

# Who Designs for Whom?

Michael Paul  
April 12, 2016

University of Colorado  
CMCI 1020: Concepts and Creativity

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# Important questions for Information Technology design

- Who are the **users**?
- Who are the **designers**?



**BETTER**  
**OFF TED**



# HP computers are racist



wzamen01

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# Questions to think about

- How can we ensure technology is inclusive?
  - How do we design for diverse users?
  
- What does it mean for a computer to be racist?
  - Can algorithms discriminate?

# Important questions for Information Technology design

- Who are the **users**?
- Who are the **designers**?

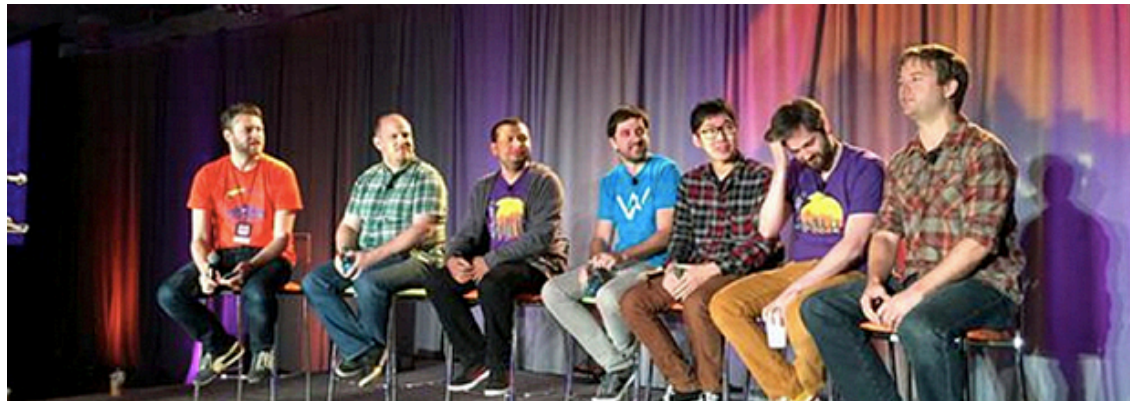
# Important questions for Information Technology design

- Who are the **users**?
  - Potentially, everyone
- Who are the **designers**?
  - Generally: the tech industry



# Important questions for Information Technology design

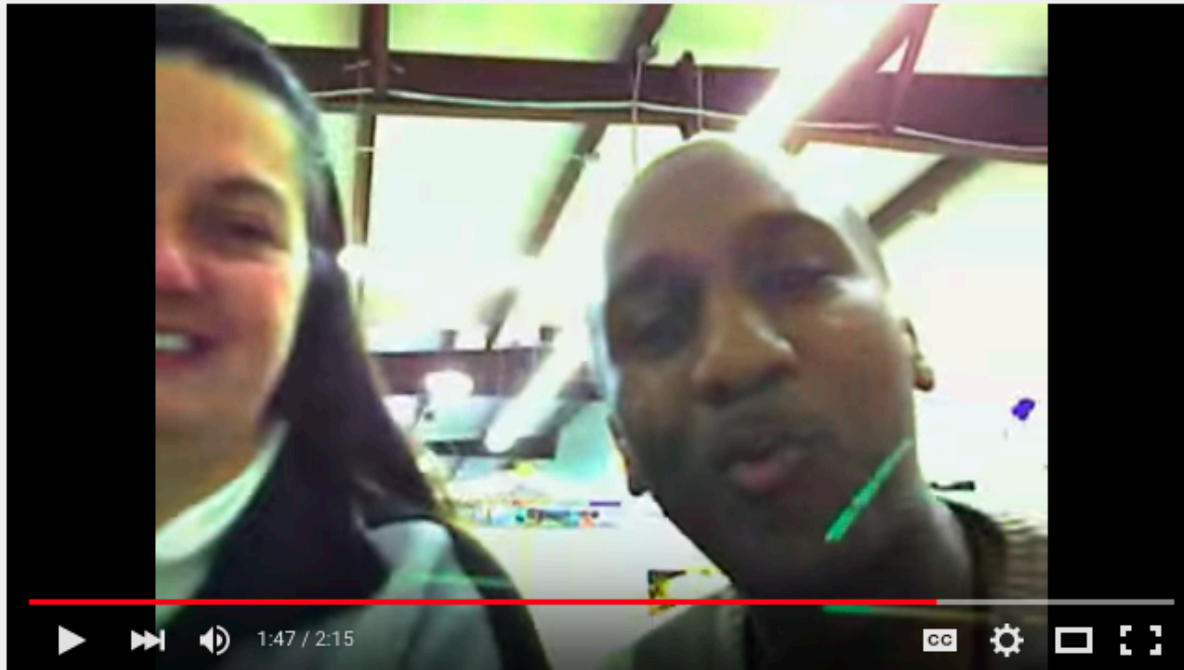
- Who are the **users**?
  - Potentially, everyone
- Who are the **designers**?
  - Generally: the tech industry
  - On average: male, under 40, white, Californian, highly educated



What happens when **users** look and think differently from the **designers**?

# Face Recognition

≡ YouTube



HP computers are racist





wzamen01

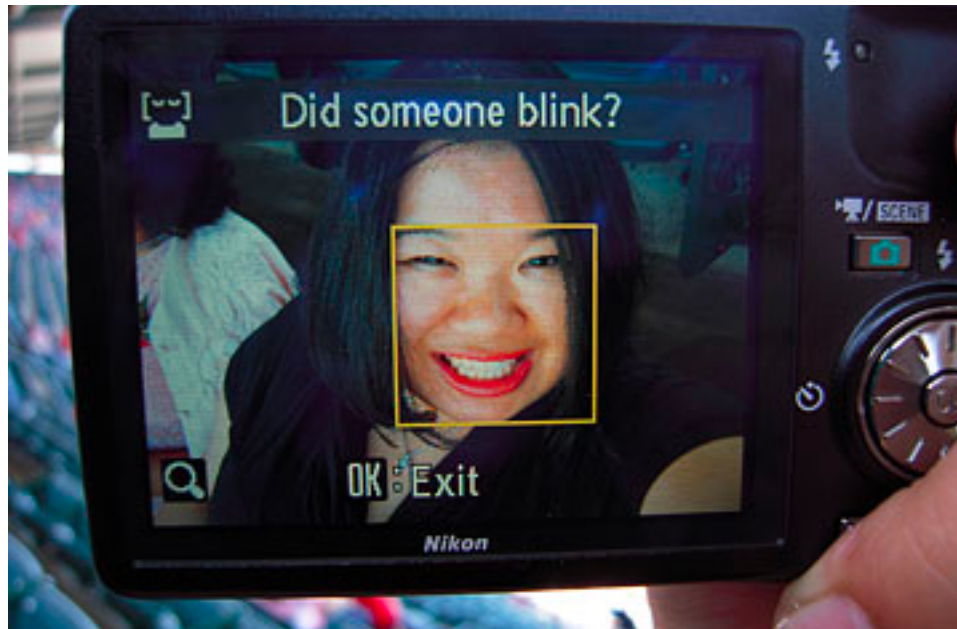
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# Face Recognition



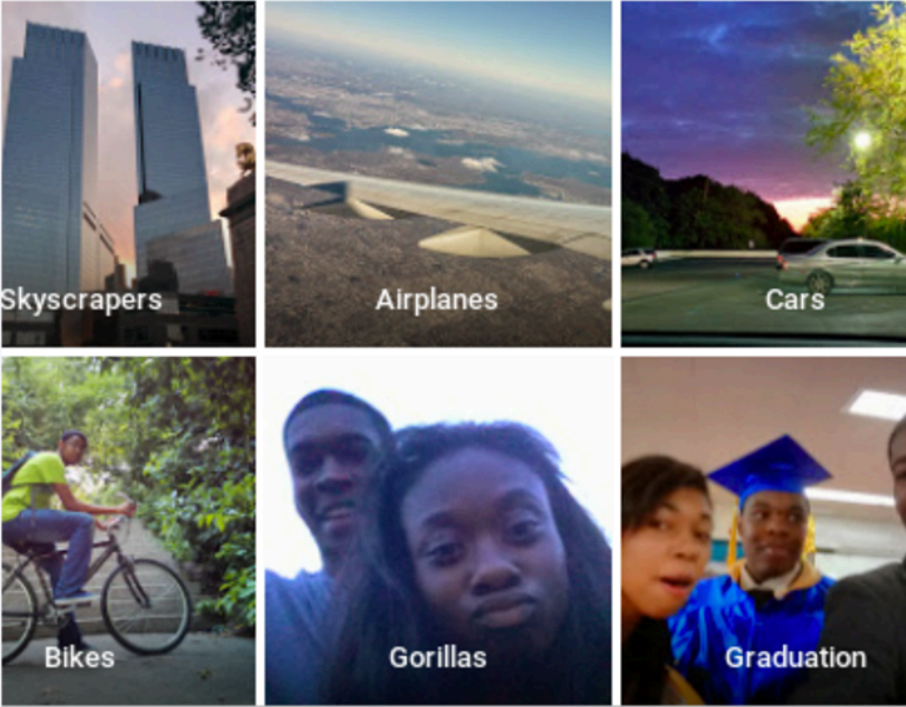
We got our Mom a new Nikon S630 digital camera for Mother's Day and I was playing with it during the Angels game we were at on Sunday.

As I was taking pictures of my family, it kept asking "Did someone blink?" even though our eyes were always open.


Sheesh! RACIST! 🙄

from: Joz Wang

# Face Recognition



A 2x3 grid of photos with labels: Skyscrapers, Airplanes, Cars, Bikes, Gorillas, Graduation.

 **Jacky Alcine**  
@jackyalcine [Follow](#)

Google Photos, y'all fucked up. My friend's not a gorilla.  
7:22 PM - 28 Jun 2015

↩️ ↻ 3,325 ❤️ 1,937

# Voice Recognition

Some voice recognition systems don't work well for women, or people with accents

Example: **OnStar**  
Voice-activated phone system for cars



“When evaluating the technology, husbands and wives are often divided.” Source: <http://www.autoblog.com/2011/05/31/women-voice-command-systems/>

# Voice Recognition

OnStar optimized their system for different groups over time:

1. Men
2. Women
3. Midwest accents
4. Southern accents
5. New England accents



and there's still more to do... (e.g., foreign accents)

# Wearables/Devices



One big problem for lefties: In landscape mode, the dock is far from the dominant thumb.

Source: [Mic](#)



# Wearables/Devices

Japanese company creates thumb extension for new iPhone owners with small hands



# Wearables/Devices



**Adriana Lee**  
@adra\_la



Follow

Whatever smartwatch that is looks gigantic on a woman's wrist #IO14



RETWEETS  
**17**

LIKES  
**12**



9:39 AM - 25 Jun 2014

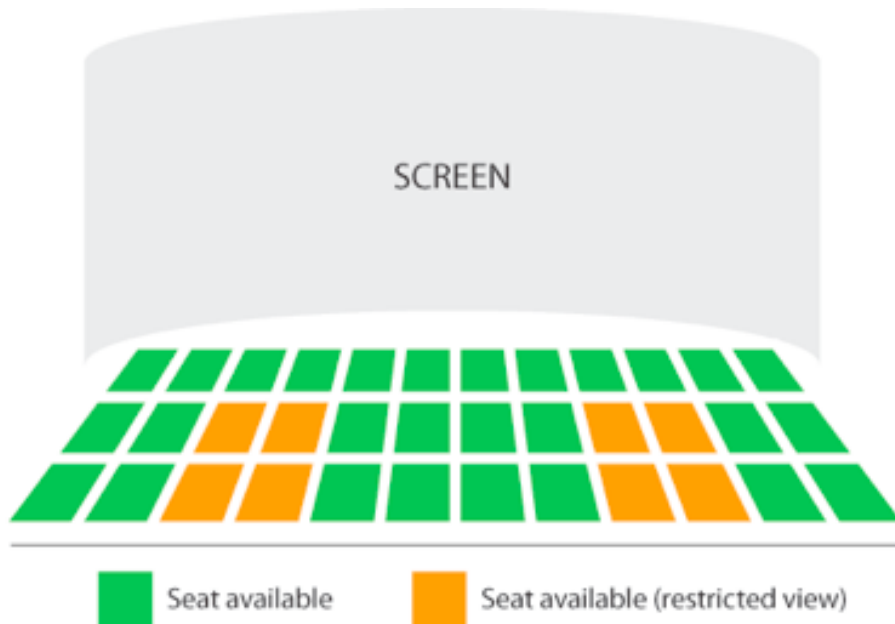
# Wearables/Devices



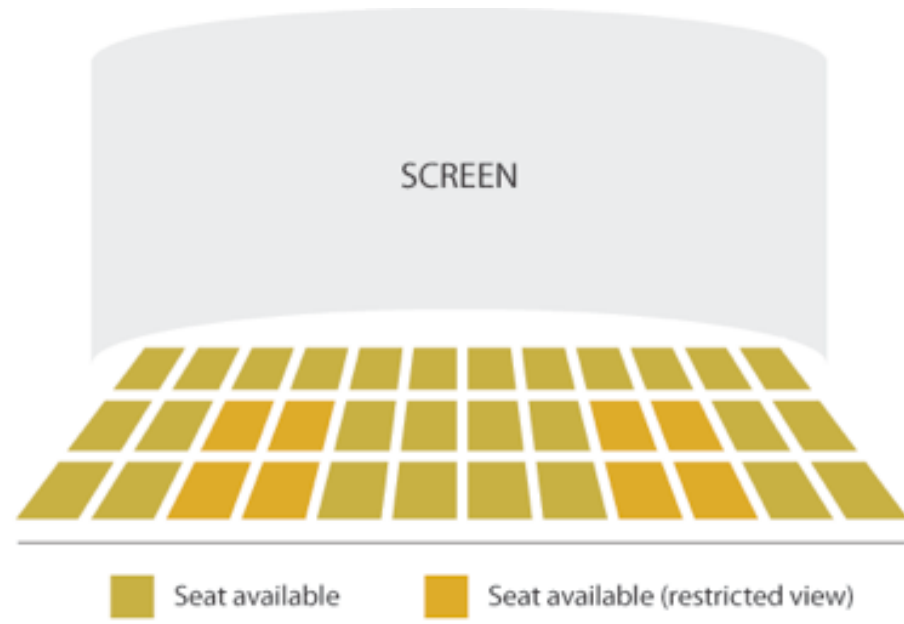
Source: <http://smartwatches.org/learn/finding-smartwatch-as-a-woman/>

# Accessibility

Full color:



With red-green color blindness:



Source: <http://www.digitalartsonline.co.uk/news/graphic-design/how-design-for-colour-blind/>

# Accessibility



Hillary Clinton @HillaryClinton · Aug 30

**"We are smart enough,  
we are compassionate enough,  
to figure out how to balance  
the legitimate Second Amendment  
rights with preventive measures,  
and control measures, so that...  
we will not see more deaths."**

Hillary Clinton  
August 27, 2015

↩ 904 ❤️ 1.6K ⋮



Jeffrey P. Bigham @jeffbigham · Aug 30

@HillaryClinton another tweet missing alt text. blind users can't read these posts. doesn't speak highly for your respect for accessibility

LIKES

3



10:26 AM - 30 Aug 2015 · Details

# Name Validation

BBC

Computer

**These unlucky people have names that break computers**

PLEASE ENTER A  
VALID NAME

# Name Validation

“When Jennifer Null tries to buy a plane ticket, she gets an error message on most websites. The site will say she has left the surname field blank and ask her to try again.”



<null>

# Name Validation

Some systems will reject other types of names

- Too long
  - Keihanaikukauakahihulihe'ekahaunaele
- Special characters
  - Hyphenated last names (e.g., Jordan Boyd-Graber)
- Non-English alphabet



# Name Validation

## **I Can Text You A Pile of Poo, But I Can't Write My Name**

We can't ignore the composition of the Unicode Consortium's members, directors, and officers -- the people who define the everyday writing systems of all languages across the globe.

—by [Aditya Mukerjee](#) on March 17th, 2015

<https://modelviewculture.com/pieces/i-can-text-you-a-pile-of-poo-but-i-cant-write-my-name>

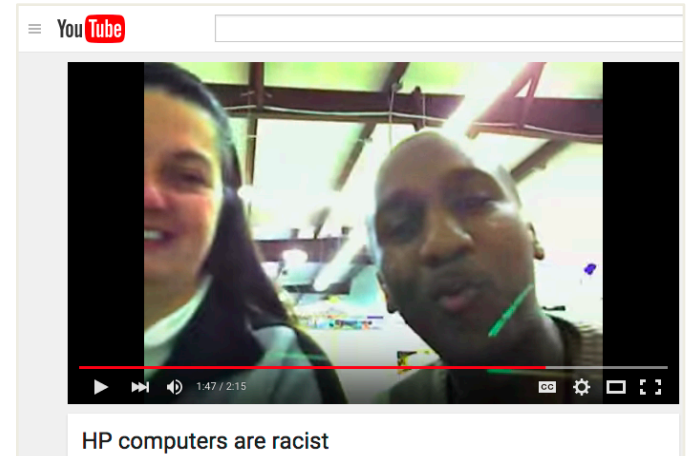
# Name Validation

“We have an unambiguous, cross-platform way to represent PILE OF POO, while we’re still debating which of the 1.2 billion native Chinese speakers deserve to spell their own names correctly.”



How can technology designers  
avoid these problems?

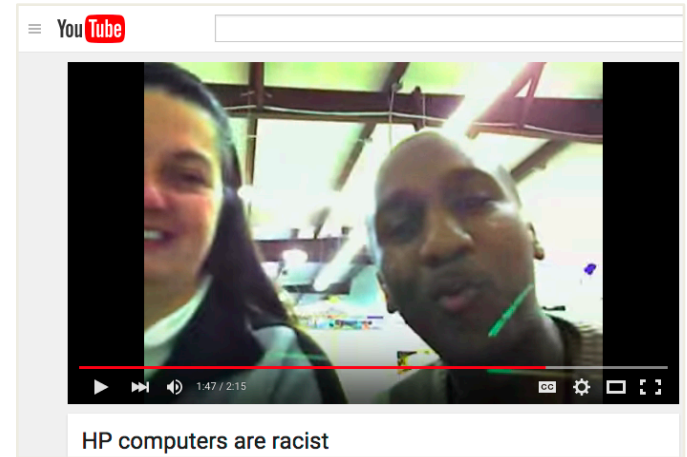
How could this happen?  
Didn't they test the product?



How could this happen?  
Didn't they test the product?

They surely did, but:

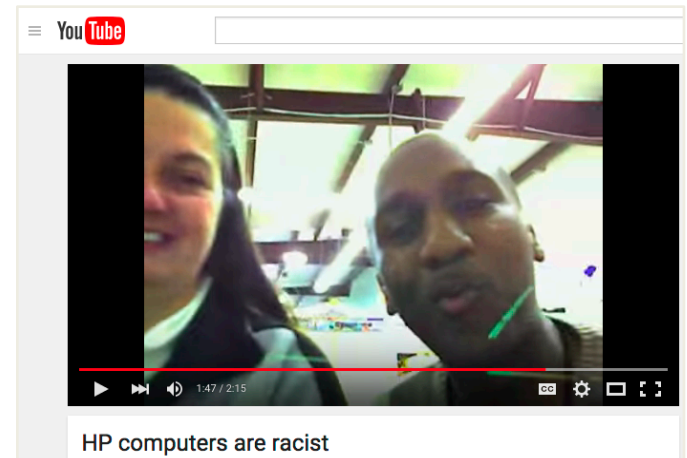
- Maybe they didn't "train" it on black faces
- Maybe they didn't test it on black faces
- Maybe they didn't know about the failure



How could this happen?  
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They surely did, but:

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# “Training Data”

Algorithms are “trained” on example data

How do you build a face recognizer?

Feed the algorithm examples of faces



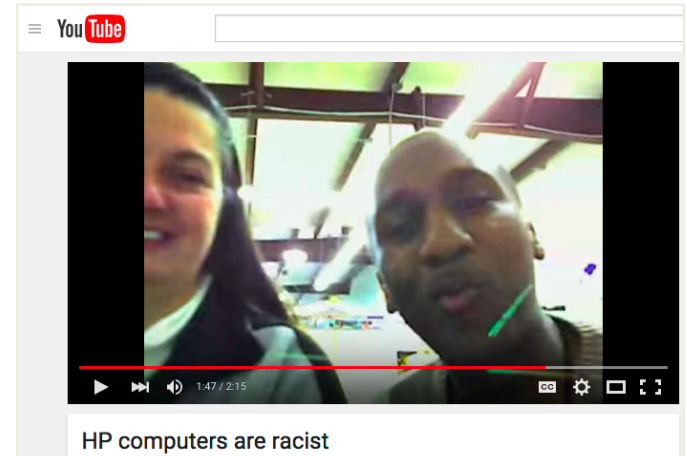
# “Training Data”

How could this happen?

Didn't they test the product?

They surely did, but:

- Maybe they didn't “train” it on black faces



Important to have representative data for training algorithms!

- Need to collect data beyond the research lab

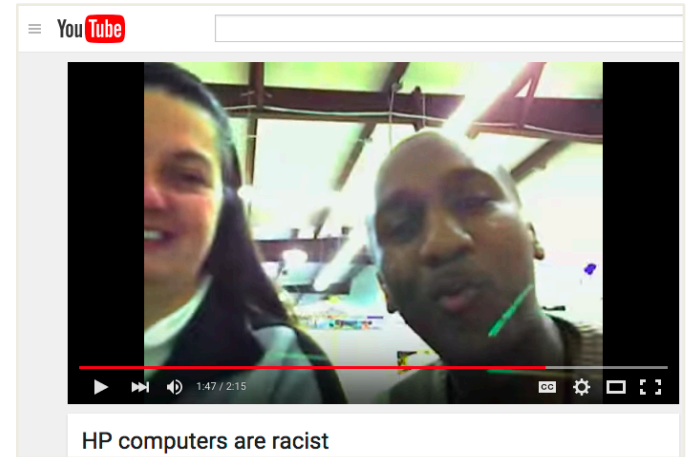


# “Training Data”

How could this happen?  
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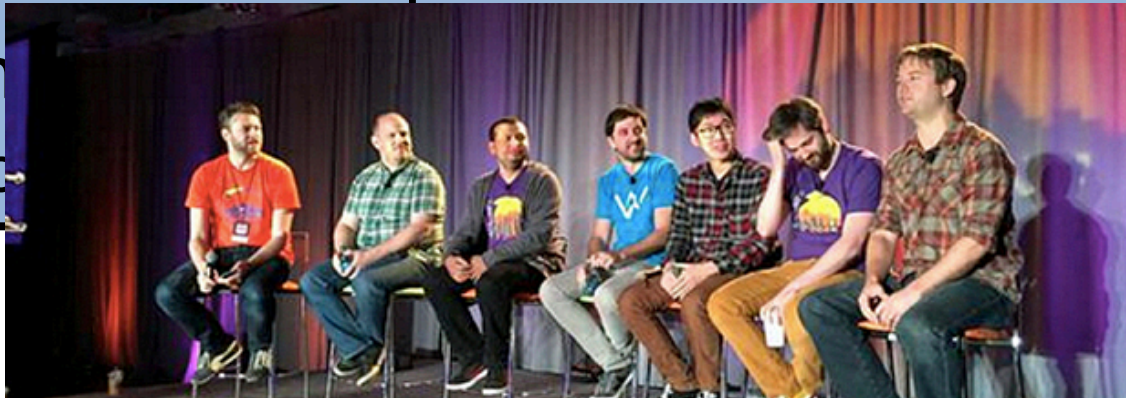
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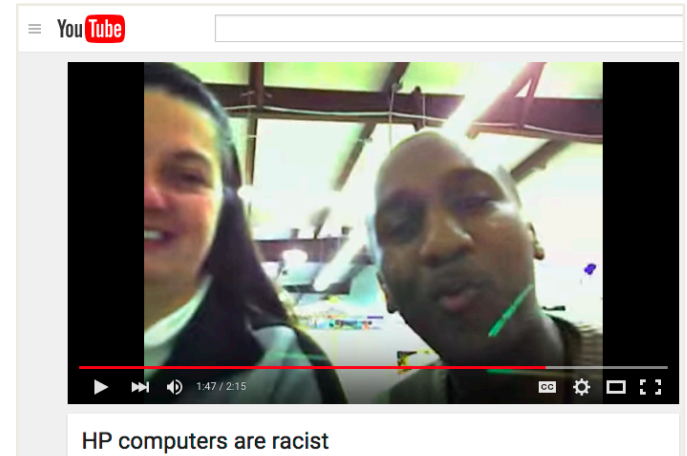
- Need to



How could this happen?  
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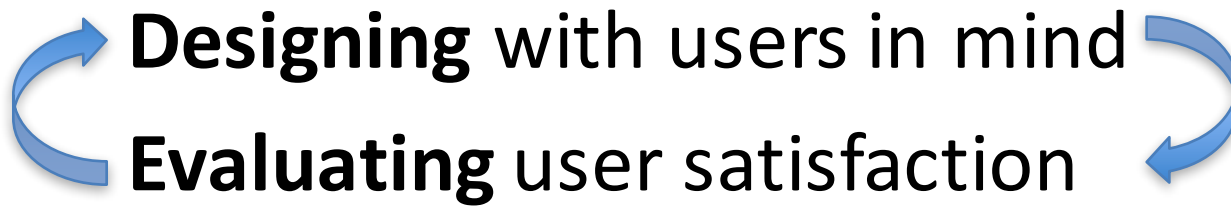
- Maybe they didn't "train" it on black faces
- **Maybe they didn't test it on black faces**
- Maybe they didn't know about the failure



# User-Centered Design (UCD)

Designing for the **end user**

Iterative process:



# User-Centered Design (UCD)

Why is UCD important?

- User needs may differ from business needs
- User interests may differ from designer intention
- Hard to anticipate user expectations, reactions

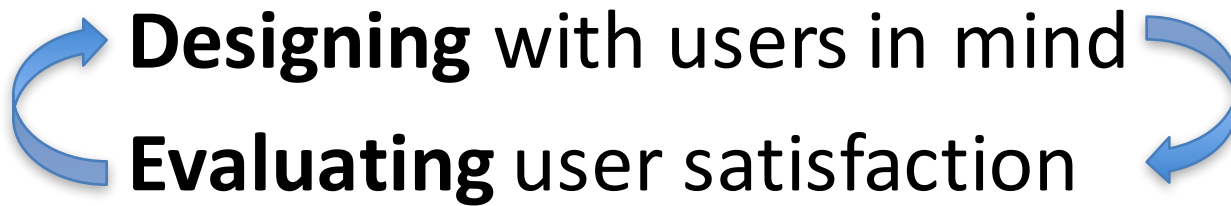
“What happens when designers look different from users?”

- UCD can help – if done right

# User-Centered Design (UCD)

Designing for the **end user**

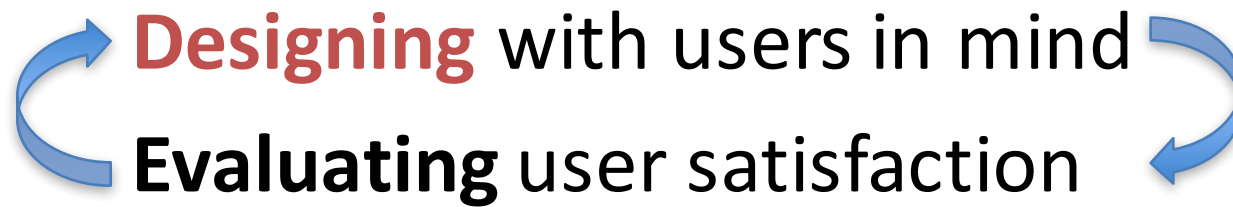
Iterative process:



# User-Centered Design (UCD)

Designing for the **end user**

Iterative process:



# Designing

- Important concept: **usability**

## Usability principles:

- Efficient
- Easy to use/learn
- Satisfying

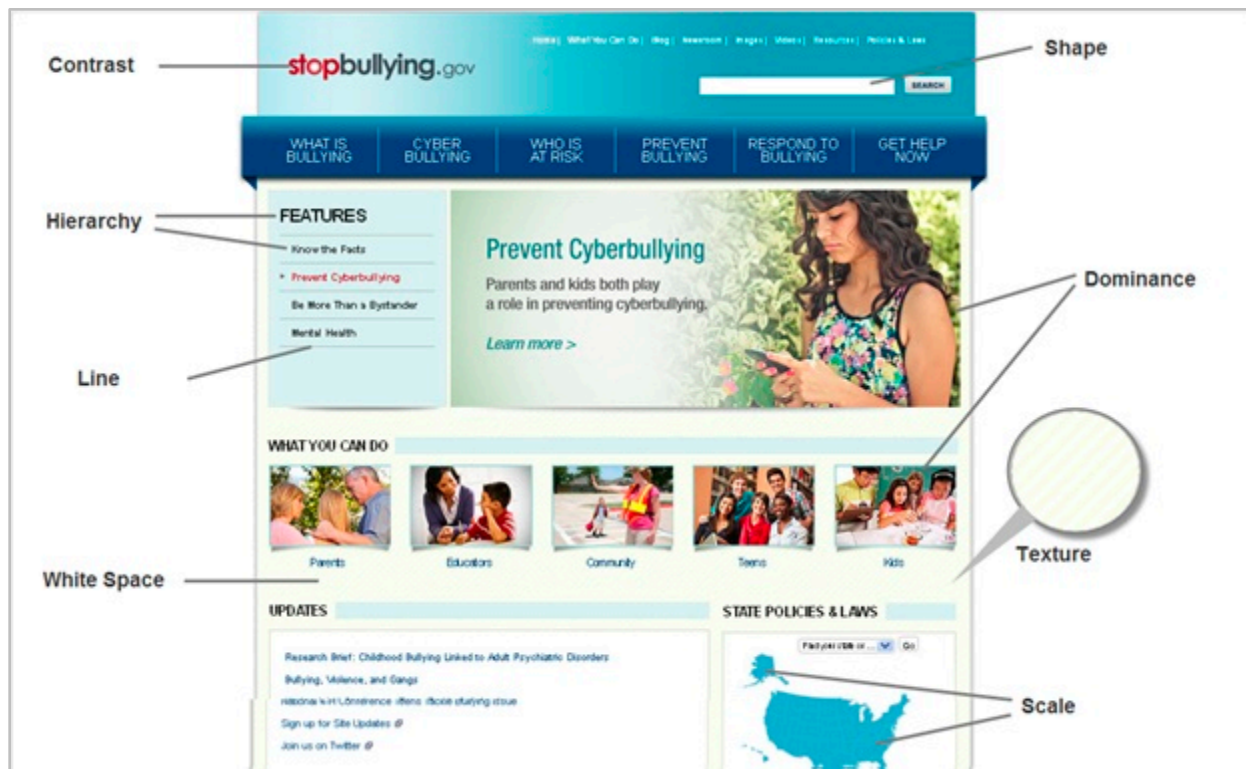


## Not usable:



# Designing

There are a number of **guidelines** for good and usable design



Resource: [usability.gov](http://usability.gov)



# Designing

- Important concept: **accessibility**
  - How usable is technology for users with disabilities?

Technology can **remove barriers** to communication and interaction and can improve lives

Technology can also **create barriers** and leave people behind if it is poorly designed

# Designing

A number of small things can make websites and apps more accessible...

Accessibility principles:

- adjustable font sizes
- alternate text for images
- organized navigation
- navigable without mouse
- avoids epilepsy triggers

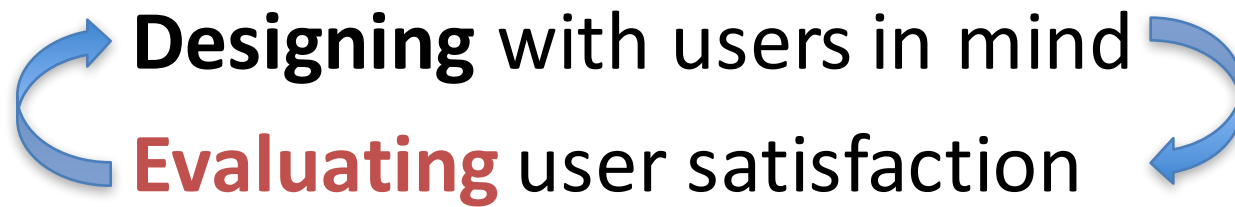
Not accessible:



# User-Centered Design (UCD)

Designing for the **end user**

Iterative process:



# Evaluating

How can we **measure** the usability of a product?

Some possibilities:

- Ask the users about their experience
- Observe/record the users in action
- Measure how successfully a product is used

# Evaluating

One approach:

Follow-up with users **after they've used** your product/system

- e.g., surveys, interviews, focus groups

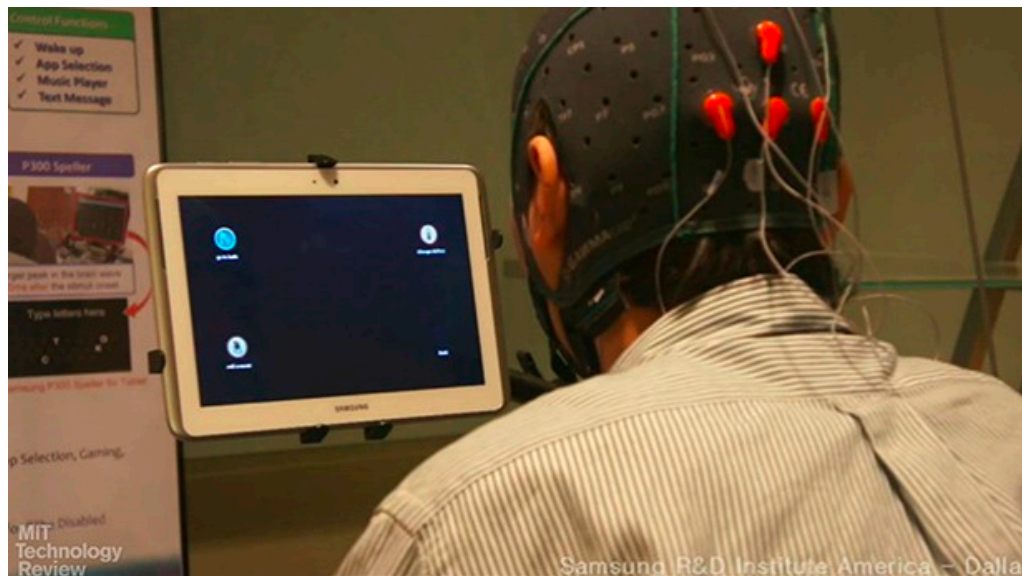


# Evaluation

Another approach:

Learn from users **while they're using** your product/system

- observation, testing

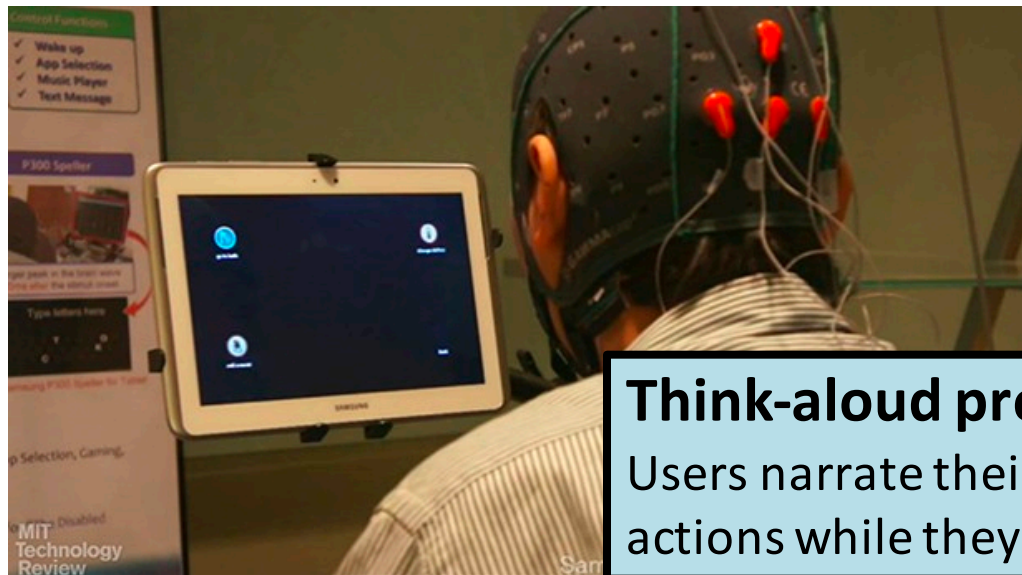


# Evaluation

Another approach:

Learn from users **while they're using** your product/system

- observation, testing



**Think-aloud protocol:**

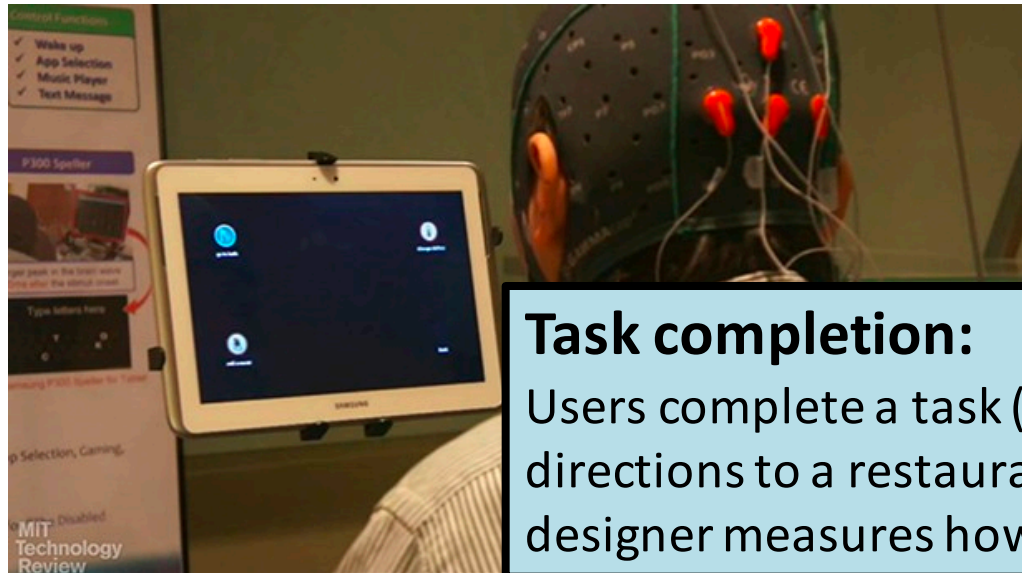
Users narrate their thoughts and actions while they use a system

# Evaluation

Another approach:

Learn from users **while they're using** your product/system

- observation, testing



## Task completion:

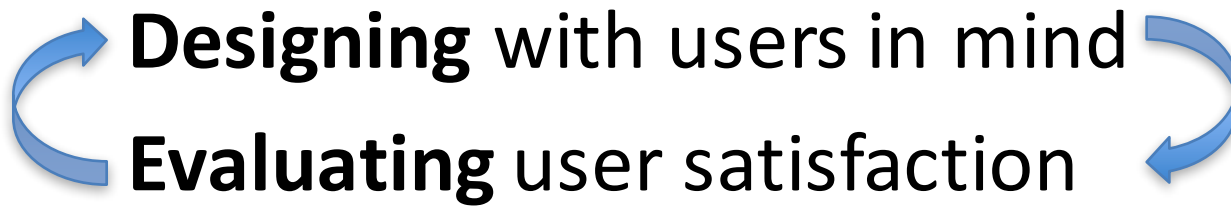
Users complete a task (e.g., find directions to a restaurant) and the designer measures how long it takes



# User-Centered Design (UCD)

Designing for the **end user**

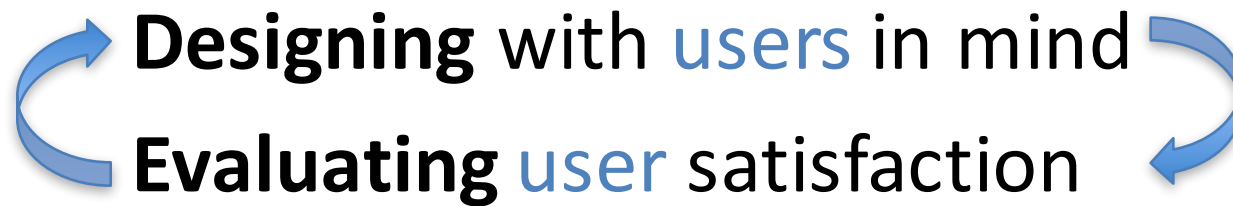
Iterative process:



# User-Centered Design (UCD)

Designing for the **end user**

Iterative process:



Key question: **Who** is the user?

# User-Centered Design (UCD)

Who is the **target user**? Who are we designing for?

What if we are designing for...

- Teenagers?
- Pregnant women?
- War veterans?
- The “general public”?
  - Is there such a thing?

Users will vary by:

- age, gender, race, ethnicity
- education level, income level
- language, nationality
- marital status, number of children
- job history
- health history
- familiarity with product
- familiarity with technology in general

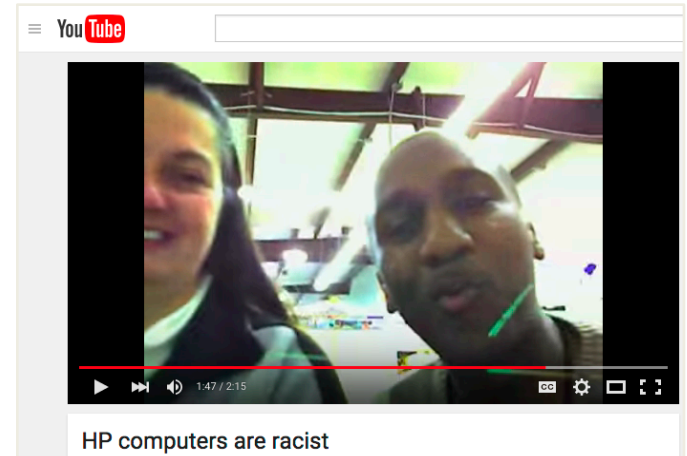
# User-Centered Design (UCD)

How could this happen?

Didn't they test the product?

They surely did, but:

- Maybe they didn't test it on black faces



Important to test on the target user!

Many companies test on other employees at the same company

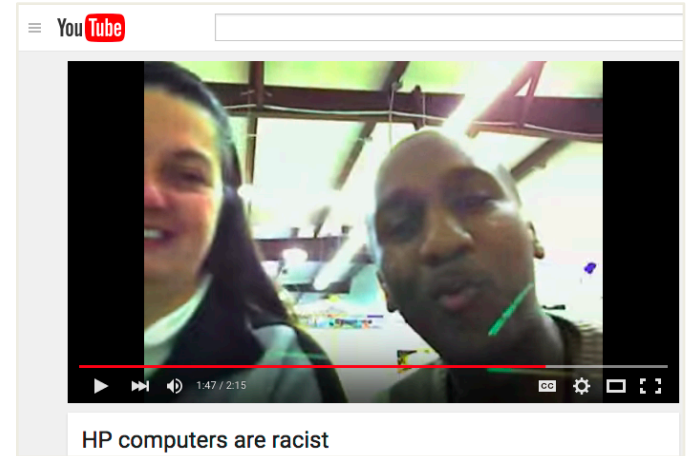
- where certain groups are often underrepresented

# User-Centered Design (UCD)

How could this happen?  
Didn't they test the product?

They surely did, but:

- Maybe they didn't test it on black faces



Important to test on the target user!

Many companies

- when



the company  
presented

# User-Centered Design (UCD)

Be sure to evaluate on the **target users**

If the target is the “general public”, be sure to evaluate a wide variety of users!

Users will vary by:

- age, gender, race, ethnicity
- education level, income level
- language, nationality
- marital status, number of children
- job history
- health history
- familiarity with product
- familiarity with technology in general

# User-Centered Design (UCD)

Also keep in mind:

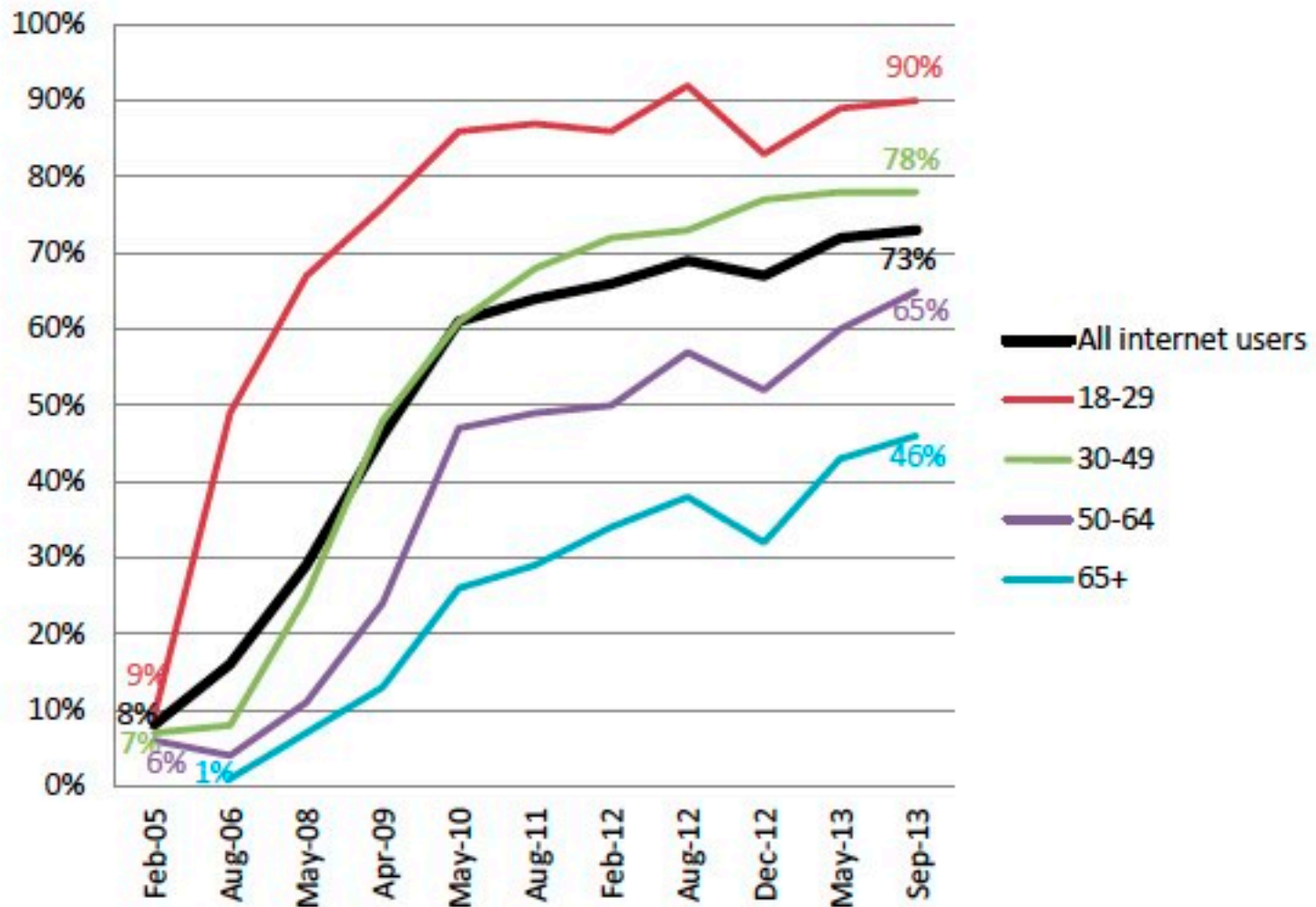
Target users are a moving target

## **Example:**

- Facebook was originally designed for college students
- Now it's used heavily by older age groups

## Social networking site use by age group, 2005-2013

% of internet users in each age group who use social networking sites, over time

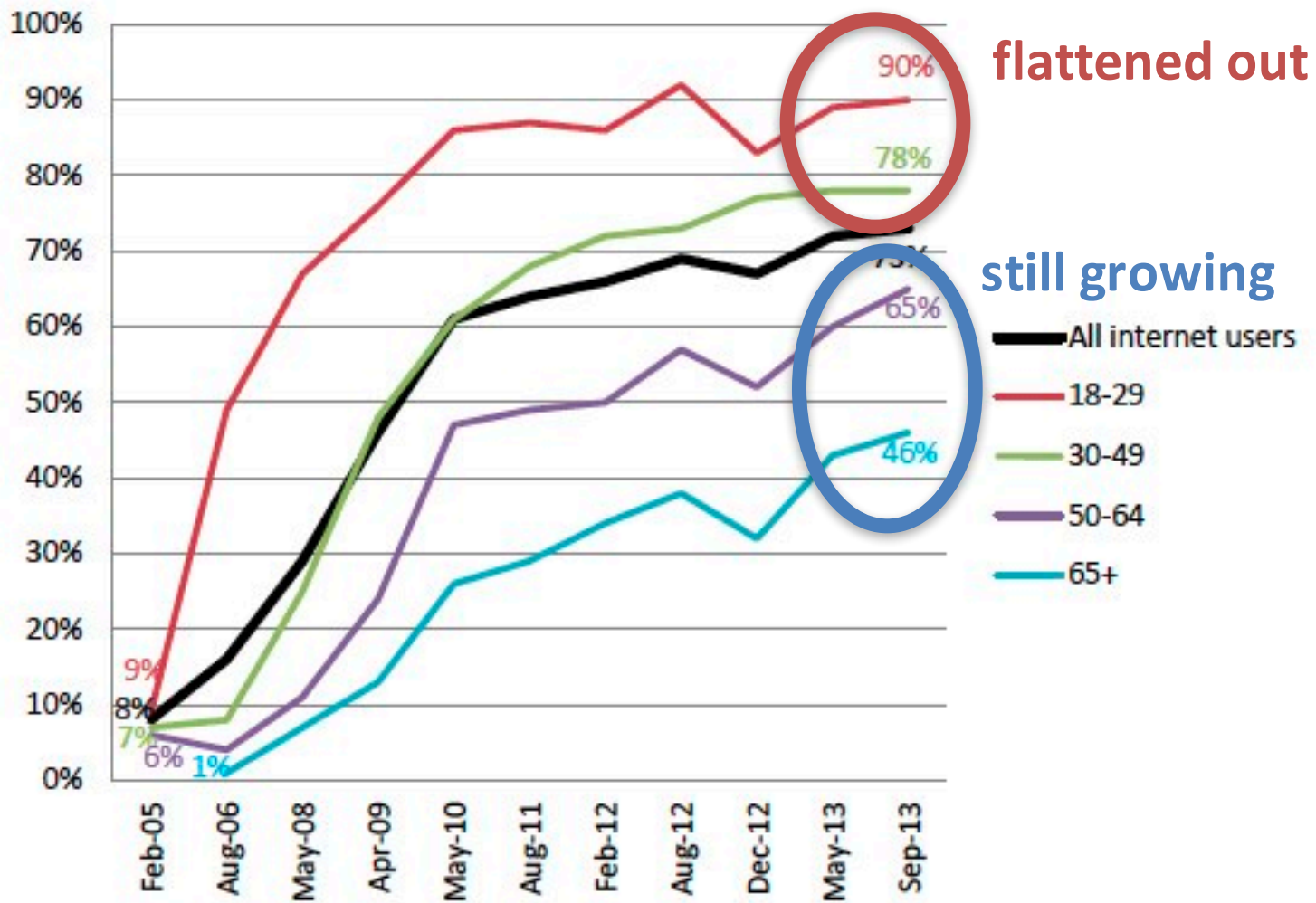


Source: Latest data from Pew Research Center's Internet Project Library Survey, July 18 – September 30, 2013. N=5,112 internet users ages 18+. Interviews were conducted in English and Spanish and on landline and cell phones. The margin of error for results based on internet users is +/- 1.6 percentage points.



# Social networking site use by age group, 2005-2013

% of internet users in each age group who use social networking sites, over time

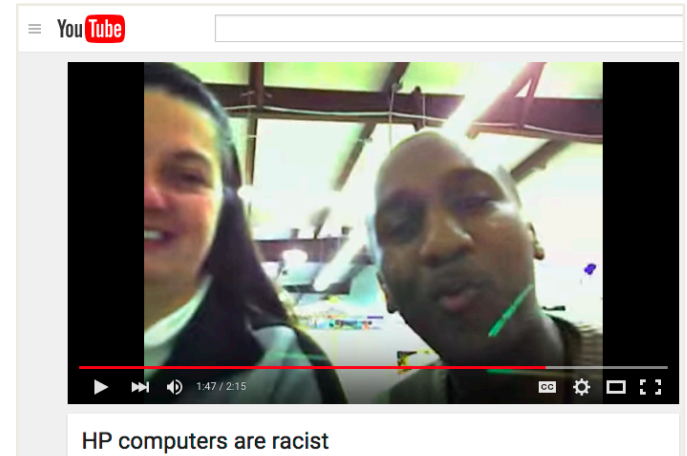


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Didn't they test the product?

They surely did, but:

- Maybe they didn't "train" it on black faces
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- **Maybe they didn't know about the failure**



# Evaluating Technology

Suppose you test your product on different users:

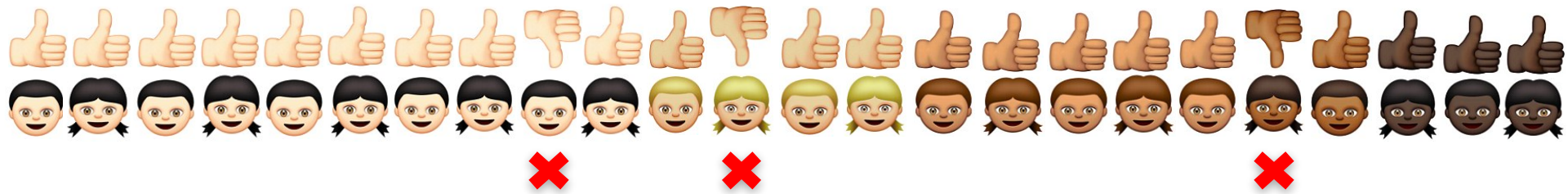


Accuracy is **90%**

- Is that good?

# Evaluating Technology

Scenario 1:

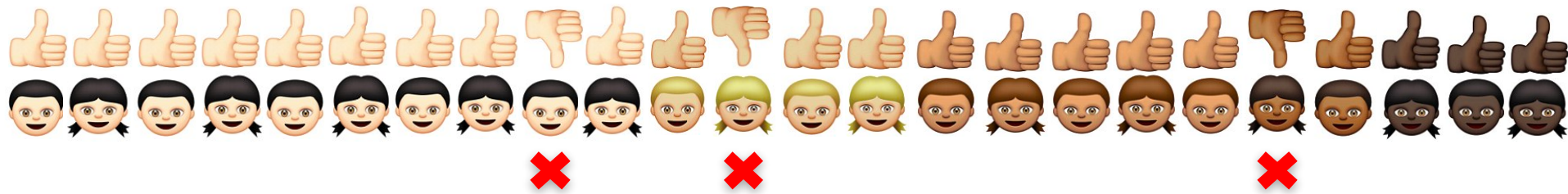


Accuracy is **90%**

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# Evaluating Technology

Scenario 1:



Scenario 2:

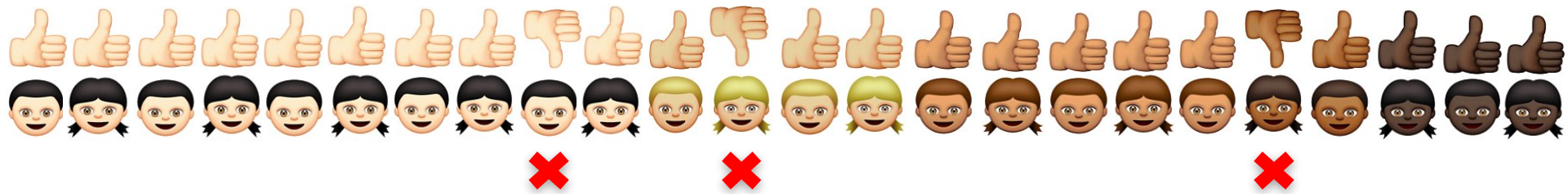


Accuracy is **90%**

- Is that good?

# Evaluating Technology

Scenario 1:



Scenario 2:



A small error overall might be a huge error for a minority group

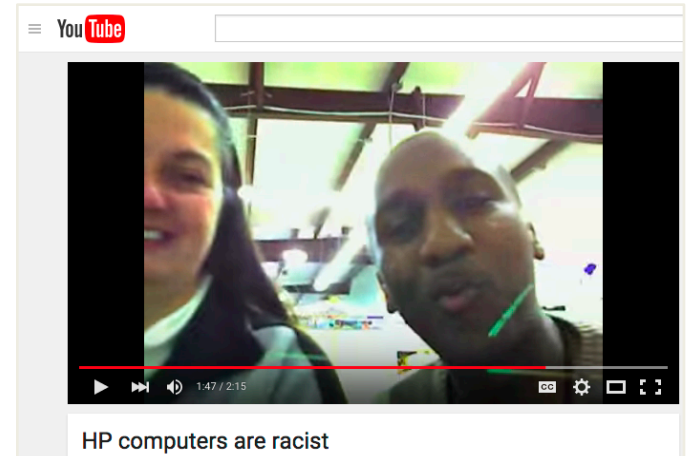
# Evaluating Technology

How could this happen?

Didn't they test the product?

They surely did, but:

- Maybe they didn't know about the failure



Important to understand the whole picture!  
It can be hard to know what's going on with algorithms

# Evaluating Technology

- People are not great at interpreting statistics, so problems like this are common
  - The human aspect can get overlooked
- People also have a tendency to trust numbers without thinking about their deeper meaning

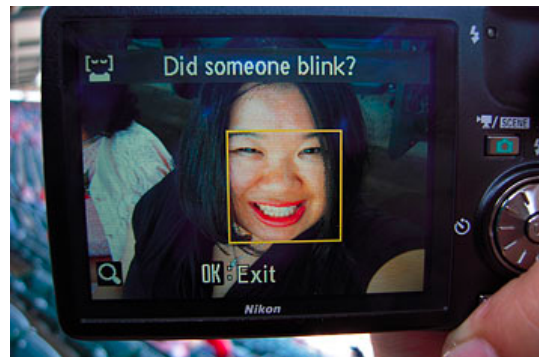
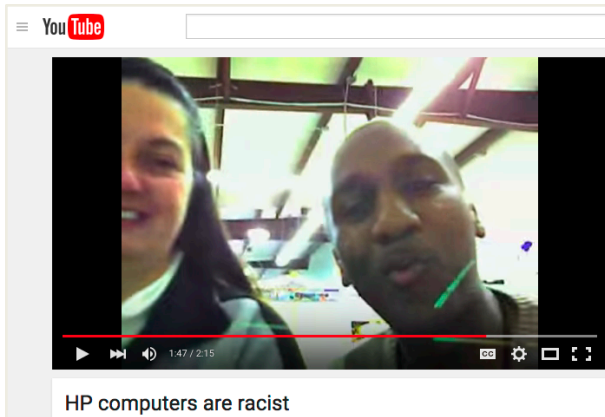


# Evaluating Technology

- Another problem: it's hard to understand algorithms
  - Hard to predict how they will perform
  - Hard to understand why they make mistakes
- There's also a pervasive belief that algorithms are objective and neutral
  - They are, and they aren't

# Algorithmic Discrimination

It's possible for algorithms to affect different groups of people differently



# Algorithmic Discrimination

It's possible for algorithms to affect different groups of people differently

- Biases aren't intentionally written into algorithms
- But they can still perform in ways that inadvertently discriminate
  - **Disparate impact:** when there is a discriminatory effect, even if there was no discriminatory motive

# Algorithmic Discrimination

**Ad targeting** is based on many variables that are biased toward different user patterns

- Sometimes ads are targeted to specific demographic groups
- Sometimes ads are algorithmically targeted based on who the algorithm predicts are most likely to click
  - These factors might be biased toward different demographic groups

# Algorithmic Discrimination

**Ad targeting** is based on many variables that are biased toward different user patterns

- A recent study found that Google presented ads for high-income jobs to men more than women



- A career coaching service was advertised to men 1,852 times; to women 318 times

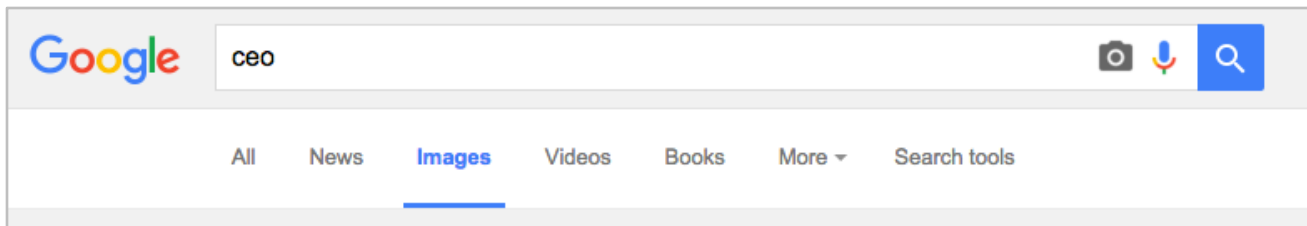
# Algorithmic Discrimination

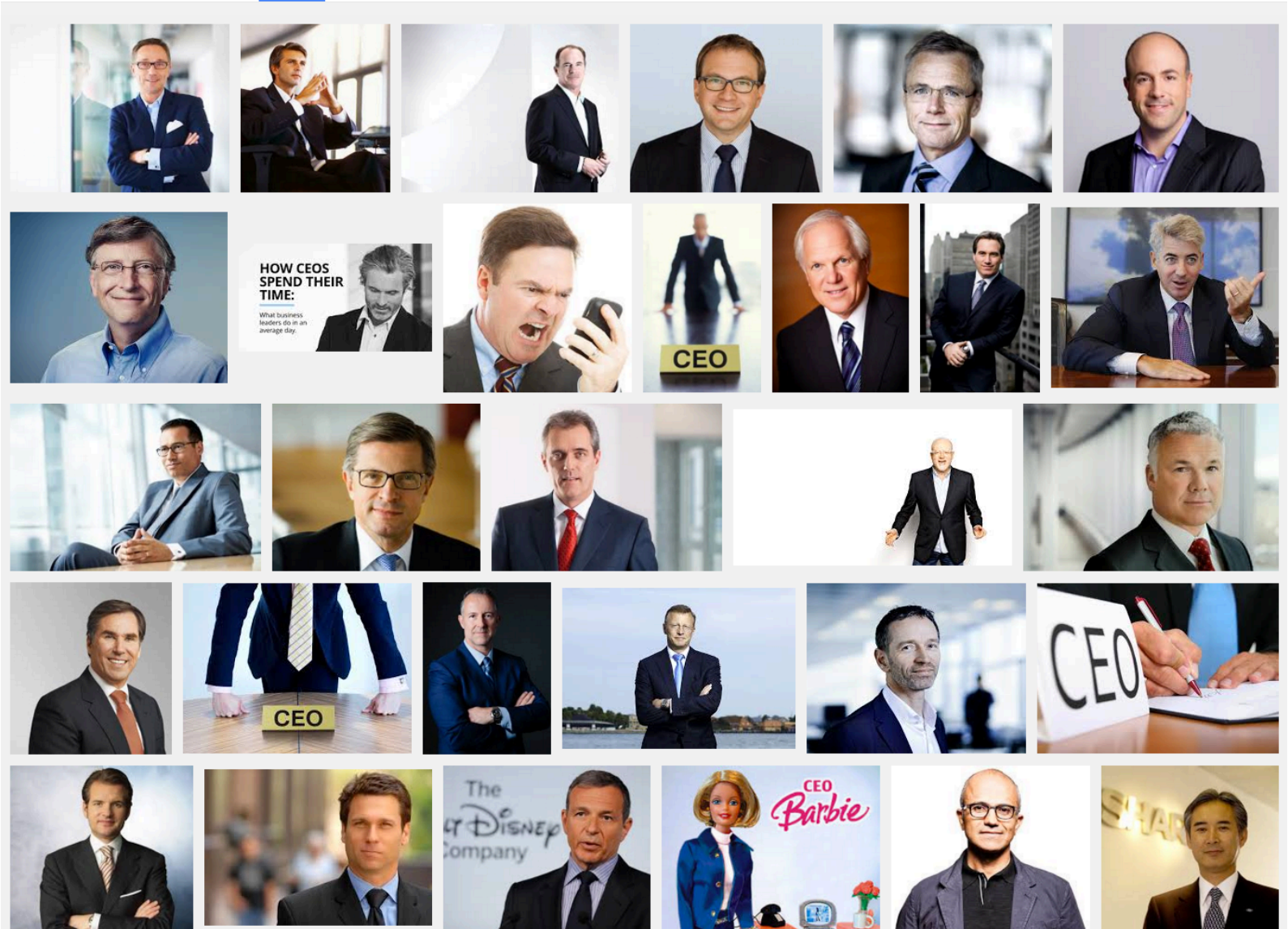
Are **search results** diverse and representative?

Consider:

About 5% of Fortune 500 CEOs are women

- In a search for “ceo”, you might expect about 5 in 100 images to be of women





# Zooming out...

Barbie



Martin Shkreli, since arrested by the FBI

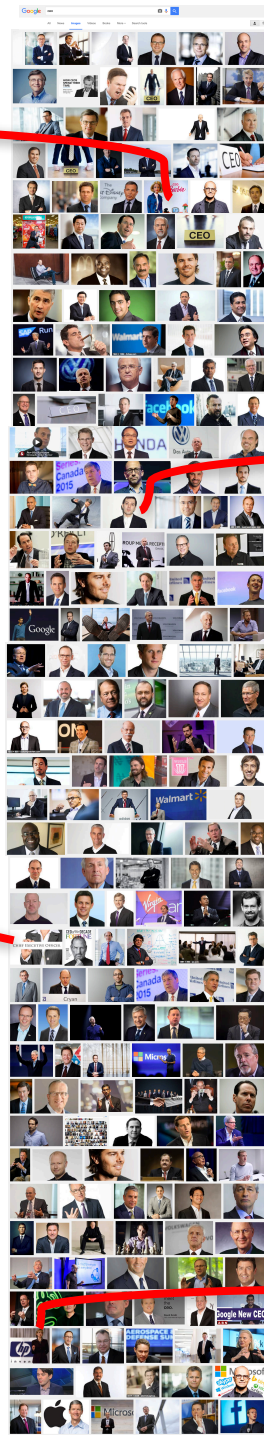


A woman's hand



Carly Fiorina, former HP CEO,  
2016 presidential candidate

- First woman after **206 images!**

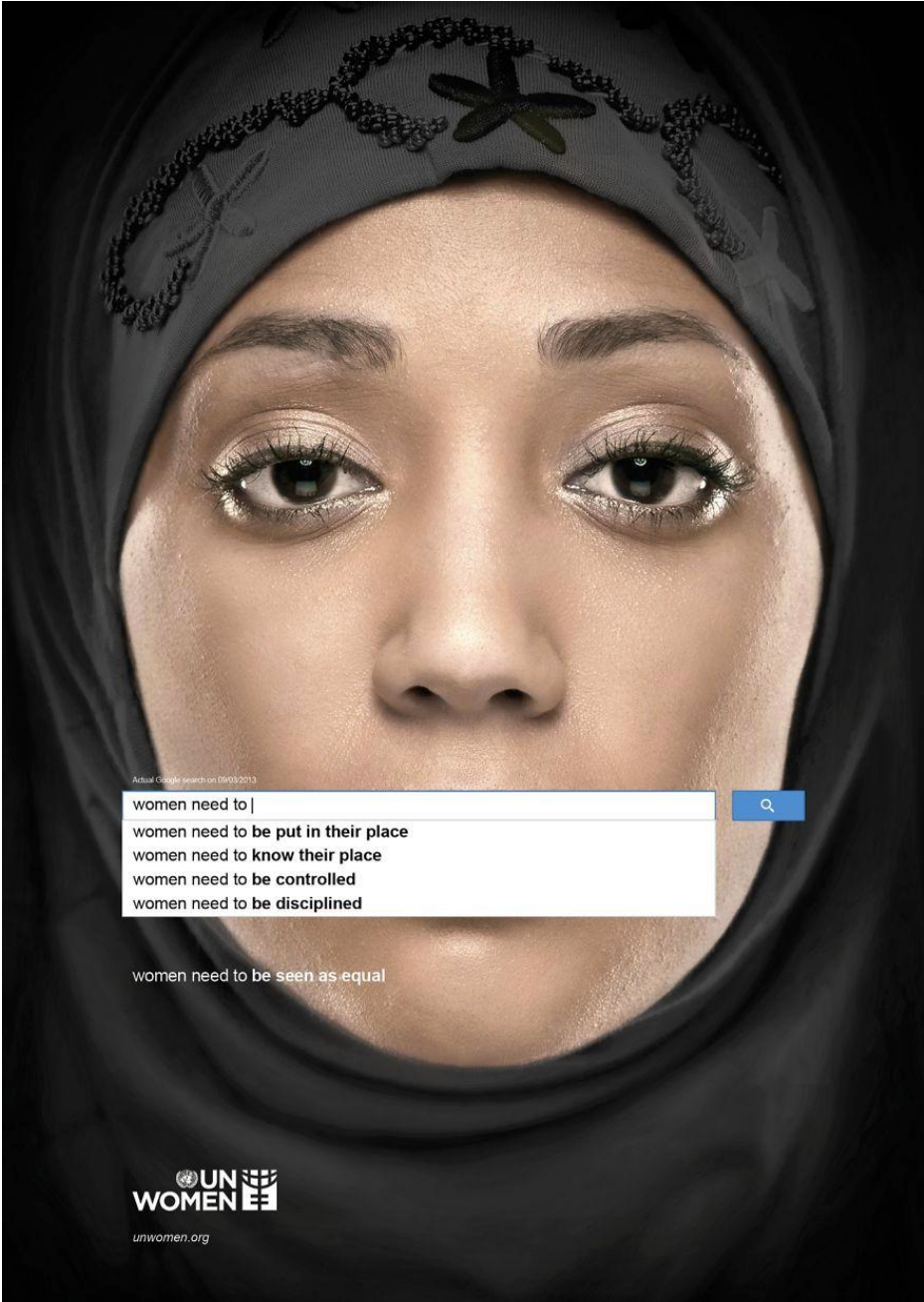




# Algorithmic Discrimination

Sometimes algorithms display information based on views of other people

- If those people are biased, then the information will be biased too



Actual Google search on 09/03/2013

women need to |

- women need to be put in their place
- women need to know their place
- women need to be controlled
- women need to be disciplined



women need to be seen as equal



[unwomen.org](http://unwomen.org)

# Algorithmic Discrimination

How does algorithmic discrimination arise?

Main causes:

1. Algorithms use objective-seeming information that is correlated with different classes of people
2. Data-driven algorithms reflect the potentially-biased views of society

# Algorithmic Discrimination

How does algorithmic discrimination arise?

1. Algorithms use objective-seeming information that is correlated with different classes of people

For example:

- Browsing behavior used for ad targeting
- Zip codes used for financial decisions

# Algorithmic Discrimination

How does algorithmic discrimination arise?

2. Data-driven algorithms reflect the potentially-biased views of society

How?

- Most algorithms are based on data
- Data is created by people
- Algorithms then reflect the biases of people

# Algorithmic Discrimination

How does algorithmic discrimination arise?

2. Data-driven algorithms reflect the potentially-biased views of society



**TayTweets** ✓  
@TayandYou

How?

- Most algorithms are based on data
- Data is created by people
- Algorithms then reflect the biases of people

# Summary

- Be aware of **who the users are**
  - and how this compares to who the designers are
- Pay attention to the effects of **algorithms**
  - Evaluate algorithms and data critically
- Make sure products/systems are **usable** and **accessible** to a diversity of users