

CSE 252D: Advanced Computer Vision

Manmohan Chandraker

Lecture 0: Introduction



Virtual classrooms

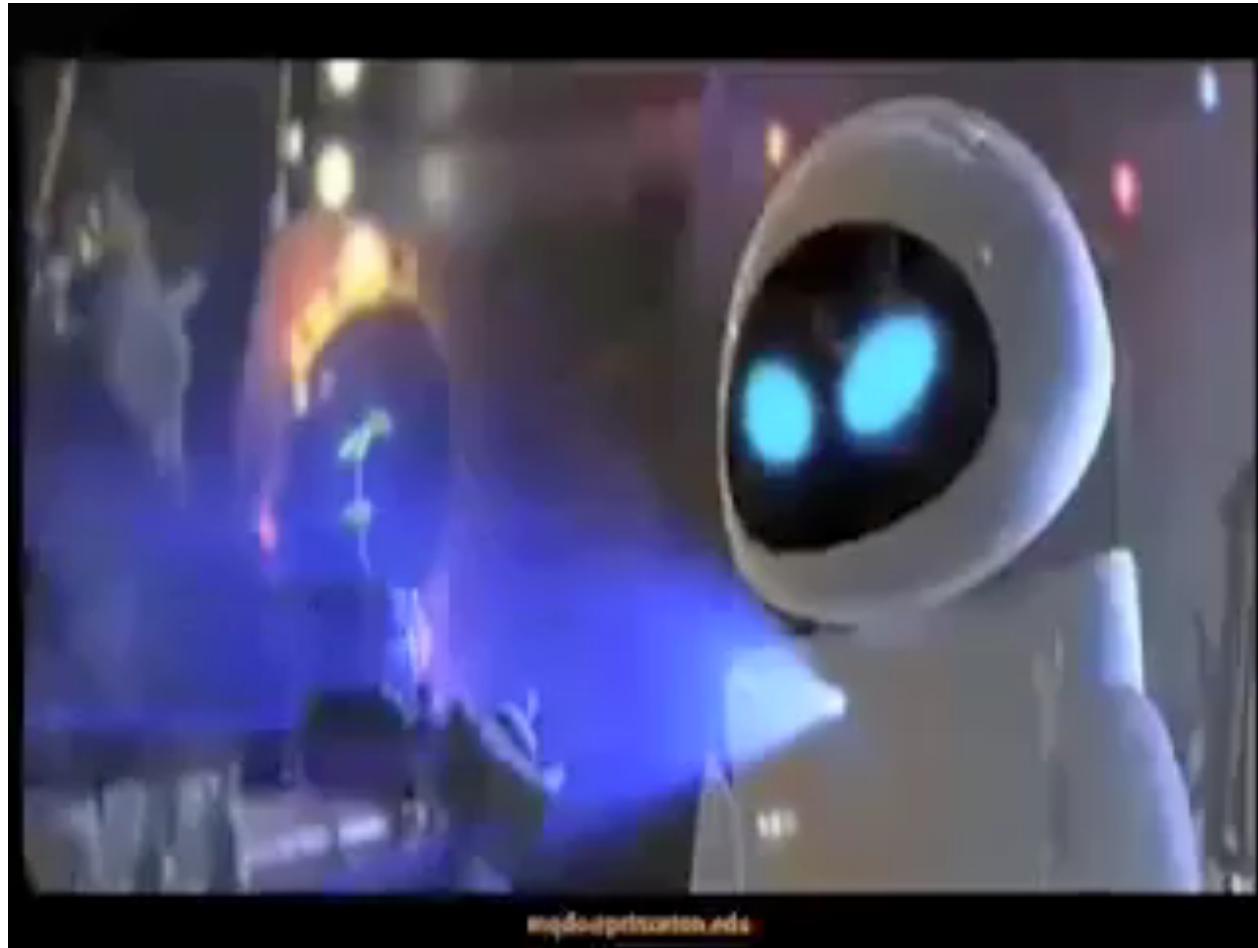
- Virtual lectures on Zoom
 - Only host shares the screen
 - Keep video off and microphone muted
 - But please do speak up (remember to unmute!)
 - Slides uploaded on webpage just before class
- Virtual interactions on Zoom
 - Ask and answer plenty of questions
 - “Raise hand” feature on Zoom when you wish to speak
 - Post questions on chat window
 - Happy to try other suggestions!
- Lectures recorded and upload on Kaltura
 - Available under “My Media” on Canvas

Enrollment logistics

- To enroll if you are on the waitlist
 - Send “**Request to enroll**” email to instructor if on waitlist
 - Include CV, courses, project experience relevant to computer vision
- While on the waitlist
 - You are welcome to attend lectures even if on waitlist
 - To limit TA workload, we can grade only enrolled students
 - Most should be able to enroll eventually
- Canvas
 - All enrolled and waitlisted students should have access
- All announcements will be posted on Piazza
 - Send email to TA (CC instructor) if cannot access Piazza

Computer Vision

Defining computer vision



Wall-E: Fact and Fiction (Minh Do, Princeton University)

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Studying computer vision

- Vision is a fundamental interface to the world



Humans



Machines



Virtual
(Cyborgs!)



Studying computer vision

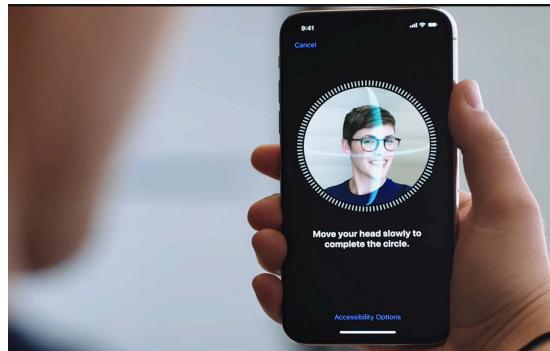
- Vision is a fundamental interface to the world
- We are all users of computer vision



Transportation



Design



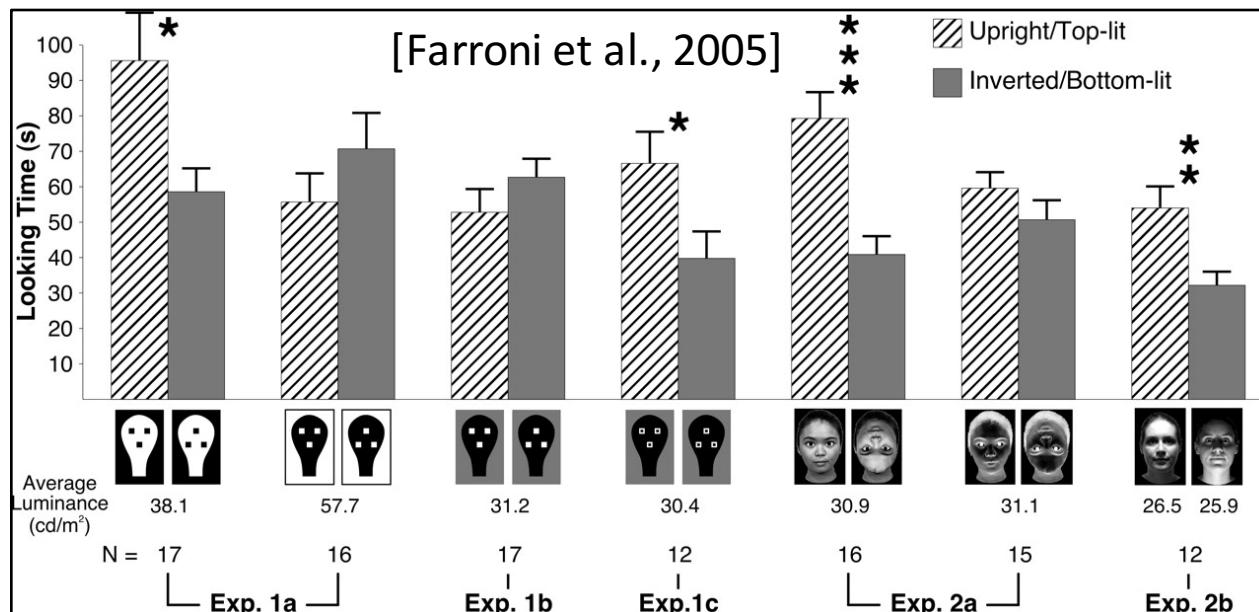
Communication



E-Commerce

Studying computer vision

- Vision is a fundamental interface to the world
- We are all users of computer vision
- Deep and attractive scientific problems
 - How do we recognize objects?
 - Why do newborn babies respond to face-like shapes?



Defining computer vision

MASSACHUSETTS INSTITUTE OF TECHNOLOGY
PROJECT MAC

Artificial Intelligence Group
Vision Memo. No. 100.

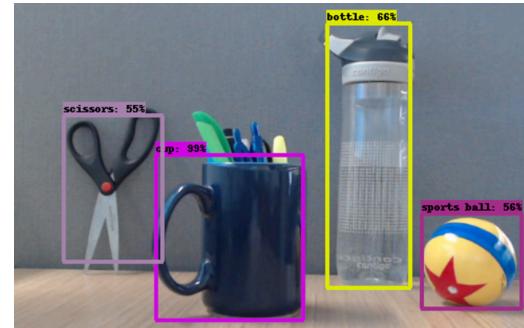
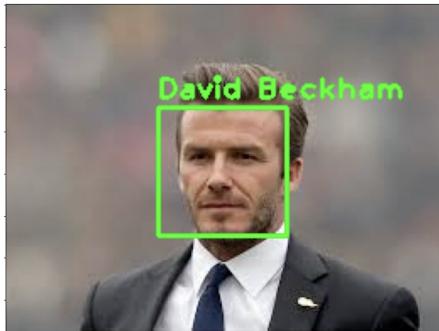
July 7, 1966

THE SUMMER VISION PROJECT

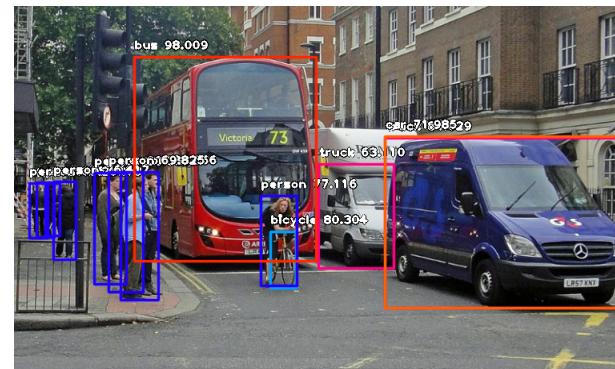
Seymour Papert

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".

Defining computer vision



Our problems became more complex



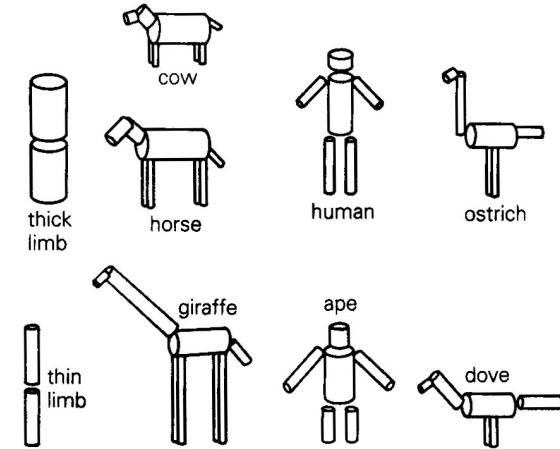
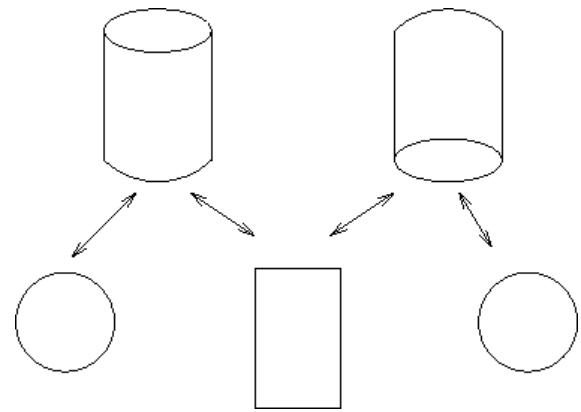
Defining computer vision



Our ambitions became higher



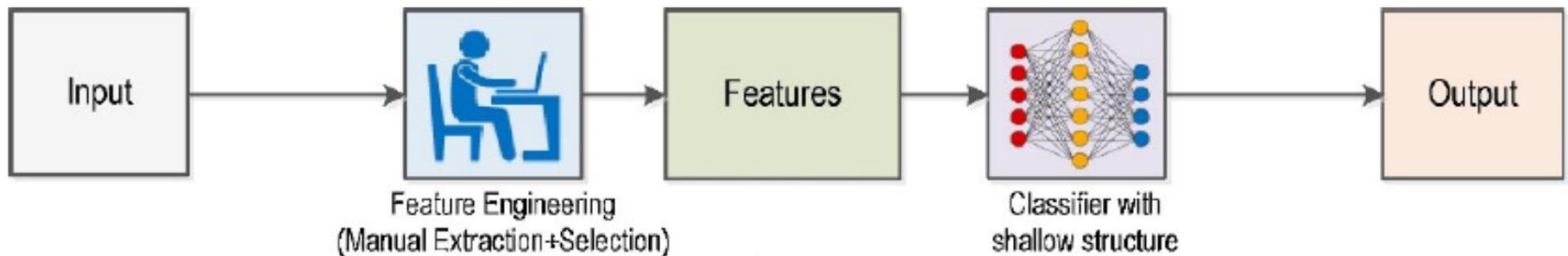
Defining computer vision



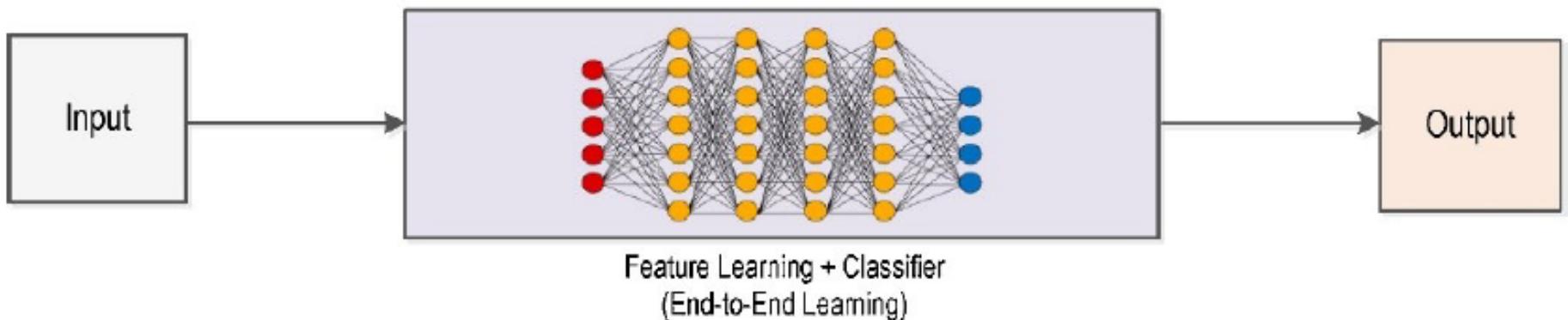
Our concepts changed over time



Defining computer vision



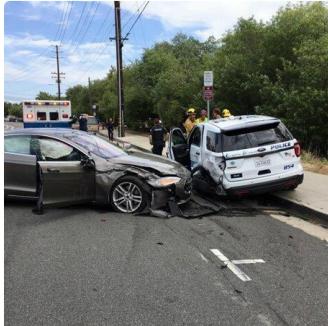
Our tools advanced over time



Gaining perspective on computer vision

 **Laguna Beach PD PIO**
@LBPD_PIO_45

This morning a Tesla sedan driving outbound Laguna Canyon Road in "autopilot" collides with a parked [@LagunaBeachPD](#) unit. Officer was not in the unit at the time of the crash and minor injuries were sustained to the Tesla driver. #lagunabeach #police #tesla



♡ 321 12:11 PM - May 29, 2018(i)

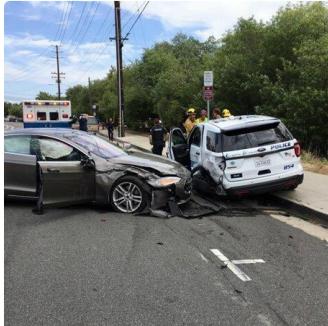
💬 466 people are talking about this>

Important for Autopilot:
Do not hit a *police* car!

Gaining perspective on computer vision

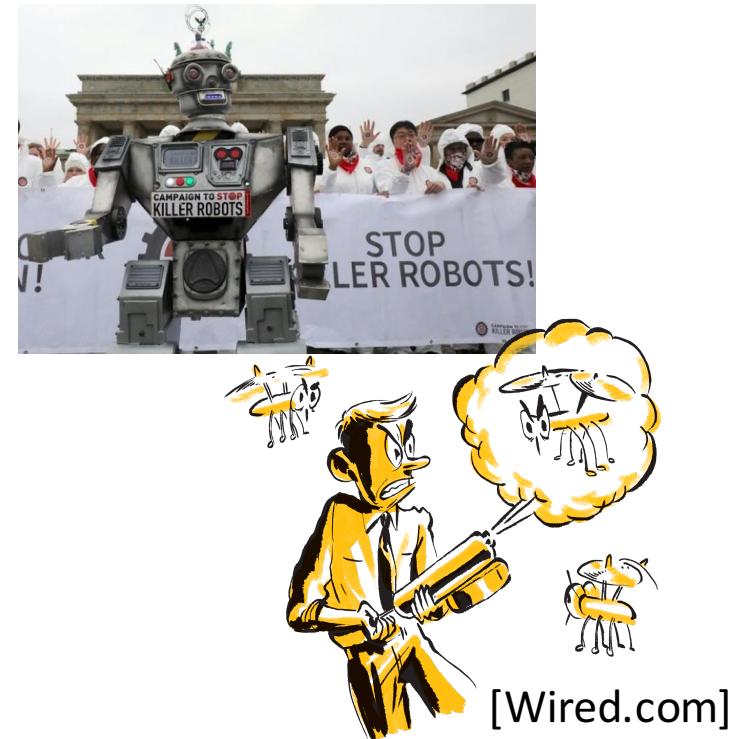
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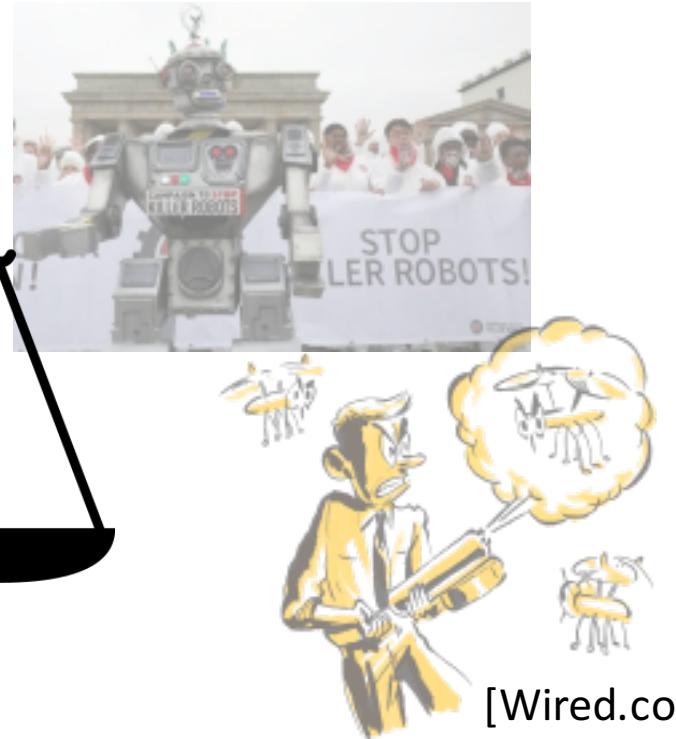
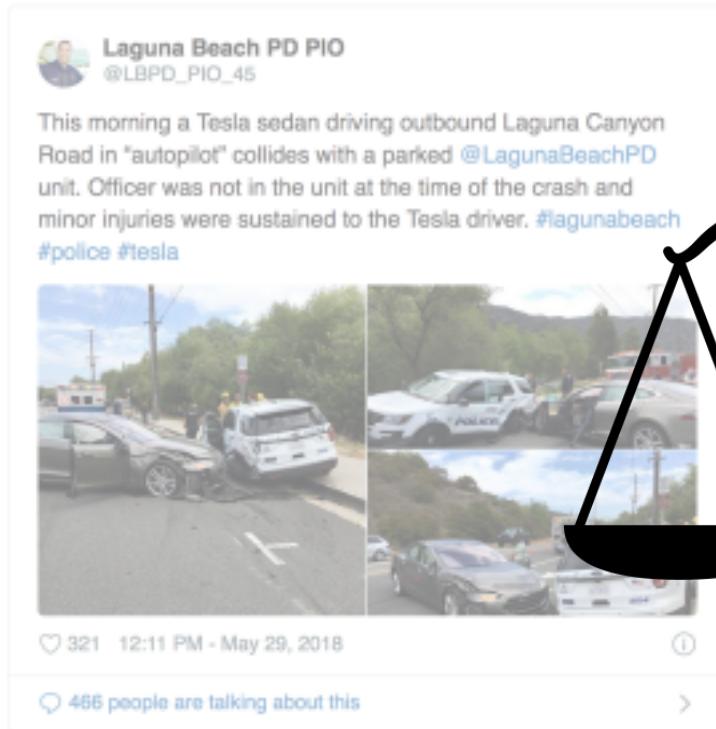
466 people are talking about this >



Important for Autopilot:
Do not hit a *police* car!

Important for Robots:
Stay clear of humans!

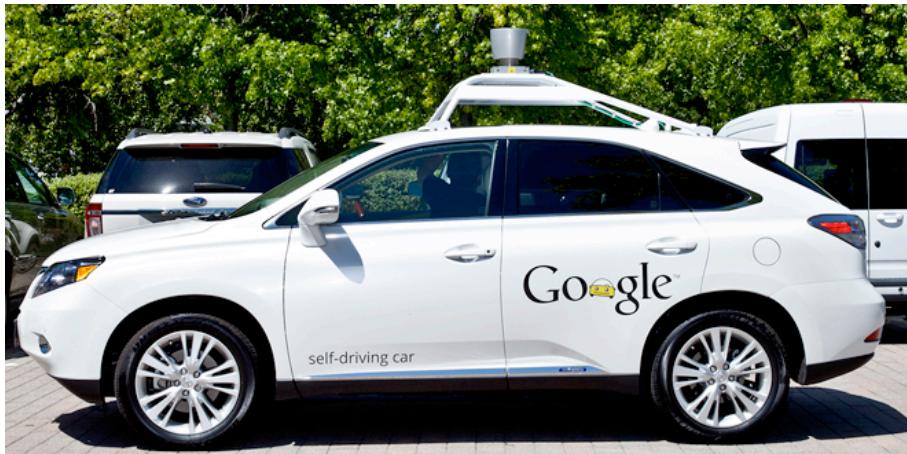
Gaining perspective on computer vision



Important for Autopilot:
Do not hit a *police* car!

Important for Robots:
Stay clear of humans!

Autonomous driving

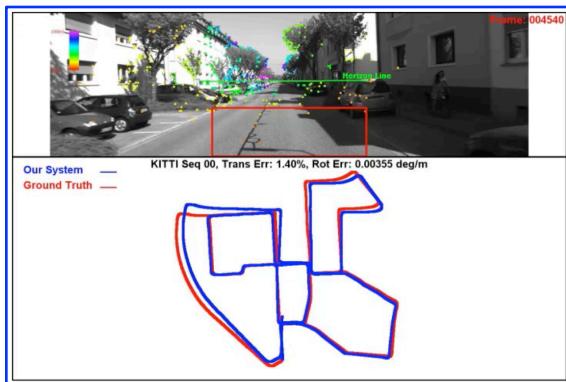


Autonomous driving

Where is our car?

Structure from Motion

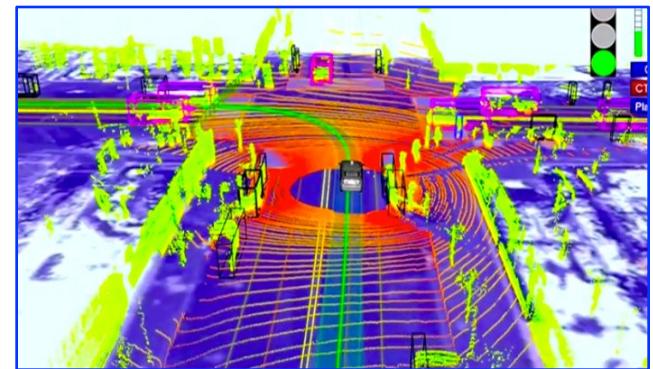
Visual SLAM



What is a safe path?

Behavior prediction

Path planning



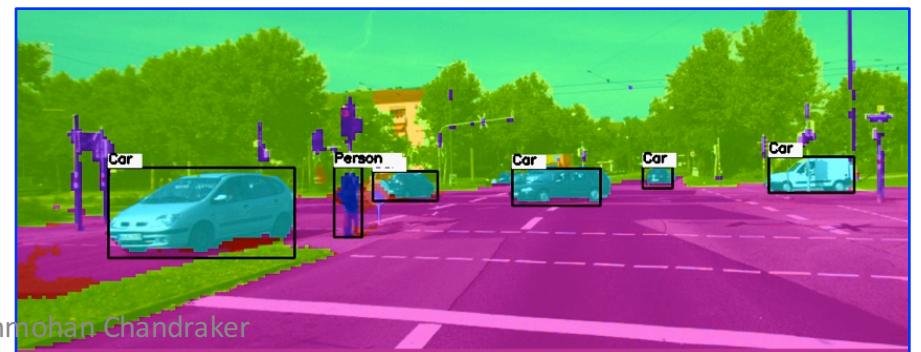
Where are other agents?

Object detection

3D localization



Where are scene elements?
Semantic segmentation



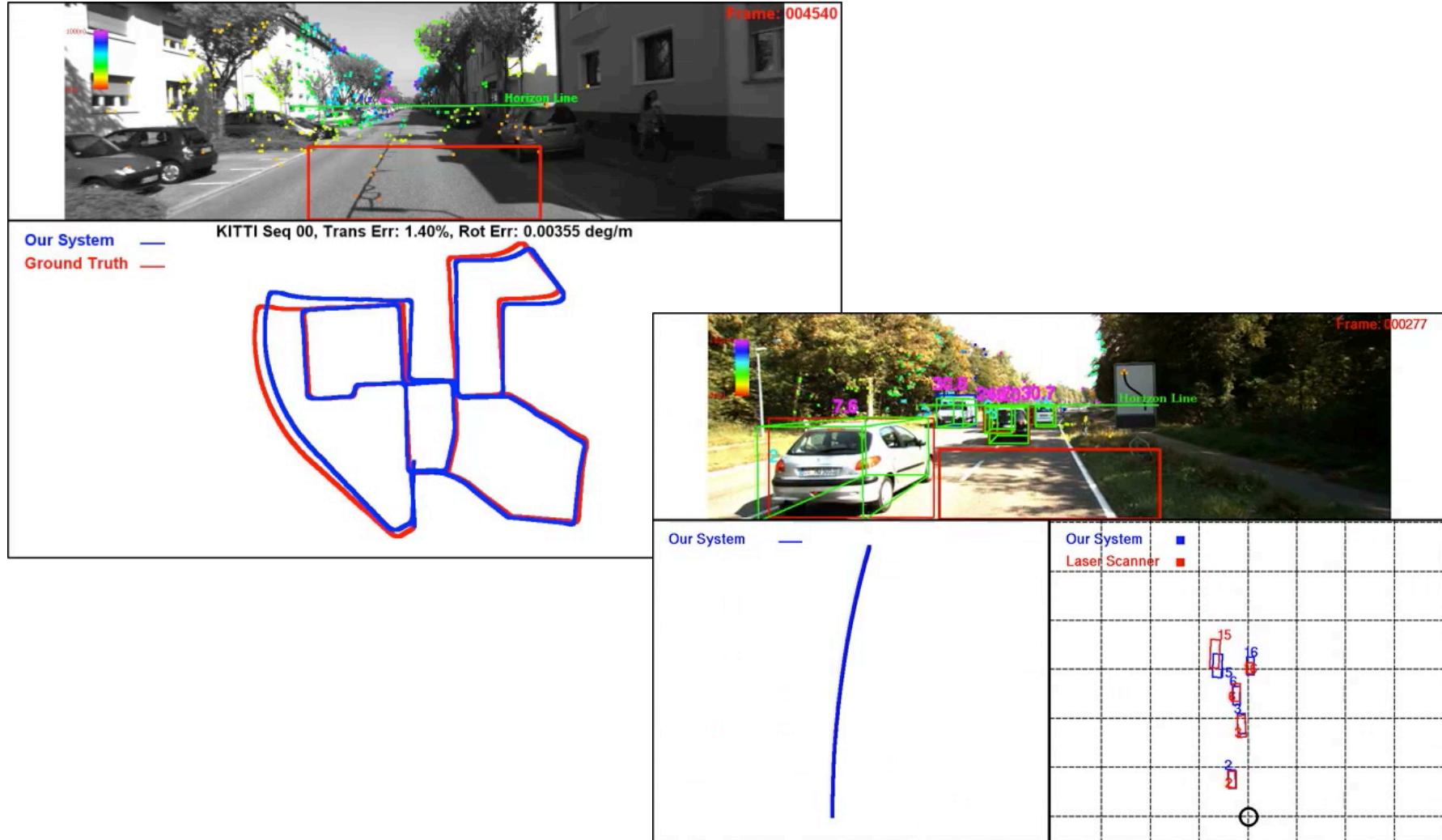
A few topics in this course

- Structure and Motion
- Faces and Humans
- Objects and Stuff
- Material and Lighting
- Bias and Adaptation

Structure and Motion



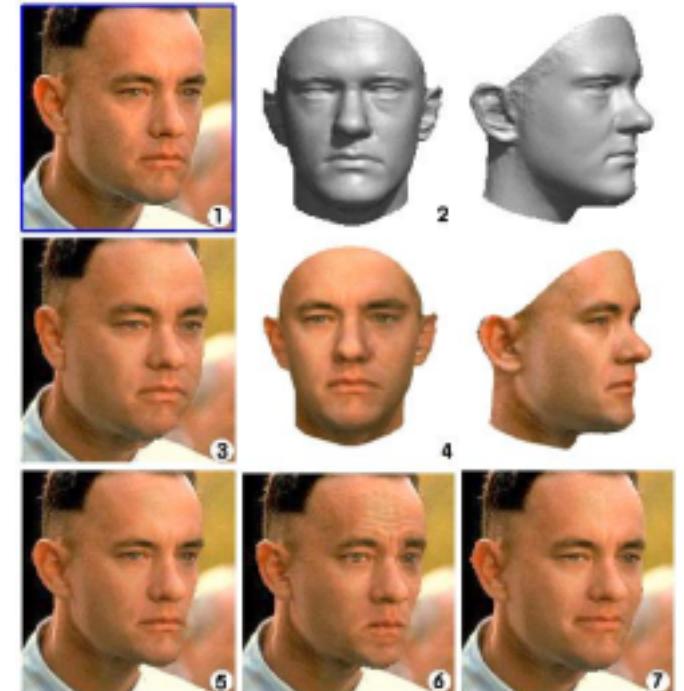
Structure and Motion



Faces and Humans

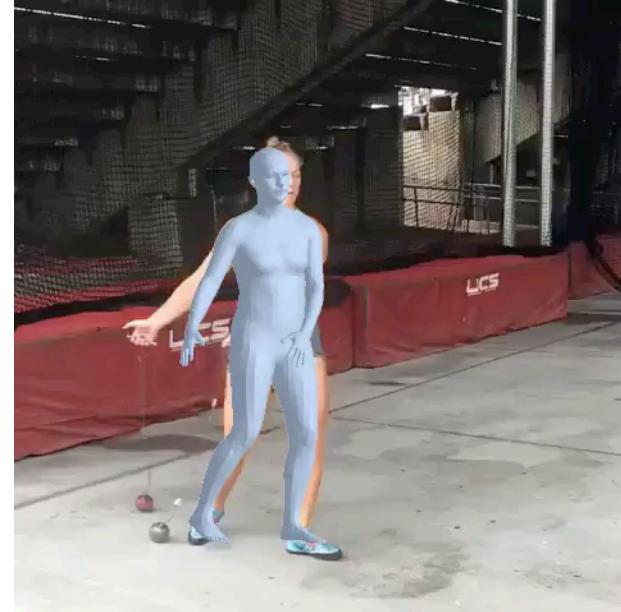


[Microsoft blog, 2018]

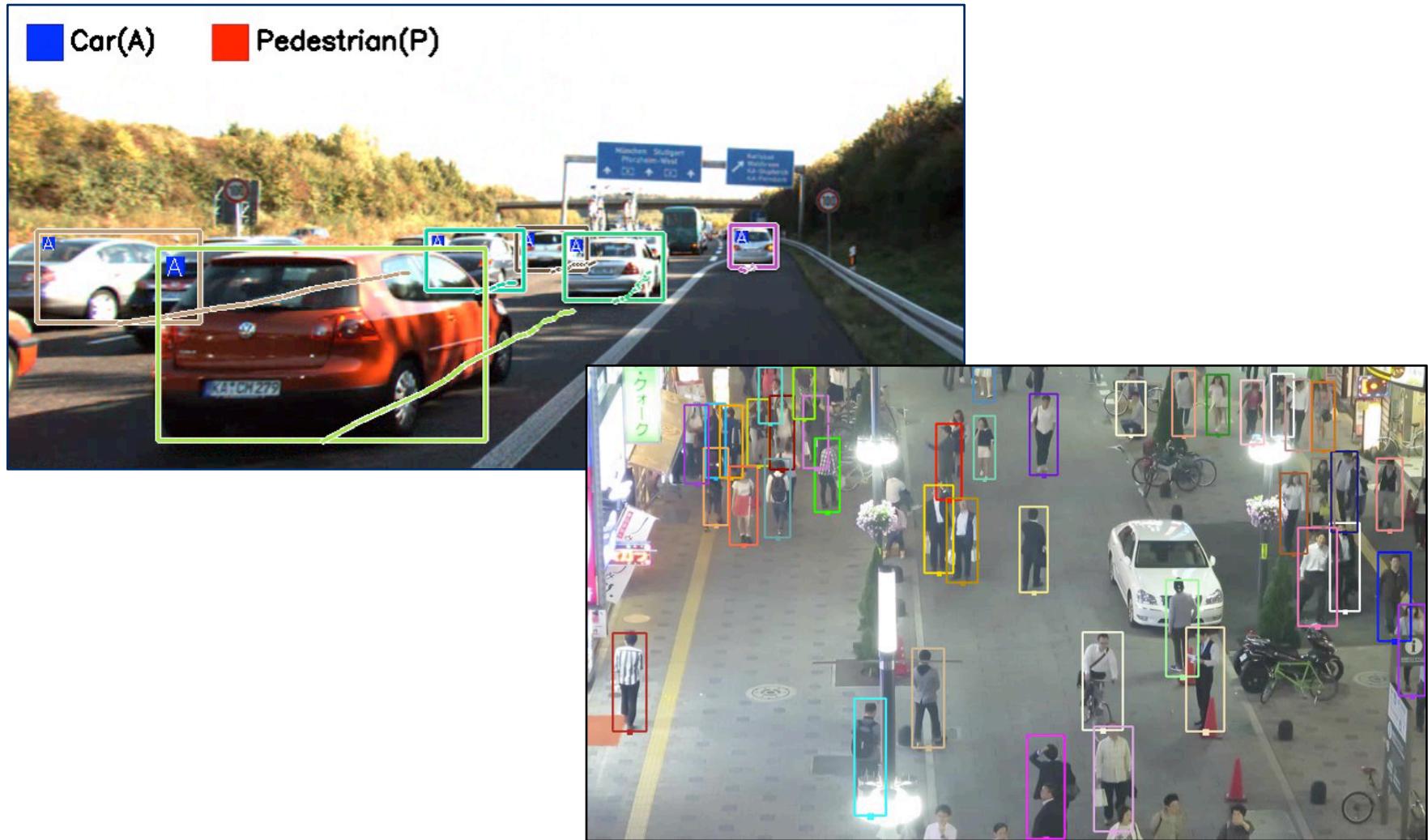


[Blanz and Vetter]

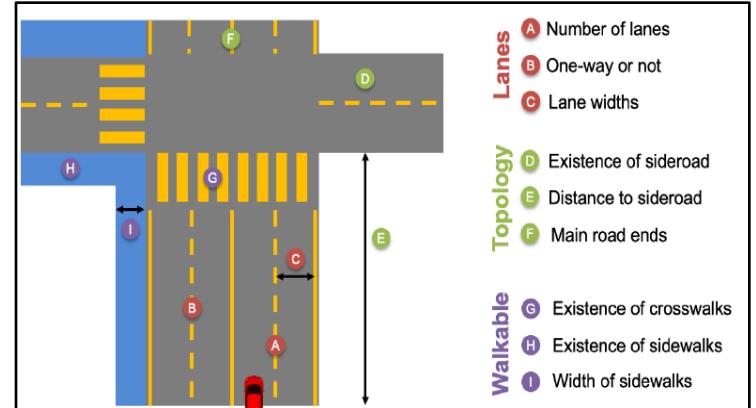
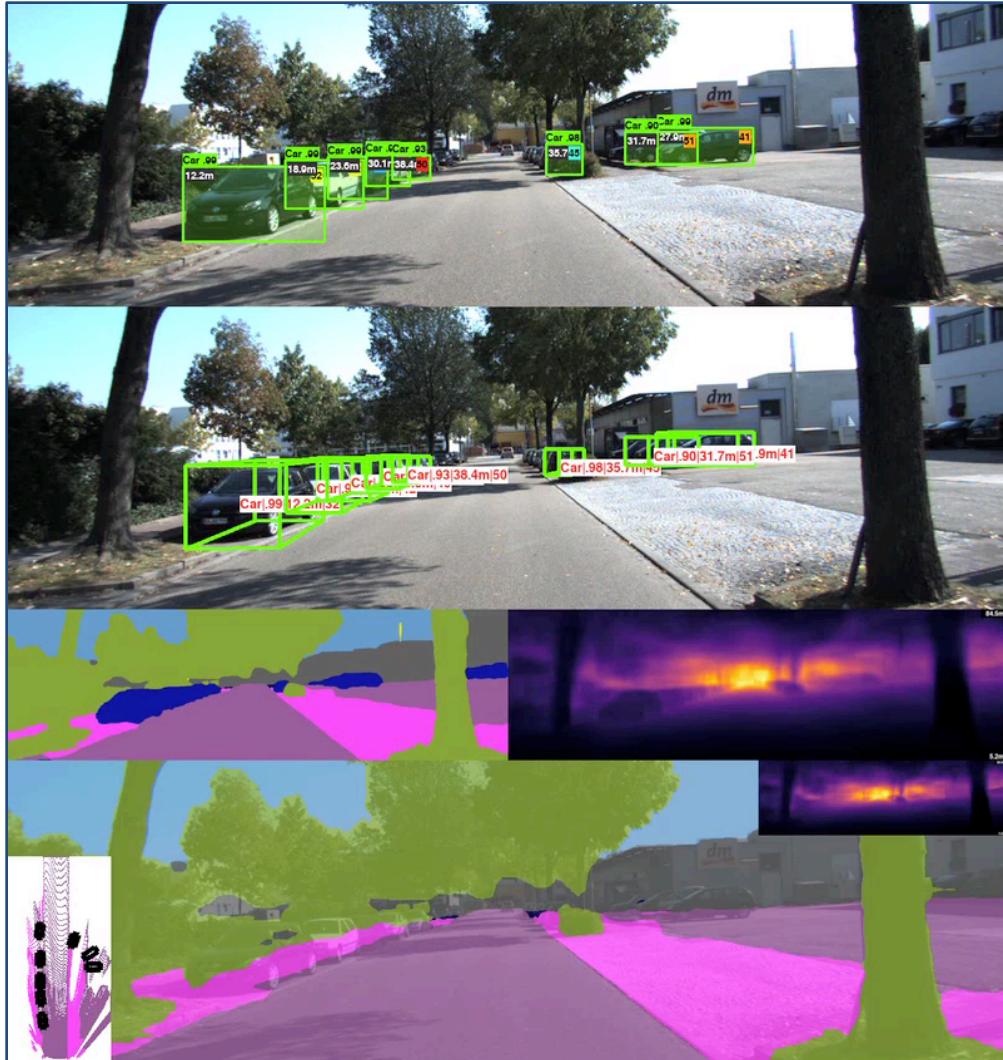
Faces and Humans



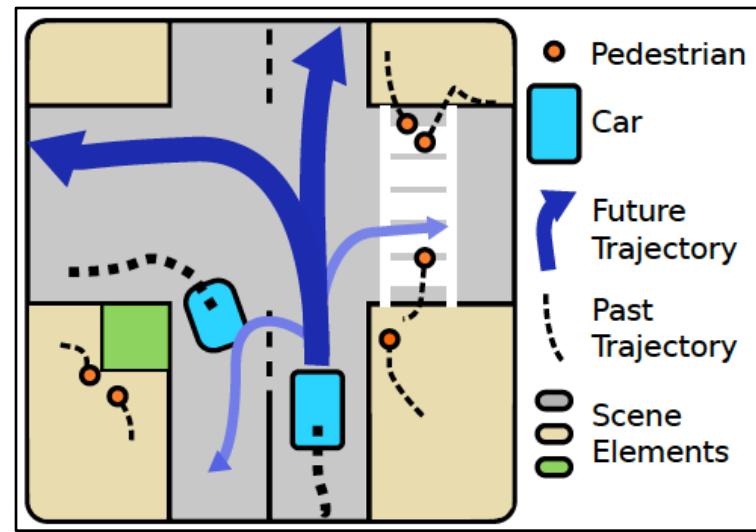
Objects and Stuff



Objects and Stuff



[Wang et al., CVPR 2019]



[Lee et al., ICCV 2017]

Material and Lighting

Input image



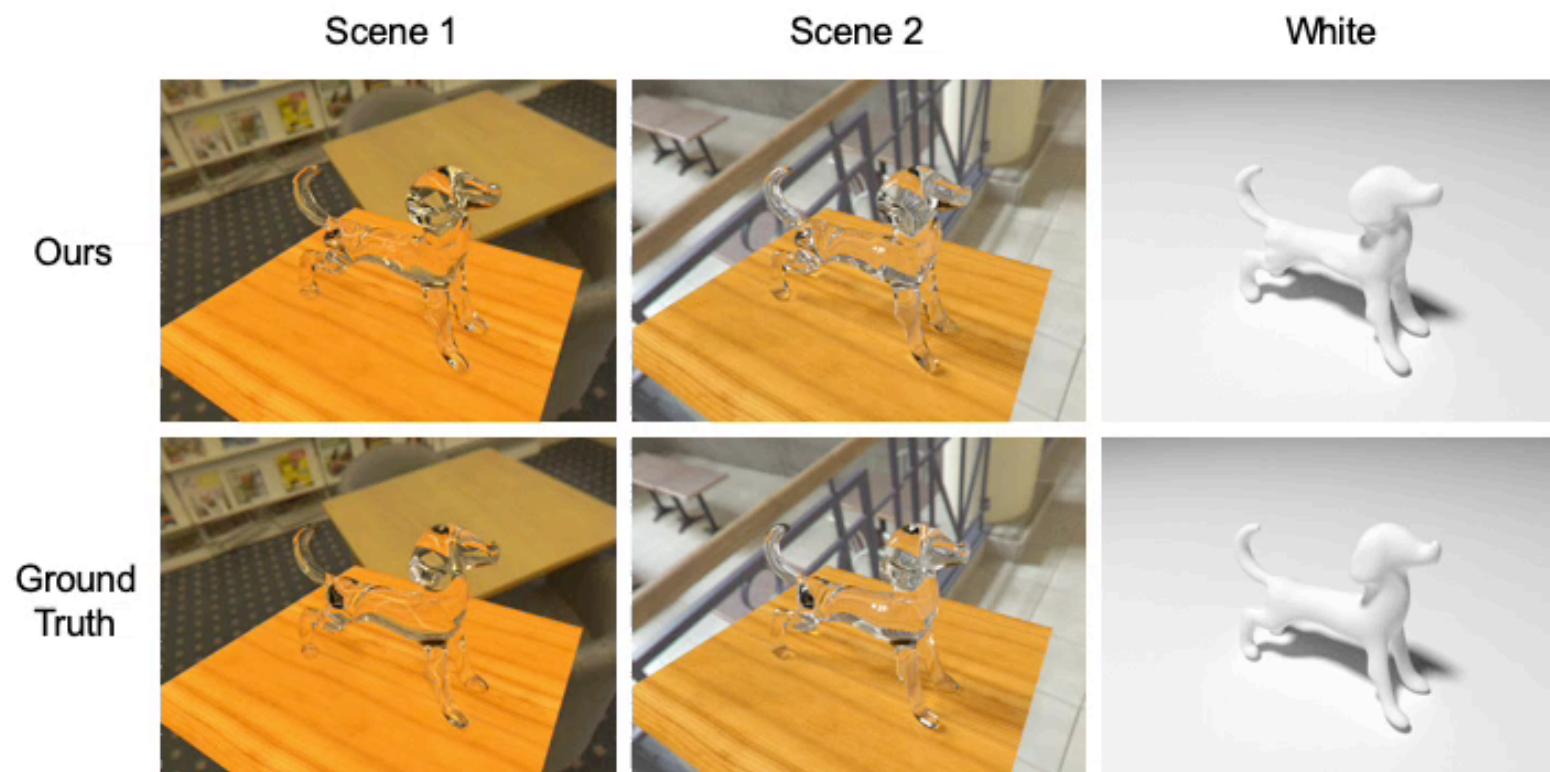
Augmented image



Material and Lighting

Dog

Rendering Results under Different Scenes



*Ours results are reconstructed from 10 views

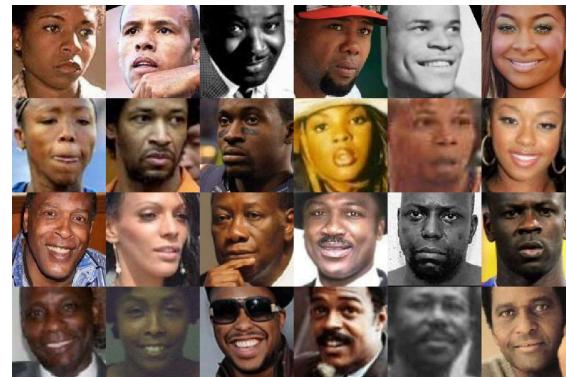
Bias and Privacy

Proportion in datasets: 80%



Caucasian

Proportion in datasets: 10%



African-American

Proportion in datasets: 5%



East-Asian

Training on biased data without domain adaptation

High accuracy

Low accuracy



Low accuracy



Using domain adaptation to address dataset bias

High accuracy

High accuracy

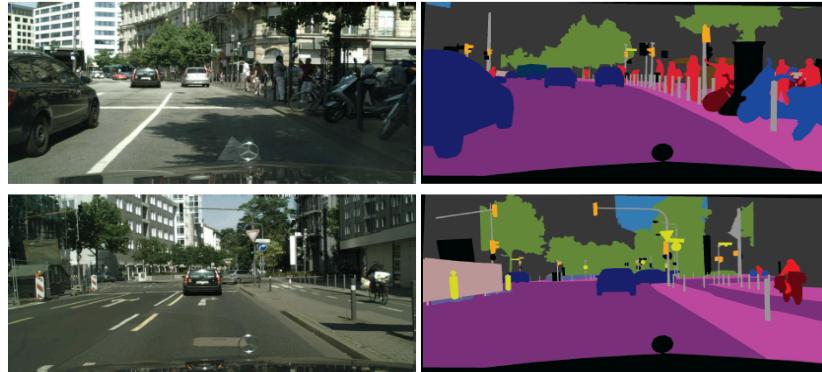


High accuracy



Bias and Privacy

Source domain: good weather, **with labels**

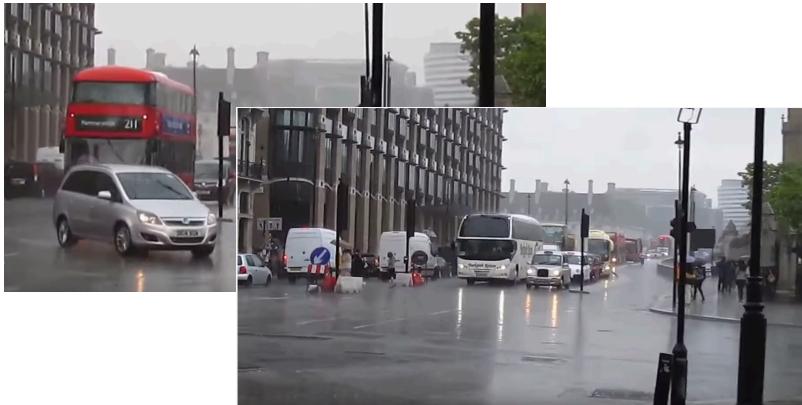


Train on source, **apply on target**



Labels require **1.5 hours** per image!

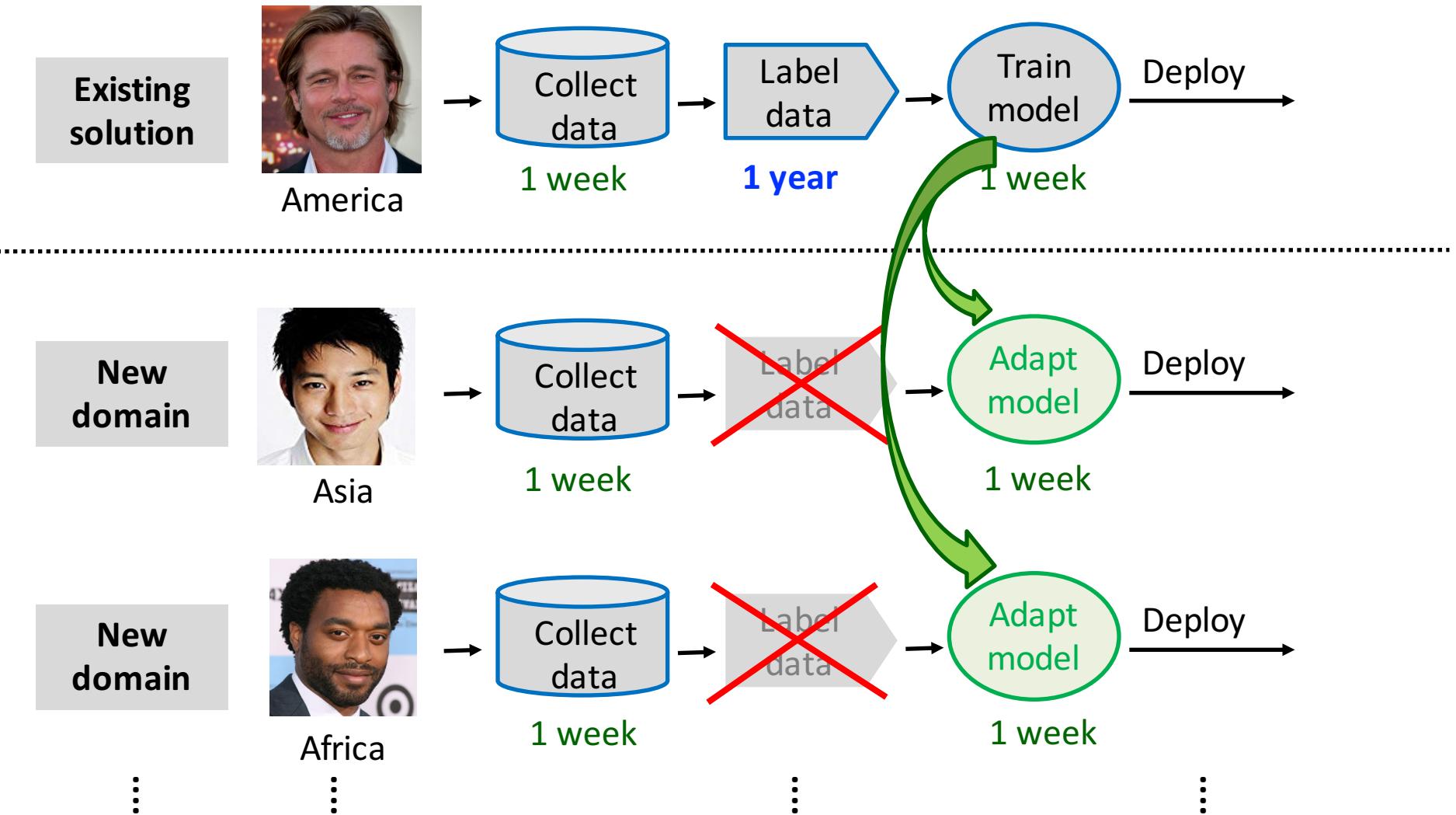
Target domain: rainy weather, **no labels**



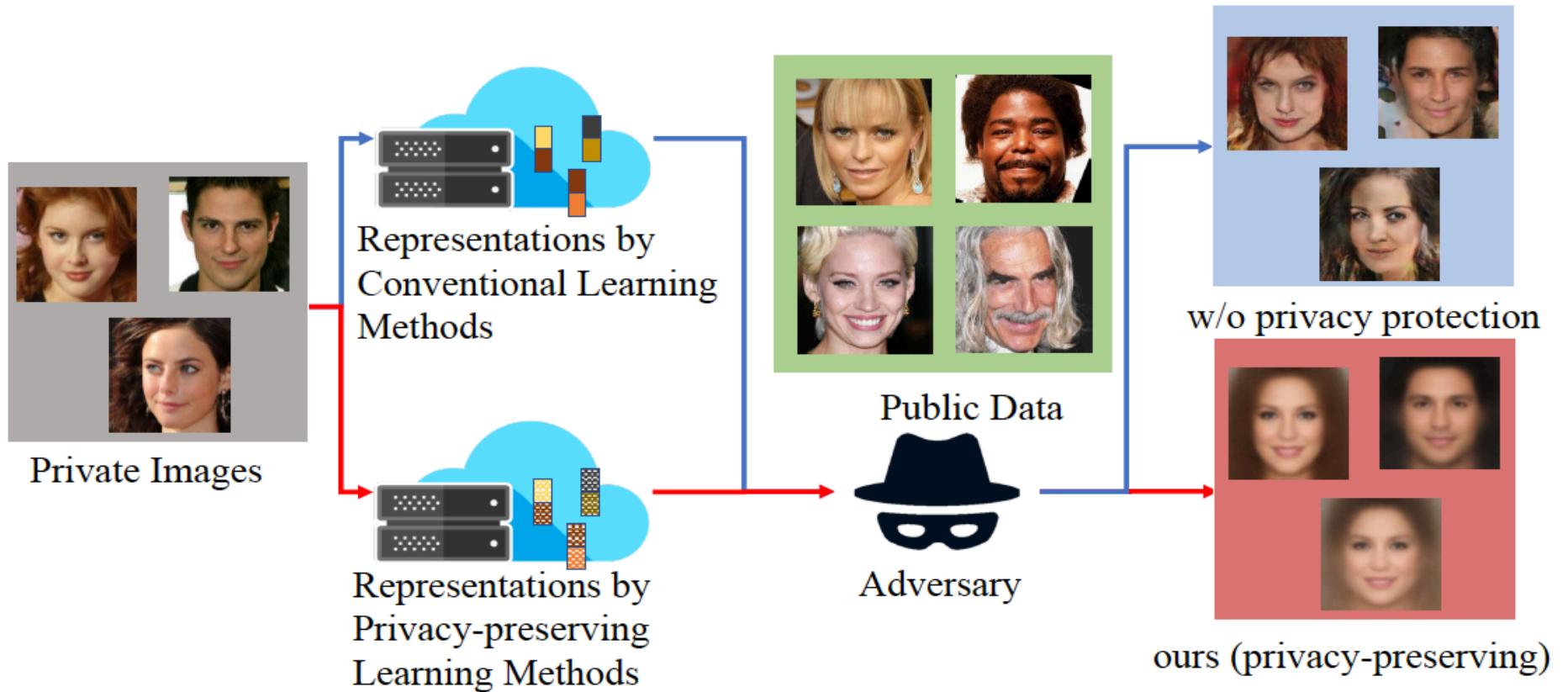
Train on source, **adapt to target**



Bias and Privacy



Bias and Privacy



My Interests

Scene understanding for self-driving

Distillation networks for fast and accurate object detection

Distillation for compressed CNN (student) to mimic uncompressed CNN (teacher), to achieve greater accuracy at the same speed.

Learning to simulate

Reinforcement learning for simulations.
Ensuring diversity and coverage.

Learning to predict uncertain future behavior

DESIRE: Deep Stochastic IOC RNN Encoder-Decoder

Monocular SFM

CVPR 2015

Large-scale, real-time, monocular SFM.
Accuracy comparable to stereo systems.

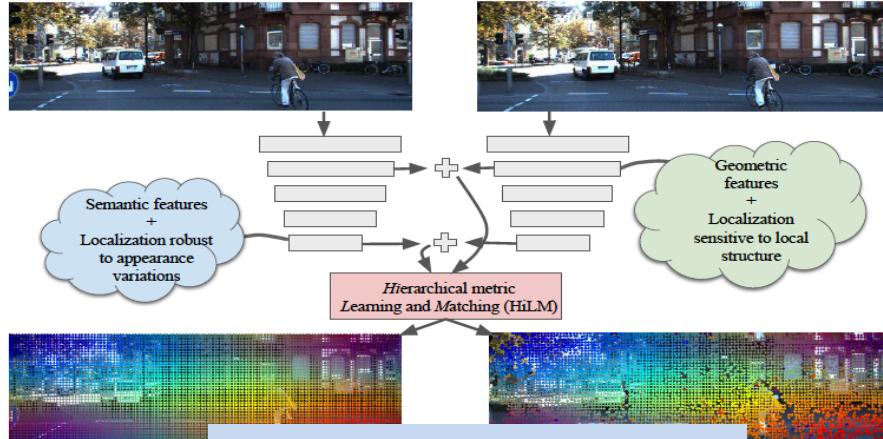
CVPR 2017

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- Deep CVAE (autoencoder) to generate diverse hypotheses.
- RNN to rank predictions based on motion, scene and interactions.
- Deep inverse reinforcement learning for long-term future rewards.

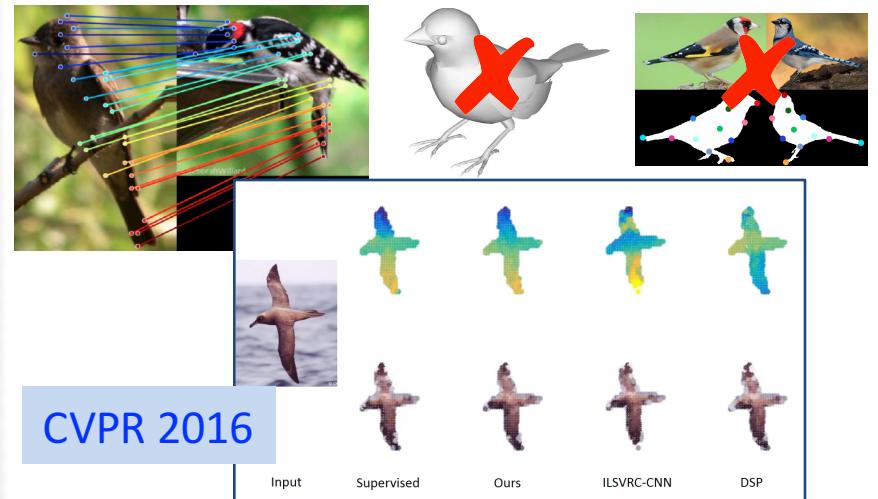
Geometric and semantic 3D reconstruction

Metric learning for correspondence



NeurIPS 2016, ECCV 2018

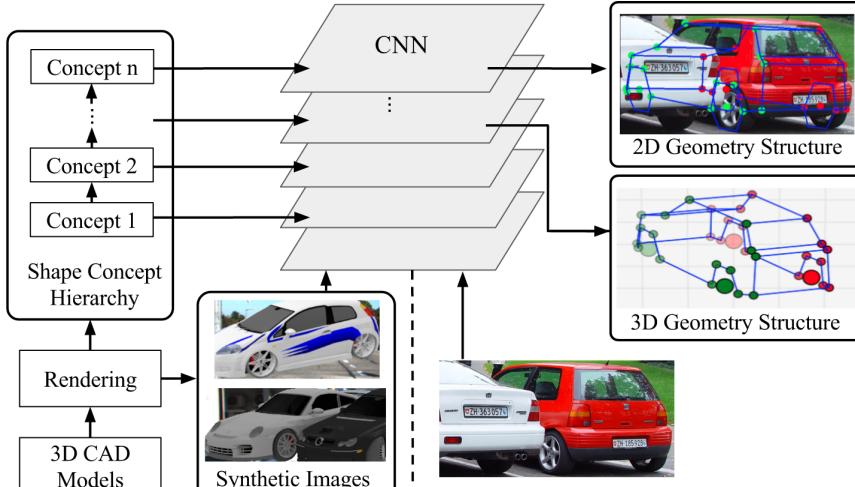
Weakly supervised semantic reconstruction



CVPR 2016

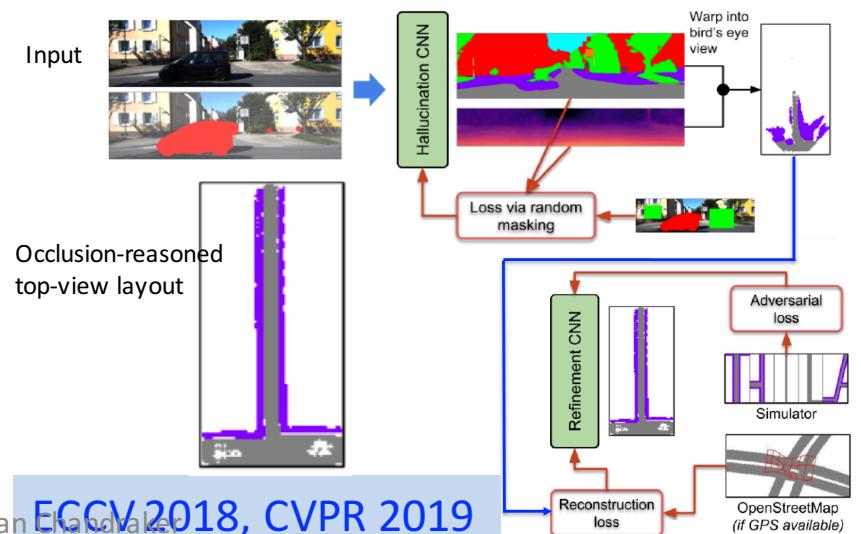
Input Supervised Ours ILSVRC-CNN DSP

Deep supervision for occlusion-reasoned parts



CVPR 2017, PAMI 2018

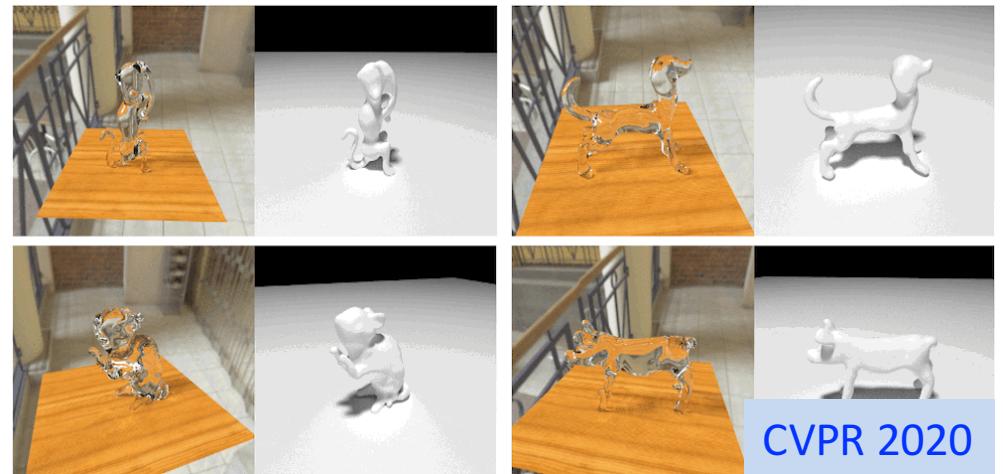
Occlusion reasoning and large transformations



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ECCV 2018, CVPR 2019

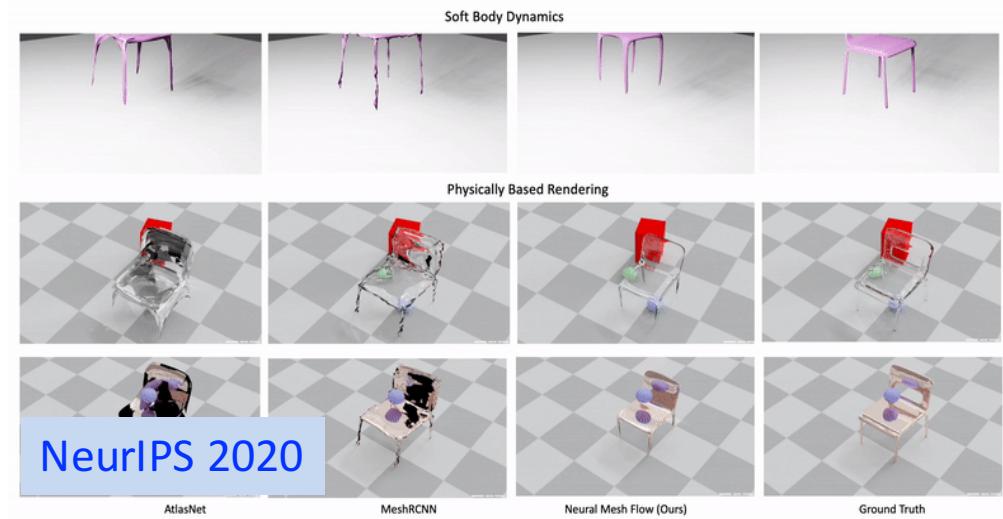
Physically-based learning for shape and material



Complex refractions and reflections



CVPR 2020



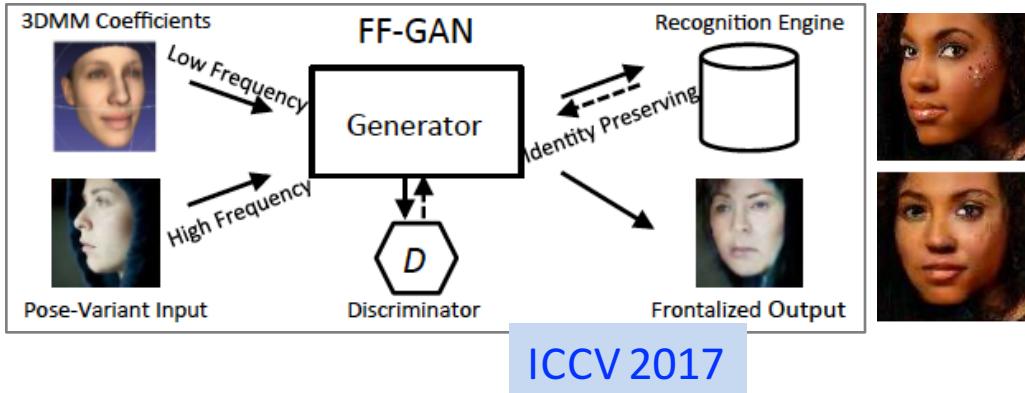
Complex materials and lighting

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Physically realizable representations

Unsupervised adaptation to new domains

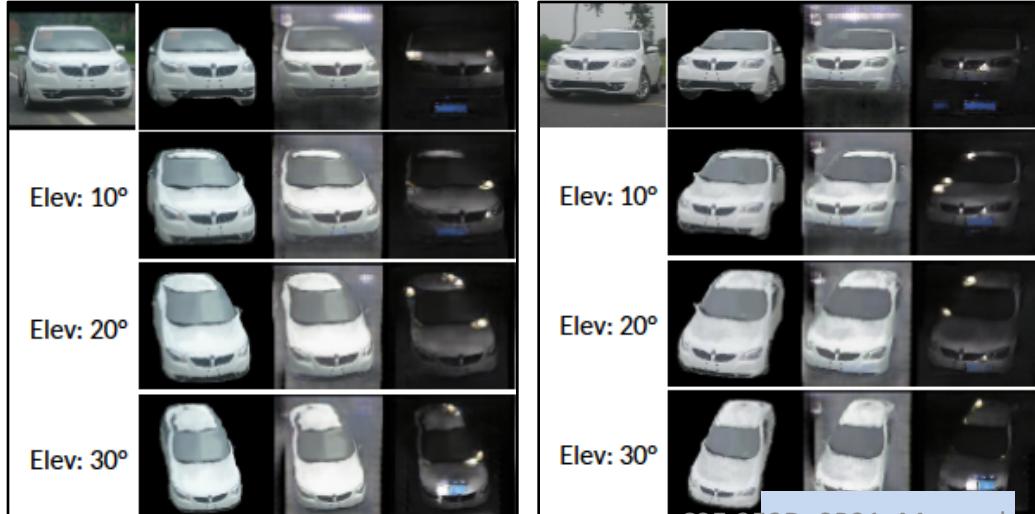
Face recognition for profile inputs



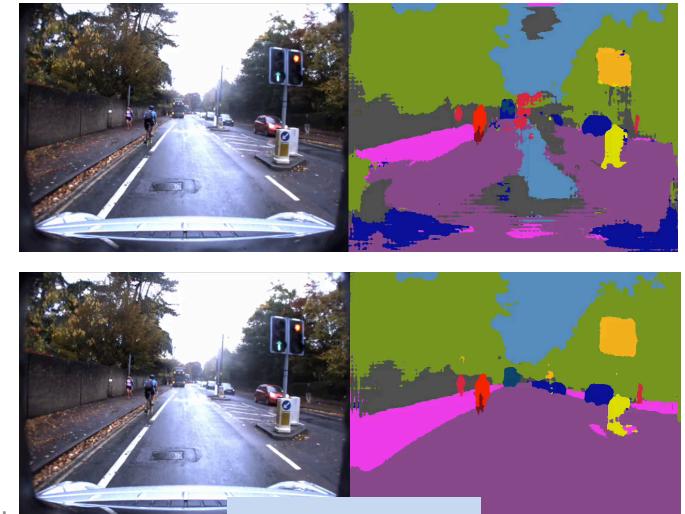
Segmentation across geographies



Car recognition across camera and lighting conditions



From rainy to good weather



Overall goals for the course

- Introduce the state-of-the-art in computer vision
- Study principles that make them possible
- Get understanding of tools that drive computer vision
- Enable one or all of several such outcomes
 - Pursue higher studies in computer vision
 - Join industry to do cutting-edge work in computer vision
 - Gain appreciation of modern computer vision technologies

Course Details

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Course details

- Each class will cover topics in computer vision
- Examples of topics
 - Correspondence
 - Optical flow
 - Structure from motion
 - Face recognition
 - Human pose estimation
 - Material and lighting
 - Semantic segmentation
 - Object detection
 - Action recognition
 - Domain adaptation

Course details

- Topics structured into a few modules
 - Background
 - Structure and Motion
 - Faces and Humans
 - Objects and Stuff
 - Material and Lighting
 - Bias and Adaptation

Course details

- “Lightning” presentations
 - Provide a broad view of the field
 - An important skill to digest and present literature
 - Four students to present in one class
 - Papers to be assigned by instructor
 - Order of presentation: alphabetic (Googledoc will be posted)
- Send recorded presentation video 3 days before class
 - Will share PPT template
 - Well-practiced and fluent presentation
 - Incorporate feedback from instructor or TA
 - Include question to class
 - Ask and answer questions after presentation

Course details

- Presentation format:
 1. Motivation and problem description
 2. Prior work
 3. Method overview
 4. Method analysis
 5. Experiments
 6. Future work and discussion

Course details

- Class webpage:
 - <http://cseweb.ucsd.edu/~mkchandraker/classes/CSE252D/Spring2021/>
- Instructor email: mkchandraker@eng.ucsd.edu
- TA: Yu-Ying Yeh (Email: yuyeh@eng.ucsd.edu)
- Grading
 - 10% presentation
 - 60% assignments
 - 30% final exam
 - Ungraded quizzes
- Aim is to learn together, discuss and have fun!

Take-home message

- Computer vision is a key branch of AI
- Enables several modern applications around us
- A lot of highly visible and high-impact activity
- Huge industry interest
- This is a great time to study computer vision!