CSC420: Intro to Image Understanding Introduction

Sanja Fidler

January 7, 2019



The Team

Instructor:



Sanja Fidler (fidler@cs.toronto.edu)

• Office: DH 3084

• Office hours: Monday 12-1pm, or by appointment

TA:



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Course Information

• Class time: Monday at 1-3pm

Location: NE 2110

 Tutorials: TUT0101 on Monday 4-5pm (DH 4001), TUT0102 on Monday 5-6pm (IB 210), demos and Q&A, we'll do it on demand

• Class Website:

http://www.cs.toronto.edu/~fidler/teaching/2019/CSC420.html

- The class will use Piazza for **announcements** and **discussions**: https://piazza.com/utoronto.ca/winter2019/csc420
- Your grade will not depend on your participation on Piazza.
 It's just a good way for asking questions, discussing with your instructor, TAs and your peers

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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

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: 40%

- For the project you will need to do
 - Short project proposal
 - Project report
 - Project presentation (oral)
- I will be asking questions about relevant part of the class material during project presentations which will influence the grade

Term Work Dates

| Term Work | Post Date | Due Date |
|----------------------|-----------|---------------------|
| Assignment 1 | Jan 17 | Jan 24 |
| Assignment 2 | Jan 31 | Feb 7 |
| Assignment 3 | Feb 14 | Feb 21 |
| Assignment 4 | Mar 7 | Mar 14 |
| Project Report | | First week of April |
| Project Presentation | | First week of April |

- All dates are for 2019
- Dates are approximate (depend on what material we cover in class)

Programming Language?

- Your assignments / project can be in Python, Matlab, C++
- As long as it compiles, runs, and you know how to defend it, we're happy
- HOWEVER, most code and examples we will provide during the class will be in Python, 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

| Week nb. | Date | Topic |
|----------|----------|-----------------------|
| 1 | Jan 7 | Intro |
| 2 | Jan 14 | Linear filters, edges |
| 3 | Jan 21 | Image features |
| 4 | Jan 28 | Keypoint detection |
| 5 | Feb 4 | Matching |
| 6 | Feb 11 | Grouping |
| 7 | Feb 18 | Stereo, multi-view |
| 8 | Feb 25 | Stereo, multi-view |
| 9 | March 4 | Object recognition |
| 10 | March 11 | Object detection |
| 11 | March 18 | Neural Networks |
| 12 | March 25 | Segmentation |
| 13 | April ? | Project Presentations |

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



[text adopted from A. Torralba]

- What does it mean to see?
 - To know what is where by looking Marr, 1982
 - Understand where things are in the world





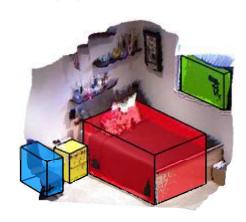
• What does it mean to see?

[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

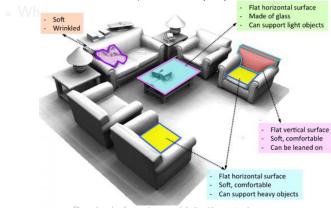




• What does it mean to see?

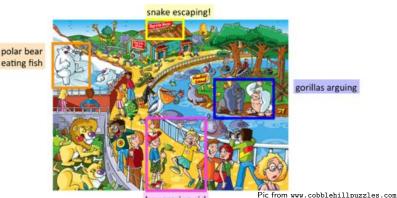
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[text adopted from A. Torralba]

- What does it mean to see?
 - 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?



polar bear

• 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



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



Q: What is the largest object? A: bed



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

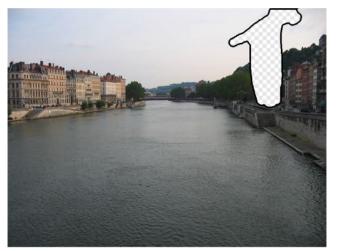


Amnon Shashua's Mobileye autonomous driving system

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



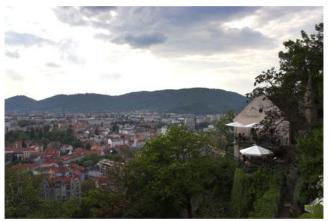
Scene Completion using Millions of Photographs, Hays & Efros, SIGGRAPH 2007



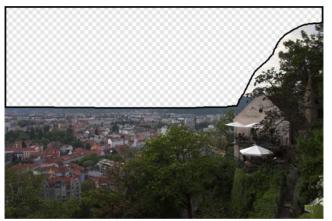
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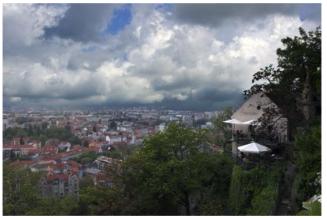
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Scene Completion using Millions of Photographs, Hays & Efros, SIGGRAPH 2007

Change style of images









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

• Change style of videos



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

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

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 Swie in the style of the impressionistic painting that inspired the film. We document how the technique can be driven within the framework of an inerative 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/ (try it!)

Figure out what people are wearing





http://clothingparsing.com (try it!)

• How Fashionable Are You?



LOS ANGELES, CA 466 FANS 288 VOTES 62 FAVOURITES

TAGS CHIC EVERDAY FALL

COLOURS WHITE-BOOTS NOVEMBER 10, 2014 GARMENTS

White Cheap Monday Boots Chilli Beans Sunglasses Missguided Romper Daniel Wellington Watch

COMMENTS

Nice!! Love the top! cute

Figure: An example of a post on http://www.chictopia.com. We crawled the site for 180K posts.

• How Fashionable Can You Become?



Figure: Examples of recommendations provided by our model. The parenthesis we show the fashionability scores.

[S. Zhu, C.C Loy, D. Lin, R. Urtasun, S. Fidler. In submission.]



The lady's upper-clothes contain the pattern of flowers

 $[S.\ Zhu,\ C.C\ Loy,\ D.\ Lin,\ R.\ Urtasun,\ S.\ Fidler.\ In\ submission.]$



The woman is wearing a blue short-sleeved T-shirt and blue jeans

[S. Zhu, C.C Loy, D. Lin, R. Urtasun, S. Fidler. In submission.]



A woman wearing a black overcoat and white shorts

 $[S.\ Zhu,\ C.C\ Loy,\ D.\ Lin,\ R.\ Urtasun,\ S.\ Fidler.\ In\ submission.]$



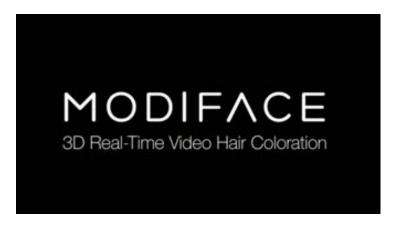
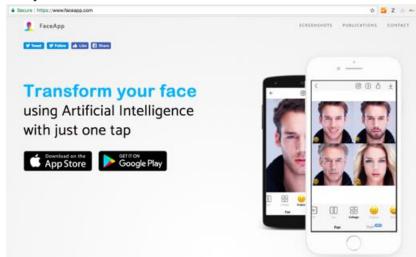


Figure: Modiface: Toronto-based startup









• Generate image captions automatically



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

• Generate image captions automatically

A man with a colorful umbrella walking down a street.



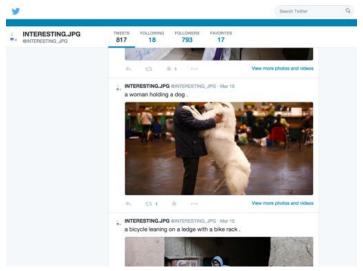
[Source: L. Zitnick, NIPS'14 Workshop on Learning Semantics]

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Generate image captions automatically



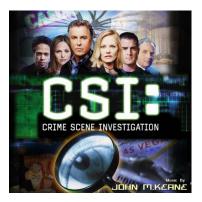
 $[Kiros,\,Salakhutdinov,\,Zemel.\,\,Unifying\,\,Visual-Semantic\,\,Embeddings\,\,with\,\,Multimodal\,\,Neural\,\,Language\,\,Models.\,\,2014]$

• Have a computer do math for you



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

Fingerprint recognition





[Source: S. Lazebnik]

You can do some movie-like Forensics



Figure: Source: Nayar and Nishino, Eyes for Relighting



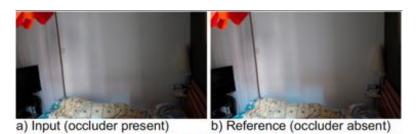
[Source: N. Snavely]



Figure: Source: Nayar and Nishino, Eyes for Relighting

[Source: N. Snavely]

Some more CSI

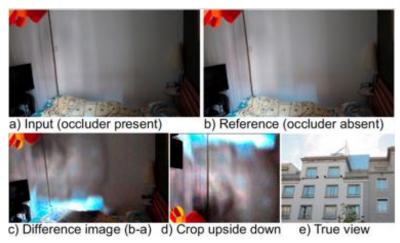


• Can you see something on the wall?

Torralba & Freeman, CVPR'12

Why study Computer Vision?

Some more CSI



Why study Computer Vision?

Object recognition (in mobile phones)





[Source: S. Seitz]

Why study Computer Vision?

• Games, games & games: 3D Pose Estimation with Depth Sensors



[Source: Microsoft Kinect]

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".

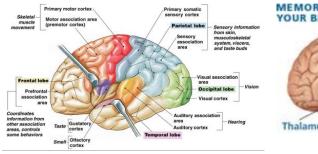
50 years and thousands of PhDs later...

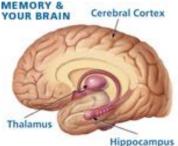
Popular benchmarks: KITTI, PASCAL, Cityscapes, MS-COCO

| | Method | Setting | Code | Moderate | Easy | Hard | Runtime | Environment | Compare |
|---|-----------------|---------|------|----------|---------|---------|---------|---------------------------------|---------|
| 1 | THU CV-AI | | 1000 | 91.97 % | 91.96% | 84.57 % | 0.38 s | GPU @ 2.5 Ghz (Python) | 0 |
| 2 | DH-ARI | | | 91.48 % | 90.87 % | 82.25 % | 46 | 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++) | 0 |
| 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++) | .0 |

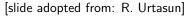
Reasoning demo: http://vqa.cloudcv.org/

 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!





All this is dog...





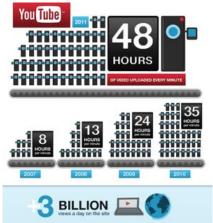


Biederman, 1987

[slide credit: R. Urtasun]

Lots of data to process:

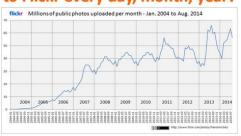
- Thousands to millions of pixels in an image
- 100 hours of video added to YouTube per minute [source: YouTube]
- Over 6 billion hours of video are watched each month on YouTube – almost an hour for every person on Earth [source: YouTube]



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?



| 1 | Instagram | | | | | | |
|----------|--------------------------------------|-----|------|-------------------------------|--|--|--|
| 200M | 65% ⁺ People Outside U.S. | 20B | 1.6B | 60M Average Photos Per Day | | | |

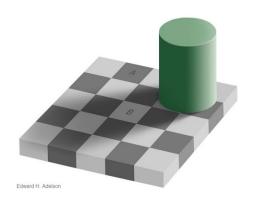
Exploit so Much Data!



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. Adeison

• 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