Introduction



16-385 Computer Vision Spring 2020, Lecture 1

Overview of today's lecture

- Teaching staff introductions
- What is computer vision?
- Course fast-forward and logistics

Teaching staff introductions

Instructor: Ioannis (Yannis) Gkioulekas

I won't hold it against you if you mispronounce my last name



Originally from Greece



National Technical University of Athens (2004-2009)



Harvard University (2009-2017)



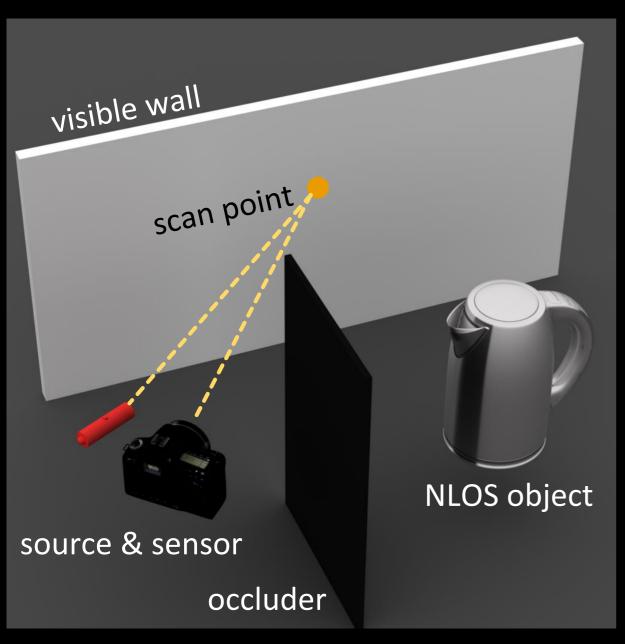
Carnegie Mellon University (2017-now)



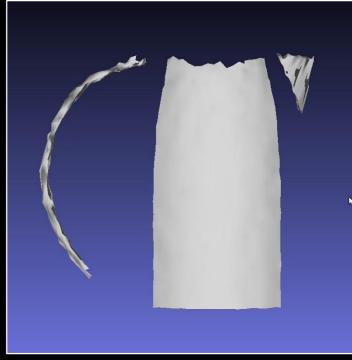
Yannis at Harvard in 2011

My website: http://imaging.cs.cmu.edu/~igkioule See also: http://imaging.cs.cmu.edu/

Looking around corners



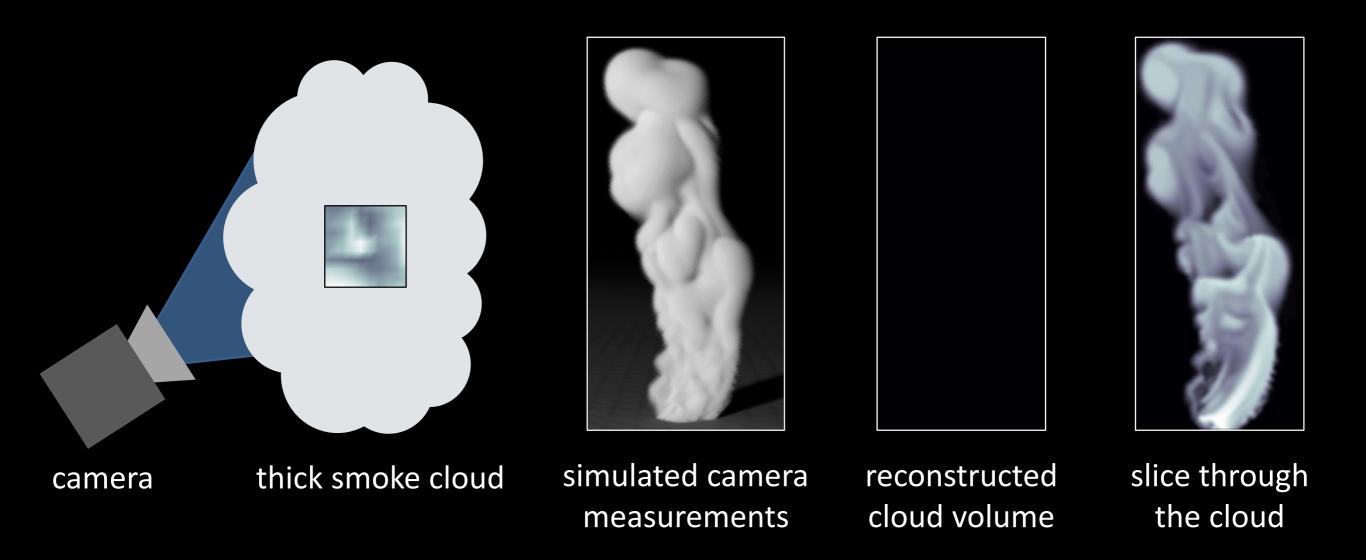




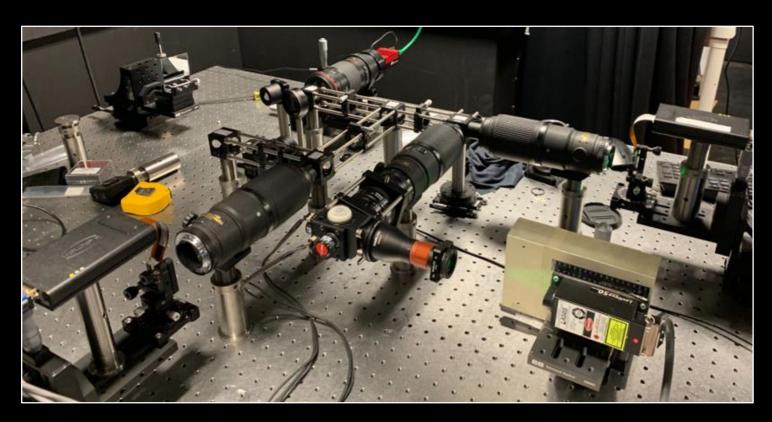
what a regular camera sees

what we can reconstruct

Looking inside deep scattering objects

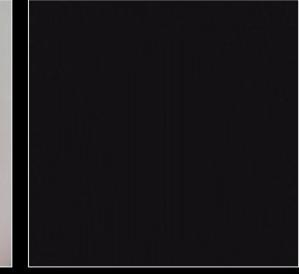


Seeing light in flight

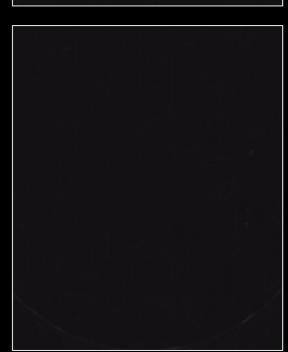


camera for capturing video at 10¹⁵ frames per second

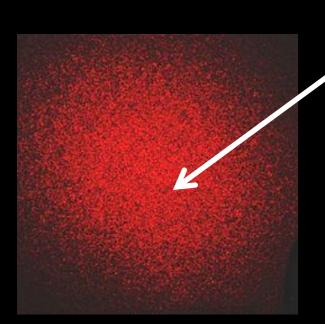






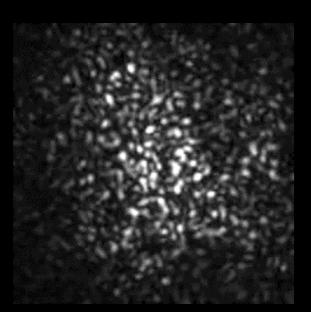


Rendering wave effects

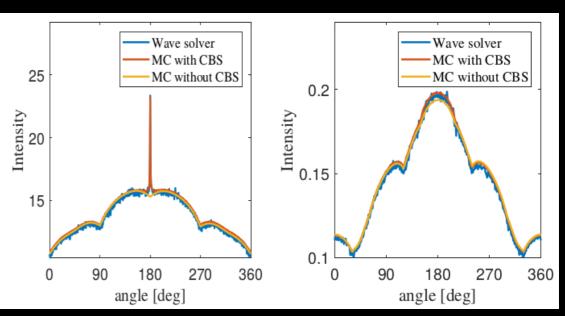


speckle: noiselike pattern

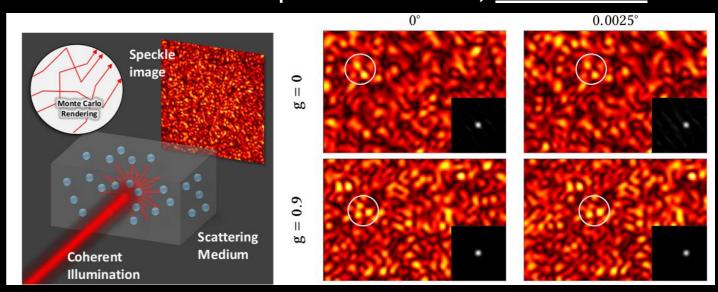
what real laser images look like



what real laser videos look like

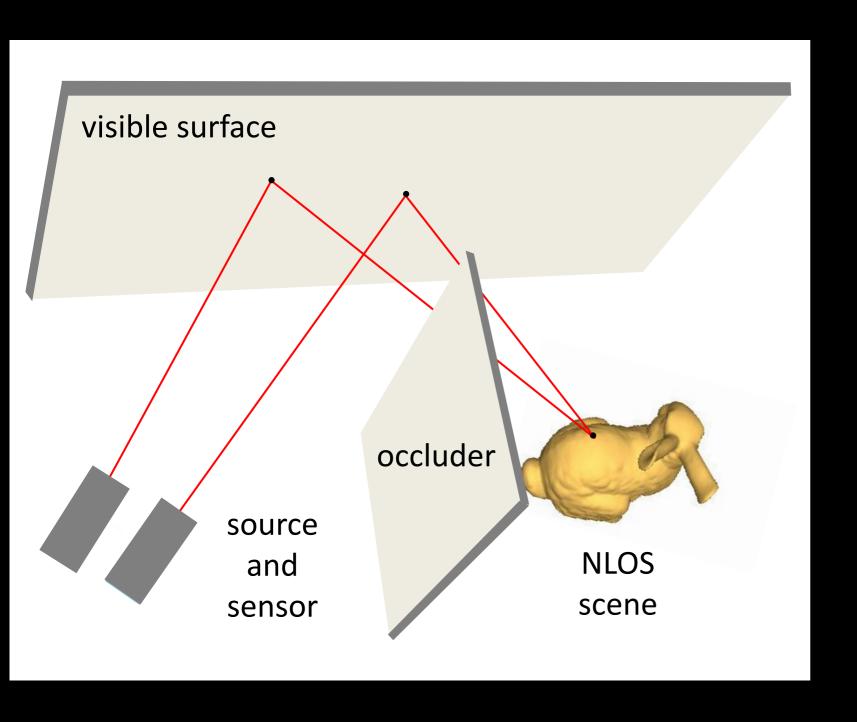


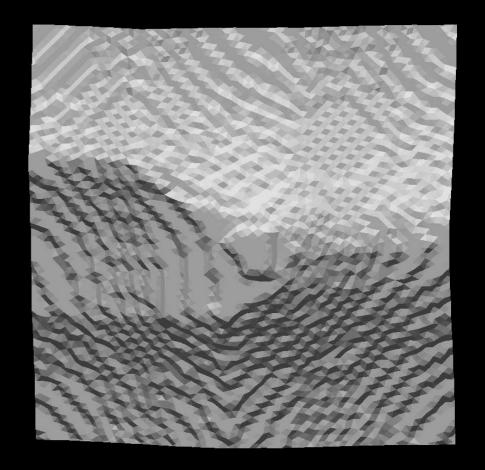
match wave equation solvers, **10**5x faster



reproduce physical effects like memory effect

Differentiable rendering





reconstruction evolution
http://imaging.cs.cmu.edu/

TA: Anand Bhoraskar

Master of Science in Computer Vision (MSCV)

Research Interests:

Simultaneous Localization and Mapping (SLAM), Deep Learning, Object Detection/Tracking

Current Area of Research:

Long term mapping for SLAM for dynamic environments

Past Research:

Video Stabilization, Object Tracking



TA: Prakhar Kulshreshtha (PK)

Master of Science in Computer Vision (MSCV)

Research Interests:

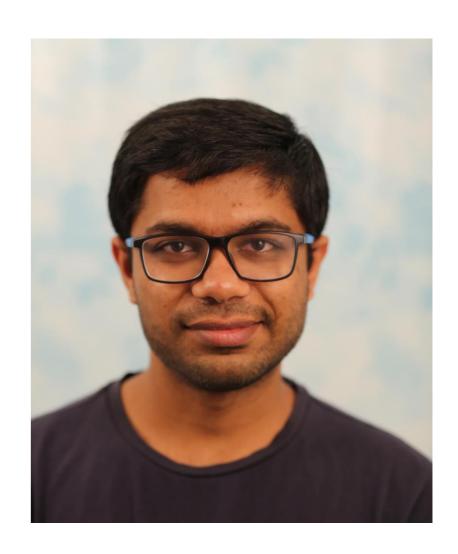
- Deep Learning for Detection and Instance Segmentation
- SLAM
- Online Learning in Computer Vision

Current area of research:

 Long-term SLAM for Dynamic environments (under Prof. Michael Kaess)

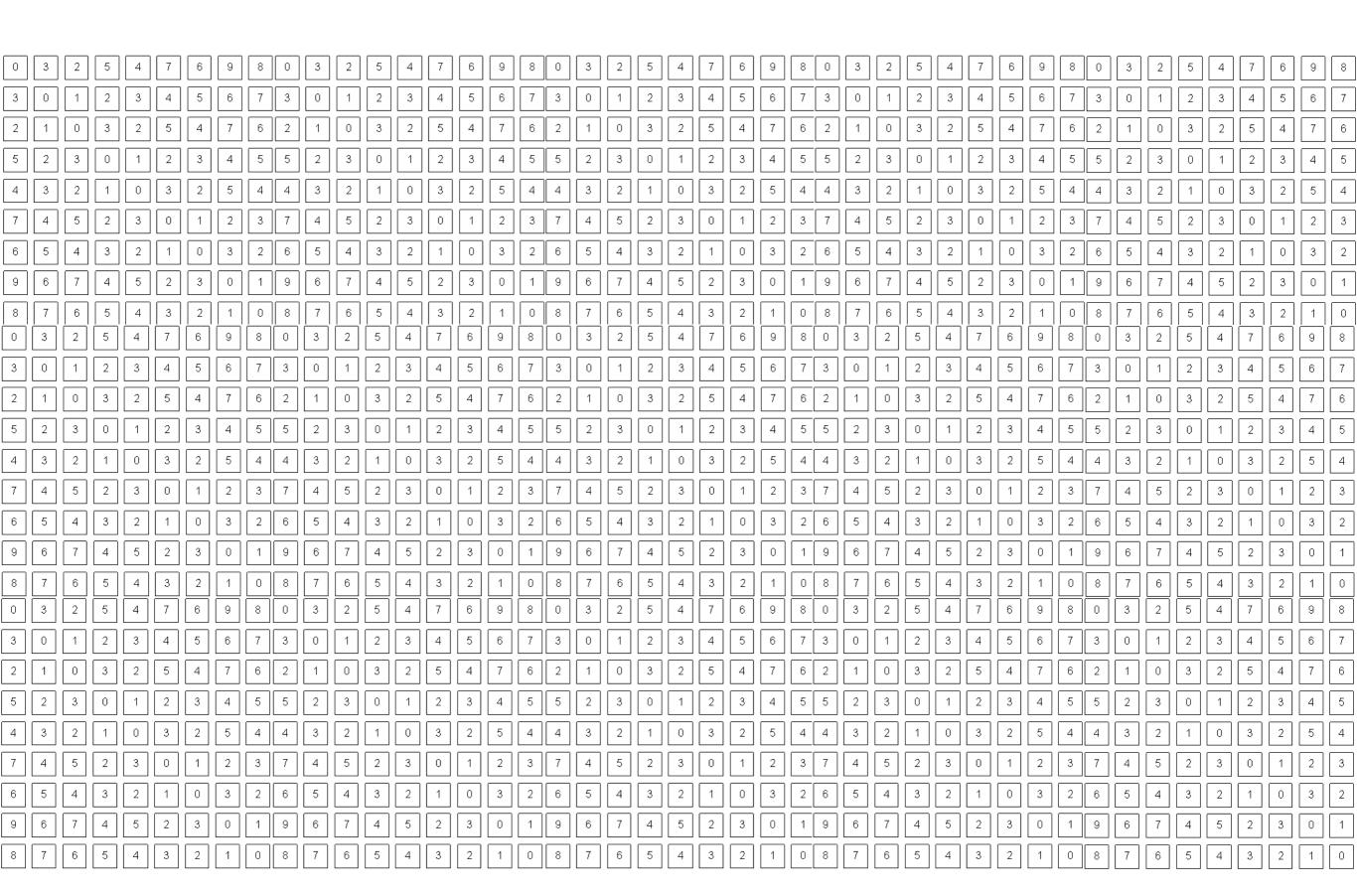
Past Reseach:

- Instance segmentation for quality estimation of food grains on a smartphone
- Text Intelligence in smartphone keyboard apps
- Online Face Clustering



What is computer vision?



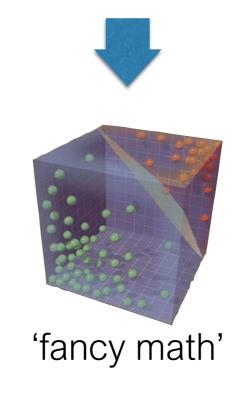


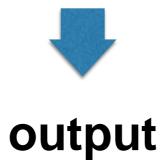
What a computer sees



The goal of computer vision is to give computers (super) human-level perception

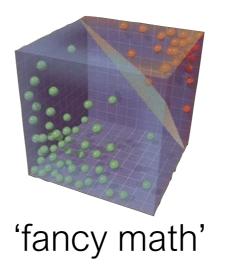
representation

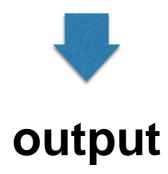




representation

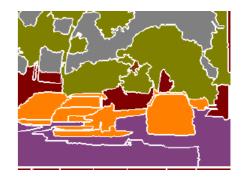






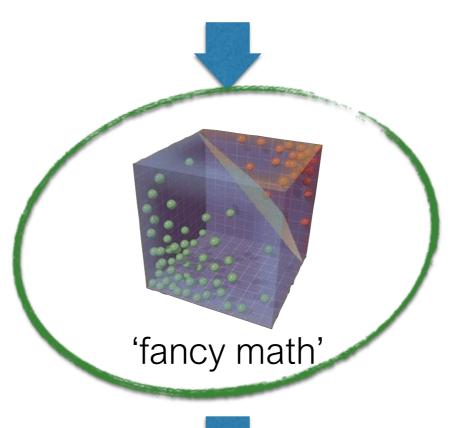


what should we look at? (image features)

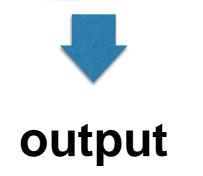


what can we understand? (semantic segmentation)

representation

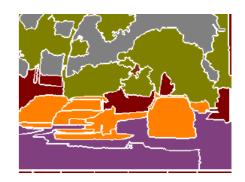


easy to get lost in the techniques

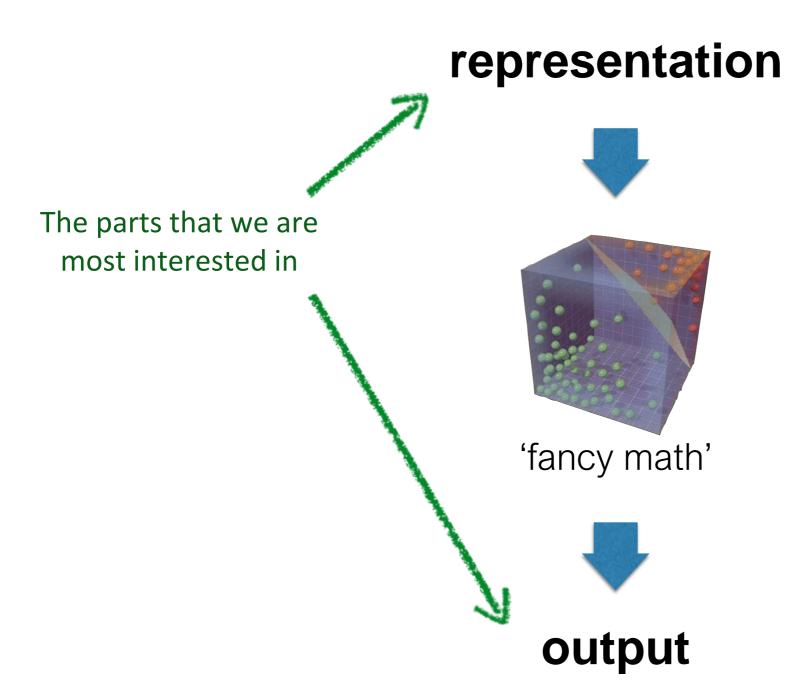




what should we look at? (image features)

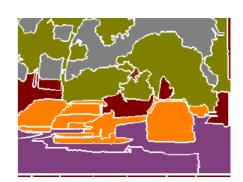


what can we understand? (semantic segmentation)





what should we look at? (image features)



what can we understand? (semantic segmentation)

Important note:

In general, computer vision does not work

Important note:

In general, computer vision does not work

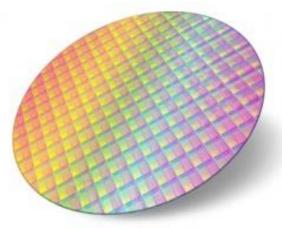
(except in certain situations/conditions)

Applications of computer vision

Machine vision

Automated visual inspection





Object Recognition



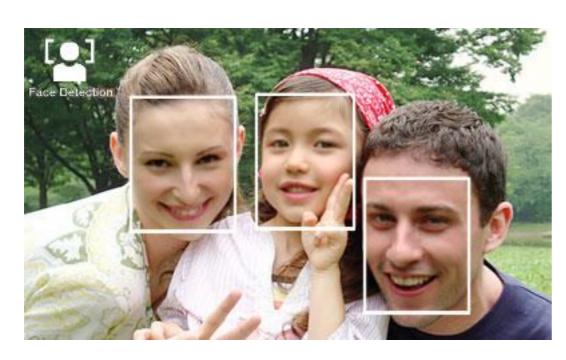
Toshiba Tech IS-910T

2013

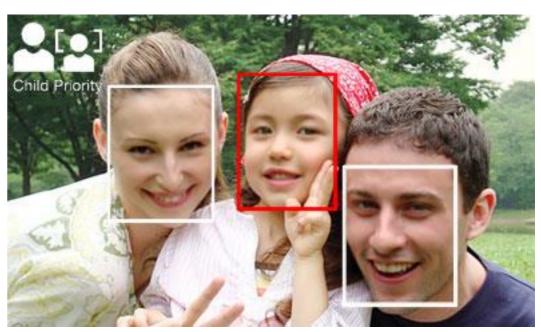


DataLogic LaneHawk LH4000

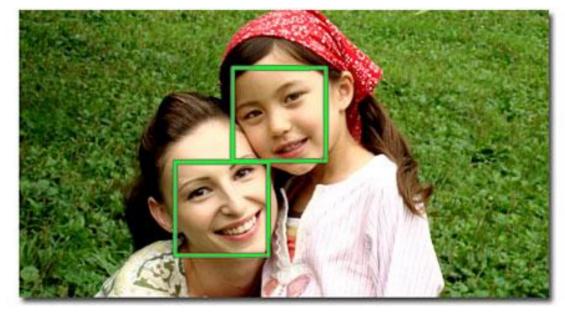
Face detection



Sony Cyber-shot

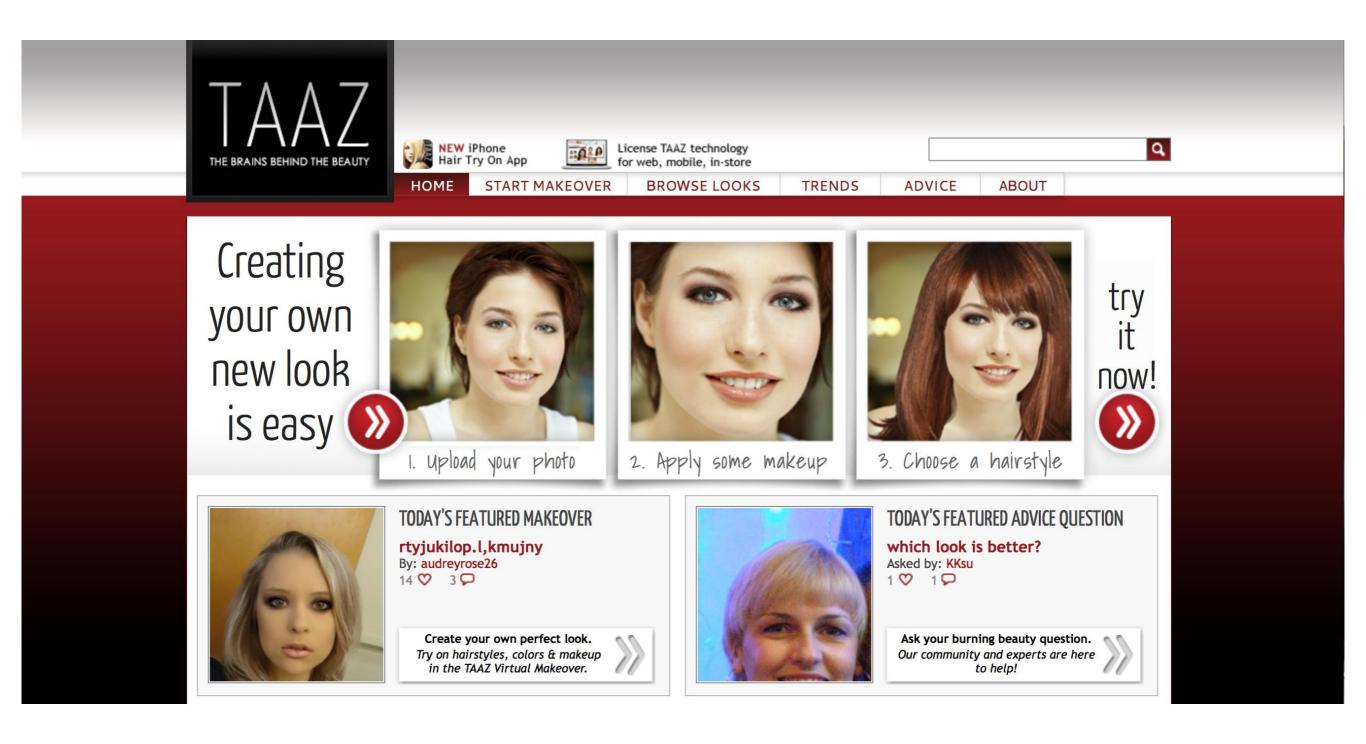


Age recognition



Smile recognition

Face makeovers









Word Lens



www.QuestVisual.com

First-down line





BMW 5 series

BMW night vision





"Around view" camera

Infinity EX





The system converts image data taken by 4 super-wide angle cameras, to display a virtual image of the vehicle from above.

Vision in Cars

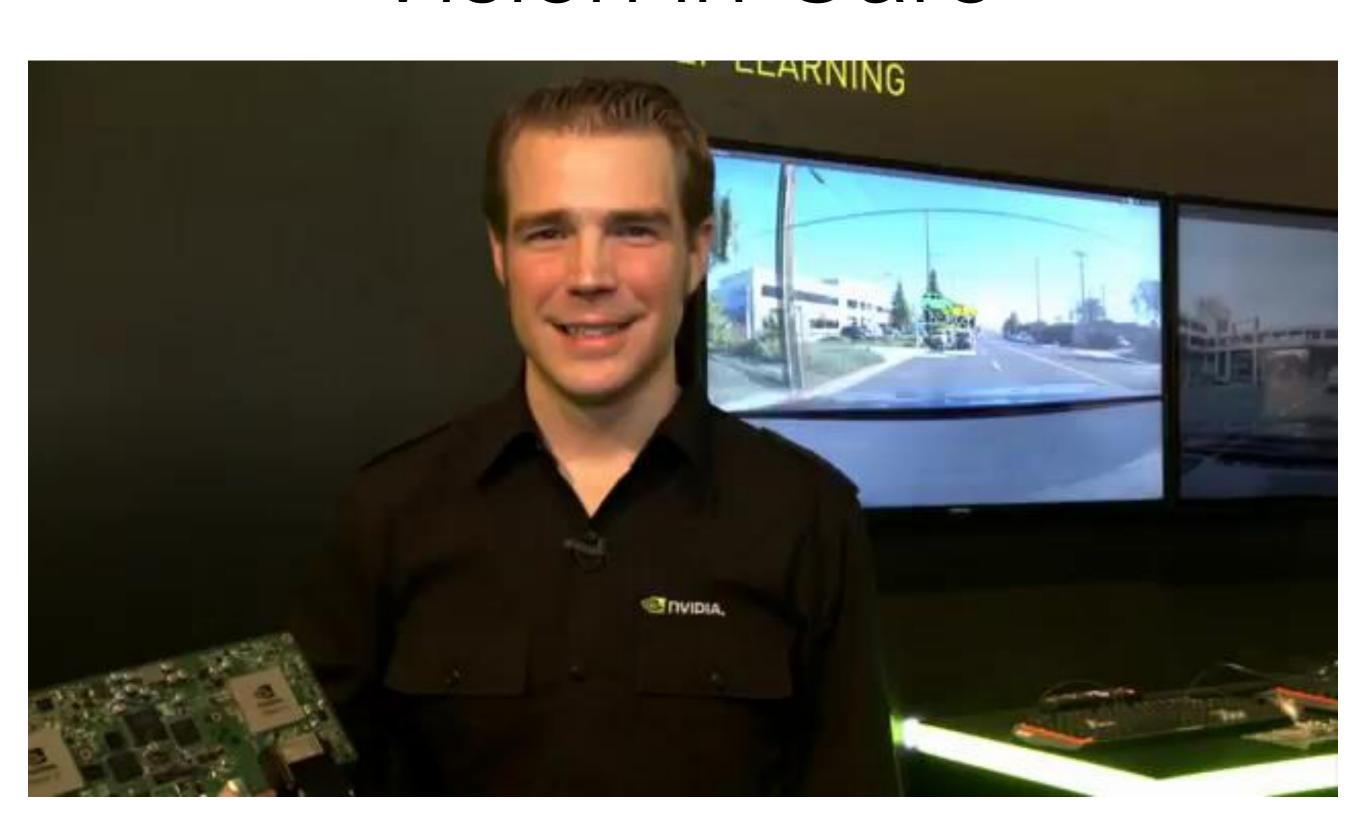
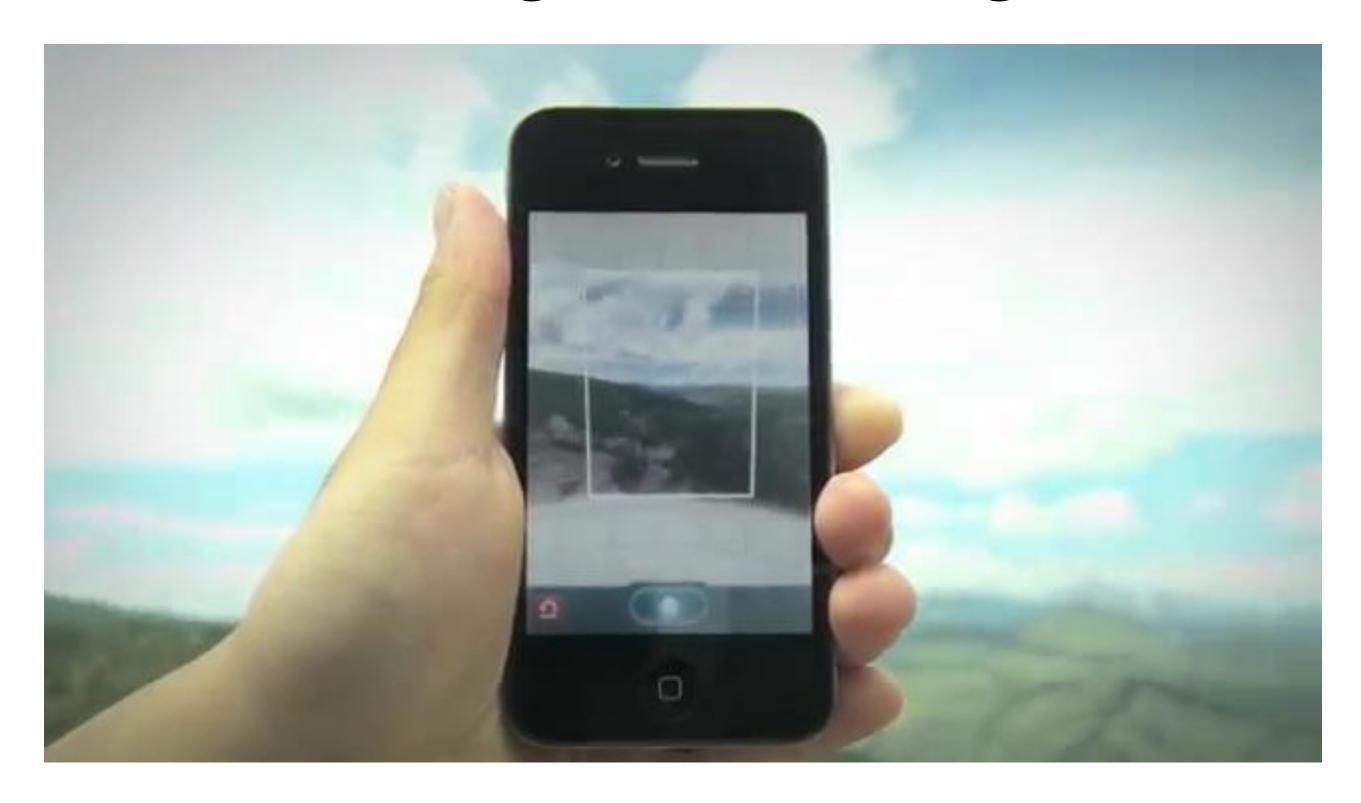
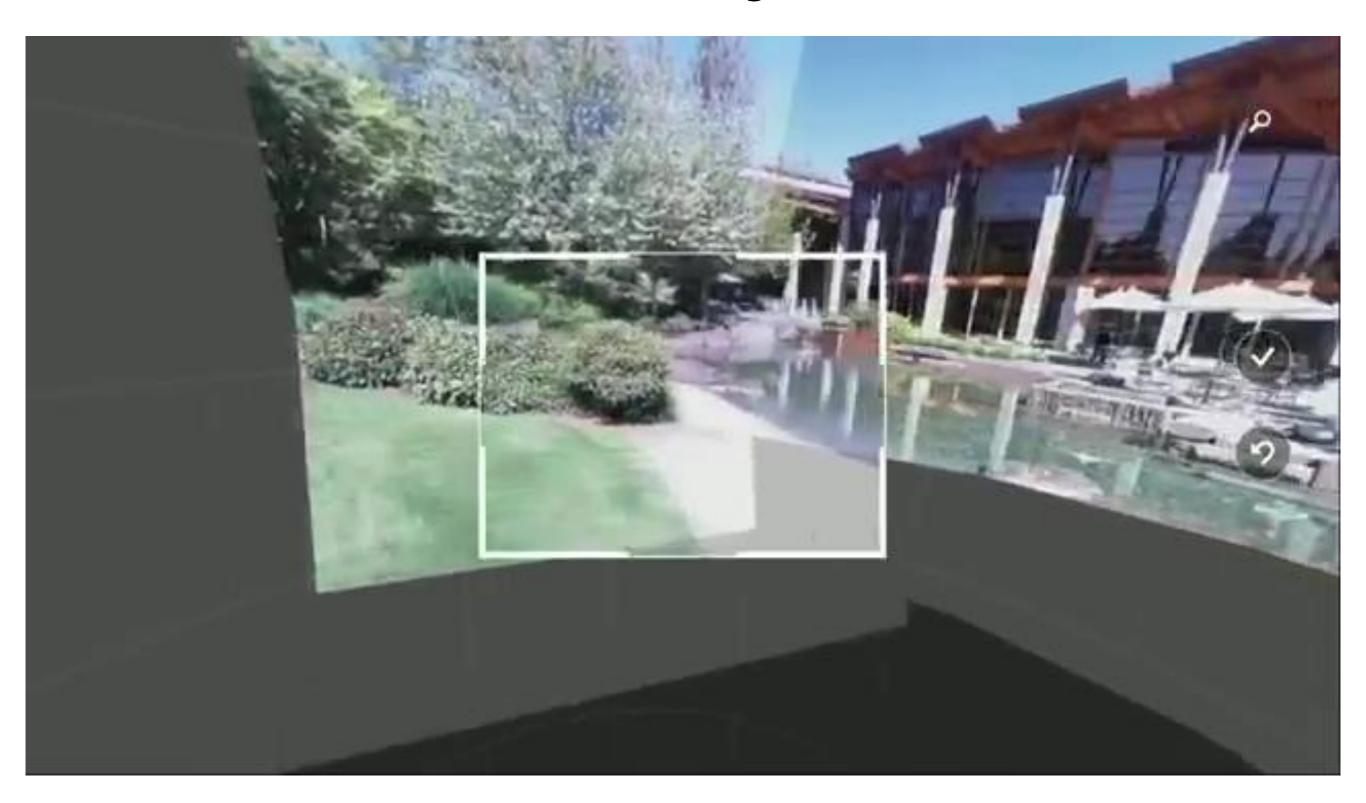


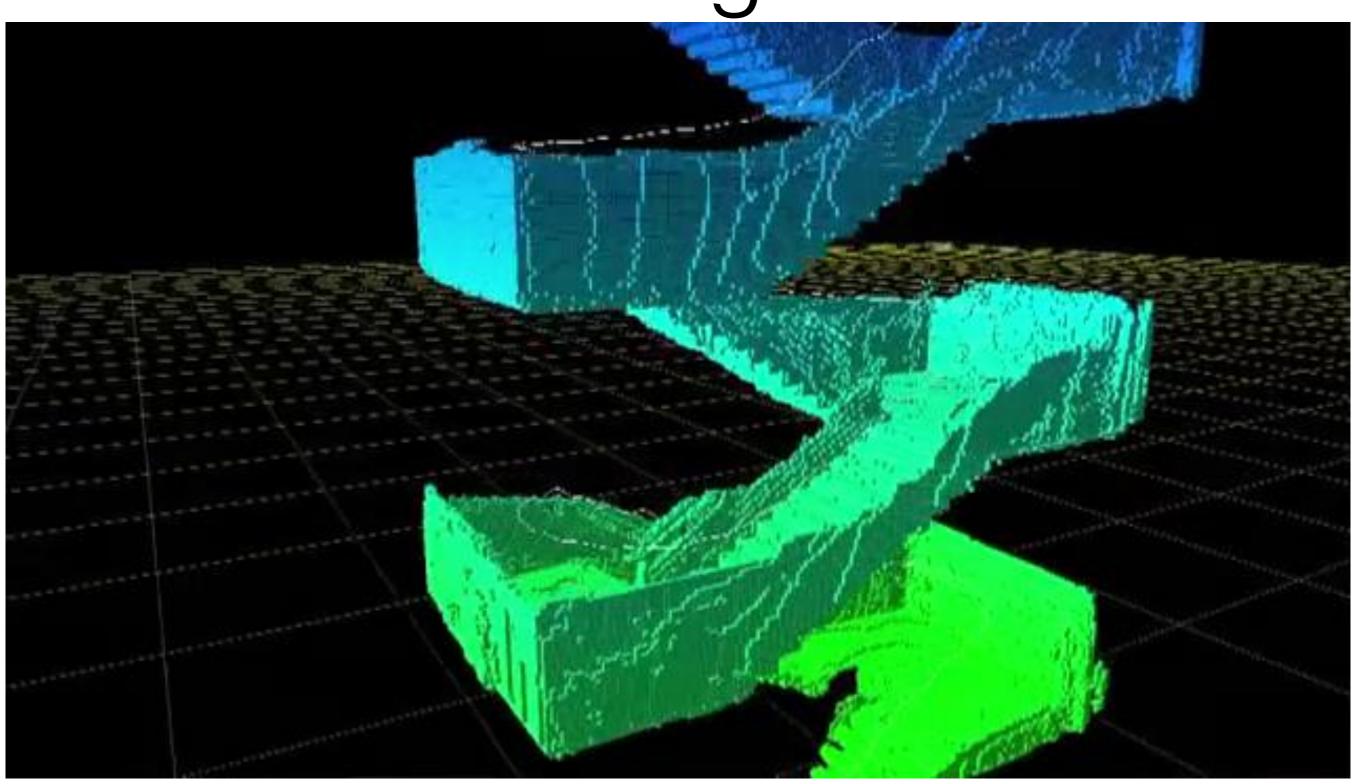
Image stitching



Photosynth



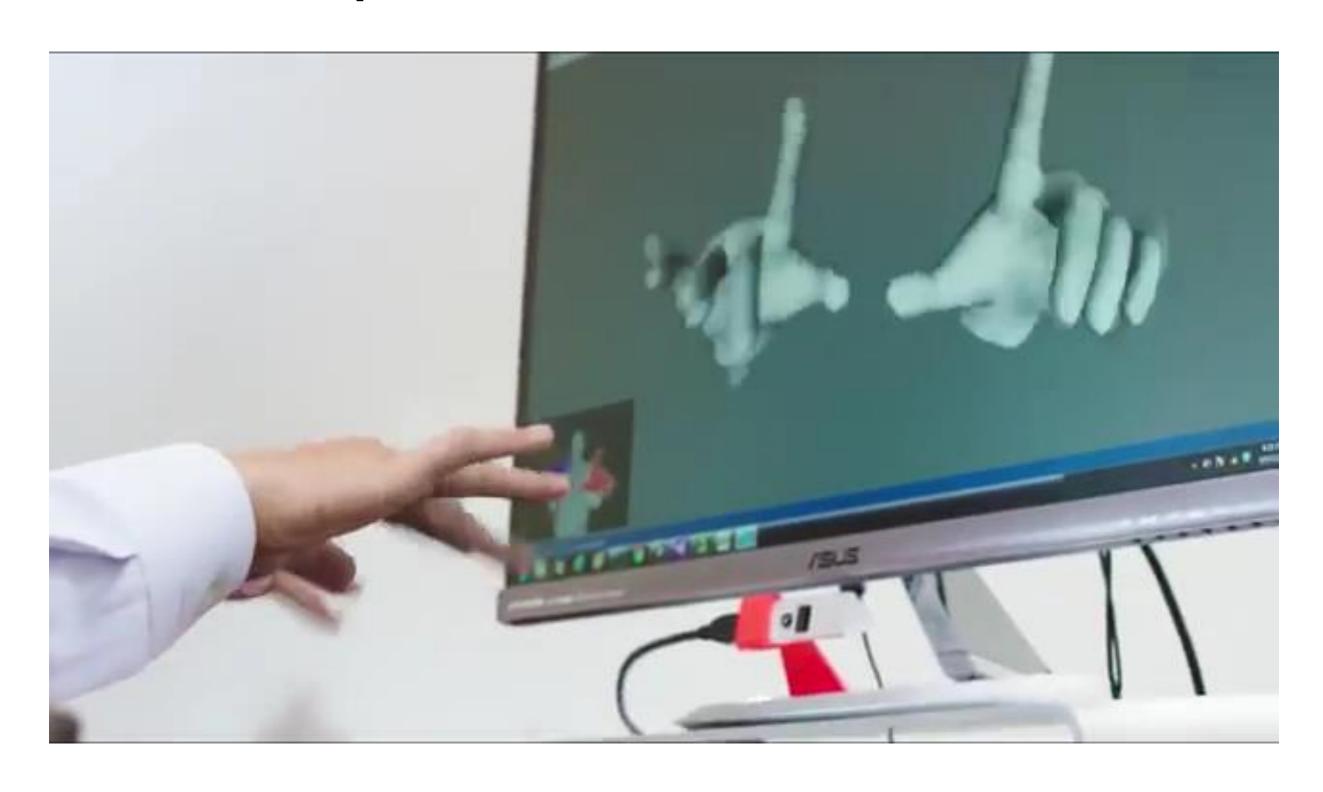
Tango



Virtual Fitting



Computer Vision for VR



Deep Face



Deep Dream





Facebook video style transfer 2016

Face2Face: Real-time Face Capture and Reenactment of RGB Videos

Justus Thies¹, Michael Zollhöfer², Marc Stamminger¹, Christian Theobalt², Matthias Nießner³

> ¹University of Erlangen-Nuremberg ²Max-Planck-Institute for Informatics ³Stanford University

> > CVPR 2016 (Oral)

It's a good time to do computer vision

Industry aggressively hiring CV faculty from universities







CMU



Toronto



UBER







UCSD Columbia

NYU











Stanford CMU Stanford



NYU

GTech











Toronto



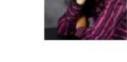
CalTech



amazon.com



UW







NYU





Zillow'





Oculus VR

CMU





CMU CMU GTech







CMU



MIT



Toronto UW









Industry aggressively hiring CV graduates, or even students!

(strong dominant industrial presence at conferences for recruitment)

facebook research Research Areas Publications People Programs Downloads Careers Blog Control Advancing computer vision technologies at CVPR 2017

By: Facebook Research

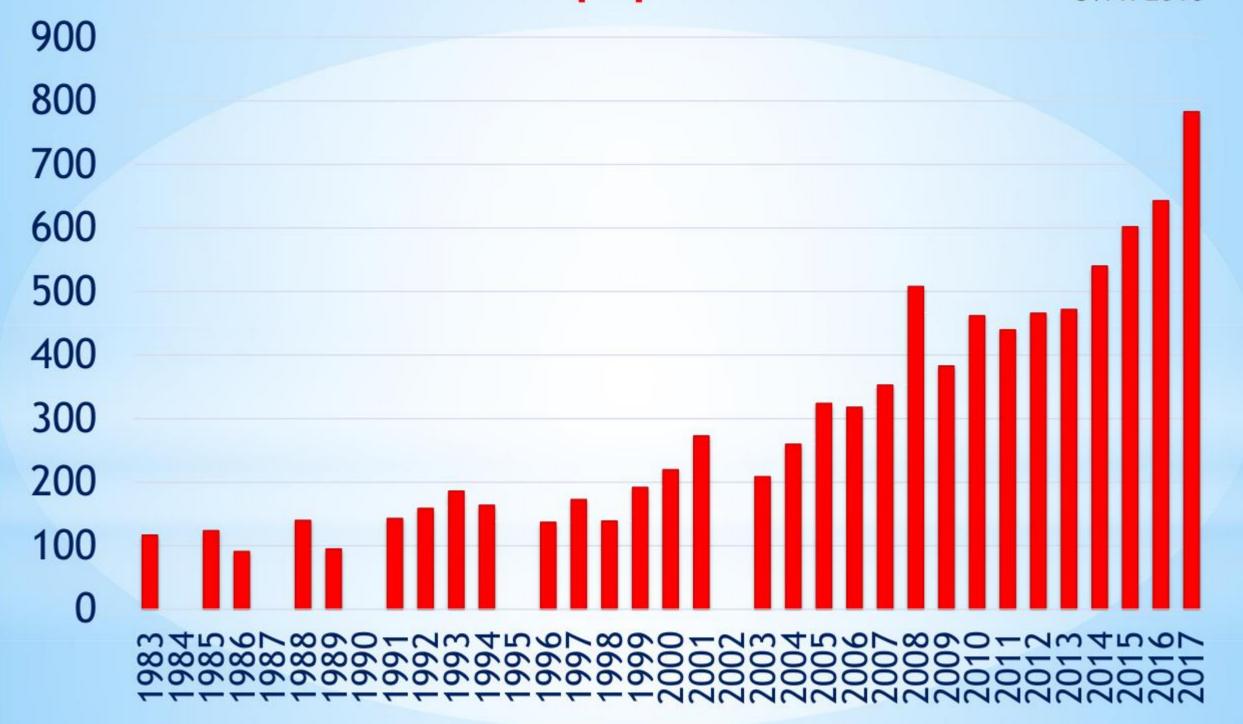




CVPR GROWTH

Number of papers at CVPR

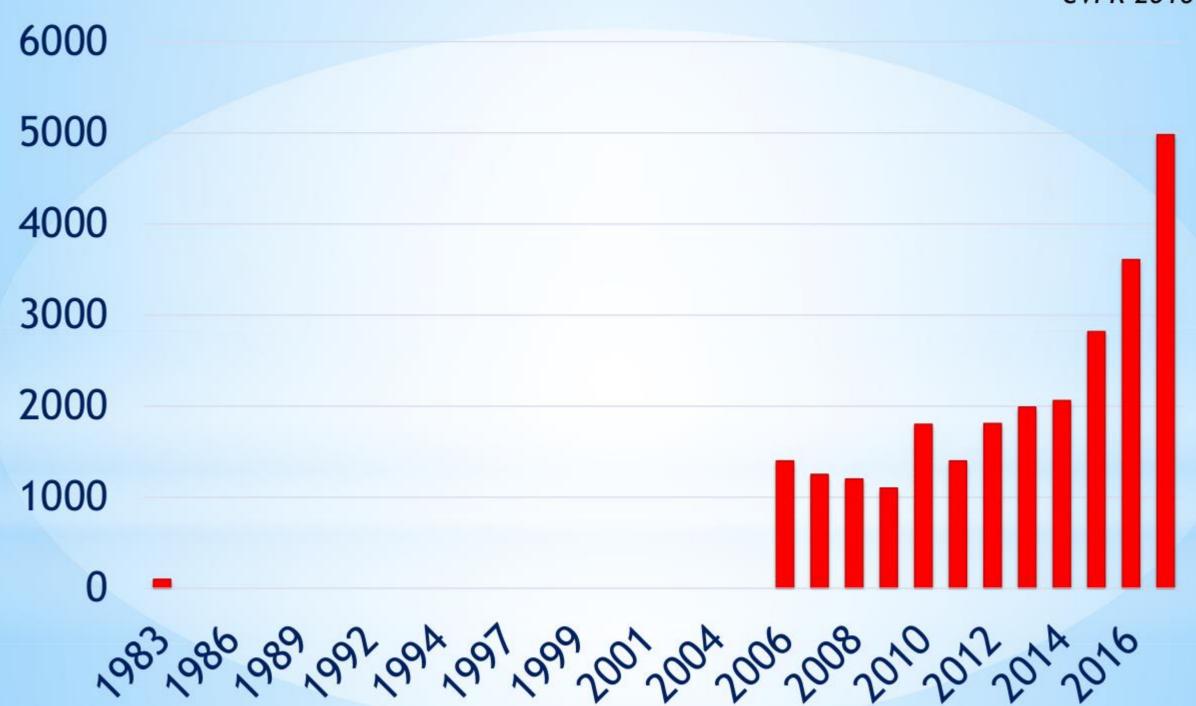
Original slide courtesy of CVPR 2016

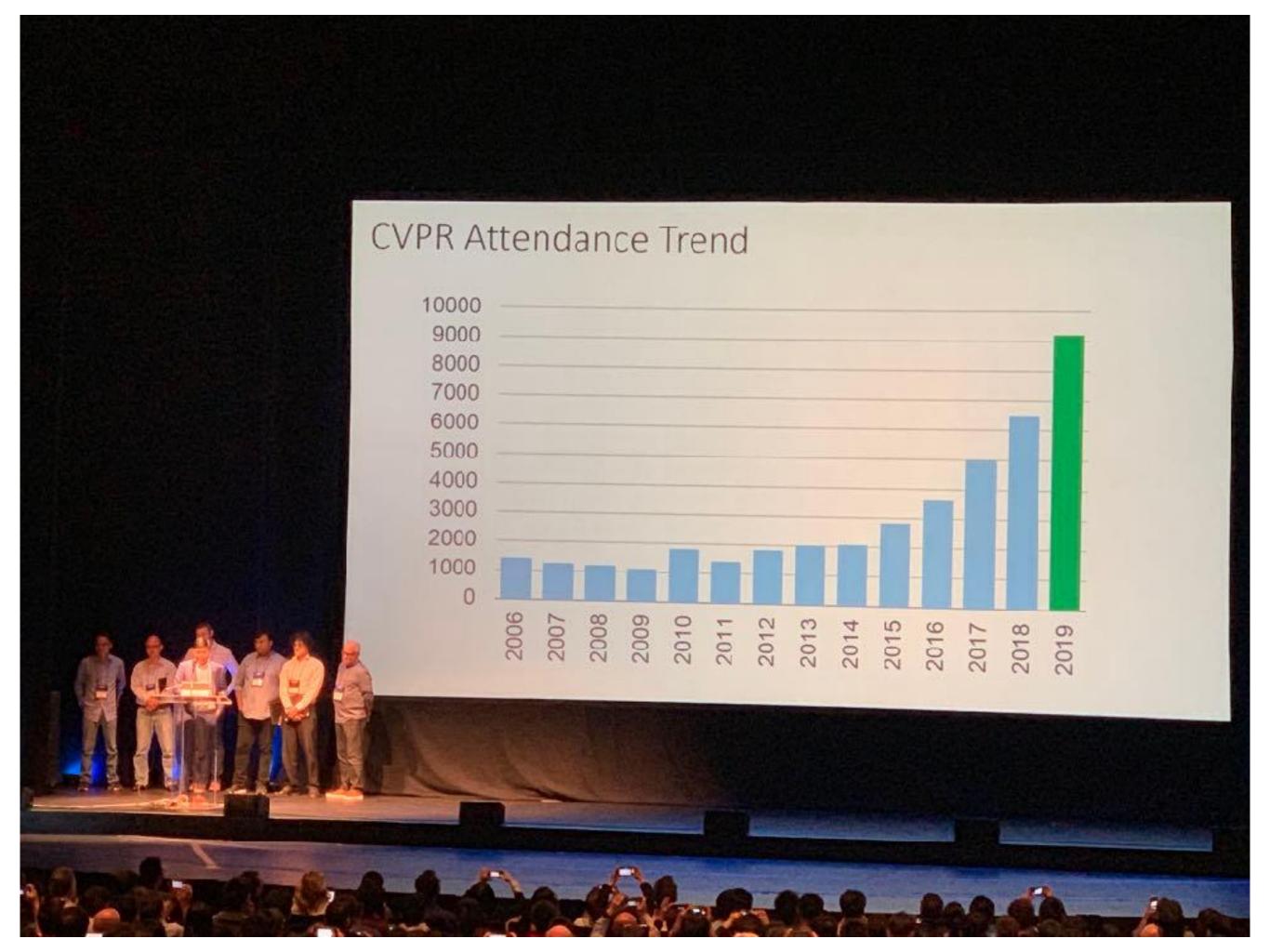


CVPR GROWTH

Number of attendees at CVPR

Original slide courtesy of CVPR 2016





Computer vision at CMU

Dedicated courses for each subject we cover in this class:

- Physics-based Methods in Vision
- Geometry-based Methods in Computer Vision
- Computational Photography
- Visual Learning and Recognition
- Statistical Techniques in Robotics
- Sensors and sensing
- ... plus an entire department's worth of ML courses.

CVPR 2019: CMU was the second most common <u>academic</u> affiliation among authors (can you guess the first?)

Master in Computer Vision at CMU



Carnegie Mellon THE ROBOTICS INSTITUTE

Master of Science - Computer Vision

MSCV

August 2016 - December 2017 (16-month program)

Computer vision is the study of acquiring and interpreting visual imagery. As computer vision shifts from research to development, there is a critical need for developers with expertise in this field.

GOALS

Offer a comprehensive set of courses
 Facilitate hands-on research and development projects
 Expose students to current and emerging
 state-of-the-art Computer Vision applications
 Prepare students for careers in Computer Vision

COURSES

Intoduction to Computer Vision Introduction to Machine Learning Mathematical Fundamentals for Robotics Visual Learning and Recognition Geometry-based Methods in Computer Vision

Electives (choose 2)

Human Communication and Multimodal Machine Learning
The Visual World as seen by Neurons and Machines
Comprehensive Sensing and Sparse Optimization
Large Scale Learning using Images and Text
Big Data approaches in Computer Vision
Human Motion Modeling and Analysis
Statistical Techniques in Robotics
Physics-based Methods in Vision
Probabilistic Graphical Models
Statistical Machine Learning
Convex Optimization
Vision Sensors

Project and Seminar Courses

MSCV Seminar MSCV Project I MSCV Project II

ADMISSION AND APPLICATION

Requirements: Undergraduate (B.S. or equivalent) in engineering, computer science or applied mathematics

Application Materials

- · Résumé · General GRE
- · TOEFL / IELTS (Foreign Students only)
- Statement of Purpose (1 to 2 pages)
- Letters of Recommendation (3 Required)
- Undergraduate/Graduate (as applicable) Transcripts

Only online applications will be accepted. Early application deadline: December 3, 2015 Final application deadline: December 15, 2015

FOR INDUSTRY SPONSORSHIPS PLEASE CONTACT JULIE GOLDSTEIN (JGOLDS@CS.CMU.EDU), 412-268-4017

Carnegie Mellon University 5000 Forbes Avenue, Pittsburgh, PA 15232 ms-cv@ri.cmu.edu

www.ri.cmu.edu/MSCV





Srinivasa Narasimhan MSCV Program Directi



Martial Hebert MSCV Spiritual Guru



J. Andrew (Drew Bagnell



Fernando De la Torre Frade



Abhina



Kris M. Kitani



Simon



Deva Kannan Ramanan



Yaser Ajma

Course logistics

Website



http://www.cs.cmu.edu/~16385/

(includes links to Canvas and Piazza)

Assignments Canvas

https://canvas.cmu.edu/courses/14118

Discussion¬es plazza

https://piazza.com/class/k53x5h48my264d

(you should sign up here on your own)

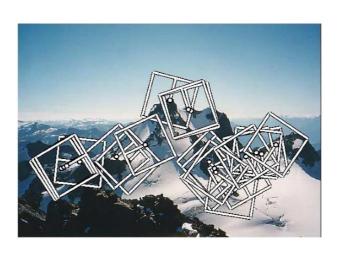
Image processing:

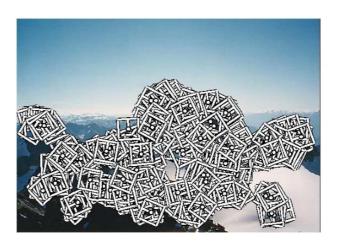
- Basics of filtering.
- Image pyramids.
- Gradients and lines.
- Hough transforms.



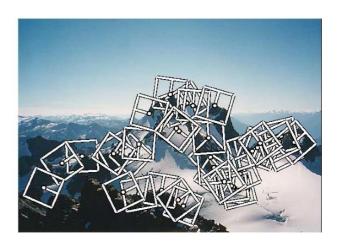
Feature detection and correspondences:

- Corner detection.
- SIFT et al.
- Feature descriptors.
- RANSAC.



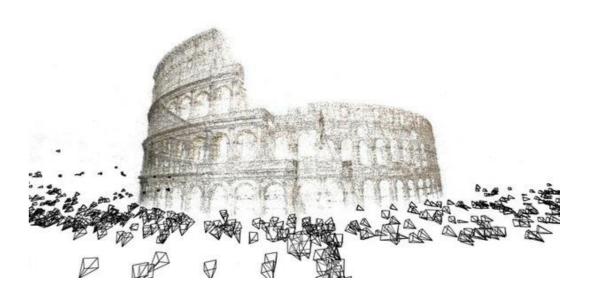






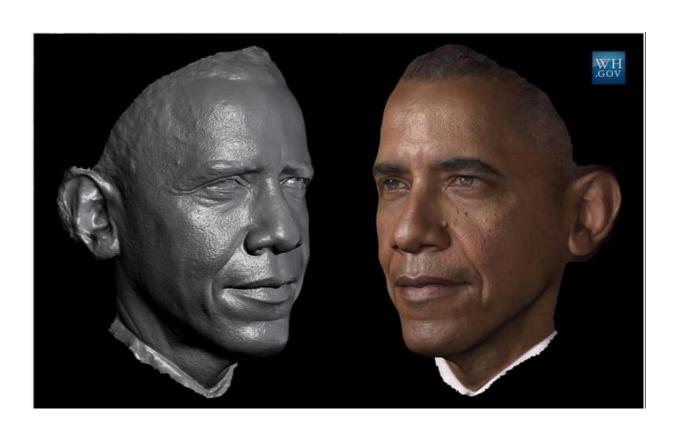
Transformations and geometry:

- Homographies and image alignment.
- Camera models.
- Fundamental matrix.
- Epipolar geometry and stereo.
- Structure from motion.



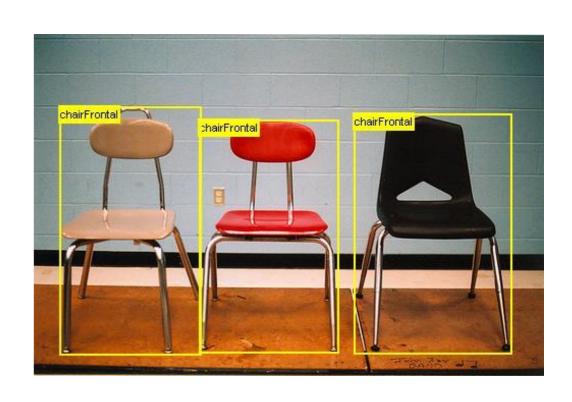
Physics-based vision:

- Reflectance and image formation.
- Radiometry.
- Shape from shading.
- Photometric stereo.
- Color.



Objects, faces, and learning:

- Basics of probability.
- K-means, KNN, PCA, SVM.
- Bag of words.
- Viola-Jones face detection.
- Perceptron, backpropagation.
- Convolutional neural networks.



Dealing with motion:

- Optical flow (LK, HS).
- Image registration.
- Kalman Filtering.
- Tracking (KLT, Mean-Shift).



Special topics:

- Computational photography.
- ???



Grading

- Seven two-week programming assignments: 70%
- Twelve weekly take-home quizzes: 27%
- Class and Piazza participation: 3%

Take-home quizzes:

- New this year.
- Two-three theory questions.
- Replace mid-term and final.

Participation:

- Be around for lectures.
- Post on Piazza discussions.
- Ask and answer questions.

Programming assignments

Assignment 1 Hough Transform

Assignment 2 Homography

Assignment 3 Stereo

Assignment 4 Photometric Stereo

Assignment 5 Bag of Words

Assignment 6 Convolutional Neural Nets

Assignment 7 Lucas-Kanade Tracking

- a lot of programming in Matlab and Python.
- hours and hours of programming.
- days and days of debugging.
- generous grading policy (like grad school)
- take advantage of extra credit

Programming assignments

Assignment 1 Hough Transform

Assignment 2 Homography

Assignment 3 Stereo

Assignment 4 Photometric Stereo

Assignment 5 Bag of Words

Assignment 6 Convolutional Neural Nets

Assignment 7 Lucas-Kanade Tracking

seriously, a lot of

programming, so start early!

- a lot of programming in Matlab and Python.
- hours and hours of programming.
- days and days of debugging.
- generous grading policy (like grad school)
- take advantage of extra credit

Schedule

- <u>Tentative</u> schedule on course website.
- Likely to change.
- Always check course website and Piazza for updates!

Date	Topics	Slides	Assignments
M, Jan 13	Introduction		
W, Jan 15	Image filtering		
M, Jan 20	No class (Martin Luther King day)		
W, Jan 22	Image pyramids and Fourier transform		PA1 out
M, Jan 27	Hough transform		TQ1 out
N, Jan 28	Feature and corner detection		
M, Feb 3	Feature descriptors and matching		TQ1 due, TQ2 out
W, Feb 5	2D transformations		PA1 due, PA2 out
M, Feb 10	Image homographies		TQ2 due, TQ3 out
W, Feb 12	Camera models		
M, Feb 17	Two-view geometry		TQ3 due, TQ4 out
W, Feb 19	Stereo		PA2 due, PA3 out
M, Feb 24	Structure from motion		TQ4 due, TQ5 out
W, Feb 26	Radiometry and reflectance		
M, Mar 2	Photometric stereo and shape from shading		TQ5 due, TQ6 out
W, Mar 4	Color		PA3 due, PA4 out
M, Mar 9	No class (spring break)		
W, Mar 11	No class (spring break)		
M, Mar 16	Image processing pipeline		TQ6 due, TQ7 out
W, Mar 18	Introduction to recognition		PA4 due, PA5 out
M, Mar 23	Bag of works		TQ7 due, TQ8 out
W, Mar 25	Neural networks		
M, Mar 30	Convolutional neural networks		TQ8 due, TQ9 out
W, Apr 1	Optimization		PA5 due, PA6 out
M, Apr 6	Faces		TQ9 due, TQ10 out
W, Apr 8	Optical flow		
M, Apr 13	Alignment		TQ10 due, TQ11 out
N, Apr 15	Tracking		PA6 due, PA7 out
M, Apr 20	Temporal models and SLAM		TQ11 due, TQ12 out
W, Apr 22	Graph-based methods		
M, Apr 27	Segmentation		TQ12 due
W, Apr 29	Wrap-up and discussion		PA7 due

Leniency

Late days for programming assignments:

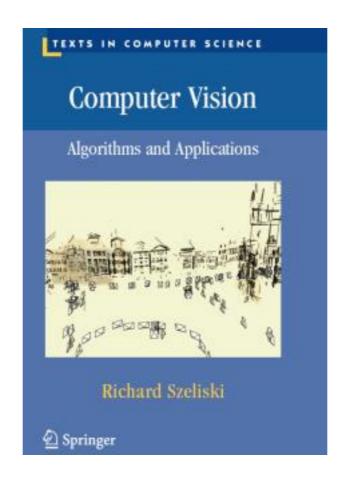
- 10% reduction of points per late day
- 6 free late days total
- use them wisely... save for later (harder) assignments!

Option to skip take-home quizzes:

- you only need to submit 9 out of 12 quizzes
- late quizzes will not be graded

Book

We will be posting readings after each lecture



PDF online

http://szeliski.org/Book/

Prerequisites

We assume familiarity with calculus, linear algebra, basic probability, and programming.

Formal prerequisites:

 "Mathematical Foundations of Electrical Engineering" (18-202) and "Principles of Imperative Computation" (15-122)

OR

 "Matrix Algebra with Applications" (21-240) and "Matrices and Linear Transformations" (21-241) and "Calculus in Three Dimensions" (21-259) and "Principles of Imperative Computation" (15-122)

If you are missing a prerequisite but still want to enroll, let me know and we'll discuss it.

Contact information and office hours

- Feel free to email us about administrative questions.
 - o please use [16385] in email title!
- Technical questions should be asked on Piazza.
 - we won't answer technical questions through email.
 - you can post anonymously if you prefer.
- Office hours will be determined by poll.
 - feel free to email Yannis about additional office hours.
 - you can also just drop by Yannis' office (Smith Hall (EDSH) Rm 225).

Yannis will announce office hours for this week.

Please take the course survey before the next lecture!



(also posted on Piazza)