

























Google Developer Student Club at William & Mary

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## Good Shitty No Data



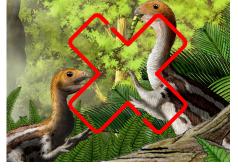








































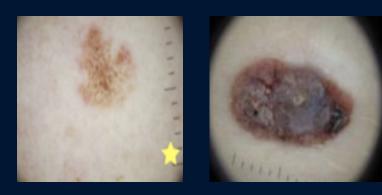






## **Detecting Skin Cancer**

#### **Cancerous**



#### Non-cancerous





129,450 images 21 dermatologists

## **Detecting Skin Cancer → Hidden Paths**

Want: Image of Cancer Cell → Classify as Cancer Cell

$$X \rightarrow Y$$

Actual: Image of Cancer Cell ← Ruler in Image → Classify as Cancer Cell

$$X \leftarrow Z \rightarrow Y$$

# "The hidden confounder creates a hidden class" and thus a "shortcut solution."

—Professor Shao, NASEC

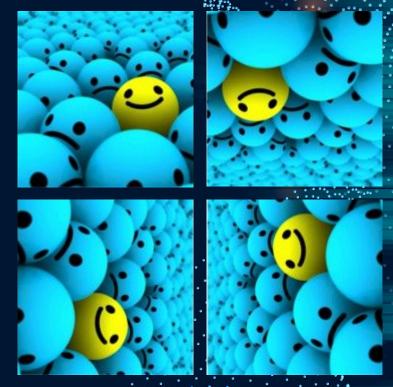
#### **Detecting Skin Cancer -> Stakeholders**

- Patients
- Doctors
- Healthcare providers

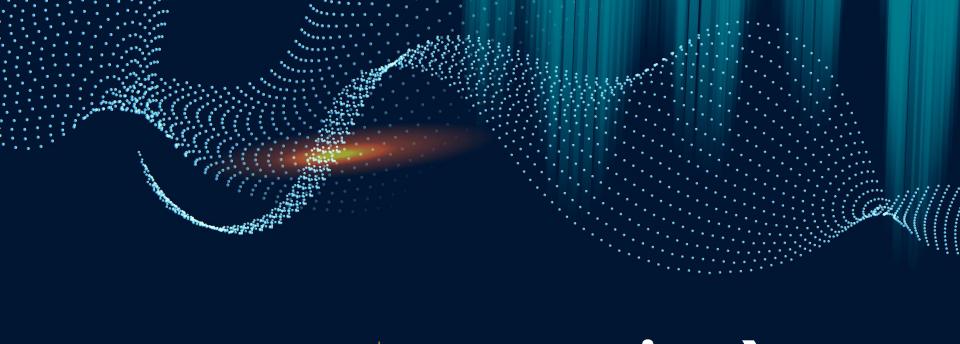


## **Detecting Skin Cancer → Recommendations**

- To Dataset Curators
  - Careful curation of data
- To Coders
  - o Invariant risk minimization
- To Policymakers
  - Require medical AI to be tested by multiple companies in many different areas
  - Require Explainable AI



Data Augmentation



2 Human Bias → Model Bias

## Convert sign language to speech

Judge criminal cases in an unbiased way

Analyze resumes to hire the best candidate

**Self-reflect** 

Win jeopardy against top jeopardy players

**Count crowds** 

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## Human bias → Model bias

#### **Facial Recognition**

- Biases about gender, race, etc. are coded into the model.
- Facial recognition software struggles most on people who are:
  - Darker-skinned
  - Female-identifying
  - Very young
  - Very old
- Privacy concerns



## **Facial recognition**

#### **ERROR RATE**

	Microsoft	IBM
Lighter Male Faces	0.0%	0.3%
Darker Female Faces	20.8%	34.7%

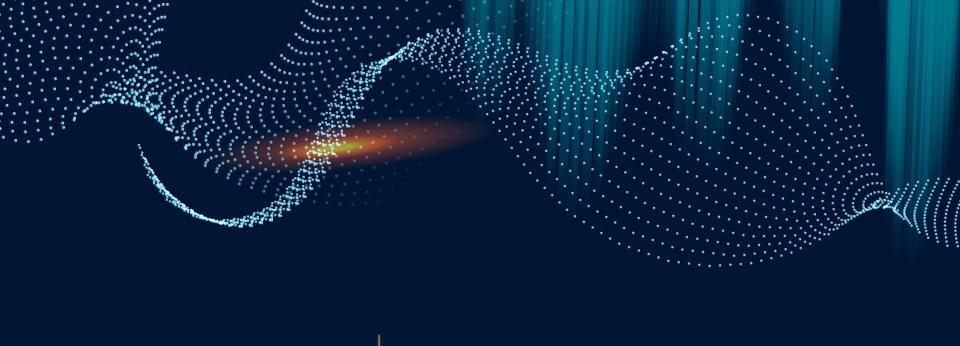
#### **Facial Recognition**

87%

One model's accuracy in a sports venue study

40%

Half of the models in the same study



Weapons of math destruction

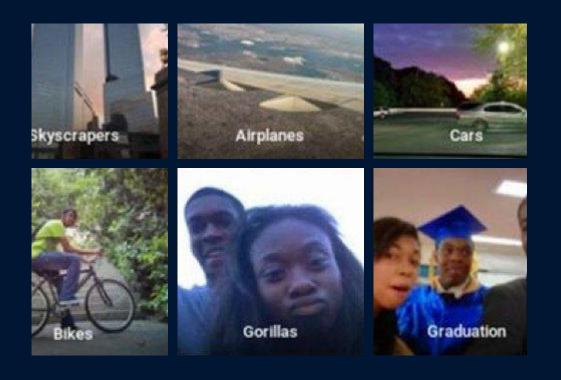
## **Google Photos' Auto-tagging Feature**

- Deep neural networks
- Identify patterns, label images





## What's wrong with this picture???



#### Sources

- NASEC 2021
- Towards Data Science
- Nanonets
- Face in Video Evaluation
- Reuters
- <u>Technology Review</u>
- Harvard Science in the News
- U.S. Government Accountability Office
- Machine Learning Research Conference



