Today: Outline

- Free-choice Lectures
- Ethics in ML and DL
- Project Breakout Session
- Announcement:
 - Practice problems available
 - Thu is a free-choice lecture
- Reminders:
 - Exam Jun 22 in class (and ~12 hrs before for remote only students)
 - Team Registration Deadline due today
 - Regrade requests for PS1 due today
 - Be sure your scores are updated on webpage http://cs-people.bu.edu/sbargal/cs523/grades.html

Deep Learning

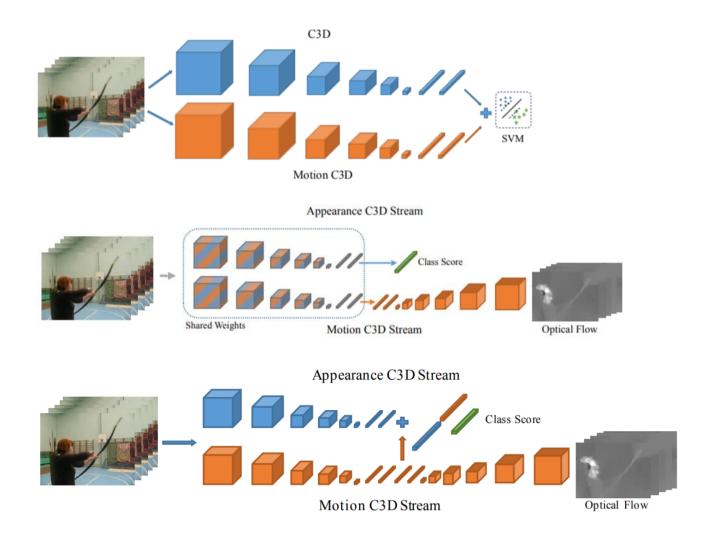
CS 523 Summer1 2021

You can use your random ID to access your row of grades for the course.

Grades**

Rand ID	PS1 (/100)	PS2 (/100)	PS3 (/100)	Par (/5)	Bonus (/2)	Pre-lec1 (/1)	Pre-lec2 (/1)	Pre-lec3 (/1)	Exam (/100)	Project (/100)
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23	71.5	<u>~</u>	-	u u	74	1	0.75	1	=	±
48	73.5	-	ā	=	ē.	1	1	1	-	7
44	79	-	-	=	-	1	1	0.75	-	-
45	85	-	2	_	-	1	1	1	_	2

Multiple Modalities & Auxiliary Tasks



Pre-lecture Material

Grad-CAM

Which of the following is true? [Multiple choices allowed]

#pin

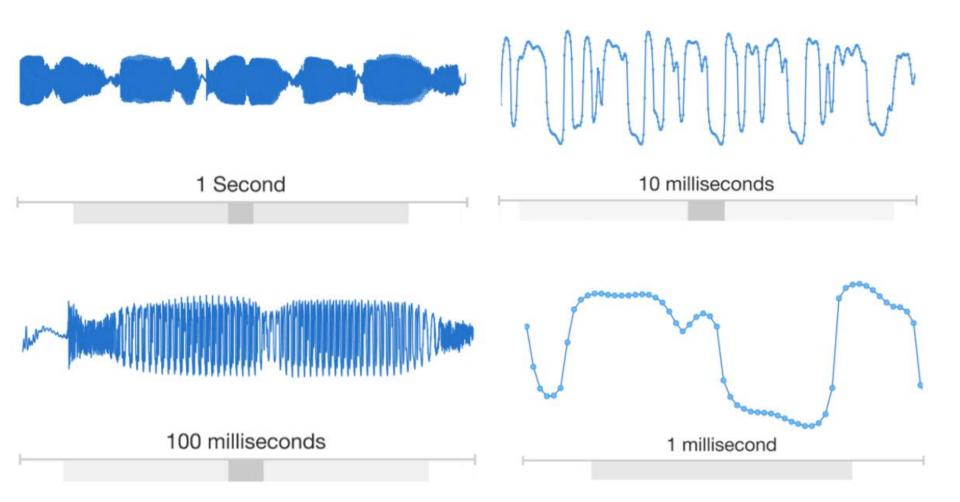
- Grad-CAM is a an interpretability method for deep learning models
- Grad-CAM provides transparency into why a model makes a specific prediction
- All applications presented in the paper are for visual tasks (takss that use images and/or videos)
- One can compute a Grad-CAM based saliency map without the knowledge of network parameters
- Grad-CAM can be used as a tool to diagnose model bias



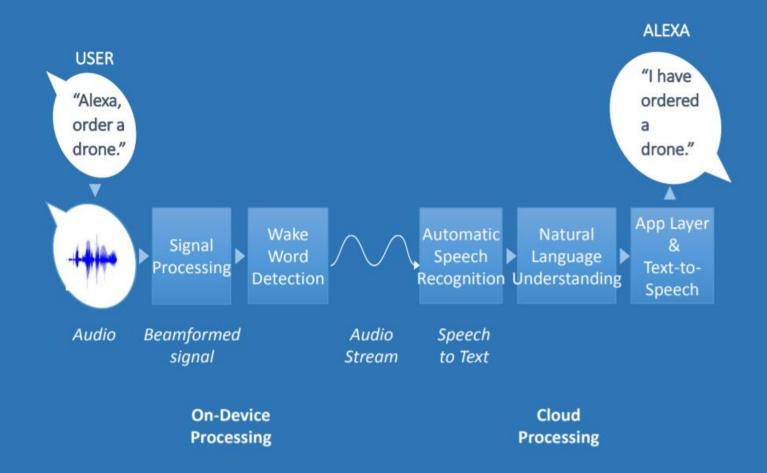
Audio

Free-choice Lecture 1

Raw Audio



The Alexa Pipeline





Vision + Language

Free-choice Lecture 2

What are vision-language problems?

Input → output might be...

- A. Image \rightarrow text
- B. Text \rightarrow image
- C. Image + text \rightarrow text

Examples...

- A. Image Captioning
- B. Sentence to Image Retrieval
- C. Visual Question Answering

Image Captioning





A woman in a white shirt and denim overalls is walking six dogs in the park.

Generating natural language sentences given the input image

Sentence to Image Retrieval

A woman in a white shirt and denim overalls is walking six dogs in the park.





Retrieving an image given the input natural language sentence

Sentence to Image Retrieval

A woman in a white shirt and denim overalls is walking six dogs in the park.





Retrieving an image given the input natural language sentence

Visual Question Answering



How many dogs are being walked?



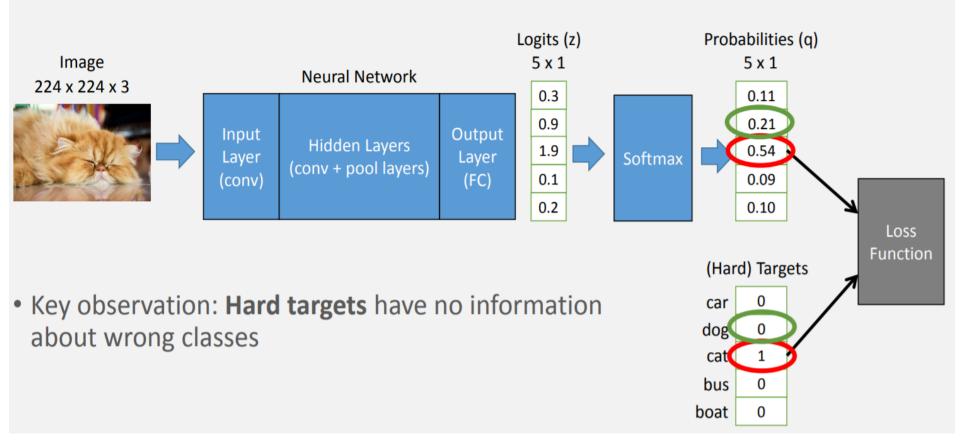
Generating an answer to a visual question, often framed as a classification problem



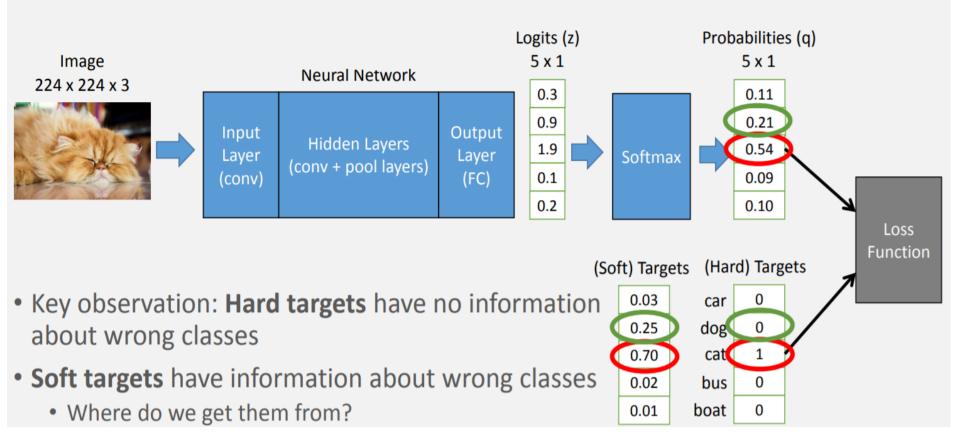
Knowledge Distillation

Free-choice Lecture 3

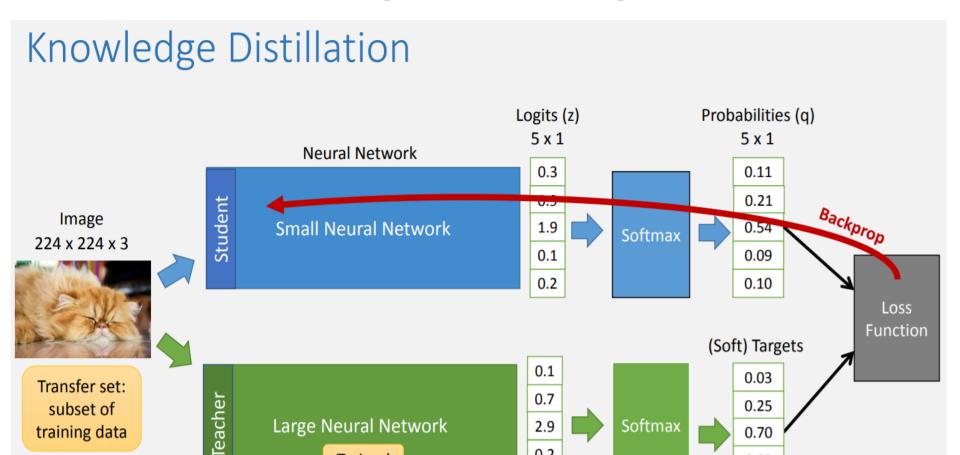
Training a neural network



Training a neural network



Where do we get soft targets from?



Trained

0.2

0.1

0.02

0.01

Ethics in Machine Learning
Kate Saenko



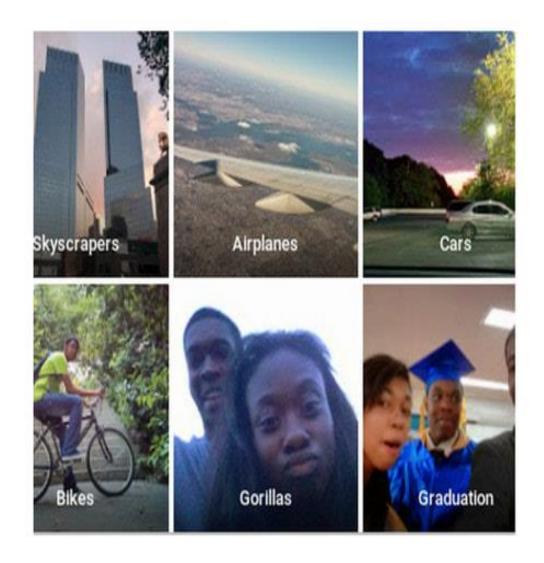
- Job Loss
- Algorithmic Bias
- Transparency
- Al Supremacy
- Fake news and videos
- Autonomous weapons
- Self-driving cars
- Privacy and surveillance





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Bias can lead to offensive or unfair results...



When Hiring for Al ...

Who codes matters

Are we creating full-spectrum teams with diverse individuals who can check each other's blind spots?

How we code matters

Are we factoring in fairness as we're developing systems?

Why we code matters
 We now have the opportunity to unlock even greater
 equality

Example of ML (un)fairness: COMPAS

- Criminal justice: recidivism algorithms (COMPAS)
- Predicting if a defendant should receive bail
- Unbalanced false positive rates: more likely to wrongly deny a black person bail

ProPublica Analysis of COMPAS Algorithm

	White	Black
Wrongly Labeled High- Risk	23.5%	44.9%
Wrongly Labeled Low- Risk	47.7%	28.0%

 Suppose we are a bank trying to fairly decide who should get a loan i.e. Who is most likely to pay us back?

 Suppose we have two groups, A and B (the sensitive attribute)
 This is where discrimination could occur

 The simplest approach is to remove the sensitive attribute from the data, so that our classifier does *not* know the sensitive attribute

Table 2: To Loan or Not to Loan?

Age	Gender	Postal Code	Req Amt	A or B?	Pay
46	F	M5E	\$300	A	1
24	М	M4C	\$1000	В	1
33	М	МЗН	\$250	Α	1
34	F	M9C	\$2000	A	0
71	F	M3B	\$200	А	0
28	М	M5W	\$1500	В	0

Table 3: To Loan or Not to Loan? (masked)

Age	Gender	Postal Code	Req Amt	A or B?	Pay
46	F	M5E	\$300	?	1
24	М	M4C	\$1000	?	1
33	М	МЗН	\$250	?	1
34	F	M9C	\$2000	?	0
71	F	M3B	\$200	?	0
28	М	M5W	\$1500	?	0

 However, if the sensitive attribute is correlated with the other attributes, this isn't good enough

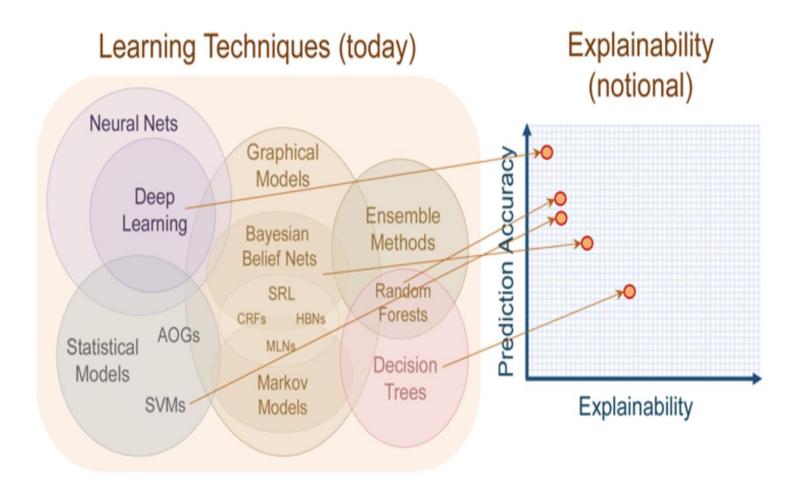
 It is easy to predict race if you have lots of other information (e.g. home address, spending patterns)

More advanced approaches are necessary

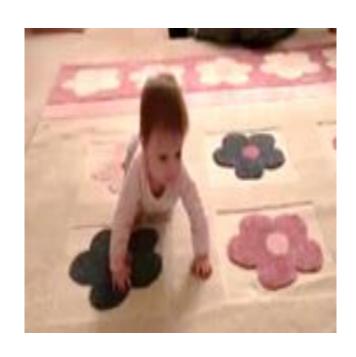
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Accuracy vs. explainability



Sample Misclassification



Ground Truth:BabyCrawling

Classified as: Pushups

 Explainablity would tell us "why", or at least highlight pixels responsible for the prediction

Why is an algorithm predicting "Pedestrians Crossing the road" very well?

 Because of the periodic motion of the legs? If so, then we would have a problem in the following test scenario where the legs of the pedestrians are completely occluded.

Train Test





E.g.: dataset bias leads to higher errors on 'novel' data... Can an explanation point to such bias?

Training

Test

Most cows are black/brown



Most sheep are white



Prediction: "cow" 76%

Explanation





True class: "sheep"

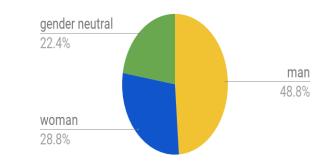
Gender bias in captioning models (Hendricks et al. 2018)

Evidence for "man"

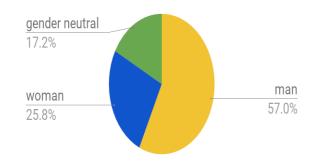


Baseline: A man sitting at a desk with a laptop computer.

Ground truth captions

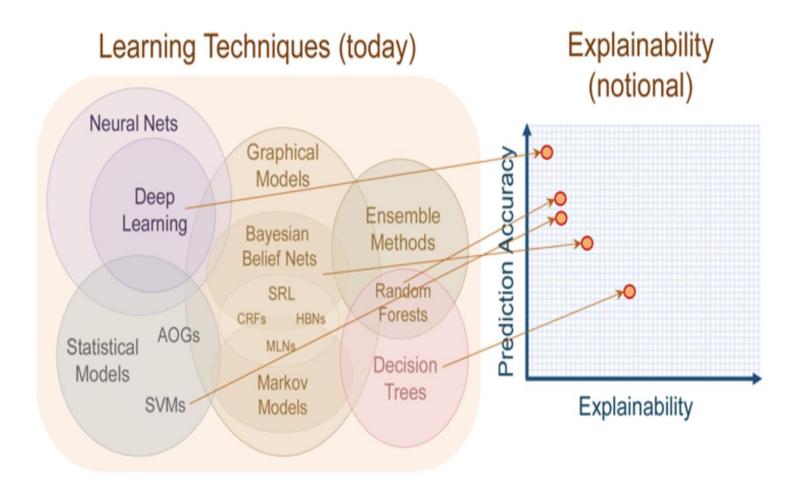


Generated captions

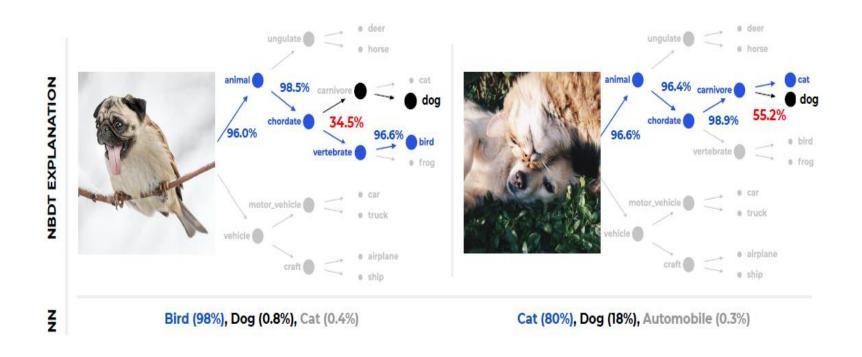


Hendricks et al. "Women Also Snowboard: Overcoming Bias in Captioning Models." ECCV 2018
Zhao et al. "Men also like shopping: Reducing gender bias amplification using corpus-level constraints."
EMNLP 2017

Accuracy vs. explainability



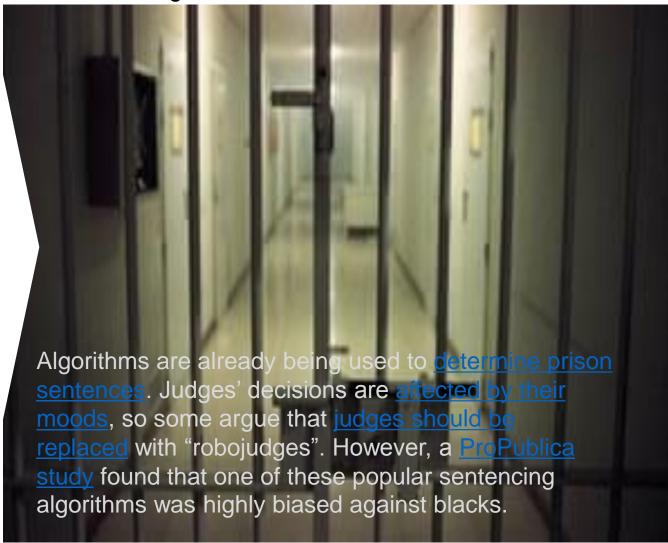
NBDT: Neural-Backed Decision Trees



A. Wan, L. Dunlap, D. Ho, J. Yin, S. Lee, H. Jin, S. Petryk, S. A. Bargal, J. E. Gonzalez. NBDT: Neural-Backed Decision Tree. *International Conference on Learning Representations (ICLR)*, 2021.

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If we start trusting algorithms to make decisions, who will have the final word on important decisions? Will it be humans, or algorithms?

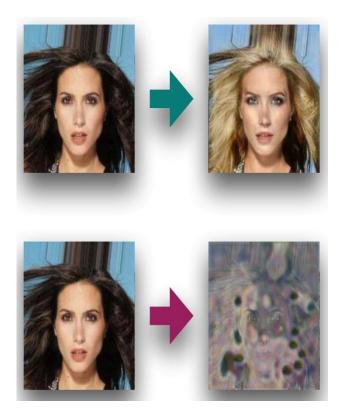


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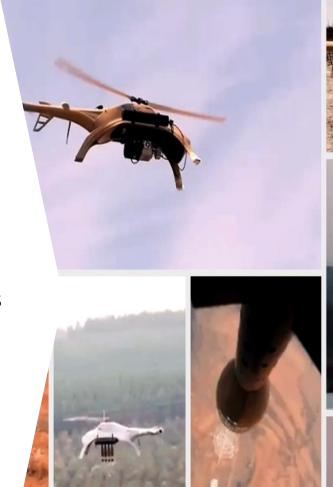
https://www.youtube.com/watch?v=VWrhRBb-1Ig

Disrupting Deepfakes



N. Ruiz, S. A. Bargal, S. Sclaroff. Disrupting DeepFakes: Adversarial Attacks Against Conditional Image Translation Networks and Facial Manipulation Systems. Workshop on Adversarial Machine Learning in Computer Vision at IEEE/CVF Conference on Computer Vision and Pattern Recognition (CVPR), 2020.

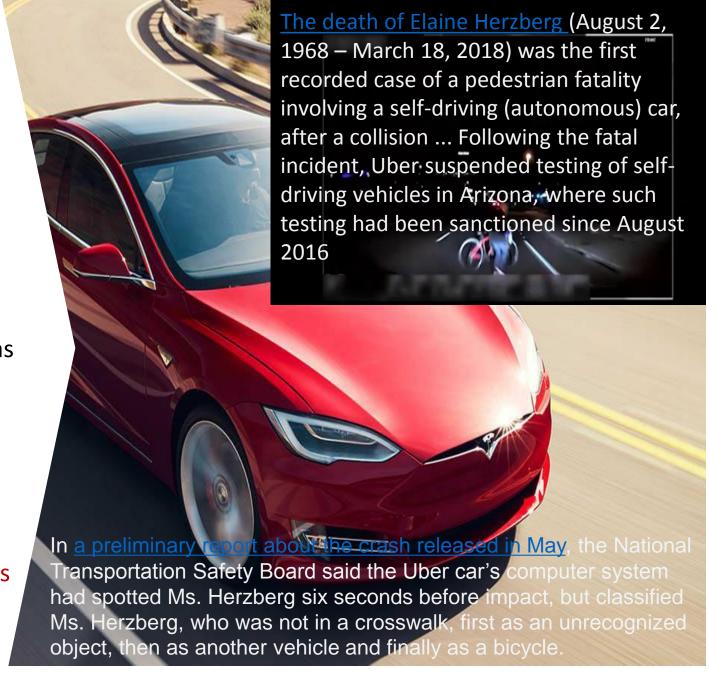
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https://www.albawaba.com/news/china-selling-autonomous-weaponized-drones-saudi-arabia-and-pakistan-1321951

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One in two American adults is in a law enforcement face recognition network-- https://www.perpetuallineup.org/



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