

Learning from Image

Master DataScience

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kristian.hildebrand@bht-berlin.de





https://en.wikipedia.org/wiki/Colonel_Meow

Introduction to image processing



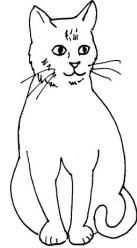


Image features and
Image retrieval

Introduction to image
processing

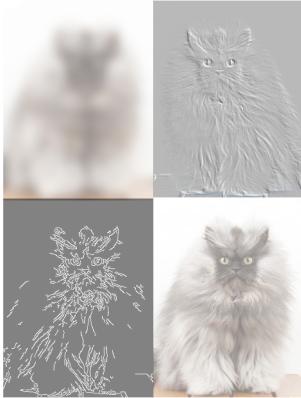




Image retrieval



Image classification and
object detection
(SVM, CNN, YOLO)

Introduction to image
processing





Image features and
Image retrieval



Image classification and
object detection
(SVM, CNN, YOLO)

Introduction to image
processing

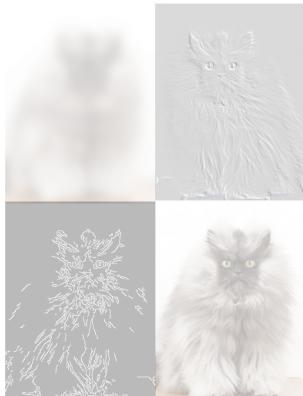




Image features and
Image retrieval



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Introduction to image
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Generative Adversarial
Networks



Image features and
Image retrieval

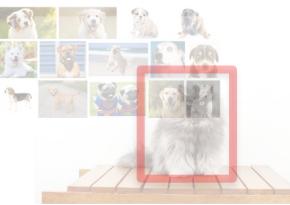
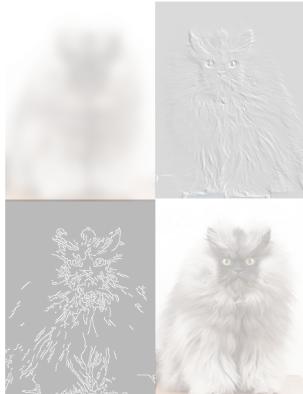


Image classification and
object detection
(SVM, CNN, YOLO)

Introduction to image
processing



Deep Reinforcement Learning



Generative Adversarial
Networks

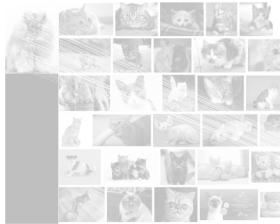
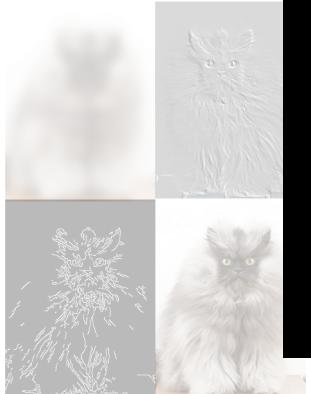


Image features and
Image retrieval



Image classification and
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(SVM, CNN, YOLO)

Introduction to image
processing



After 120 minutes of training

It plays like an expert

Generative Adversarial
Networks



Deep Reinforcement Learning

It's not data science



Image features and
Image retrieval

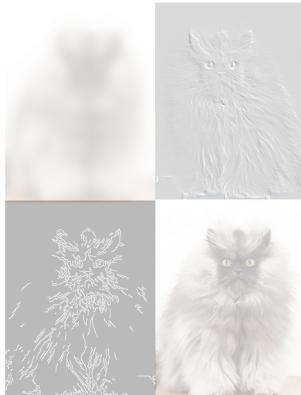


?



Image classification and
object detection
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Introduction to image
processing



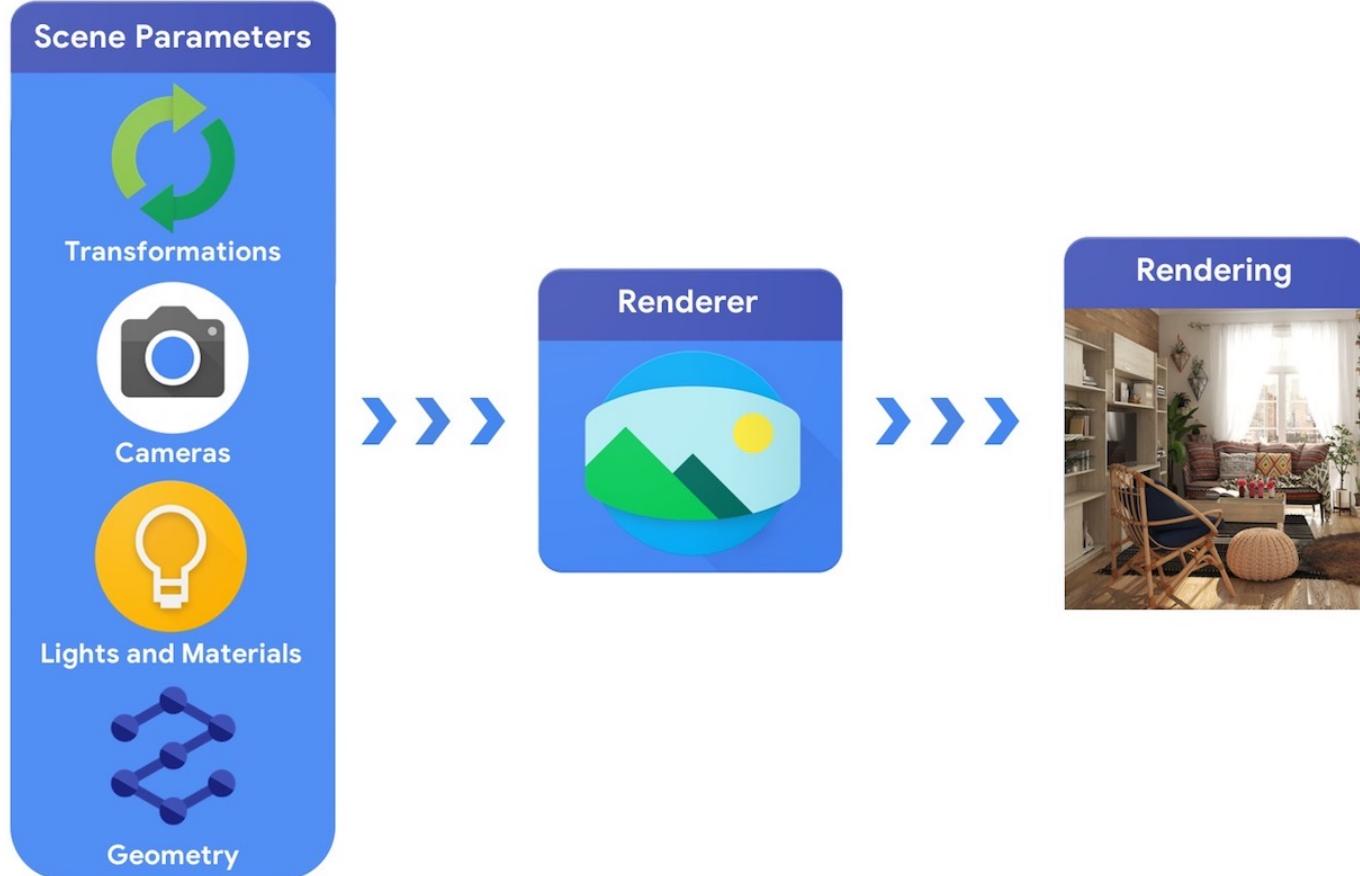
Deep Reinforcement Learning

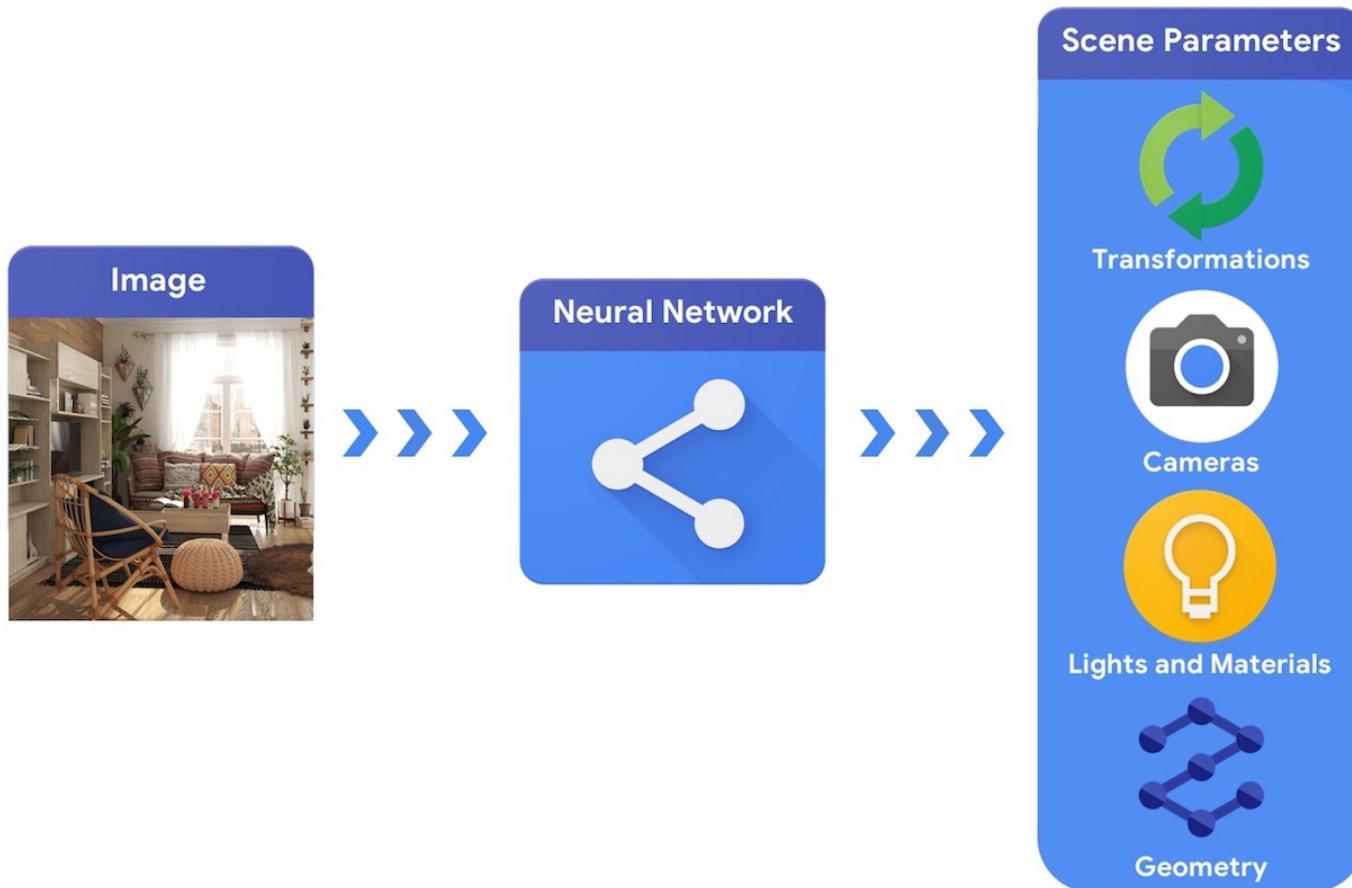
Generative Adversarial
Networks



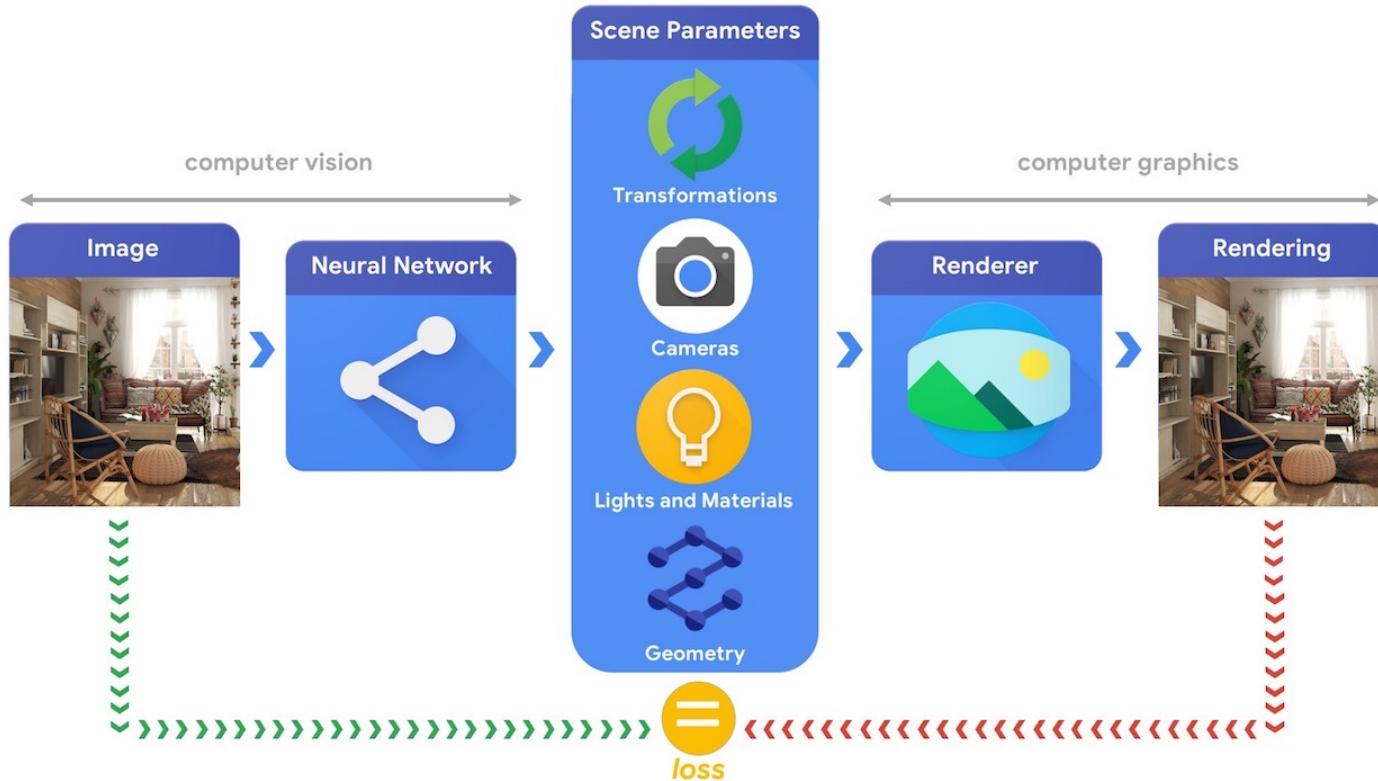
It's ~~cat~~ science

Why Computer Vision topics in
Data Science curriculum?

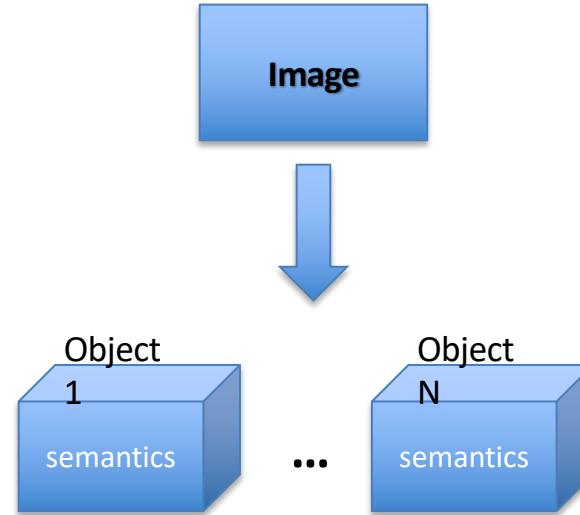
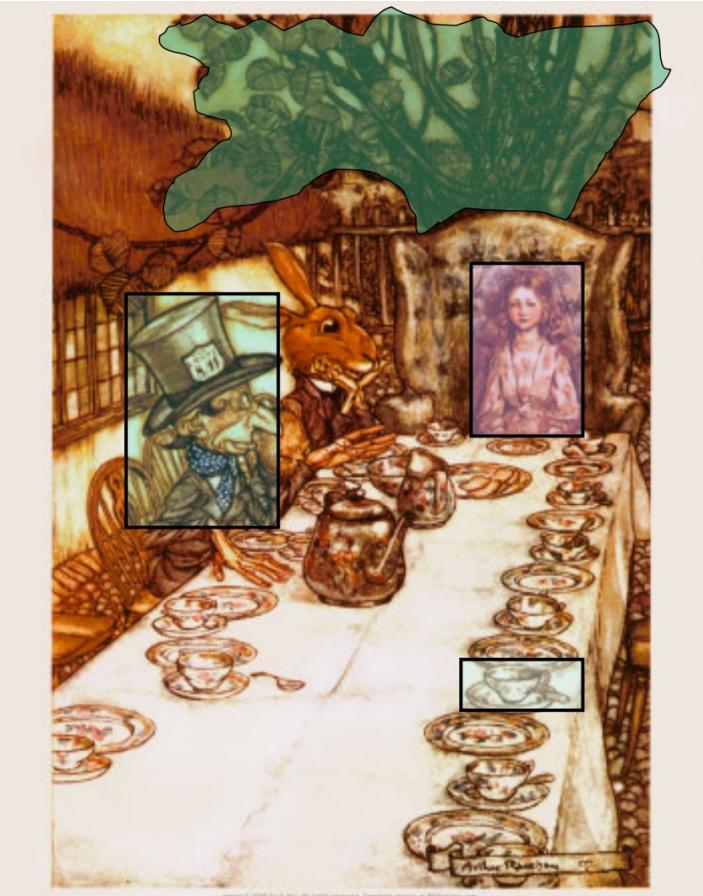




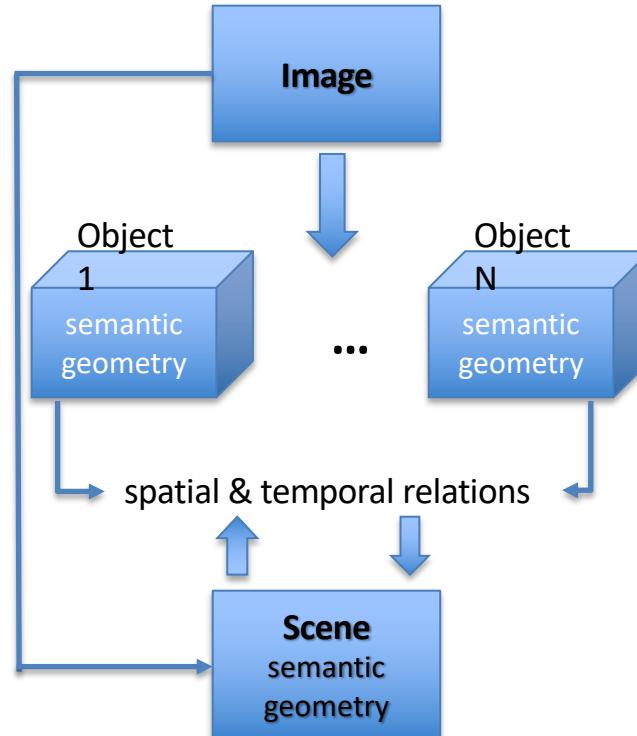
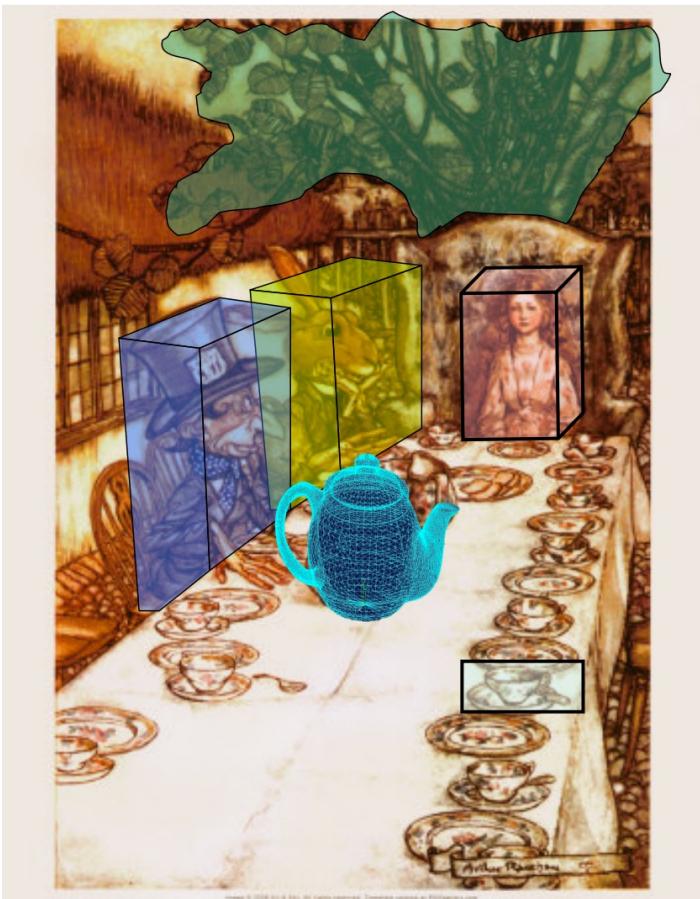
Goal of Computer Vision



Goal of Computer Vision



Goal of Computer Vision



Computer Vision Pipeline



- Information extraction:
 - features, 3D structure, motion flows etc.
- Interpretation:
 - recognize objects, scenes, events etc.

Computer Vision and Applications



<http://web.stanford.edu/class/cs231a>

Computer Vision and Applications

- Fingerprint Biometrics



- Augmentation with 3D Graphics



2d3
sensing



EosSystems

Photomodeler

- 3D Object prototyping

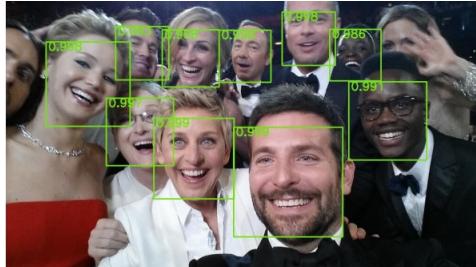
Computer Vision and Applications

- New features detector/descriptors
- CV leverages machine learning



Computer Vision and Applications

- ## ■ Face detection

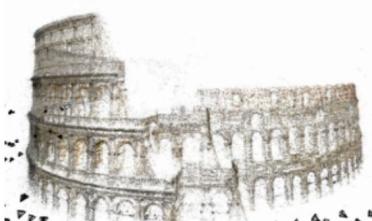


© http://www.ukprogressive.co.uk/

- ## ▪ Panoramic Photography



- #### ▪ 3D modeling of landmarks



Computer Vision and Applications

- Efficient SLAM/SFM
- Large scale image repositories
- Deep learning
- Better clouds
- More bandwidth
- Increase computational power

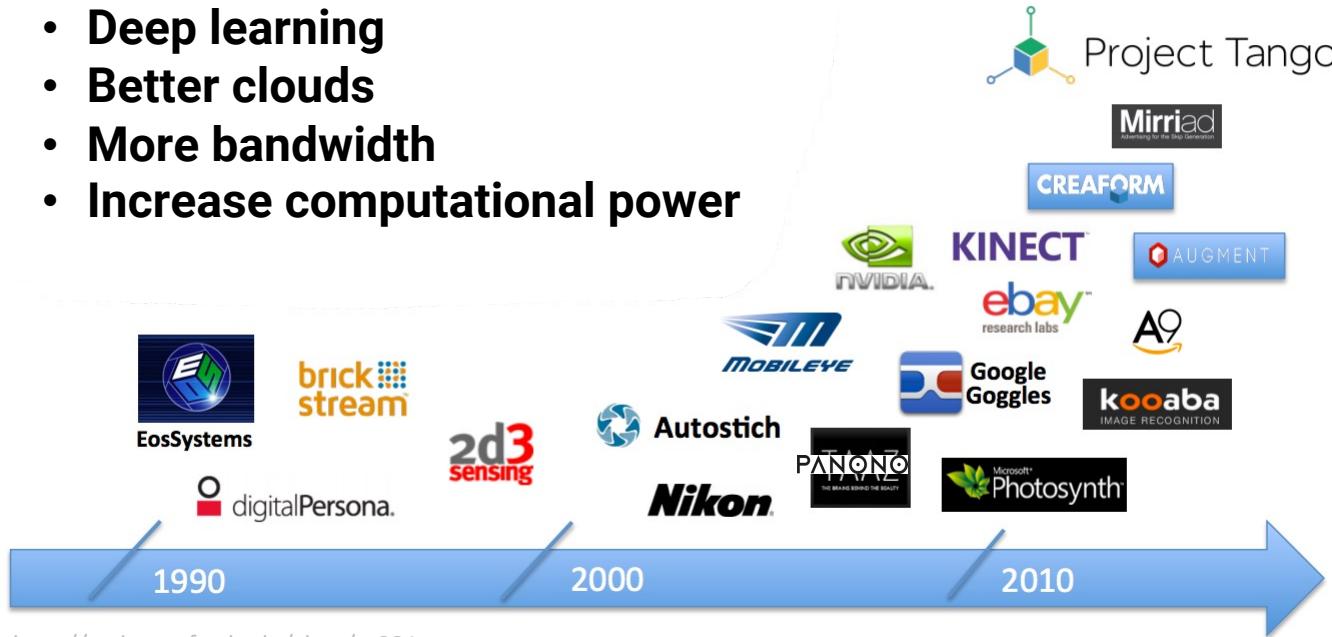


Image search engines



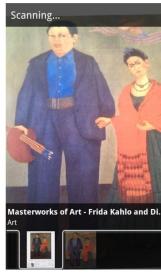
<http://web.stanford.edu/class/cs231a>

Computer Vision and Applications

- Visual search and landmarks recognition



Google Goggles



- Augmented reality

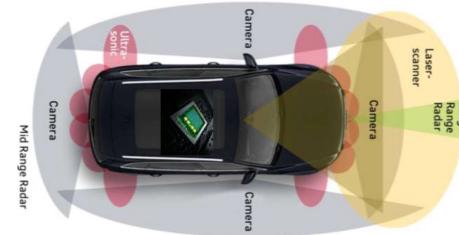


Computer Vision and Applications

- Motion sensing and gesture recognition



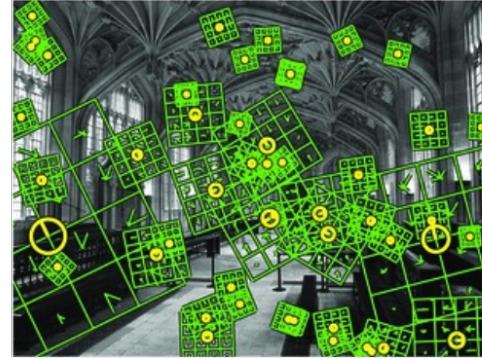
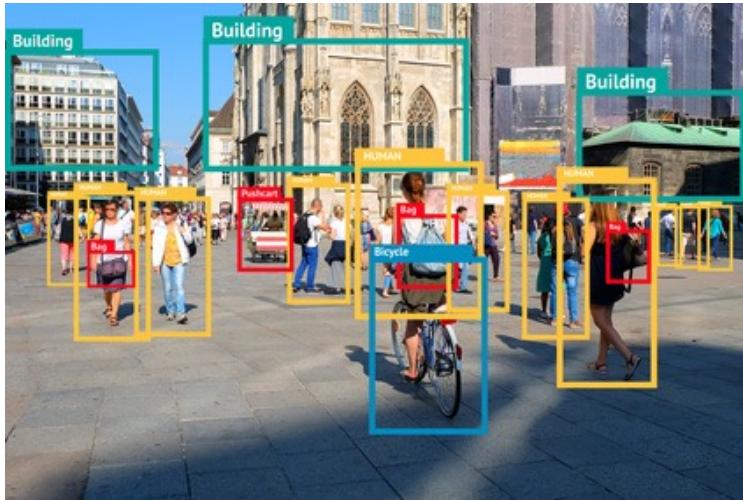
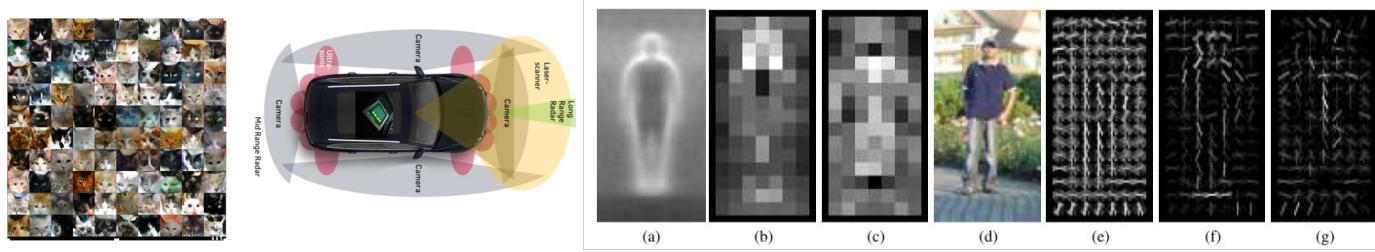
- Autonomous Driving



- Personal robotics



What I'd like you to learn here?



Computer Vision topics

Have fun coding a cool project demo!

Optional: Reading scientific papers (parallel to the lecture material)

“You can’t teach people everything they
need to know...

The best you can do is to position them
where they can find what they need to
know when they need to know it.”

*Seymour Papert, MIT, Lego Mindstorms
Erfinder*

Syllabus

Date	Lecture topic	Assignment due date
09.10.23	Introduction	Assignment 0
16.10.23	Image Processing (Filter, Transforms)	
23.10.23	Image Basics / Image features	Assignment 1
30.11.23	Feature Matching	
06.11.23	Image Retrieval	
13.11.23	Neural Networks recap	Assignment 2
20.11.23	ConvNets, ConvNet Architectures	
27.11.23	Object Detection (YOLO, MobileNet)	
04.12.23	Machine Learning, Supervised Image Classification	
	Fine Tuning and Domain Adaptation	Assignment 3
11.12.23	Autoencoder and Embeddings	
18.12.23	Transformers	
08.01.24	Transformers and Images	
15.01.24	Unsupervised, Generative Methods	Assignment 4
22.01.24	GANs and Diffusion Models II	
29.01.24	State of the Art / Recent Developments	
12.02.24	Projektpäsentation / Dokumentation	

About me

Contact and Information

- Kristian Hildebrand
 - Computer Graphics and Interactive Systems
- Email
 - khildebrand@beuth-hochschule.de
- Website
 - <http://hildebrand.beuth-hochschule.de>
 - Course material + news via **Moodle**
- Room
 - B332
- Office
 - Haus Gauß, Room B218
 - Consultation hours as needed
 - Available on DSM-Students Slack-Workspace

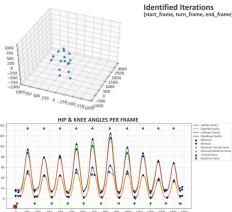
Kristian Hildebrand

Since 10/2015	Professor Graphics and Interactive Systems Beuth University for Applied Science
2014 – 2015	Principal Research Engineer, DISDAR GmbH, Berlin
2008 – 2013	Research Assistant, TU Berlin
2012	Disney Research, Visiting Researcher, Zürich
2006 – 2008	Software developer, art+com AG, Berlin
1999 – 2005	Diploma, Max-Planck-Institut Informatik, Saarbrücken Computer Science, UBC, Vancouver Mediensystems, Bauhaus University Weimar

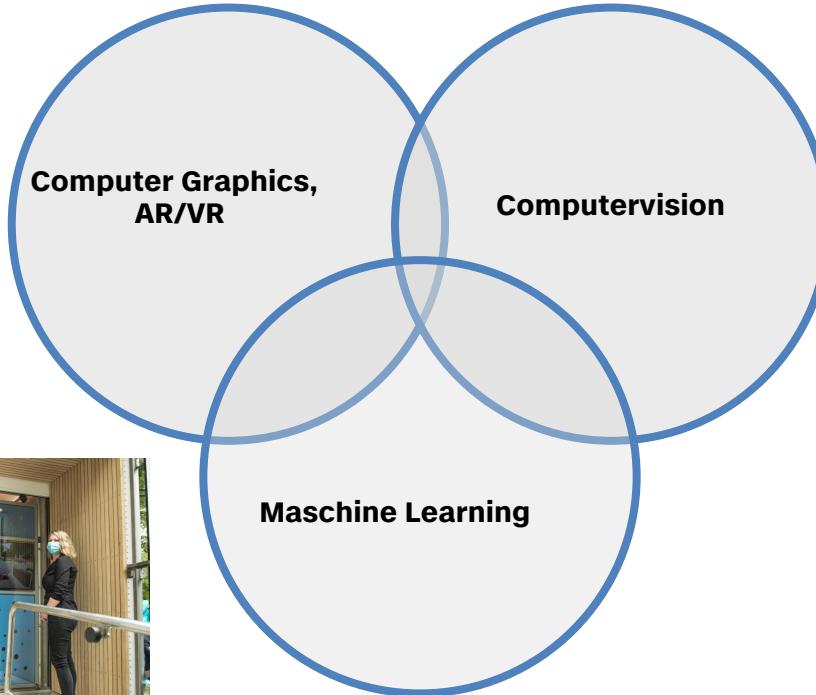
Intelligent Interactive Systems



BewARE



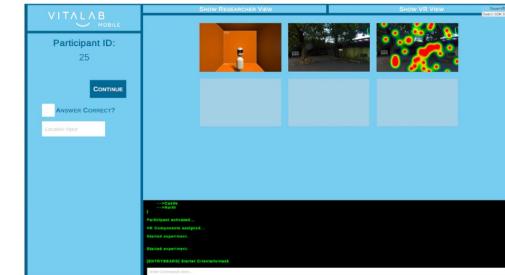
VITALAB
MOBILE



Interaction-Based
Redirected Walking

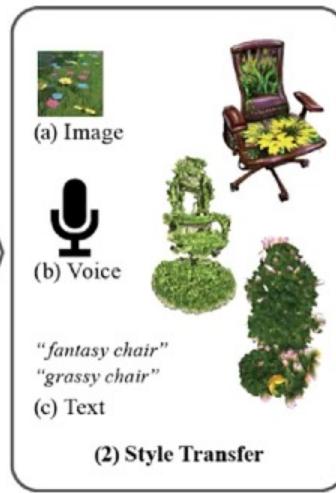
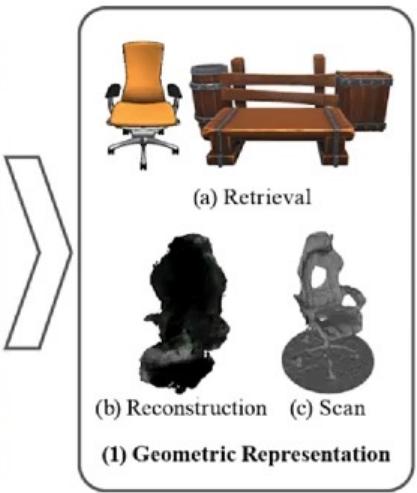


VR Evaluation Toolkit



Style-aware Augmented Virtuality Embeddings (SAVE)

Online Submission ID: 1157

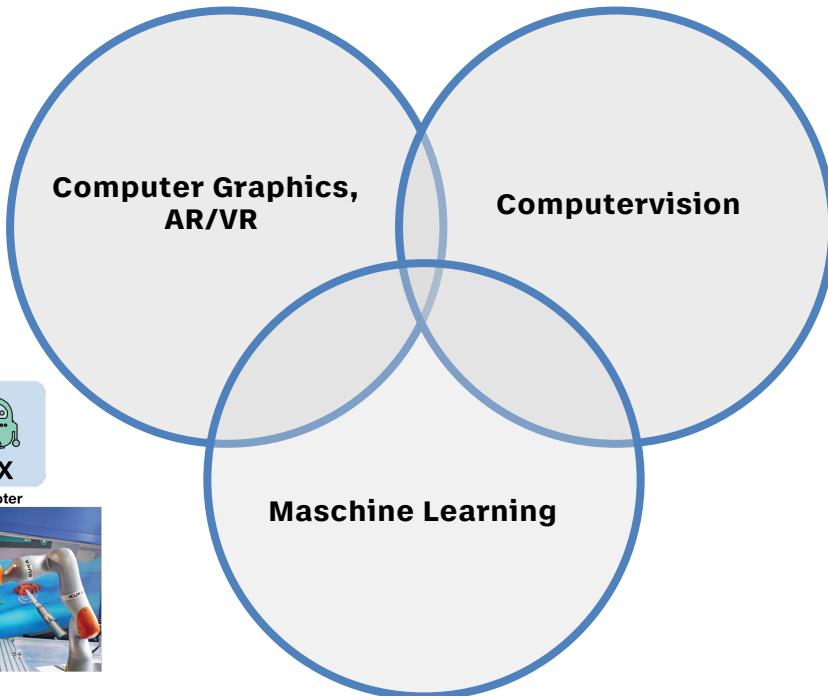
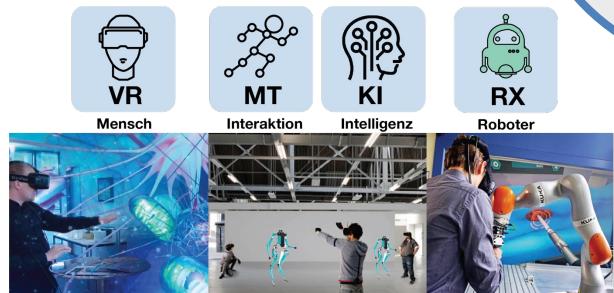




B WD 6936

Intelligent Interactive Systems

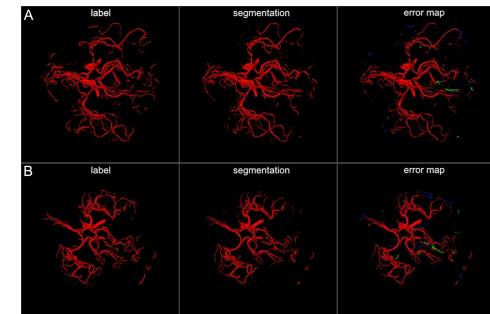
HUMAN.VR.LAB
Human-Robot Interaction
Virtual Reality Laboratory



SynthNet



Medical Imaging





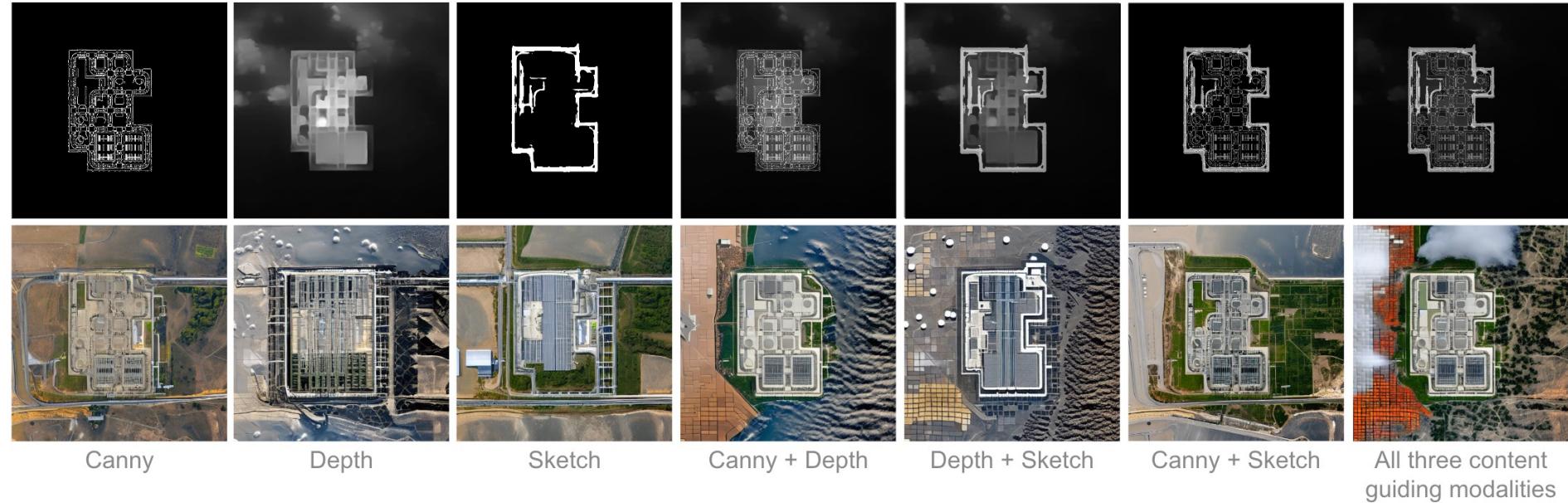
Draw me something

and I'll tell you, what I think it is!



[Start again!](#)

Multimodally-guided Satellite Imagery Synthesis



Using Game Engines and Machine Learning to Create Synthetic Satellite Imagery for Tabletop Verification Excercise.

Johannes Hoster, Sara Al-Sayed, Felix Biessmann, Alexander Glaser, Kristian Hildebrand, and Vy Nguyen. INMM & ESARDA Joint Annual Meeting, Vienna, May 2023

CAD Models to Real-World Images: A Practical Approach to Unsupervised Domain Adaptation in Industrial Object Classification



CAD Models to Real-World Images: A Practical Approach to Unsupervised Domain Adaptation in Industrial Object Classification.

Dennis Ritter, Mike Hemberger, Marc Höning, Volker Stopp, Eric Rodner, Kristian Hildebrand. Adapting to Change: Reliable Learning Across Domains - ECML-PKDD 2023 International Workshop

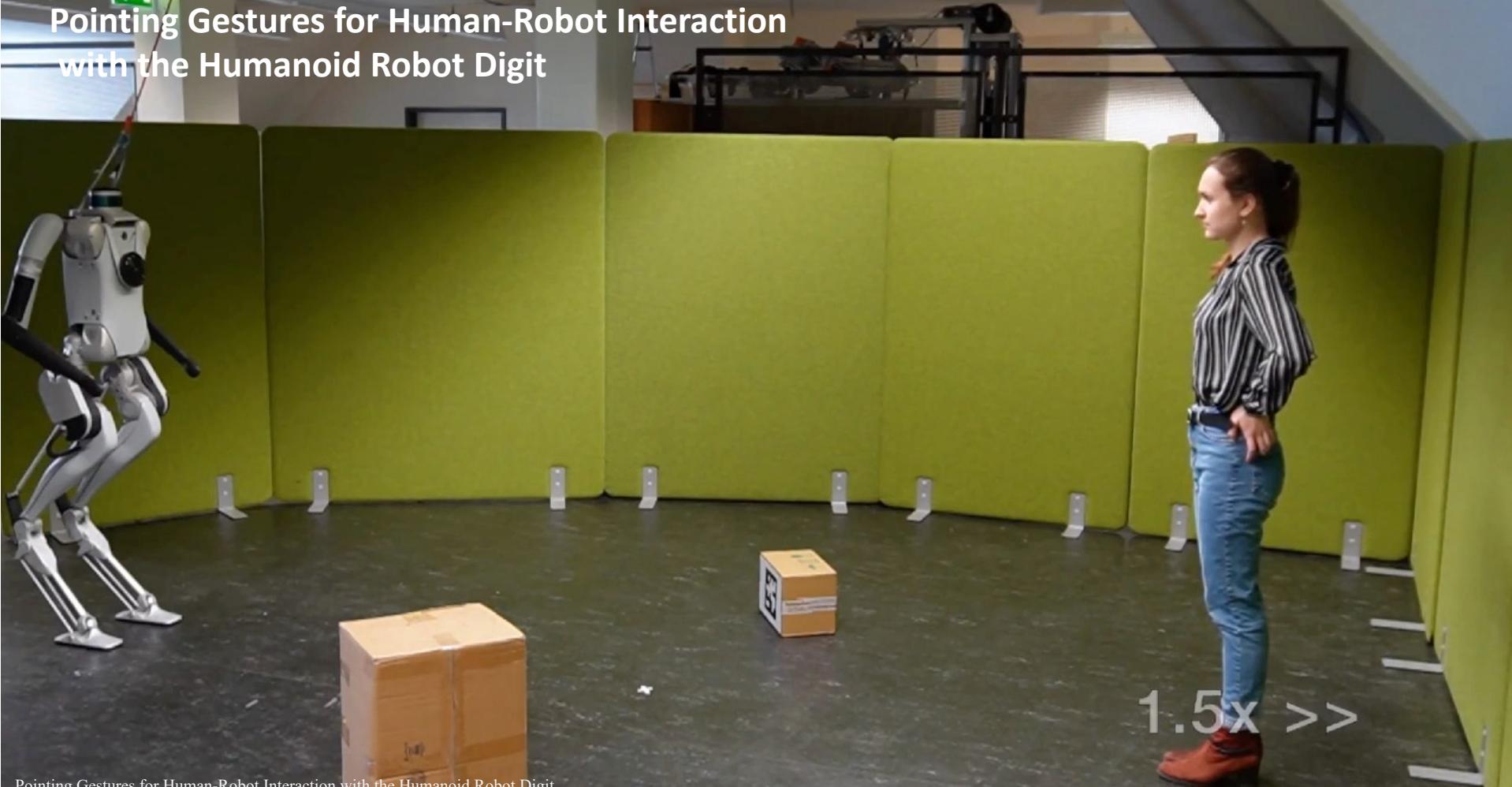
<https://github.com/dritter-bht/synthnet-transfer-learning>

Stabilized human pose estimation by temporal guidance

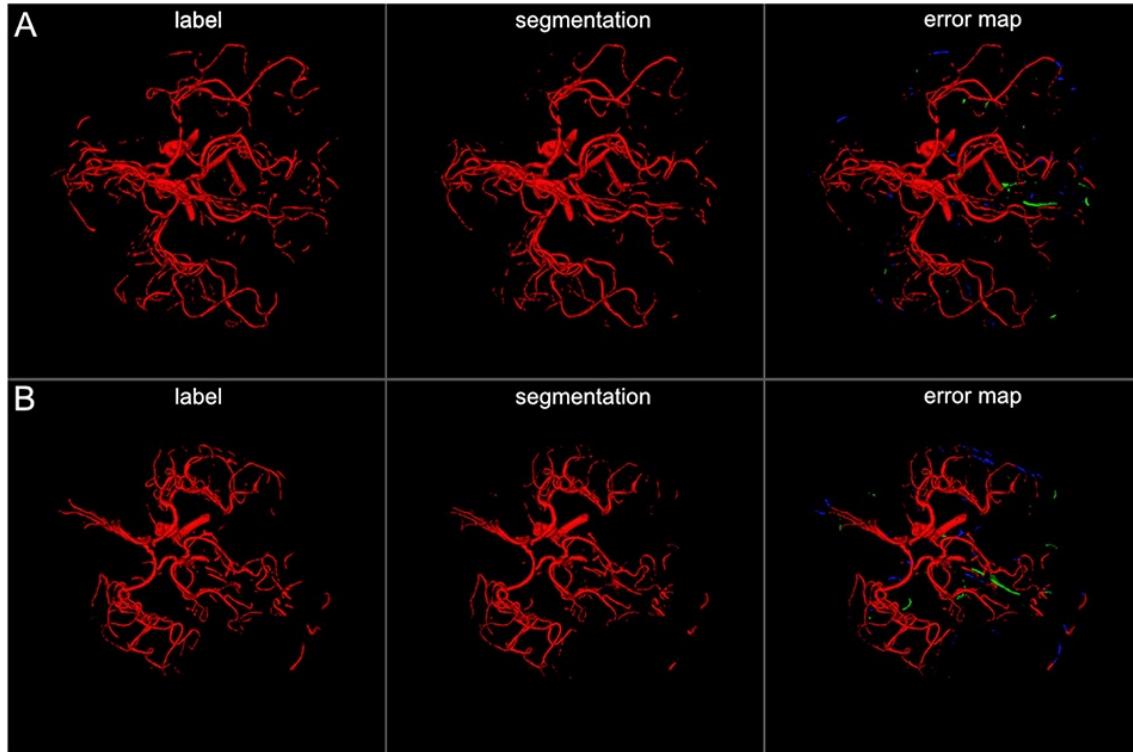




Pointing Gestures for Human-Robot Interaction with the Humanoid Robot Digit



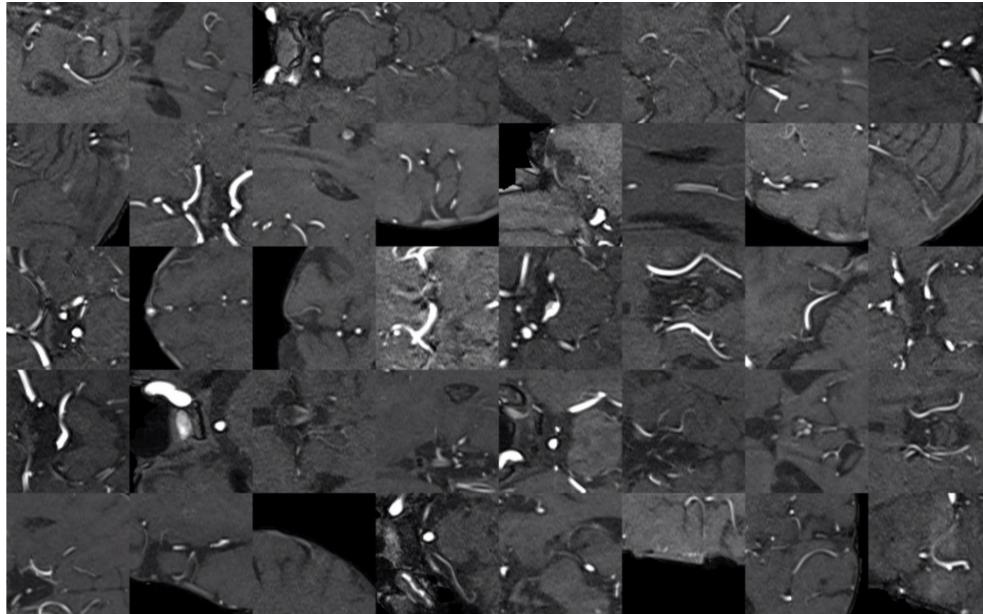
U-net Deep Learning Framework - Results



Model architecture/Performance measure	U-net
Dice coefficient	0.891
95% Hausdorff distance (voxels)	47.277
Averaged Hausdorff distance (voxels)	0.342

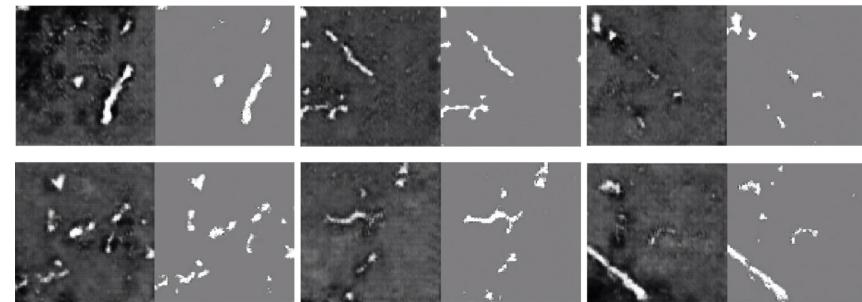
U-net Deep Learning Framework for High Performance Vessel Segmentation in Patients with Cerebrovascular Disease *Frontiers in Neuroscience*. 2019.
Livne, Michelle and Rieger, Jana and Aydin, Orhun and Taha, Abdel and Akay, Ela and Kossen, Tabea and Sobesky, Jan and Kelleher, John and
Hildebrand, Kristian and Frey, Dietmar and Madai, Vince

Generating MRI using Generative Adversarial Networks



Tabea Kossen

real data



synthetic data

Organization Facts

Facts

- 4 Assignments: **done alone**
- **Questionnaires – easy weekly questions to recap**
- Course Project: done in groups (2-4 people)
- Frameworks: Python + Numpy, scikit-learn, opencv, pytorch
- GPU resources will be needed
 - Who has, who needs?
- **Please bring your laptop**

Rules I

- **There are no stupid questions! Ask – It is not always a receiver problem! ☺**
 - **Attendance mandatory for Part 1**
 - **Grading**
 - Exercises + Questions: 50%
 - Project (implementation): 50%
 - Project Presentation / Evaluation / Documentation (pdf) (33% of project grade)
- ← **Why?**
- ← **Why?**
**Get familiar with
evaluating your work!**

Rules II

- **Questions**

- Every Week there is a set of three questions about the previous content ($\max(\text{sum}(0.5 \text{ points}, \text{week}), 10)$)
 - 13 weeks, only 10 count

Rules III

- Reuse existing resources, but
 - **Must give references**
 - Example: you may use illustrations in your talk taken from other slides, but add a note such as [image: taken from Alexa et al.]
 - For the assignments and your projects rather complete implementations are available on the web – **use them for inspiration, do not copy!**
 - **If in doubt, ask me**
- **Two days before presentation – send me your slides!**

Important information

- Assignments due in time
- Assignments need to run out of the box
 - Don't copy paste code (communicate but implement alone)
- Advice for data path

```
Assignments/
├── lfi-01/
│   ├── your_code.py
│   └── your_code2.py
├── lfi-02/
│   └── your_code.py
├── lfi-03/
│   └── your_code.py
└── lfi-04/
    ├── your_code.py
    └── your_code2.py

Data/

# please refer to your data dir in a relative path
.../Data/lfi-01/...
```

Possible to use
colab

QUIZ

- Who had image processing in bachelor program?
- What is an image?
- Pixel? Voxel?
- Feature / SIFT / Corner detectors?
- Convolution

QUIZ

- Linear algebra
 - Transformation
 - PCA / SVD?
- Machine Learning
 - SVM?
 - K-means? Kd-Tree
 - Implemented your own basic neural network from scratch?
 - ConvNet / Auto-Encoder / GAN?