## CSCI 497P/597P: Computer Vision

|  | AC | AD | AE | AF | AG | AH | Al | AJ | AK | AL | AM | AN | AO | AP | AQ | AR | AS | AT |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 80 | 102 | 115 | 127 | 119 | 118 | 110 | 111 | 107 | 110 | 111 | 115 | 117 | 120 | 124 | 122 | 119 | 120 | 120 |
| 81 | 10 | 115 | 126 | 118 | 117 | 109 | 110 | 106 | 110 | 111 | 118 | 119 | 122 | 126 | 124 | 121 | 122 | 124 |
| 82 | 101 | 105 | 112 | 103 | 108 | 108 | 109 | 110 | 107 | 108 | 113 | 115 | 118 | 121 | 117 | 115 | 113 | 0 |
| 83 | 101 | 105 | 114 | 105 | 112 | 112 | 113 | 116 | 115 | 116 | 121 | 123 | 126 | 129 | 125 | 123 | 123 | 114 |
| 84 |  | 103 | 11: | 104 | 111 | 111 | 11 | 11 | 117 | 11 | 12 | 125 | 128 | 131 | 127 | 12 | 125 | 118 |
| 85 | 118 | 123 | 113 | 113 | 109 | 116 | 118 | 120 | 120 | 120 | 120 | 121 | 124 | 125 | 123 | 121 | 117 | 113 |
| 86 | 120 | 125 | 115 | 115 | 113 | 122 | 124 | 128 | 128 | 128 | 128 | 129 | 129 | 130 | 131 | 129 | 125 | 123 |
| 87 | 119 | 124 | 114 | 114 | 114 | 122 | 12 | 130 | 13 | 130 | 138 | 3: | 132 | 13. | 13 | 131 | 12 |  |
| 88 | 112 | 113 | 112 | 111 | 115 | 128 | 130 | 126 | 127 | 123 | 122 | 124 | 125 | 124 | 121 | 117 | 117 | 11 |
| 89 | 114 | 115 | 114 | 115 | 121 | 134 | 135 | 134 | 135 | 131 | 130 | 132 | 130 | 129 | 129 | 125 | 125 | 126 |
| 90 | 113 | 11 | 11 | 11 | 121 | 134 | 138 | 136 | 137 | 133 | 132 | 134 | 133 | 132 | 131 | 127 | 127 |  |
| 91 | 112 | 110 | 104 | 114 | 129 | 135 | 139 | 138 | 131 | 129 | 126 | 123 | 123 | 123 | 122 | 121 | 123 | 115 |
| 92 | 114 | 112 | 108 | 118 | 135 | 140 | 144 | 143 | 139 | 137 | 134 | 131 | 131 | 131 | 130 | 129 | 128 | 127 |
| 93 | 113 | 111 | 107 | 117 | 135 | 143 | 147 | 146 | 141 | 139 | 136 | 133 | 133 | 133 | 132 | 131 | 131 |  |
| 94 | 119 | 111 | 116 | 131 | 138 | 145 | 144 | 137 | 130 | 128 | 125 | 123 | 116 | 119 | 117 | 118 | 122 | 120 |
| 95 | 121 | 115 | 120 | 135 | 143 | 150 | 149 | 142 | 138 | 136 | 133 | 131 | 124 | 127 | 125 | 126 | 130 | 128 |
| 96 | 12 | 112 | 11 | 136 | 146 | 153 | 152 | 145 | 140 | 138 | 135 | 133 | 12 | 12 | 127 | 128 | 132 | 31 |
| 97 | 117 | 112 | 131 | 141 | 147 | 149 | 141 | 137 | 133 | 131 | 126 | 121 | 122 | 115 | 113 | 118 | 123 | 123 |
| 98 | 121 | 118 | 137 | 146 | 152 | 154 | 149 | 145 | 141 | 139 | 136 | 131 | 132 | 125 | 123 | 128 | 133 | 134 |
| 99 | 120 | 116 | 137 | 149 | 155 | 157 | 151 | 147 | 143 | 141 | 137 | 132 | 131 | 124 | 122 | 127 | 134 |  |
| 100 | 100 | 124 | 143 | 145 | 146 | 147 | 142 | 138 | 134 | 130 | 124 | 124 | 126 | 118 | 115 | 118 | 115 | 110 |
| 101 | 106 | 130 | 148 | 150 | 151 | 152 | 150 | 146 | 142 | 140 | 134 | 134 | 136 | 130 | 127 | 130 | 126 | 121 |
| 102 | 10 | 130 | 151 | 153 | 154 | 155 | 152 | 148 | 144 | 141 | 135 | 13 | 135 | 12 | 12. | 136 | 12 |  |
| 103 | 95 | 132 | 145 | 145 | 147 | 146 | 140 | 134 | 132 | 130 | 126 | 123 | 116 | 110 | 109 | 113 | 105 | 108 |
| 104 | 104 | 141 | 153 | 153 | 153 | 152 | 145 | 142 | 140 | 140 | 136 | 135 | 128 | 122 | 121 | 124 | 119 | 121 |
| 105 | . 03 | 140 | 155 | 155 | 153 | 152 | 148 | 144 | 142 | 141 | 135 | 133 | 12 | 120 | 121 | 12 | 122 |  |
| 106 | 109 | 131 | 140 | 145 | 144 | 143 | 142 | 134 | 134 | 129 | 120 | 109 | 106 | 106 | 112 | 108 | 106 | 10 |

Lecture 1: Course Overview; Images

## A few more words about Spring quarter madness.

## Course Overview

What does this course cover, and why?

## Visual Perception

What can you tell me about this image? or, What might you want to know about this image?

# Levels of Vision 

What can you tell me about this image? or, What might you want to know about this image?

"low-level"



What would a noise-free image look like?
Where are the edges?
Where are there straight lines?
Which patches of this image are distinctive?
How would the unblurred background appear?
Is this the same scene as another image?
How far from the camera is each point?
What is the 3D shape of the subject?
Is the image level? Which way is up?
Which groups of pixels "belong" together?
Which pixels belong to the subject?
What is the subject?
What breed of dog is the subject?
What is the dog's emotional state?

## Levels of Vision

Subfield/topic

Example(s)
"low-level"

Filtering / Image Processing

Feature detection
Computational Photography Image matching / stitching

> Geometric vision

Segmentation

Semantic understanding

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## In this course...

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## Course Overview

5ish projects / major topics:

- Image filtering

- Feature detection
- Panorama stitching
- Depth estimation via stereo vision
- Image recognition via deep learning


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"high-level"


## Socrative

Socrative instructions:

1. Go to socrative.com (or open Socrative Student app)
2. Click "Login" button, then Login as Student
3. Enter CSCI497P in the Room Name field
4. You're in!

Go ahead and answer my multiple choice question. Hint: the correct answer is D!

## Breakout rooms, Norms

- In small groups, spend 4 minutes introducing yourselves and agree on 1-3 norms for this class.
- Can be anything, but thinking about Zoom etiquette may be useful this quarter.
- Can relate to your expectations of me as well as of each other.
- One member of the group: submit your norms to the open-ended poll on Socrative.


## Vision is hard?



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WHEN A USER TAKES A PHOTO, THE APP SHOULD CHECK WHETHER THEY'RE IN A NATONAL PARK...

xkcd
9/24/2014

## Introducing: Flickr PARK or BIRD



Zion National Park Utah by Les Haines (cc) EY

convolution + pooling layers


Secretary Bird by Bill Gracey (cc) EY-NC-ND

fully connected layers


Nx binary classification
flickr
10/20/2014

## Why is vision hard?



Illumination

## Why is vision hard?



Intra-class variation


Background clutter


Motion (Source: S. Lazebnik)


Occlusion

## Why is vision hard?

The state of Computer Vision and Al: we are really, really far.


But for me it is also one of those examples that make me sad about the outlook for Al and for Computer Vision. What would it take for a computer to understand this image as you or I do? I challenge you to think explicitly of all the pieces of knowledge that have to fall in place for it to make sense. Here is my short attempt:

- You recognize it is an image of a bunch of people and you understand they are in a hallway
- You recognize that there are 3 mirrors in the scene so some of those people are "fake" replicas from different viewpoints.
- You recognize Obama from the few pixels that make up his face. It helps that he is in his suit and that he is surrounded by other people with suits.
- You recognize that there's a person standing on a scale, even though the scale occupies only very few white pixels that blend with the background. But, you've used the person's pose and knowledge of how people interact with objects to figure it out.
- You recognize that Obama has his foot positioned just slightly on top of the scale. Notice the language l'm using: It is in terms of the 3D structure of the scene, not the position of the leg in the 2D coordinate system of the image.
- You know how physics works: Obama is leaning in on the scale, which applies a force on it. Scale measures force that is applied on it, that's how it works $=>$ it will over-estimate the weight of the person standing on it.
- The person measuring his weight is not aware of Obama doing this. You derive this because you know his pose, you understand that the field of view of a person is finite, and you understand that he is not very likely to sense the slight push of Obama's foot.
- You understand that people are self-conscious about their weight. You also understand that he is reading off the scale measurement, and that shortly the over-estimated weight will confuse him because it will probably be much higher than what he expects. In other words, you reason about implications of the events that are about to unfold seconds after this photo was taken, and especially about the thoughts and how they will develop inside people's heads. You also reason about what pieces of information are available to people.
- There are people in the back who find the person's imminent confusion funny. In other words you are reasoning about state of mind of people, and their view of the state of mind of another person. That's getting frighteningly meta.
- Finally, the fact that the perpetrator here is the president makes it maybe even a little more funnier. You understand what actions are more or less likely to be undertaken by different people based on their status and identity.


## What is an image?

## Computationally speaking...



| 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 |
| ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: | ---: |
| 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 |
| 255 | 255 | 255 | 20 | 0 | 255 | 255 | 255 | 255 | 255 | 255 | 255 |
| 255 | 255 | 255 | 75 | 75 | 75 | 255 | 255 | 255 | 255 | 255 | 255 |
| 255 | 255 | 75 | 95 | 95 | 75 | 255 | 255 | 255 | 255 | 255 | 255 |
| 255 | 255 | 96 | 127 | 145 | 175 | 255 | 255 | 255 | 255 | 255 | 255 |
| 255 | 255 | 127 | 145 | 175 | 175 | 175 | 255 | 255 | 255 | 255 | 255 |
| 255 | 255 | 127 | 145 | 200 | 200 | 175 | 175 | 95 | 255 | 255 | 255 |
| 255 | 255 | 127 | 145 | 200 | 200 | 175 | 175 | 95 | 47 | 255 | 255 |
| 255 | 255 | 127 | 145 | 145 | 175 | 127 | 127 | 95 | 47 | 255 | 255 |
| 255 | 255 | 74 | 127 | 127 | 127 | 95 | 95 | 95 | 47 | 255 | 255 |
| 255 | 255 | 255 | 74 | 74 | 74 | 74 | 74 | 74 | 255 | 255 | 255 |
| 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 |
| 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 | 255 |

usually one byte per pixel

A grayscale (black-and-white) image is a 2D array of numbers.

## What is an image?

Mathematically speaking...
A grayscale (black-and-white) image is a function $f$, from $\boldsymbol{R}^{2}$ to $\boldsymbol{R}$.


## What's the difference?

The computational representation is a sampled version of the (ideal) mathematical representation.

(we can also still write a step function that represents the sampled version)

## Transforming Images

Written as a function, we can transform the image function to create altered functions (images):

(increase brightness)

$$
g(x, y)=f(x, y)+20
$$


(flip horizontally)

$$
g(x, y)=f(-x, y)
$$

## Real images aren't perfect

- Real images are not only sampled, but they often contain noise.


How could we denoise f?

