

# Introduction to the Tutorial Sessions

# About Me - Phillip

- PhD student in the QUVA lab, working on:
  - Intersection between Causality and DL
  - Causal Representation Learning
- Alumni MSc AI at UvA
- Co-organizing the Deep Learning 1 since 2020
  - Teaching the implementation side of DL



# Schedule

- Werkcollege sessions split into:
  - Tuesday 17.00-18.00 – Tutorial session on one of the notebooks
  - Tuesday 18.00-19.00 – TA hour in one large group; Q&A
  - Fridays – TA hours in smaller groups; Q&A for assignments, lectures, etc.
- TA hours by respective TAs of your group
- Live tutorial sessions by Phillip

Monday

Tuesday

Wednesday

Thursday

Friday

15-17 Hoorcollege dr. Y.M. Asano	SP L1.02
17-19 Werkcollege Phillip Lippe MSc	SP L1.02

9-11 Hoorcollege dr. Y.M. Asano	SP L1.02
11-13 Werkcollege E: F:	SP G0.18A SP D1.115
13-15 Werkcollege A: B:	SP D1.114 SP D1.115
15-17 Werkcollege C: D: G:	SP G0.18A SP G0.18B SP D1.113

# Assignments

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- 3 Practicals on different topics aligned with the lectures
  - Published on ans-delft, code submission via Canvas
  - Repository for assignment code:  
[https://github.com/uvadlc/uvadlc\\_practicals\\_2022](https://github.com/uvadlc/uvadlc_practicals_2022)
- Practical 1: MLPs and Backpropagation
- Practical 2: CNNs, Transformers and Graph NNs
- Practical 3: Generative Models
- Compute Resources:
  - Lisa cluster
  - Google Colab

# Assignments and Tutorials

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- But there is so much more to explore... ⇒ Tutorial notebooks
  - Teaching the concepts from the lectures from an implementation perspective
  - Get familiar with PyTorch and PyTorch Lightning
  - Test your understanding by playing around with the models
- The notebooks are ungraded, but relevant for assignments and exam

# Tutorial Notebooks

- Jupyter notebooks where we implement and train our own models
- Ready to be run on Google Colab or locally by yourself
- Accessible via <https://uvadlc-notebooks.readthedocs.io/en/latest/>
- Integrated in PyTorch Lightnings' [documentation](#) and Google's [Dev library](#)

## Tutorial 7: Graph Neural Networks

Status **Finished**

Filled notebook:  Repo  View On Github  Open in Colab

Pre-trained models:  Repo  View On Github  GDrive  Download

Recordings:  YouTube  Part 1  YouTube  Part 2

In this tutorial, we will discuss the application of neural networks on graphs. Graph Neural Networks (GNNs) have recently gained increasing popularity in both applications and research, including domains such as social networks, knowledge graphs, recommender systems, and bioinformatics. While the theory and math behind GNNs might first seem complicated, the implementation of those models is quite simple and

# Tutorial Notebooks

- How to interact/learn from the notebooks:
  - Reading
  - Running code yourself
  - Video recordings
  - Live tutorial sessions on Tuesdays

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# Why PyTorch?

- There is a great variety of other DL frameworks out there



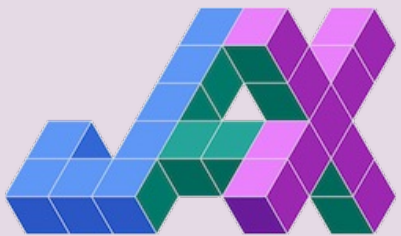
## TensorFlow

- Production-level code
- Used in companies
- Large community



## PyTorch

- Most popular in research
- Simple, easy to debug
- Large community



## JAX

- Function-oriented
- JIT-compiled, very fast
- Currently on the rise

## In this course:

- Main framework: PyTorch
- All notebooks provided in both PyTorch and JAX
- If you already know PyTorch, give JAX a try!

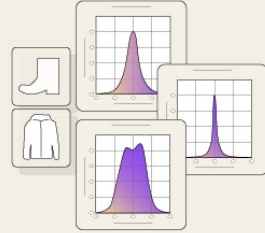


# Tutorial schedule



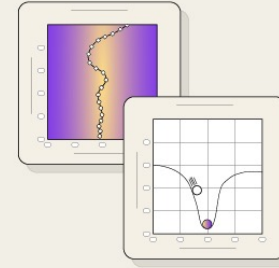
## Week 1

Introduction to PyTorch



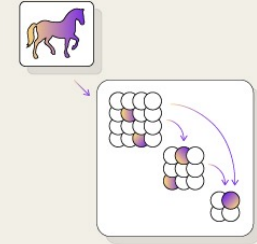
## Week 2

Activation Functions



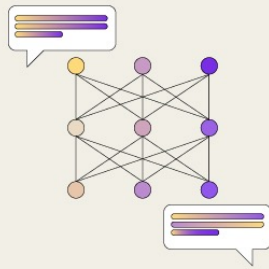
## Week 3

Optimization & Initialization



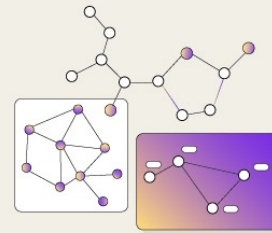
## Week 4

Inception, ResNet & DenseNet



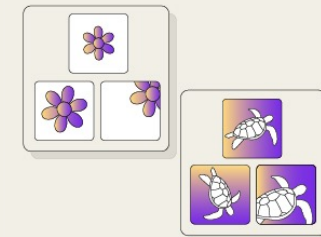
## Week 5

Transformers & Attention



## Week 6

Graph Neural Networks



## Week 7

Self-Supervised (and Causal) Representation Learning

Figure credit:  
PyTorch Lightning team

# Tutorial-Lecture Alignment

Week	Lecture - Tuesday	Tutorial - Tuesday	Lecture - Friday
Week 1	Introduction to DL	Introduction to PyTorch	Modular Learning
Week 2	DL Optimizations 1	Activation Functions	DL Optimizations 2
Week 3	Convolutional NNs	Optimization & Initialization	Modern ConvNets
Week 4	Transformers	Inception, ResNet and DenseNet	Graph Neural Networks
Week 5	Generative Modelling	Transformers and Attention	Deep Variational Inference
Week 6	Neural Modelling of 3D	Graph Neural Networks	Deep Learning for Physics
Week 7	Self-Supervised Learning 1	Self-Supervised (and Causal) Representation Learning	Self-Supervised Learning 2

# Plan for today

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- First look at notebooks
- [Tutorial 2: Introduction to PyTorch](#)
- Afterwards: Introduction to Assignment 1
  
- Friday: [Working with the Lisa cluster](#)