### Master Seminar: Practical Applications of Deep Learning

Joseph Bethge, Ziyun Li, Hendrik Rätz, Jona Otholt, Gregor Nickel, PD Dr. Haojin Yang Multimedia and Machine Learning Group Chair of Internet Technologies and Systems Hasso Plattner Institute, University of Potsdam

# Content

- Teaching team
- Topics
- Important information



Joseph Bethge, M.sc

- Research background
  - 2010~2013 Bachelor Degree (Hasso-Plattner-Institute)
  - 2014~2017 Master Degree (Hasso-Plattner-Institute)
  - 2017~ PhD Student at Hasso-Plattner-Institute
- Research interests
  - Computer vision, deep learning, binary neural networks
  - Meanwhile promoting reproducibility and Open Source code



Hendrik Rätz, M.Sc.

#### Research background

- 2014-2017 Bachelor Degree (Hasso Plattner Institute)
- 2017-2021 Master Degree (Hasso Plattner Institute)
- 2021 PhD Student at Hasso Plattner Institute
- Research interests
  - Computer vision, self-supervised learning, text/handwriting recognition



Jona Otholt, M.Sc.

#### Research background

- 2015-2018 Bachelor Degree (Hasso Plattner Institute)
- 2018-2021 Master Degree (Hasso Plattner Institute)
- Since 2021 PhD Student at Hasso Plattner Institute
- Research interests
  - Computer vision, document analysis, unsupervised / weakly supervised learning



Gregor Nickel, M.Sc.

#### Research background

- 2013 2018 Bachelor Degree (RWTH Aachen University)
- 2017 2018 Research assistant at the Chair of Imaging and Computer Vision at RWTH Aachen University
- 2018 2020 Master Degree (RWTH Aachen University)
- 2022 PhD Student at Hasso Plattner Institute
- Research interests
  - Computer vision, optimization of the training process of deep models, efficient deep learning



Ziyun Li, M.sc

- Research background
  - 2013~2017 Bachelor Degree (Hangzhou Dianzi University)
  - 2017~2018 Master Degree (City University of Hongkong)
  - 2019~now PhD Student at Hasso-Plattner-Institute
- Research interests
  - Machine learning, semi-supervised learning, self-supervised learning, transfer learning
  - Novel class discovery, open-set recognition, out-of-distribution, label noise learning



### PD Dr. Haojin Yang

- 10/2002-01/2008 media technology (Dipl.-Ing.), University of Technology Ilmenau, Germany
- 11.2013, PhD in Computer Science (CS), Hasso Plattner Institute for Software Systems Engineering (HPI)/University of Potsdam
- 2017-present, head of Multimedia and Machine Learning Research Group, HPI
- since 07/2019, Privatdozent (PD) at HPI/University of Potsdam
  - Habilitation thesis: Deep Representation Learning for Multimedia Data Analysis
- 11/2019-10/2020, Head of Edge Computing Lab Beijing Branch, Al Labs & Video Cloud, Alibaba Group

# **Current Research Interests**

#### Decoupling the dependence of Deep Learning on high performance computing resources.

- Binary neural networks
- Deep model compression
  - Compact network design using NAS, knowledge distillation, model pruning etc.
- Efficient deep models for large language models
  - Dynamic BERT, binary BERT

#### Decoupling the dependence of Deep Learning on large-scale labeled datasets.

- Dataset synthesis
  - Text recognition, image analysis in a large-scale Art Historical Database
- Weakly supervised representation learning
- Novel class discovery, noise label

#### Edge Al

• Future paradigm: computation offloading methods, deep model partitioning, collaborative training and inference etc.

### Topic 1: QNN/BNN for Recommender Systems Motivation

# 80% of content watched on Netflix, and 60% of videos on YouTube came from recommendations.





Shuai, Zhang, et a. "Deep Learning based Recommender System: A Survey and New Perspectives." arxiv 2019.

### Topic 1: QNN/BNN for Recommender Systems Speed Up

- Quantization/Binarization reduces the runtime/computation needed
- Lower precision operations replace 32-bit floating point operations:
  - INT4 (using 4 bits)
  - Binary (using 1 bit)



Exercise 2.9 from [Müller, FMP, Springer 2015]



### Topic 1: QNN/BNN for Recommender Systems Previous Work

• Implementation in PyTorch and BITorch (<u>https://github.com/hpi-xnor/bitorch</u>)



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• Results on the Criteo Display Advertising Challenge dataset:

Model	ROC AUC	Accuracy	Size (MB)	Speed (MFLOPs)
DLRM Full Precision	0.8021	0.7884	2162	2430
binary DLRM (ours)	0.7839	0.7801	64	974

### Topic 1: QNN/BNN for Recommender Systems Goals

- Understand existing approaches for *recommender systems* based on *(binary) neural networks*
- Improve the model performance by considering quantized (4-bit) layers
- Train and evaluate a quantized/binarized DLRM for larger datasets
- Measure performance gains in practical applications using BITorch

### DOCUMENT OBJECT DETECTION



[See illustration]

728. PARCEL-GILDED REPOUSSÉ SILVEE COVERED BEAKER Cornelius Poppe, Augsburg, fl. 1705-1723 On three ball feet, loose cover with ball finial, repoussé with medallions of Roman warriors, and gilded swags of fruit and shells. (Turnovsky) Height 834, inches 729. PARCEL-GILDED REPOUSSÉ SILVER COVERED BEAKER Adolph Gaah, Augsburg, 1664-1695 Type of the preceding, repoussé with medallion heads and swags of fruit. (Tur-

novsky) Height 8 inches

730. GILDED SILVER PINEAPPLE CUP With silver floral finial and stem in the form of a tree trunk. (*Turnovsky*) *Height* 10% inches



[NUMBER 823]

823. GEORGE II FINELY CARVED MAHOGANY AND OLD GOLD SATIN DAMASK ARNCHAIR Carved balloon back with rocaille cresting, open acanthus-carved voluted arms and supports, fret-carved front rail, and shell-carved cabriole legs with clawand-ball feet; back, seat and armpads in old gold satin *damassé* with pastorales. (Conn. Private Collector) From Mallett & Son, Bath

[See illustration]

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# **DOD** - Motivation

#### Filter Documents that Contain Certain Objects



#### Preprocessing for Subsequent Analysis



# DOD - Challenge



# **DOD - Solutions**

Creation of Synthetic Datasets

**Self-Supervision** 

(Unsupervised) Domain Adaptation



Supervised Classification

Labeled Unlabeled

Known



Novel

#### Supervised Classification

Labeled Unlak

Unlabeled

Labeled Unlabeled

Novel Class Discovery





Novel





#### Supervised Classification

Labeled Unlak

Unlabeled

Labeled

**Novel Class Discovery** 

Unlabeled

#### **General Class Discovery**

#### Labeled

Unlabeled





Novel













**Balanced Dataset** 

#### Long-Tail Dataset





Goals

- Study related work on GCD and long-tail classification
- Decide on approach to tackle GCD on long-tail datasets
- Implement and evaluate approach

# **Tools and Hardware**

- Deep learning framework
  - PyTorch
- GPU servers from Multimedia & Machine Learning group

# PYTÖRCH

# **Grading Policy**

- The final evaluation will be based on:
  - Initial implementation / idea presentation, **10%** (A2.1/Zoom, 30.05.2022)
  - Final presentation, **20%** (A2.1/Zoom, 25.07.2022)
  - Report/Documentation, 12-18 pages (single column), **30%** (31.08.2022)
  - Implementation, **40%** (31.08.2022)
  - Participation in the seminar (bonus)
  - Grading

(30.09.2022)

# Enrollment/Anmelden

- 1. Registration at Studienreferat (<u>Studienreferat(at)hpi.uni-potsdam.de</u>)
- until **29.04.2022**, inform your **preferred and secondary topics** by email
  - Send email to: <u>haojin.yang@hpi.de</u>
- **03.05.2022:** Announcement of the topic and group assignment
- After: Individual weekly meeting with teaching team

# Contact

**Email**: {joseph.bethge, ziyun.li, hendrik.raetz, jona.otholt, gregor.nickel, haojin.yang}@hpi.de **Office**: H-1.11, H-1.21, H-1.22



# Thank you for your Attention!