CSC420: Intro to Image Understanding Introduction

Sanja Fidler

September 15, 2015



The Team

• Instructor:



Sanja Fidler (fidler@cs.toronto.edu)

• Office: 283B in Pratt

• Office hours: Tuesday 1.20-2.50pm, or by appointment

• TAs:



Sara Sabour (saaraa@cs.toronto.edu)



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Kaustav Kundu (kkundu@cs.toronto.edu)

Course Information

- Class time: Tuesday and Thursday at 3-4pm
- Location: BA1200
- Tutorials: demos and Q&A, we'll do it on demand
- Class Website:

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http://www.cs.toronto.edu/~fidler/teaching/2015/CSC420.html
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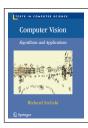
• The class will use Piazza for **announcements** and **discussions**:

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https://piazza.com/utoronto.ca/fall2015/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

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:

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

Requirements and Grading

Each student expected to complete 5 assignments and a project

Grading

• Assignments: 60% (12% each)

• Project: 40%

Assignments:

- Short theoretical questions and programming exercises
- Will be given every **two weeks** (starting with second week of class)
- You will have a week to hand in the solution to each assignment
- You need to solve the assignment alone

• Project:

- You will be able to choose from a list of projects or come up with your own project (discussed prior with your instructor)
- Need to hand in a report and do an oral presentation
- Can work individually or in pairs

Term Work Dates

Term Work	Post Date	Due Date	% of grade
Assignment 1	Sept 22	Sept 29	12%
Assignment 2	Oct 6	Oct 13	12%
Assignment 3	Oct 20	Oct 27	12%
Assignment 4	Nov 3	Nov 10	12%
Assignment 5	Nov 17	Nov 24	12%
Project Report		Dec 5	25%
Project Presentation		Dec 8	15%

• All dates are for 2015. ;)

Programming Language?

- Your assignments / project can be in Matlab, Python, 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 Matlab
- Choose wisely

Lateness

Deadline The solutions to the 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	Sept 15	Intro
2	Sept 17 & Sept 22	Linear filters, edges
3	Sept 24 & Sept 29	Image features
4	Oct 1 & Oct 6	Keypoint detection
5	Oct 8 & Oct 13	Matching
6	Oct 15 & Oct 20	Segmentation
7	Oct 22 & Oct 27	Grouping
8	Oct 29 & Nov 3	Object recognition
9	Nov 5 & Nov 12	Object detection
10	Nov 17 & Nov 19	Neural Networks
11	Nov 24 & Nov 26	Stereo, multi-view
12	Dec 1 & Dec 3	Recognition in 3D
13	Dec 8 & Dec 9	Project Presentations

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"





• What does it mean to see?

[text adopted from A. Torralba]

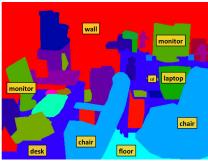
To know what is where by looking – Marr, 1982



• 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





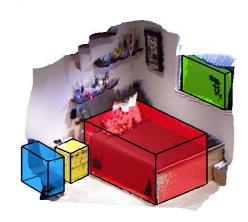
• What does it mean to see?

[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 properties?

image





• 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 properties? Flat horizontal surface Made of glass - Soft Can support light objects Wrinkled Flat vertical surface Soft, comfortable Can be leaned on Flat horizontal surface Soft, comfortable Can support heavy objects

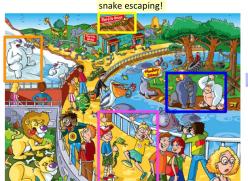
Depth pic from http://vladlen.info

• 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 properties?
- What actions are taking place?

polar bear eating fish



gorillas arguing

boy scaring girl Pic from www.cobblehillpuzzles.com

• 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?



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?



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

• Because it is challenging and fun



Jialiang Wang's (4th undergraduate year, UofT) video about his summer research in computer vision (click on the pic to see video – you'll need internet connection)

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



(click on the pic to see video)

• ... and fold your laundry





(click on each pic to see videos)

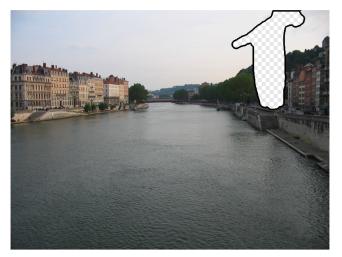
• ... and drive you to work (video)



Amnon Shashua's Mobileye autonomous driving system



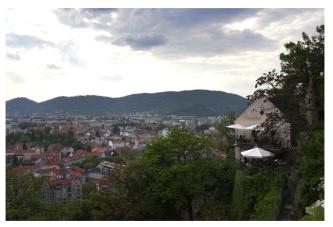
Scene Completion using Millions of Photographs, Hays & Efros, SIGGRAPH 2007



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

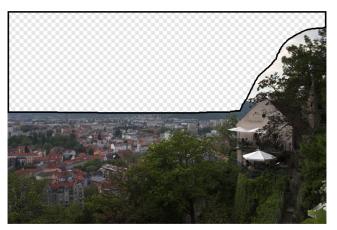


Scene Completion using Millions of Photographs, Hays & Efros, SIGGRAPH 2007



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

Allows you to manipulate your images



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

Allows you to manipulate your images



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

Change style of images









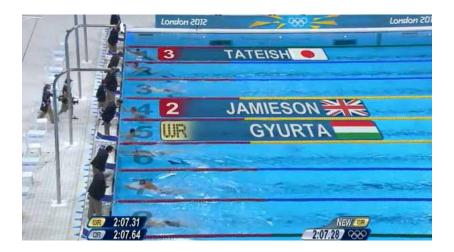
 $[\mathsf{Gatys},\ \mathsf{Ecker},\ \mathsf{Bethge}.\ \mathsf{A}\ \mathsf{Neural}\ \mathsf{Algorithm}\ \mathsf{of}\ \mathsf{Artistic}\ \mathsf{Style}.\ \mathsf{Arxiv'15}.]$

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



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

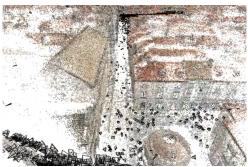




[Source: http://cvfxbook.com and http://vimeo.com/100095868]

Reconstruct the world in 3D from online photos!





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?



Current Outfit: Pink Outfit (3)

Recommendations: Heels (8) Pastel Shirts/Skirts (8) Black/Gray Tights/Sweater (5)



Current Outfit: Pink/Blue Shoes/Dress Shorts (3)

Recommendations: Black/Gray Tights/Sweater (5) Black Casual (5) Black Boots/Tights (5)



Current Outfit: Pink/Black Misc. (5)

Recommendations: Pastel Dress (8) Black/Blue Going out (8) Black Casual (8)



Current Outfit: Blue with Scarf (3)

Recommendations: Heels (8) Pastel Shirts/Skirts (8) Black Casual (8)



Current Outfit: Pink/Blue Shoes/Dress Shorts (3)

Recommendations: Black Casual (7) Black Heavy (3) Navy and Bags (3)



Current Outfit: Formal Blue/Brown (5)

Recommendations: Pastel Shirts/Skirts (9) Black/Blue Going out (8) Black Boots/Tights (8)

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

Crazy media attention!!!



QUARTZ Quartz

TECHTIMES **Tech Times**

WIRED.CO.UK

Mashable

Wired, UK

Mashable



Huffington Post, UK (video)

Huffington Post, Canada

y msn MSN. Canada Protein Protein

AOL News (video)

MailOnline

psfk

thestar.com (

YAHOO! NEWS Yahoo, Canada **Science**Daily Science Daily

Daily Mail, UK

PSFK Toronto Star





















Harper's Bazaar

FASHION

Fashion Magazine



Yahoo style Red Magazine, UK



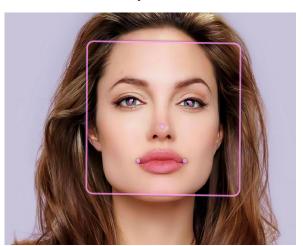


Crazy media attention!!!



From Cosmopolitan: The technology scores your facial attributes (this just keeps getting better, doesn't it) from your looks, to your age, and the emotion you're showing, before combining all the information using an equation SO complex we won't begin to go into it.

• Detect and analyze faces



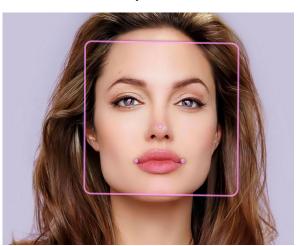
http://www.rekognition.com (try it!)



confidence : true (value : 1)
pose :roll(0.9) ,yaw(3.59) ,pitch(8.63)
race : white(0.28)
emotioin : calm:68%,happy:28%
age : 29.52 (value : 29.52)
smile : true (value : 0.65)
glasses : no glass (value : 0)
sunglasses : false (value : 0)
eye_closed : open (value : 0)
mouth_open_wide : 3% (value : 0.03)
beauty : 99.42 (value : 0.99422)

gender: female (value: 0)

• Detect and analyze faces



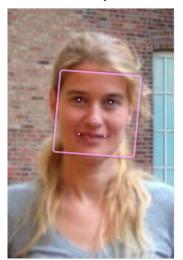
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confidence : true (value : 1)
pose :roll(0.9) ,yaw(3.59) ,pitch(6.63)
race : white(0.28)
emotion : calm:68%,happy:28%
age : 29.52 (value : 29.52)
smile : true (value : 0.65)
glasses : no glass (value : 0)
sunglasses : false (value : 0)
eye.closed : open (value : 0)
mouth open wide : 3% (value : 0.03)
beauty : 59.42 (value : 0.99422)

gender: female (value: 0)

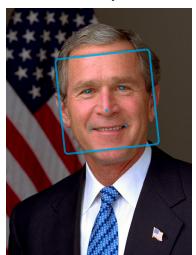
Detect and analyze faces





http://www.rekognition.com (try it!)

• Detect and analyze faces

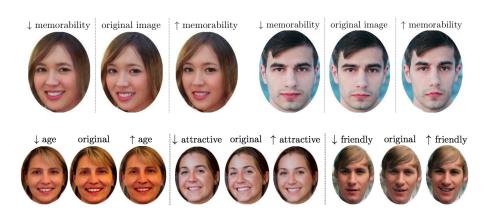


http://www.rekognition.com (try it!)



```
confidence: true (value:1)
pose roll-(6.26) yaw(-6.81) .pitch(1.66)
race: white(0.99)
emotioin: happy:92%, confused:1%
age:60.9 (value:60.9)
smile: true (value:0.87)
glasses: no glass (value:0.01)
sunglasses: false (value:0.0)
eye_closed: open (value:0)
mouth_open_wide:3% (value:0.03)
beauty:78.62 (value:0.78628)
gender: male (value:1)
```

You can make yourself look better (and competitors worse)



[Khosla, Bainbridge, Oliva, Torralba, Modifying the Memorability of Face Photographs, ICCV 2013]

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.



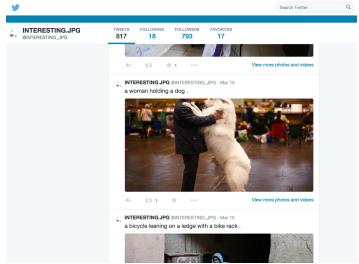
[Source: L. Zitnick, NIPS'14 Workshop on Learning Semantics]

Generate image captions automatically



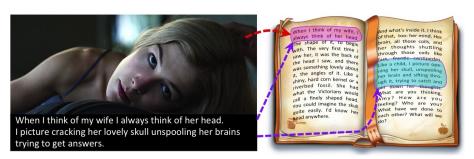
[Source: L. Zitnick, NIPS'14 Workshop on Learning Semantics]

Generate image captions automatically



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

Align movies and books



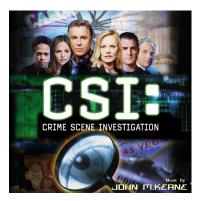
[Zhu, Kiros, Zemel, Salakhutdinov, Urtasun, Torralba, Fidler. ICCV'15.]

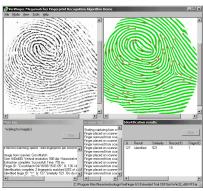
• Have a computer do math for you



Figure: Photomath: https://photomath.net/

Fingerprint recognition





[Source: S. Lazebnik]

You can do some 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.





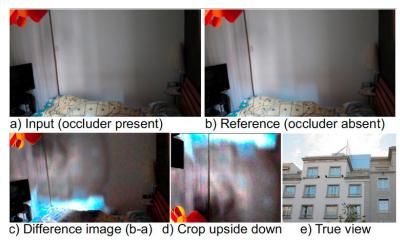
a) Input (occluder present)

b) Reference (occluder absent)

• Can you see something on the wall?

Torralba & Freeman, CVPR'12

Some more CSI



• Object recognition (in mobile phones)





[Source: S. Seitz]

Recognizing movie posters (in mobile phones)

iPhone Apps: kooaba (www.kooaba.com)



Source: S. Lazebnik

• 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".

[Slide credit: A. Torralba]

50 years and thousands of PhDs later...

Popular benchmarks:





Car

	Method	Setting	Code	<u>Moderate</u>	Easy	Hard	Runtime	Environment	Compare
1	DenseBox2			89.32 %	93.94 %	79.81 %	5 s	GPU @ 2.5 Ghz (C/C++)	
2	DJML			88.79 %	91.31 %	77.73 %	x s	GPU @ 1.5 Ghz (Matlab + C/C++)	
3	3DOP	ďď		88.64 %	93.04 %	79.10 %	3s	GPU @ 2.5 Ghz (Matlab + C/C++)	

X. Chen, K. Kundu, Y. Zhu, A. Berneshawi, H. Ma, S. Fidler and R. Urtasun: 3D Object Proposals for Accurate Object Class Detection. NIPS 2015.

Cyclist

A. Gonzalez, G. Villalonga, J. Xu, D. Vazquez, J. Amores and A. Lopez: Multiview Random Forest of Local Experts Combining RGB and LIDAR data for Pedestrian Detection. IEEE Intelligent Vehicles Symposium (IV) 2015.

| Mark |

50 years and thousands of PhDs later...

- Algorithms work pretty well
- Still some embarrassing mistakes...
- The general vision problem is not yet solved

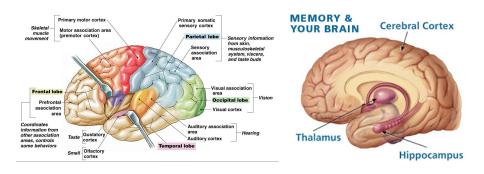




Where pink means "person"

[This pic is from 2014]

 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...





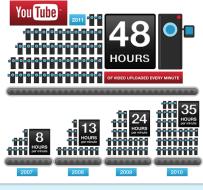


Biederman, 1987

[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?



↑ Instagram			Log in	
200M Monthly Actives	65% ⁺ People Outside U.S.	20B Photos Shared	1.6B	60M Average Photos Per Day

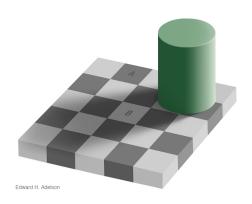
Exploit so Much Data!



Figure: Vemodalen: The Fear That Everything Has Already Been Done

[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

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



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

• Is something happening in the picture?



Figure: Torralba et al.

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



Figure: Torralba et al.

• 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?

- Some math knowledge
- Good programming skills
- Imagination
- Even better intuition
- Lots of persistence
- Some luck always helps