# Ethics: Choices, Values, & Burritos

# **Learning Goals**

- 1. Choices embed values
- 2. Define Bias & Fairness
- 3. Normative vs. Descriptive
- 4. Problem Formulation, Representation & Long Tail



# Learning Goals

- 1. Choices embed values
- 2. Define Bias, Fairness, Problem Formulation, Long Tail ...
- 3. Keep Mehran, Chris, and Juliette Busy in the Chat



### **Mid-Quarter Evaluations**

### CS 106A Mid-Quarter Evaluation \* Required Instructor Feedback These first few questions come from Mehran Sahami, Chris Piech, and Juliette Woodrow. Your anonymous feedback will be reviewed by them, so please be specific and give suggestions for improvement. What has Mehran been doing so far that has worked well for you? What should he continue doing? \* Your answer What has Mehran been doing so far that has not worked well for you? Is there anything that he should stop or start doing? Your answer What has Chris been doing so far that has worked well for you? What should he

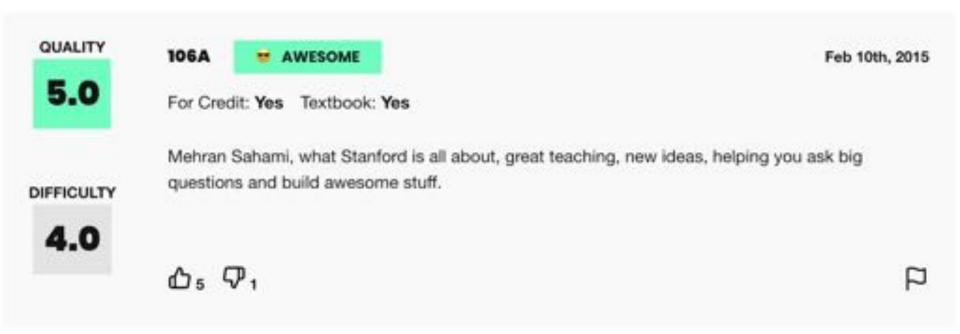


# **Assignment 5: Sneak Preview**



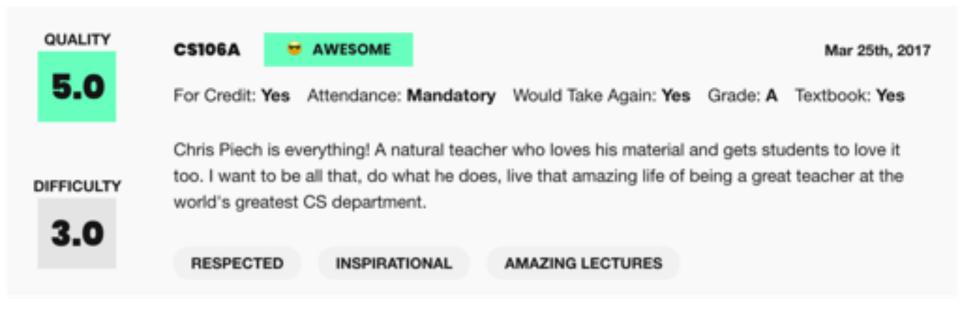


# **Assignment 5: Ratings**



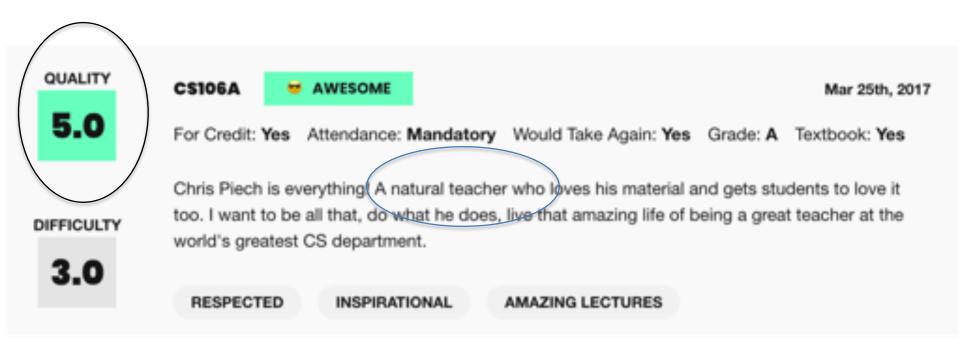


# **Assignment 5: Ratings**



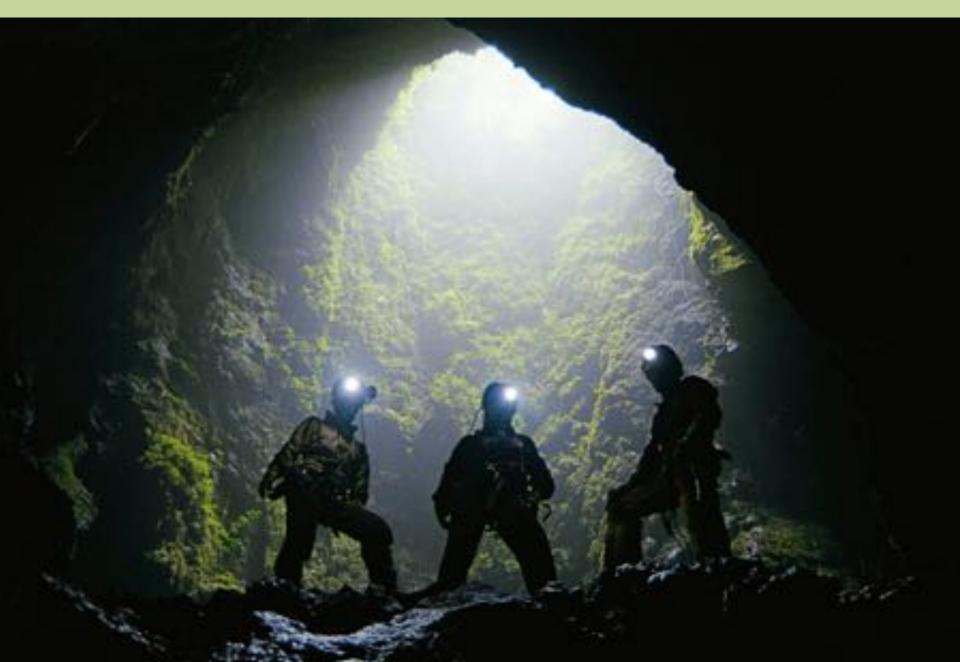


# Are These Ratings Biased?

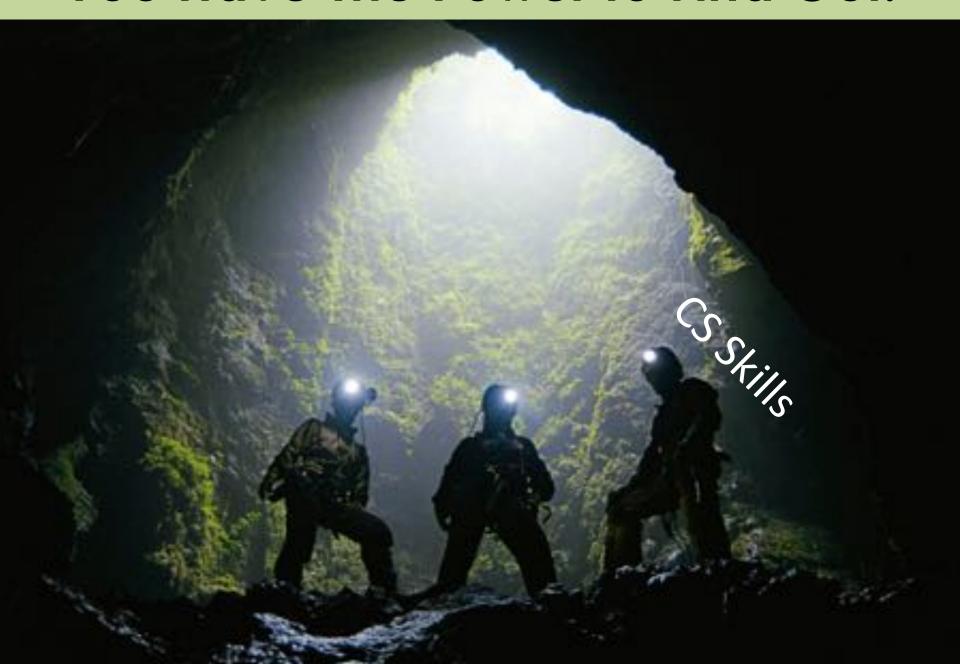




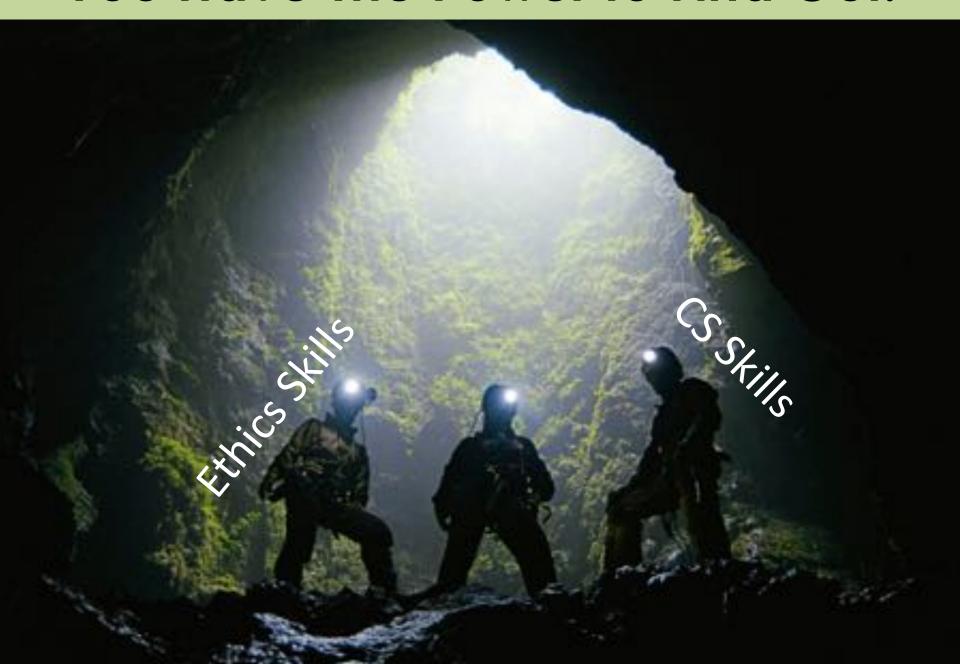
# What is in the Dataset?



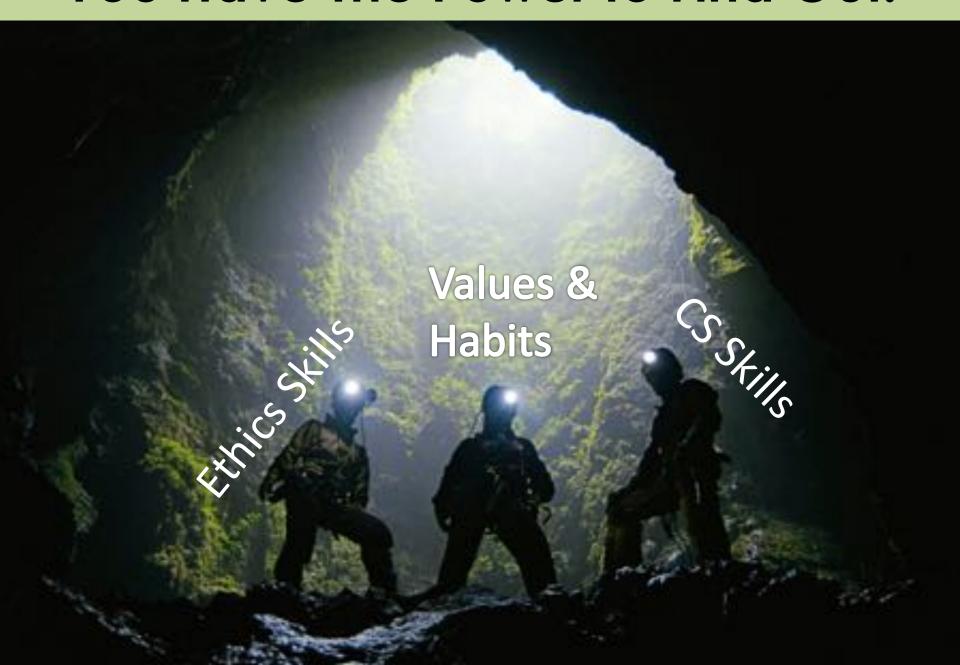
## You Have The Power to Find Out:



# You Have The Power to Find Out:



### You Have The Power to Find Out!



Bias: difference between measured results and "true" value



Bias: difference between measurement results and "true" value





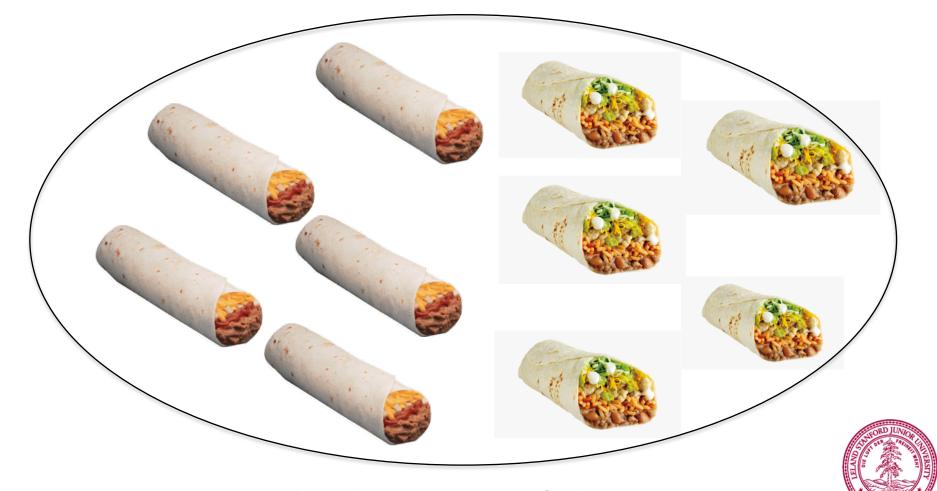
Bias: difference between measurement results and "true" value



Statistical Bias: learn more in future computer science & statistics classes!



Bias: difference between measurement results and "true" value



Piech + Sahami, CS106A, Stanford University

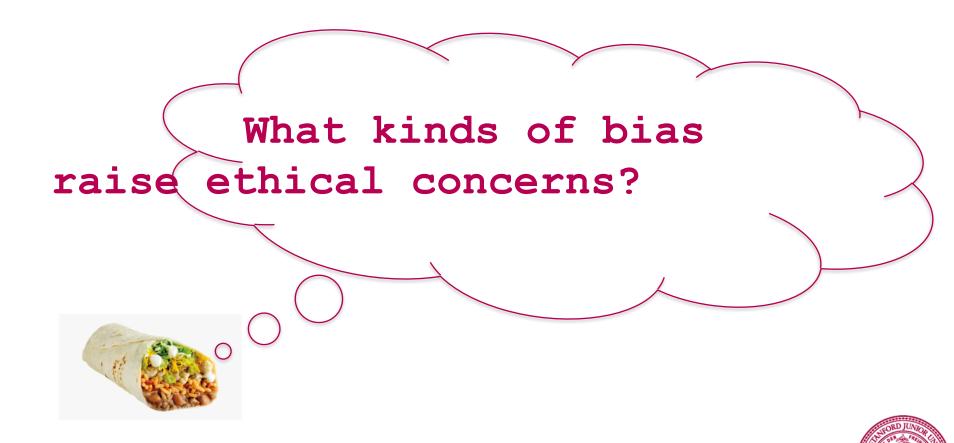
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Statistical Bias: learn more in future computer science & statistics classes!



Bias: difference between measurement results and "true" value



# **Discriminatory Bias**

Biased measurement or classification

+ use of that bias to compound existing injustice or fail to treat all as having equal moral worth

=> Unfair Bias



# Unfair Bias in Ratings

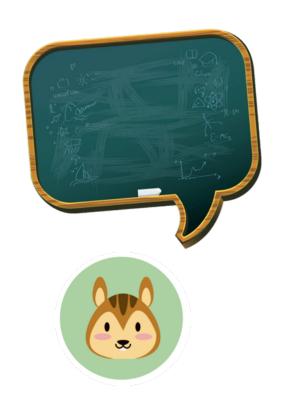
How might this definition of work in ratings?



# **Equality of Opportunity?**

Equality of Opportunity: everyone has same opportunity to develop skills needed for the job, apply for the job, and get promoted.



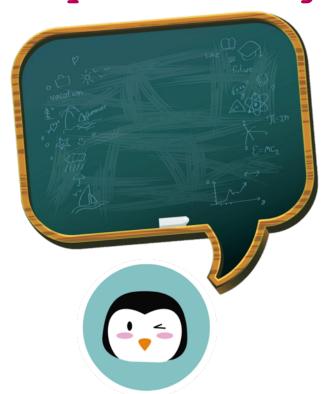




# Parity: Equality of Numbers

Parity: Everyone is equally likely to be a good teacher, so we should expect to end up with number of good teachers equal to population.

Parity-fair ratings should reflect this.



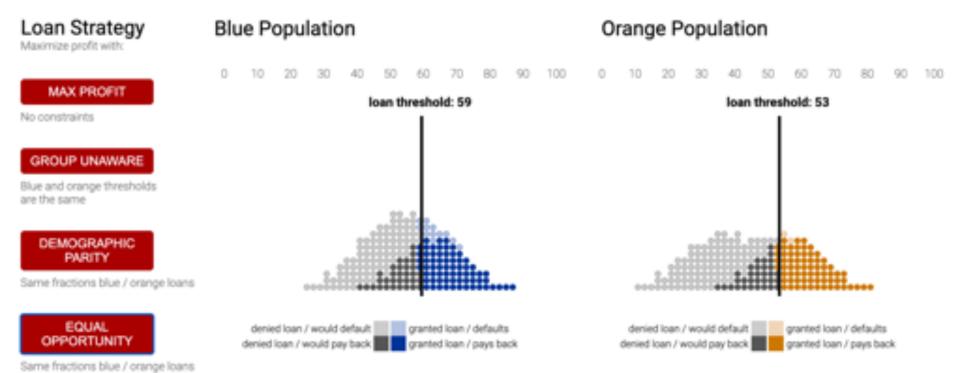


### **Measures of Fairness**

#### Simulating loan decisions for different groups

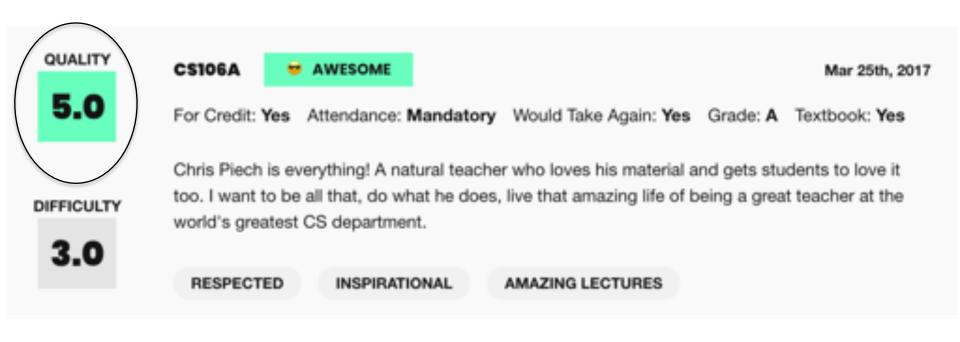
Drag the black threshold bars left or right to change the cut-offs for loans. Click on different preset loan strategies.

to people who can pay them off





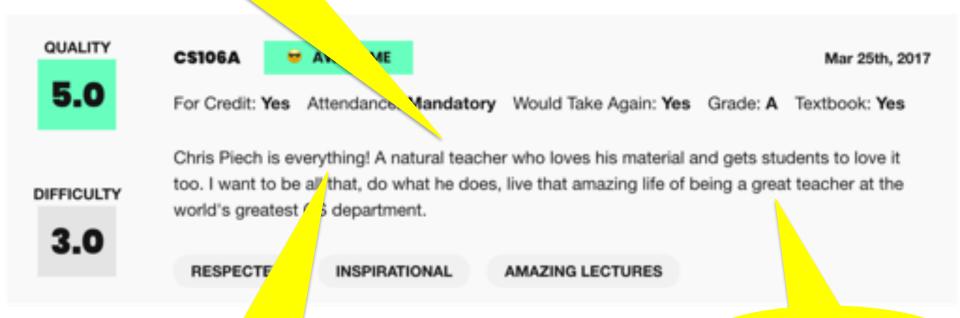
# Fairness: Beyond the Numbers





# **Evaluation Beyond the Numbers**

### NATURAL TEACHER



GREAT

#### EVERYTHING!



# Descriptive vs. Normative

#### Descriptive language

#### Normative language

- is
- was
- what people did
- what happened

- right
- wrong
- good
- bad
- should
- should not



### Descriptive or Normative?

### NATURAL TEACHER

5.0

DIFFICULTY

3.0

CS106A



Mar 25th, 2017

For Credit: Yes Attendance Mandatory Would Take Again: Yes Grade: A Textbook: Yes

Chris Piech is everything! A natural teacher who loves his material and gets students to love it too. I want to be all that, do what he does, live that amazing life of being a great teacher at the world's greatest CS department.

RESPECTED

INSPIRATIONAL

AMAZING LECTURES





"Thick" normative terms are both descriptive AND normative:

• brave : foolhardy





"Thick" normative terms are both descriptive AND normative:

- brave : foolhardy
- Cowardly :





"Thick" normative terms are both descriptive AND normative:

- brave : foolhardy
- Cowardly : cautious
- Polite : (?)
- rude : (?)
- chill
- kind
- etc





#### Thick Normative Terms & Fairness

- We compare people in many ways, not just numerically
- Thick normative terms express "loaded" judgments



#### Are These Claims Descriptive or Normative?

Mark Zuckerberg on whether Facebook would fact-check false claims about election suppression:

- 1. "We have a different policy, I think, than Twitter on this."
- 2. "You know, I just believe strongly that Facebook shouldn't be the arbiter of truth of everything that people say online."
- 3. "I think in general private companies probably shouldn't be—or, especially these platform companies—shouldn't be in the position of doing that."

#### Are These Claims Descriptive or Normative?

Not surprising that statements setting the policy for platforms would be normative. What about the programs behind the platforms themselves?

Do programs like the ones you are writing contain normative claims or values?



# How are values embedded in design?

### 1. Problem Formulation Embeds Values

#### "Sandcastle" (warm-up) problem: Finding forest fires.

We're going to start by writing a function called highlight\_fires (in the file forestfire.py) that highlights the areas where a forest fire is active. You're given a satellite image of Greenland's 2017 fires (photo credit: Stef Lhermitte, Delft University of Technology). Your job is to detect all of the "sufficiently red" pixels in the image, which are indicative of where fires are burning in the image. As we did in class with the "redscreening" example, we consider a pixel "sufficiently red" if its red value is greater than or equal to the average of the pixel's three RGB values times some intensity threshold. In this case, we have provided you with an appropriate intensity threshold of 1.05 via a constant named INTENSITY\_THRESHOLD in the file forestfire.py. Note that this is a different intensity threshold value than we used in class for the "redscreening" example, as different applications often require different intensity threshold.

When you detect a "sufficiently red" pixel in the original image, you set its red value to 255 and its green and blue values to 0. This will highlight the pixel by making it entirely

Formulating a problem means describing the desired solution as good or worthy of being done.

# Problem formulation and goal statements are normative

Formulating a problem means describing the desired solution as good or worthy of being done.

- What is the problem to be solved?
- For whom is this a problem? Who would benefit from its solution?
- Who can agree that this is a problem worth solving?



"Homeless people are sleeping here and we (who is we?) want them to stop"





"Homeless people are sleeping here and we (who is we?) want them to stop"





"Homeless people are sleeping here and we (who is we?) want them to stop" "Some people in our community don't have a place to sleep and we (who is we?) think they should"



"Homeless people are sleeping here and we (who is we?) want them to stop" "Some people in our community don't have a place to sleep and we (who is we?) think they should"



**Search Engines** 

**Ratings of Professors** 





#### **Problem formulation embeds values**



### What are the Problem(s) to be Solved?

# Ratings of Professors

What are the problem(s) to be solved?



For whom are these problems? Who would benefit from their solution(s)?

For each problem, who can agree that the problem is worth solving?

### What are the Problem(s) to be Solved?

#### Search Engines

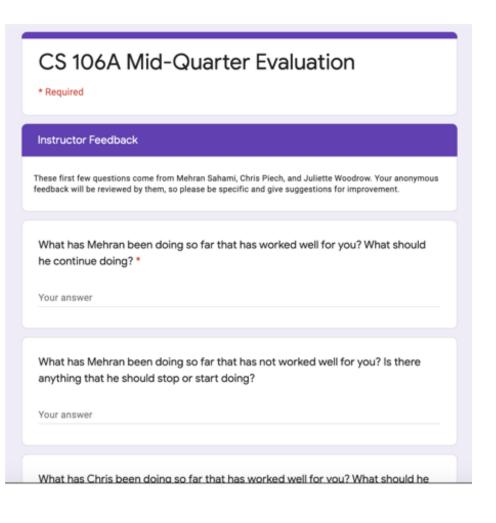




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#### 2. Choice of Data Embeds Values

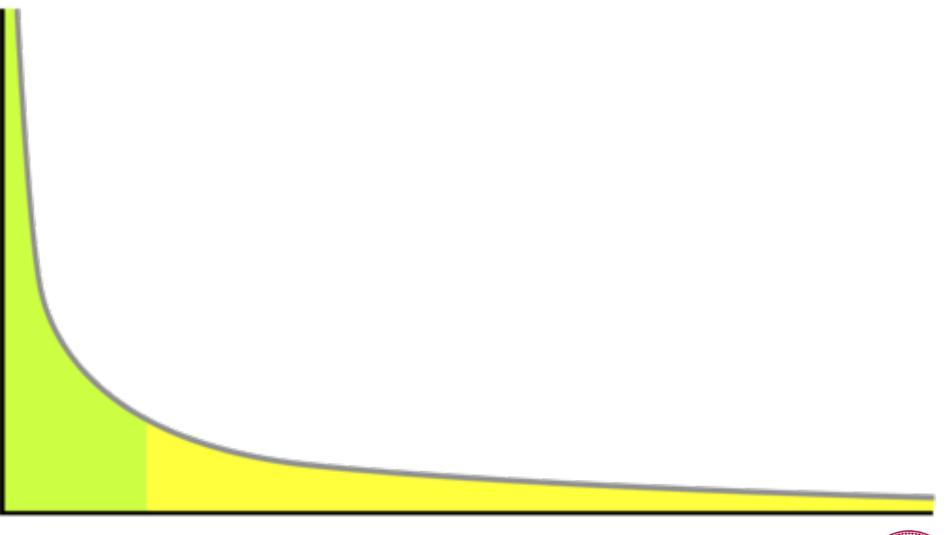


- Surveys are cheap to run
- They measure opinions

 What are other ways to measure quality of professors?



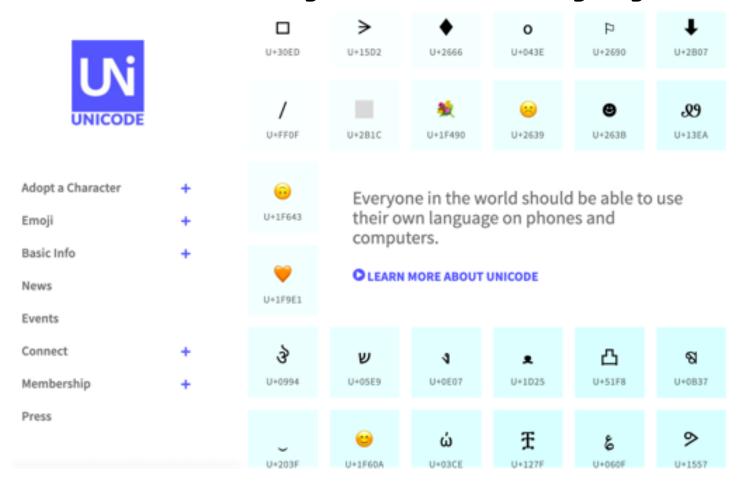
# **Underrepresentation & The Long Tail**





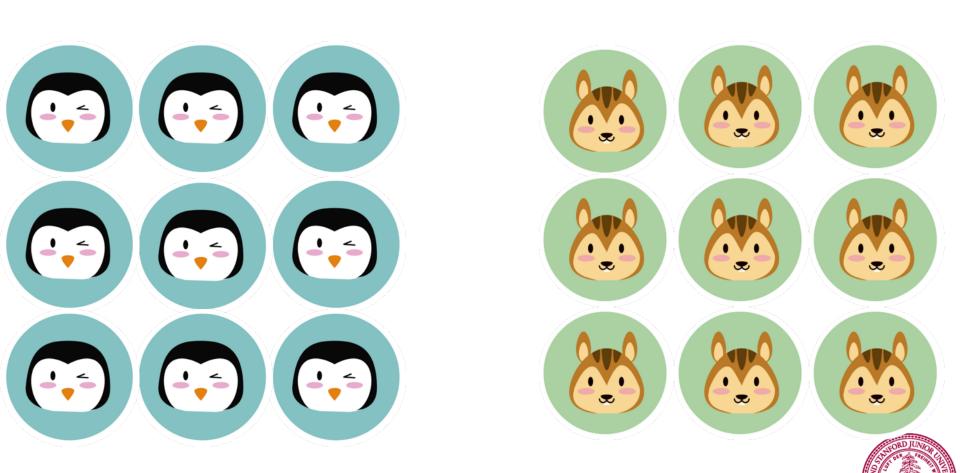
### **Unicode & Representation**

Covers over 1 million symbols, Full coverage of 90 languages, basic coverage for 200 languages



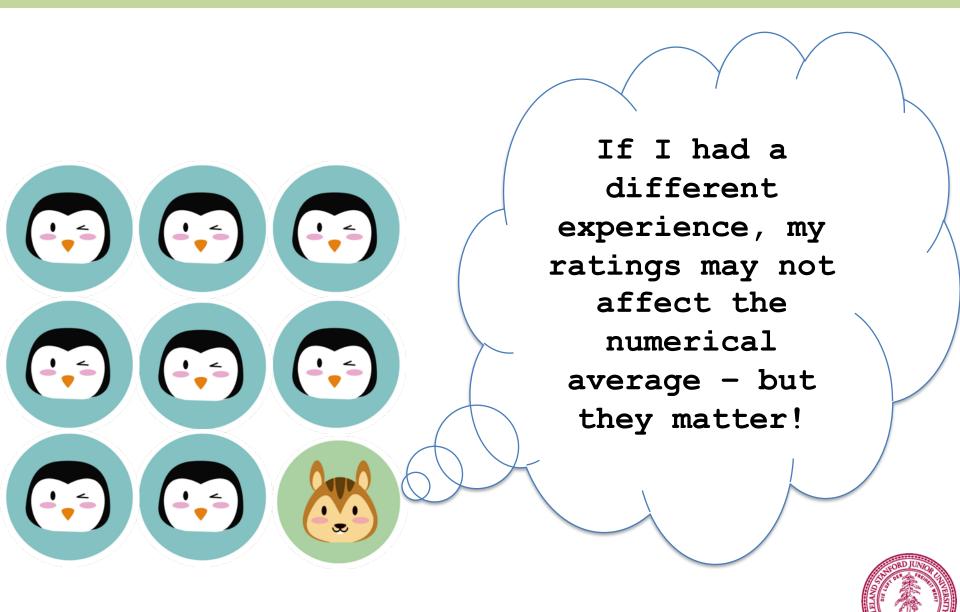


# Representation & Surveys

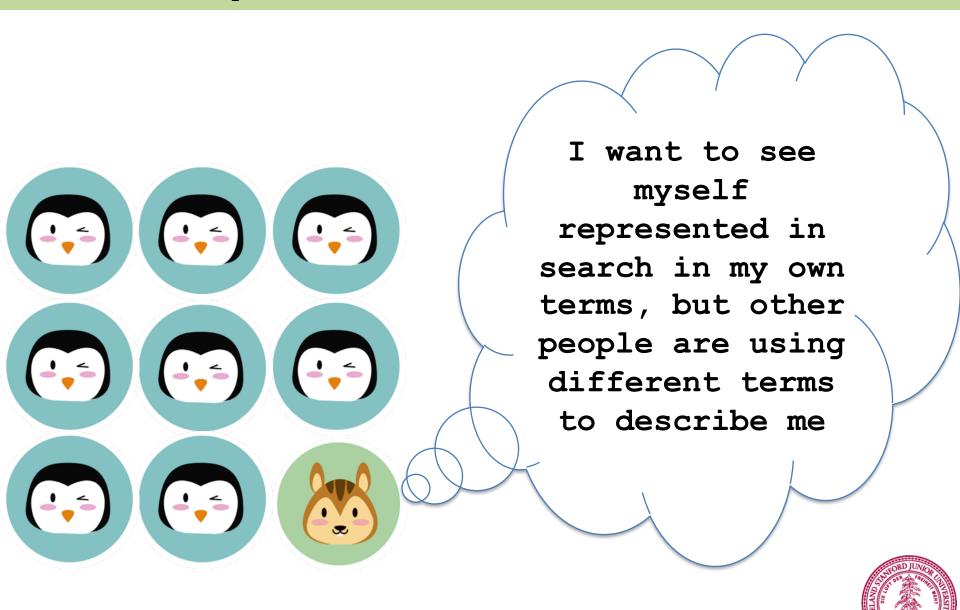


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# Representation & Survey Data



### Representation & Search



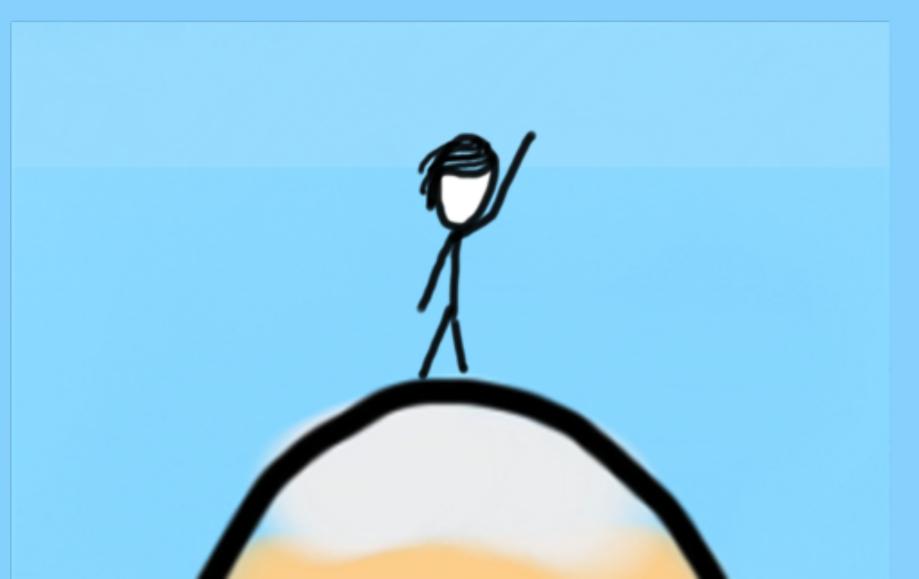
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# Questions?

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#### Discrimination

Laws and policies demean when they express that a person is of lesser moral worth in a context in which the actor has the kind of social power to lower the status of the individuals affected (Hellman, 2008,7).

Is the same true for bias in data?

