

Federal Ministry of Education and Research

## **Machine Learning Lab Course**

**Organizational Meeting** 

lecturer: Prof. Dr. Stephan Günnemann

Summer Term 2018



#### Team

- Prof. Dr. Stephan Günnemann
- Daniel Zügner

#### This is a practical course (Praktikum) for **Master** students! *Name of module: Large-Scale Machine Learning (IN2106, IN4192)*

website: ml-lab.in.tum.de

## Why attend our Machine Learning lab course?

- 1. Get the chance to **implement and apply** state-of-the-art ML algorithms
- 2. Gain **hands-on experience** working on real-world data, solving real-world tasks (e.g. by working on one of the projects by our **industry partners**).
  - Successful projects might even qualify for a subsequent master thesis.
- 3. Work on **large-scale problems** with the support of state-of-the-art **GPU computing resources.**

### Requirements

- Requirements for the lab course
  - strong programming skills (Java, Python, C++, Java, etc.)
  - strong knowledge in data mining/machine learning
  - you should have passed relevant courses (the more, the better)
    - Mining Massive Datasets
    - Machine Learning
    - Our seminars
  - self-motivation
- Additional selection criteria
  - other **relevant** experience (projects in companies, experience as a HiWi)
    - you can send an overview of your experience to us (see end of slides)

### Organization

- Groups of 3-4 students
- Each team will work on a different project, e.g. in cooperation with one of our industry partners or on a topic they have suggested themselves
- Groups are allowed (should) collaborate!
  - exchange your experience with the other groups
  - how do the other groups tackle certain problems?
- Technical aspects:
  - each group will get exclusive access to at least one high-end GPU server with
    - 4x NVIDIA GPU w/ 11GB RAM
    - 10-core CPU
    - 256 GB RAM
  - scale up your models and data!

## Organization

- Weekly meetings (around 90-120 minutes)
  - each group should briefly report their progress, open problems, and next steps
- Regular documentation of your work
  - status reports and documentation (we might set up a wiki)
  - use of a central code repository

# Grading

- The grade is based on the whole semester's performance!
  - regular completion of **documentation**
  - regular presentations/discussions during semester
  - final presentation at the end of the semester
    - overview about what you have done, how did you implement it, what are the results, what went wrong, discussion of the framework, ...
    - each member of the team needs to present some parts

### Content

- Techniques we might want to look at (if you know these, that's good!)
  - Optimization (e.g. via gradients)
  - Stochastic optimization
  - Neural networks
  - Learning with non-i.i.d. data (e.g. temporal data)
- Tasks:
  - preprocessing
  - classification
  - profiling
  - clustering/topic mining
  - recommendation
  - anomaly detection

— ...

There are **three types** of projects in this lab course:



Industry projects

Your own projects

# **Reproduction and improvement of a published model**

- Can you spot inconsistencies in a recent publication's experimental setup?
  Can you even improve their results?
- Students can choose a recent algorithm (e.g. from ICLR 2018), and aim to reproduce and improve the results in the paper.
- Given the computational resources available to the students, they can even select large-scale models and evaluate the validity of the results and claims.
- This can also be a good way to lay the **foundation** of a new algorithm for a **master thesis**.

# Industry project: Oktoberfest food classification



- Industry partner: ilass AG, maker of software for gastronomy and party tents (e.g. Oktoberfest).
- The project will be about detecting and classifying food items on images to be extracted from a video stream.
- Representative present today: Peter Vogel





- Automatic anonymization of faces in image and video data is important to protect the privacy of people.
- Blurring or completely graying out parts in images where faces are detected means a loss of information since all facial features are removed.
- Goal: develop a method for face anonymization while preserving the most relevant facial features to still recognize basic information like emotions.





### **Industry project: Siemens**

Details to be announced.





- You can submit a **brief exposé** of your project idea provided that:
  - There is a considerable challenge from a machine learning perspective, e.g.
    non-i.i.d. data (graphs, temporal data), very noisy data, new application,
  - You have a sufficiently large and challenging dataset at hand (e.g. from an open data platform),
  - The project is suitable for a group of 3-4 students.



### **Own projects: exposé**

- The **exposé** should contain
  - a brief description of the problem and why it is important,
  - a description of the dataset you plan to use
  - a rough outline of an approach you would like to pursue
- If you are a group of students, only one student should fill in the exposé and add the others' student ID
- Max, 3,000 characters
- Submit via online form (see end of slides)

## **Registration via the matching system!**

Module name: Large-Scale Machine Learning (IN2106, IN4192)

+ fill out the application form (see next slide)



- Fill out our brief online form about your experience until 14.02.2018
  - you can provide us with a list of your experience in data mining/machine learning (courses, projects, ...)
  - please send a short overview only (bullet list); not a complete CV
  - (**optional**) attach a brief exposé of your own project idea.
- Check <u>ml-lab.in.tum.de</u> for a link to the form.

