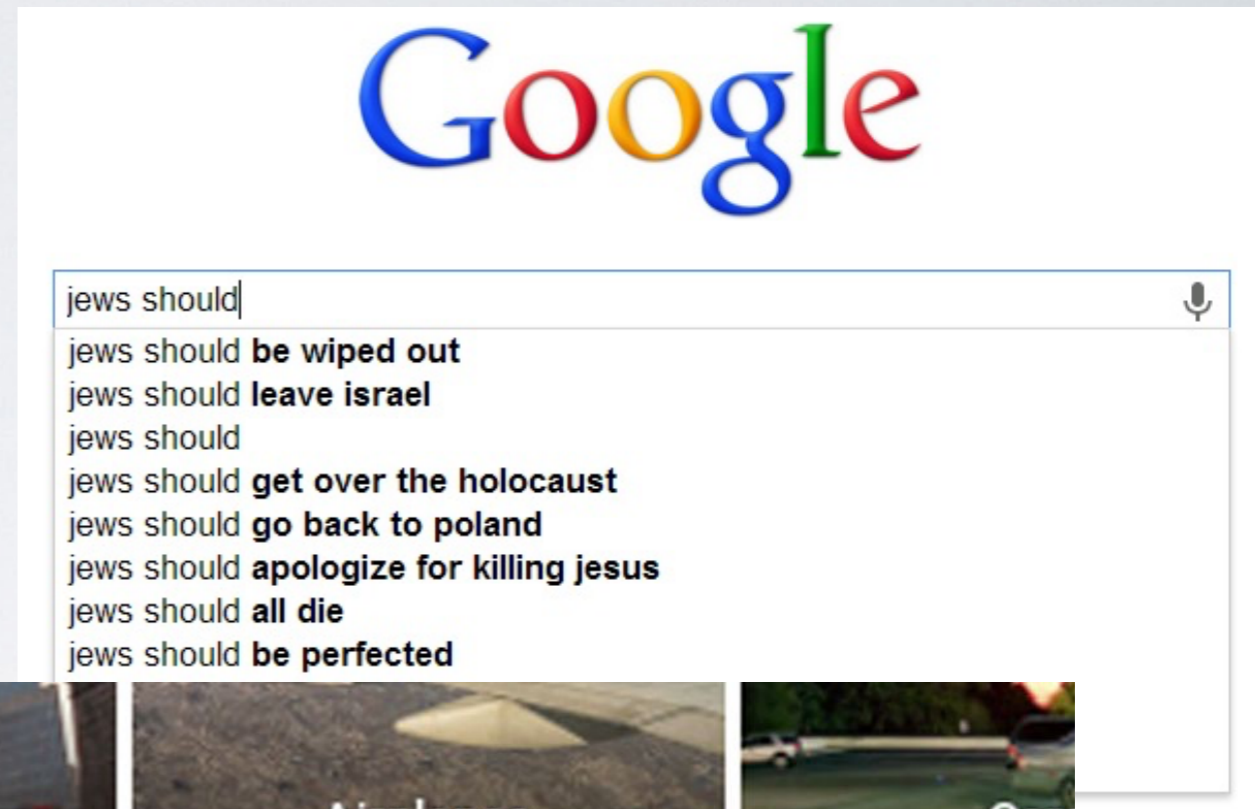
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5 HARMS OF REPRESENTATION

1. Denigration
2. Stereotyping
3. Recognition
4. Under-representation
5. Ex-nomination

DENIGRATION

- When people use culturally offensive or inappropriate labels
- Examples
 - Google photos that produce racist labels
 - Autosuggestions when people typed, “Jews should...”



STEREOTYPING

- Word embeddings are biased
- Google translate
- Problems with natural language processing

Extreme <i>she</i> occupations		
1. homemaker	2. nurse	3. receptionist
4. librarian	5. socialite	6. hairdresser
7. nanny	8. bookkeeper	9. stylist
10. housekeeper	11. interior designer	12. guidance counselor

Extreme <i>he</i> occupations		
1. maestro	2. skipper	3. protege
4. philosopher	5. captain	6. architect
7. financier	8. warrior	9. broadcaster
10. magician	11. fighter pilot	12. boss

Gender stereotype <i>she-he</i> analogies.		
sewing-carpentry	register-nurse-physician	housewife-shopkeeper
nurse-surgeon	interior designer-architect	softball-baseball
blond-burly	feminism-conservatism	cosmetics-pharmaceuticals
giggle-chuckle	vocalist-guitarist	petite-lanky
sassy-snappy	diva-superstar	charming-affable
volleyball-football	cupcakes-pizzas	hairdresser-barber

Gender appropriate <i>she-he</i> analogies.		
queen-king	sister-brother	mother-father
waitress-waiter	ovarian cancer-prostate cancer	convent-monastery

RECOGNITION

- When a group is erased or made invisible by a system
 - Purely a technical problem
- Example
 - Does a system recognize a face inside an image or video?
 - Some facial recognition couldn't process darker skin tones
 - Asians characterized as "blinking"
- Failure to recognize someone's humanity



UNDER-REPRESENTATION

- Technical issues
 - Improve accuracy
 - Scrubbing to neutral
 - But what is neutral?
- Demographics of equal representation
 - What do we mean by equal?
- Awareness



EX-NOMINATION

- Barthes' term for the phenomenon whereby the bourgeoisie hides its name (and identity) by not referring to itself as such in order to naturalize bourgeois ideology and maintain its hegemony
- Eliminate social identity
 - Need transparency of school preference ranking (not matching)
 - Criminal justice - strong cases of discrimination in stop and frisk

AN ALTERNATIVE

- The trustworthiness system of Chinese Social Credit.
- We need to keep in mind who benefits and who is harmed.

CASES OF BIAS

- Judgement on preconceived notions
- Very difficult to fix with data
- Algorithmic flaws aren't easily discoverable
 - How would a woman know to apply for a job she never saw advertised?
 - How might the black community learn that it were being over-policed by software

NEW YORK DNA

- Incomplete DNA samples
- TrueAllele - probability genotyping
- Opaque algorithms and analysis for filling in the gaps
- Similar case in Australia
 - Incorrect results in 60 criminal cases
 - Altered statistics by a factor of 10
 - Prosecutors were forced to replace 24 experts

CALIFORNIA VS. JOHNSON

EXTREME VETTING INITIATIVE

MODELS OF REGULATION

- Black box

EXAMPLES OF REGULATION

RIGHT TO EXPLANATION

- EU
- Could this lead to biased explanations?

NEW YORK CITY COUNCIL

- EU
- Could this lead to biased explanations?

NEW YORK ACTUARIAL RISK

- EU
- Could this lead to biased explanations?

WHERE DOES ALL THIS COME FROM?

- Data that the models were trained on
 - Sometimes human labelled
 - Sometimes machine labelled

WHERE ARE WE NOW?

- Harms of allocation
- Allocation = resources
- An economic view
 - Home loans, etc.

	Denigration	Stereotype	Recognition	Under-representation	Ex-nomination
NY DATA					
California vs. Johnson					
Extreme vetting Initiative					
Stable marriage - schools					
Stop and Frisk					