

KI Labor - Wintersemester 2021

Computer Vision 2



Schedule

Datum	Thema	Inhalt	Präsenz
01.10.21	Allg.	Organisation, Teamfindung	Nein
08.10.21	CV	Vorstellung CV	Nein
15.10.21	CV	Q&A Sessions	Nein
22.10.21	cv	Sprintwechsel, Vorstellung Assignment	Ja
29.10.21	CV	Q&A Sessions	Nein
05.11.21	CV / NLP	Abgabe CV, Vorstellung NLP	Ja
12.11.21	NLP	Q&A Sessions	Nein
19.11.21	NLP	Sprintwechsel, Vorstellung Assignment	Ja
26.11.21	NLP	Q&A Sessions	Nein
03.12.21	NLP	Q&A Sessions	Nein
10.12.21	NLP / RL	Abgabe NLP, Vorstellung RL	Ja
17.12.21	RL	Q&A Sessions	Nein
14.01.22	RL	Sprintwechsel, Vorstellung Assignment	Ja
21.01.22	RL	Q&A Sessions	Nein
28.01.22	RL	Abgabe RL, Abschluss KI Labor	Ja



Agenda for today

- 1. Transfer learning
- 2. Assignment
- 3. Assignment ideas



Transfer learning



Transfer learning is the idea to utilize knowledge acquired for one task to solve related ones





Transfer learning differs from traditional ML in that it draws on previously learned tasks

Select Source Model Choose a pre-trained source model from available models released by research institutions that is suitable for the task of interest

Reuse Model

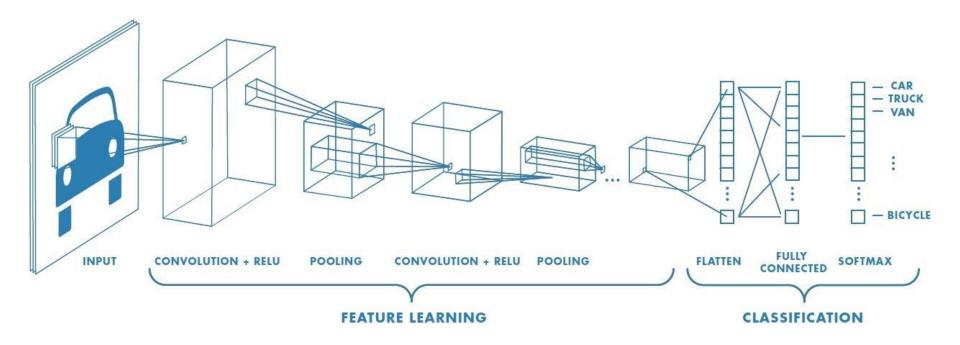
Use the pre-trained model as a starting point and leverage the model knowledge for the task of interest

Tune Model

Refine the model on data that is available for the task of interest



We can apply this idea to CNNs by freezing parts of the network and fine-tune the rest





Assignment



Assignment

- Topic: Transfer Learning
- Open Ended Assignment
- Minimum requirements
 - Choose a CV Dataset for fine-tuning
 - fine-tuning datasets have to be different from the dataset used for pretraining the model
 - Compare "pre-trained and fine-tuned" vs "training from scratch"
 - Understand and explain
 - the datasets used
 - the network architecture
 - results



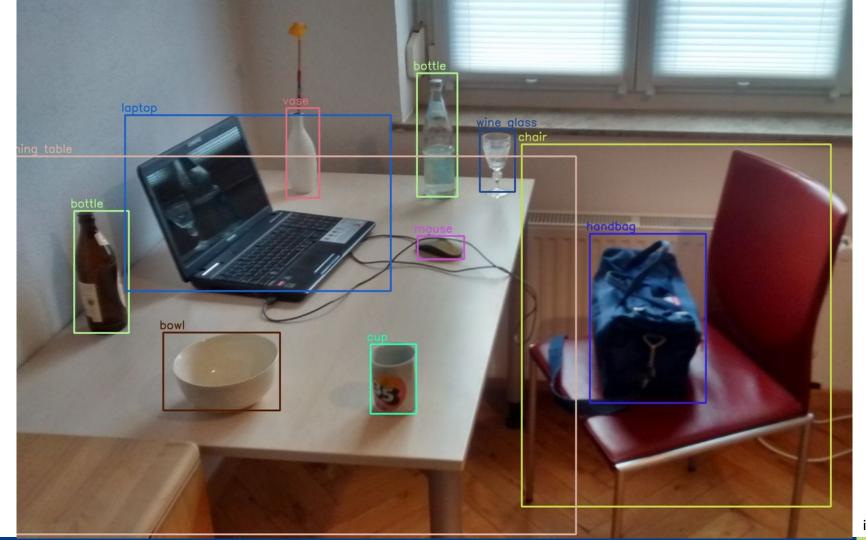
Assignment

- Many methods available
 - Image classification
 - Object detection
 - Object segmentation
 - Depth Estimation
 - o etc.
- Many datasets available
 - <u>TensorFlow Datasets</u>
 - PyTorch vision Datasets



Transfer learning for object detection







YOLO - You Only Look Once

- by Joseph Chet Redmon (2016)
- fun paper: https://arxiv.org/abs/1612.08242
- originally implemented in Darknet
 - an open source deep learning framework written in C and CUDA



YOLO v4 and v5 available in PyTorch & TensorFlow



Possibilities

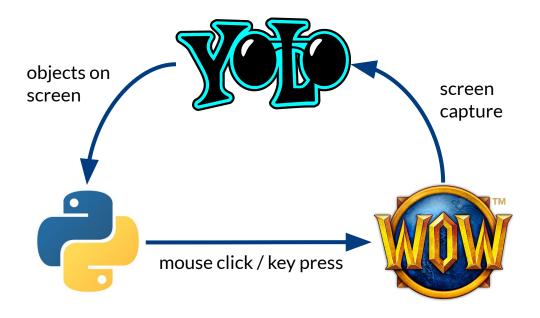
- Use pre-trained YOLO model of appropriate size (s,m,l,x)
- Fine-tune on a custom generated dataset
 - Google Open Images Dataset V6
- Generate you own mini test set!
 - Test how it works with your phone camera
 - Take ~50 photos that contain the target object
 - Annotate the photos with
 - CVAT web interface cloud version is ok for a small test set
 - <u>Labelimg</u> python & Qt offline only
 - Export in appropriate YOLO format and use in notebook



Even more fun!

- For synthetic environments where the variability of object is low you can even make your own training set in <1 hour!
 - in video games (many examples on youtube)

Example:





Even more fun!





But it's not so easy 😆

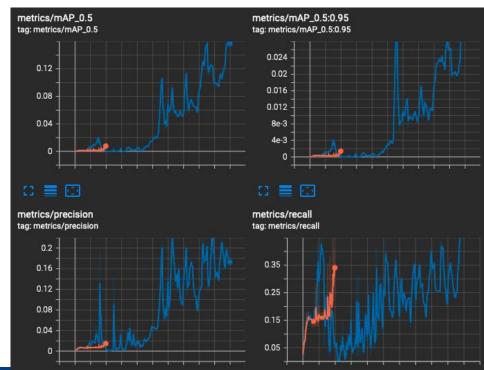




My attempt with Moorhuhn

- train: 11 screenshots
- val: 3 screenshots
- Data engineering 90% of effort
- Labeling was fast < 20min
- YOLOv5
- Relatively low precision and recall due to only 11 training images

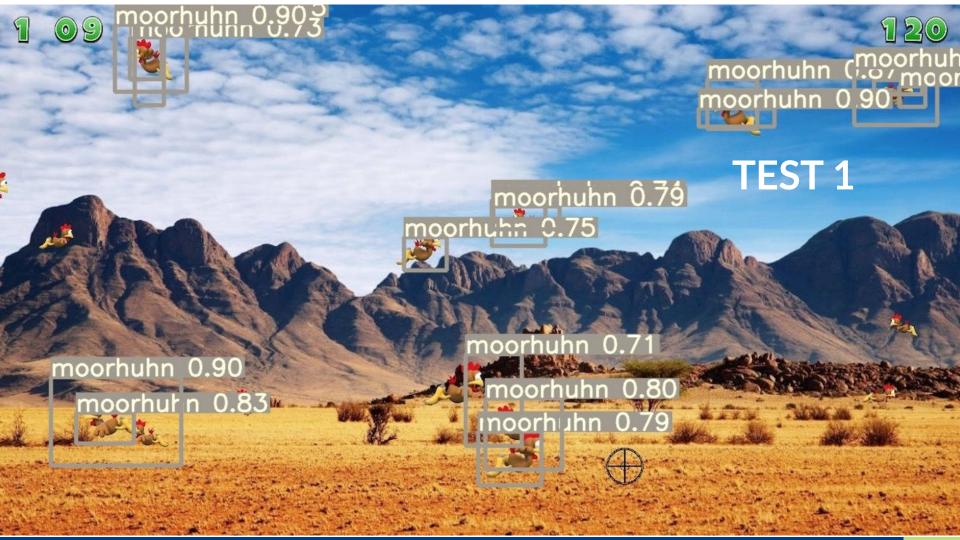




Labeled images









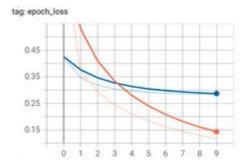


Transfer learning for image classification



A simpler alternative

- <u>TensorFlow Hub</u> contains many datasets and pretrained models for different tasks
- TF Hub Tutorial
 - MobileNetV2 pre-trained on ImageNet
 - Fine tuned on flower_photos



- PyTorch Tutorial
 - ResNet18 pre-trained on ImageNet
 - Fine tuned on ImageNet subset of ants and bees

predicted: bees predicted: ants







