CSC420: Intro to Image Understanding Introduction

Sanja Fidler

January 11, 2021



The Team

• Instructor:



Sanja Fidler (fidler@cs.toronto.edu)

- Office: all office hours will be hosted online
- Office hours: Mon 11am-12pm. Please send me an email to schedule.
- TAs:



Sayyed Nezhadi (snezhadi@cs.toronto.edu)

Frank Shen (shenti11@cs.toronto.edu)

Course Information

- Class time: Monday at 9-11am
- Location: Online
- Tutorials: TUT0101 on Monday 11am-12pm (online), TUT0102 on Monday 12-1pm (online), demos and Q&A, we'll do it on demand
- Class Website:

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http://www.cs.toronto.edu/~fidler/teaching/2021/CSC420.html
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• The class will use Quercus for announcements and discussions

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Course Information

• **Textbook**: We won't directly follow any book, but extra reading in this textbook will be useful:



Rick Szeliski

Computer Vision: Algorithms and Applications

available free online:

http://szeliski.org/Book/

• Links to other material (papers, code, etc) will be posted on the class webpage

Course Prerequisites

Course Prerequisites:

- Data structures
- Linear Algebra
- Vector calculus
- Numerical Analysis

Without this you'll need some serious catching up to do!

Knowing some basics in this is a plus:

- Python, Matlab, C++
- Machine Learning
- Neural Networks
- Solving assignments sooner rather than later

Requirements

- Each student expected to complete 4 assignments and a project
- Assignments:
 - Short theoretical questions and programming exercises
 - Will be given roughly every two weeks (starting second week of class)
 - You will have a week to hand in the solution to each assignment
 - You need to solve the assignment alone

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- Need to hand in a report and do an oral presentation
- Can work individually or in pairs

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Grading

Grade breakdown

- **Assignments**: 60% (15% each)
- Project + oral exam: $30\%(project) + 10\%(oral\ exam)$
- For the project you will need to do
 - Short project proposal
 - Project report
 - Project presentation (oral)
- Oral exam: During the project presentation, you will be asked questions about the class material

Term Work Dates

Term Work	Post Date	Due Date
Assignment 1	Jan 22	Jan 29
Assignment 2	Feb 5	Feb 12
Assignment 3	March 5	March 12
Assignment 4	March 19	March 26
Project Report		April 15
Project Presentation		TBD

- All dates are for 2021
- Dates are approximate (depend on what material we cover in class)

Programming Language?

- Your assignments / project can be implemented either in Python,
 Matlab, or C++. Python is preferred, but not a requirement.
- Most code and examples we will provide during the class will be in Python and Matlab.
- Choose wisely

Lateness

Deadline The solutions to assignments / project should be submitted by 11.59pm on the date they are due.

Anything from 1 minute late to 24 hours will count as one late day.

Lateness Each student will be given a total of **3 free late days**.

This means that you can hand in three of the assignments one day late, or one assignment three days late. It is up to the you to make a good planning of your work. After you have used the 3 day budget, the late assignments will not be accepted.

Syllabus

Tentative syllabus

Intro Linear filters, edges Image features Keypoint detection Matching Stereo, multi-view Stereo, multi-view Object recognition Object detection Neural Networks Segmentation

Introduction

Let's begin!

Introduction to Intro to Image Understanding

- What is Computer Vision?
- Why study Computer Vision?
- Which cool applications can we do with it?
- Is vision a hard problem?

• A field trying to develop automatic algorithms that would "see"





Embodied Agents

 Understand the scene in order to take actions: perception, planning, reasoning



Figure: How do I make dinner in this household?

Many simulators: Carla, Thor, House3D, VirtualHome, etc

• What does it mean to see?

[text adopted from A. Torralba]

To know what is where by looking – Marr, 1982



• What does it mean to see?

- [text adopted from A. Torralba]
- To know what is where by looking Marr, 1982
- Understand where things are in the world



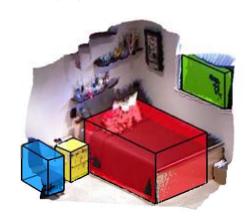


• What does it mean to see?

- [text adopted from A. Torralba]
- To know what is where by looking Marr, 1982
- Understand where things are in the world
- What are their 3D/material properties

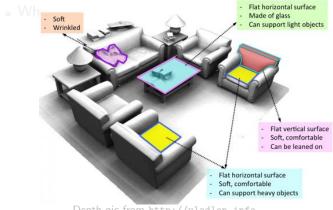
image





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• What does it mean to see?

- [text adopted from A. Torralba]
- To know what is where by looking Marr, 1982
- Understand where things are in the world
- What are their 3D/material properties?
- What actions are taking place?



gorillas arguing

Pic from www.cobblehillpuzzles.com

polar bear eating fish

• Full understanding of an image?

• Full understanding of an image? You can answer any question about it

[M. Malinowski, M. Fritz, A Multi-World Approach to Question Answering about Real-World Scenes based on Uncertain Input, NIPS, 2014]



Q: What is behind the table? A: window



Q: What is in front of the toilet? A: door



Q: What is on the counter in the corner? A: microwave



Q: What is behind the table? A: window



Q: What is in front of the toilet? A: door



Q: What is on the counter in the corner? A: microwave



Q: What is the shape of the green chair? A: horse shaped



Q: What is behind the table? A: window



Q: What is in front of the toilet? A: door



Q: What is on the counter in the corner? A: microwave



Q: What is the shape of the green chair? A: horse shaped



Q: Where is the oven? A: on the right side of the fridge

• Full understanding of an image? You can answer any question about it



Q: What is behind the table? A: window



Q: What is in front of the toilet? A: door



Q: Where is the oven? A: on the right side of the fridge



Q: What is on the counter in the corner? A: microwave



Q: What is the largest object? A: bed

Q: What is the shape of the green

Sanja Fidler



Q: Which object is red? A: toaster



Q: Which object is red? A: toaster



Q: How many drawers are there? A: 6



Q: How many doors are open A: 1



Q: How many lights are on? A: 6



Q: Which object is red? A: toaster



Q: How many drawers are there? A: 6



Q: How many doors are open A: 1



Q: How many lights are on? A: 6



Q: Can you make pizza in this room? A: yes



Q: Where can you sit? A: chairs, table, floor

Why study Computer Vision?

Why study Computer Vision?

You are curious how to one day make the robot walk your dog



http://www.cs.toronto.edu/~fidler/videos/robotsmovies.mov

Why study Computer Vision?

• ... and fold your laundry



https://www.youtube.com/watch?v=gy5g33S0Gzo



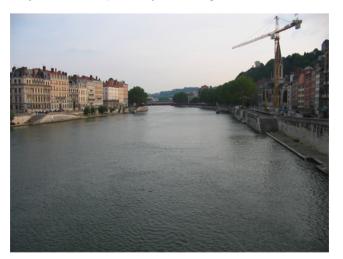
https://www.youtube.com/watch?v=KKUaVzf3Oqw

• ... and drive you to work

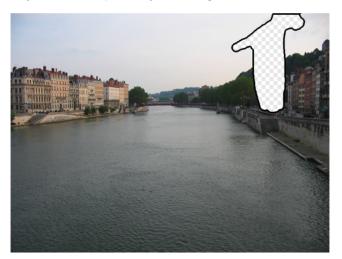


Amnon Shashua's Mobileye autonomous driving system

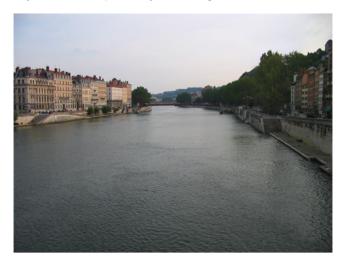
https://www.youtube.com/watch?v=4fxFDypHZLs



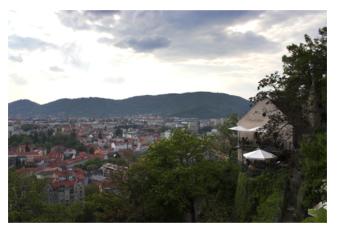
Scene Completion using Millions of Photographs, Hays & Efros, SIGGRAPH 2007



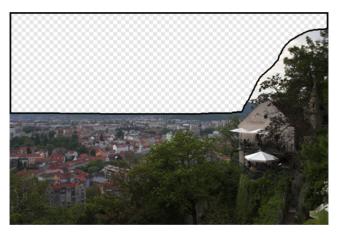
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Scene Completion using Millions of Photographs, Hays & Efros, SIGGRAPH 2007



Scene Completion using Millions of Photographs, Hays & Efros, SIGGRAPH 2007

Allows you to manipulate your images



https://www.youtube.com/watch?v=p5U4NgVGAwg

GauGan, Ming-Yu Liu et al., http://nvidia-research-mingyuliu.com/gaugan/]

• Change style of images









 $[\mathsf{Gatys},\,\mathsf{Ecker},\,\mathsf{Bethge}.\,\,\mathsf{A}\,\,\mathsf{Neural}\,\,\mathsf{Algorithm}\,\,\mathsf{of}\,\,\mathsf{Artistic}\,\,\mathsf{Style}.\,\,\mathsf{Arxiv}'\mathsf{15}.]$

• Change style of videos



https://www.youtube.com/watch?v=Khuj4ASldmU

[Ruder, Dosovitskiy, Brox. Artistic style transfer for videos, 2016]

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Why study Computer Vision?

Change style of videos

Bringing Impressionism to Life with Neural Style Transfer in Come Swim

Bhautik J Joshi* Research Engineer, Adobe Kristen Stewart Director, Come Swim David Shapiro Producer, Starlight Studios







Figure 1: Usage of Neural Style Transfer in Come Swim; left: content image, middle: style image, right: upsampled result. Images used with permission, (c) 2017 Starlight Studios LLC & Kristen Stewart.

Abstract

Neural Style Transfer is a striking, recently-developed technique that uses neural networks to artistically redraw an image in the style of a source style image. This paper explores the use of this technique in a production setting, applying Neural Style Transfer to redraw key scenes in Come Swive in the style of the impressionistic painting that inspired the film. We document how the technique can be driven within the framework of an iterative creative process to achieve a desired look, and propose a mapping of the broad parameter space to a key set of creative controls. We hope that this mapping can provide insights into priorities for future research.

execute efficiently and predictably. In a production setting, however, a great deal of creative control is needed to tune the result, and a rigid set of algorithmic constraints run counter to the need for this creative exploration. While early investigations so better map the low-level neural net evaluations to stylistic effects are underway [Li et al. 2017], in our paper we focused on examining the higher-level parameter space for Neural Style Transfer and found a set of working shortcuts to map them to a reduced but meaningful set of creative controls.

2 Realizing Directorial Intent

https://arxiv.org/pdf/1701.04928.pdf

• ... and make cool videos using a single image



http://www.cs.cmu.edu/~om3d/

3D Object Manipulation in a Single Photograph using Stock 3D Models, Kholgade, Simon, Efros, Sheikh, SIGGRAPH 2014

• Fancy visualization and game analysis in sports



• Fancy visualization and special effects in movies





[Source: http://cvfxbook.com and http://vimeo.com/100095868]

Reconstruct the world in 3D from online photos!



https://www.youtube.com/watch?v=IgBQCoEfiMs

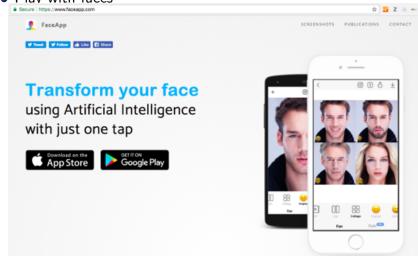
Photosynth: https://photosynth.net/

Nerf: https://www.youtube.com/watch?v=yPKIxoN2Vf0



Figure: Modiface: Toronto-based startup

Play with faces



• Play with faces



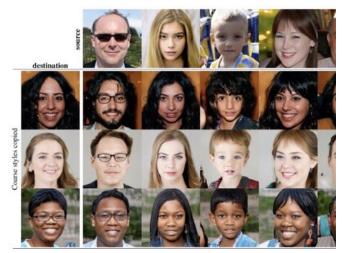
• Play with faces



Play with faces



Generate new faces



https://www.youtube.com/watch?v=kSLJriaOumA

StyleGAN, Tero Karras et al., https://github.com/NVlabs/stylegan]

Generate image descriptions automatically





[Source: L. Zitnick, NIPS'14 Workshop on Learning Semantics]

• Generate images from descriptions automatically

TEXT PROMPT an illustration of a baby daikon radish in a tutu walking a dog

AI-GENERATED IMAGES



View more or edit prompt ↓

TEXT PROMPT

an armchair in the shape of an avocado [...]

AI-GENERATED IMAGES











View more or edit prompt+

[DALL-E: https://openai.com/blog/dall-e/]

• Have a computer do math for you

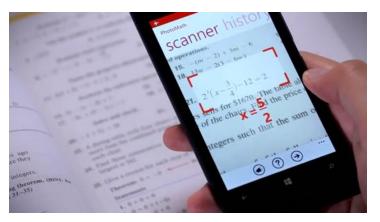


Figure: Photomath: https://photomath.net/, http://www.youtube.com/watch?v=XlbVB50mIh4

You can do movie-like Forensics



Figure: Source: Nayar and Nishino, "Eyes for Relighting"

[Source: N. Snavely]



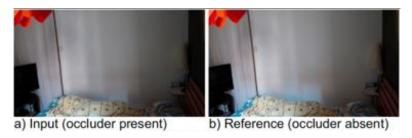
[Source: N. Snavely]



Figure: Source: Nayar and Nishino, "Eyes for Relighting"

[Source: N. Snavely]

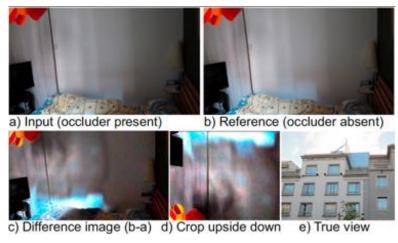
Some more CSI



• Can you see something on the wall?

Torralba & Freeman, CVPR'12

Some more CSI



Recognizing movie posters (in mobile phones)

iPhone Apps: kooaba (www.kooaba.com)



Source: S. Lazebnik

From student last year: phone app

How It All Began...

How It All Began...

MASSACHUSETTS INSTITUTE OF TECHNOLOGY PROJECT MAC

Artificial Intelligence Group Vision Memo. No. 100. July 7, 1966

THE SUMMER VISION PROJECT

Seymour Papert

The summer vision project is an attempt to use our summer workers effectively in the construction of a significant part of a visual system. The particular task was chosen partly because it can be segmented into sub-problems which will allow individuals to work independently and yet participate in the construction of a system complex enough to be a real landmark in the development of "pattern recognition".

[Slide credit: A. Torralba]

50 years and thousands of PhDs later...

Popular benchmarks: KITTI, PASCAL, Cityscapes, MS-COCO

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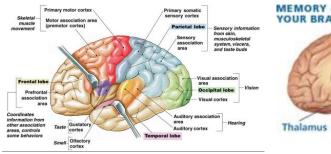
	Method	Setting	Code	Moderate	Easy	Hard	Runtime	Environment	Compare
1	THU CV-AI			91.97 %	91.96 %	84.57 %	0.38 s	GPU @ 2.5 Ghz (Python)	0
2	DH-ARI			91.48 %	90.87 %	82.25 %	4s	GPU @ 2.5 Ghz (C/C++)	0
3	HRI-SH			90.71 %	91.34 %	84.28 %	3.6 s	GPU ⊗ >3.5 Ghz (Python + C/C++)	
4	BM-NET			90.50 %	90.81 %	83.92 %	0.5 s	GPU @ 2.5 Ghz (Python + C/C++)	0
5	MVRA + I-FRCNN+			90.36 %	90.78 %	80.48 %	0.18 s	GPU @ 2.5 Ghz (Python)	0
6	TuSimple		code	90.33 %	90.77 %	82.86 %	1.6 s	GPU @ 2.5 Ghz (Python + C/C++)	

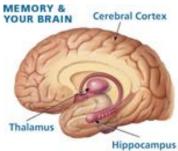
Reasoning demo: http://vqa.cloudcv.org/

Why is vision hard?

Why is vision hard?

 Half of the cerebral cortex in primates is devoted to processing visual information. This is a lot. Means that vision has to be pretty hard!





Why is vision hard?

All this is dog...

[slide adopted from: R. Urtasun]



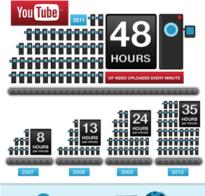


Biederman, 1987

[slide credit: R. Urtasun]

Lots of data to process:

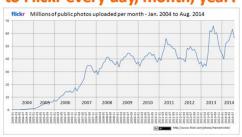
- Thousands to millions of pixels in an image
- 100 hours of video added to YouTube per minute [source: YouTube]
- Over 6 billion hours of video are watched each month on YouTube – almost an hour for every person on Earth [source: YouTube]



Lots of data to process:

- ullet \sim 5000 new tagged photos added to Flickr per minute (7M per day)
- ullet \sim 60M photos uploaded to Instagram every day [source: Instagram]

How many photos are uploaded to Flickr every day, month, year?





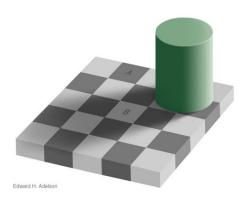
Exploit so Much Data!



Figure: Vemodalen: The Fear That Everything Has Already Been Done, https://www.youtube.com/watch?v=8ftDjebw8aA

[Source: L. Zitnick, NIPS'14 Workshop on Learning Semantics]

- Human vision seems to work quite well.
- How well does it really work?
- Let's play some games!



• Which square is lighter, A or B?



Edward H. Adelson

• Which square is lighter, A or B?



Figure: 2006 Walt Anthony

• Which red line is longer?



Figure: 2006 Walt Anthony

• Which red line is longer?



Figure: Ames room

Assumptions can be wrong



Figure: Chabris & Simons, https://www.youtube.com/watch?v=vJG698U2Mvo

- Count the number of times the white team pass the ball
- Concentrate, it's difficult!



Figure: Simons et al., http://www.perceptionweb.com/perception/perc1000/a_d_ex1.mov (more videos here: http://www.perceptionweb.com/misc.cgi?id=p3104)

• Is something happening in the picture?



Figure: Torralba et al., http://people.csail.mit.edu/torralba/courses/6.870/slides/blur.avi

• Can you describe what's going on in the video?



Figure: Torralba et al., http://people.csail.mit.edu/torralba/courses/6.870/slides/highres.avi

• Can you describe what's going on in the video?

What do I need...

What do I need to become a good Computer Vision researcher?

- Technical capabilities
- Good programming skills
- Imagination
- Even better intuition
- Lots of persistence
- Some luck always helps