

# Deep Learning

project

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# Overview

1. Assessment
2. Project I - multilayer perceptron
3. Project II - convolutional neural networks
4. Project III - recurrent neural networks
5. Project IV - generative adversarial networks

- **4 projects**
- for each project: max 25 points, 4 weeks
- groups of 2 people
- first week of delay: -5 points; second week of delay: additional -10 points (-15 points in total); exceeding 2 weeks of delay results in 0 points for the project
- grades:  $[0;50] \rightarrow 2$ ,  $[51;60] \rightarrow 3$ ,  $[61;70] \rightarrow 3.5$ ,  $[71;80] \rightarrow 4$ ,  $[81;90] \rightarrow 4.5$ ,  $[91;100] \rightarrow 5$ ,

# Timetable

1.	23.02	<b>Introduction</b>
2.	02.03	Tutorial hours
3.	09.03	Initial presentation of the first part
4.	16.03	<b>Project I (first part) deadline</b>
5.	23.03	<b>Project I (second part) deadline</b>
6.	30.03	Tutorial hours
7.	13.04	Initial presentation of the second project
8.	20.04	<b>Project II deadline</b>
9.	27.04	Tutorial hours
10.	11.05	Initial presentation of the third project
11.	18.05	<b>Project III deadline</b>
12.	25.05	Tutorial hours
13.	01.06	Tutorial hours
14.	08.06	Initial presentation of the fourth project
15.	15.06	<b>Project IV deadline</b>

# Project I - multilayer perceptron

Topic: **Multilayer perceptron (MLP) employing backpropagation algorithm**

Dataset for the first part of the project will be provided during the first class.

Dataset for the second part: MNIST

<https://www.kaggle.com/c/digit-recognizer>

# Project I - multilayer perceptron

- low-level implementation is required; you can use only some basic packages like NumPy, pandas, etc.
- MLP parameters are:
  - number of hidden layers and number of neurons in hidden layers
  - activation function
  - bias presence
  - batch size
  - number of iterations
  - learning rate
  - momentum
  - problem type: classification or regression
- during the project presentation, you will be asked to train and test your network on new (unseen before) datasets
- take care of reproducibility by initializing a random number generator with a constant seed

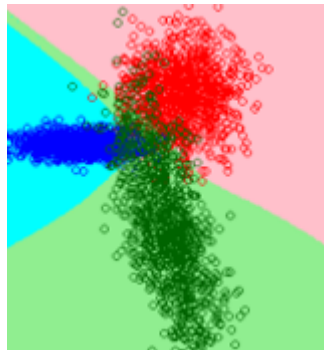
# Project I - multilayer perceptron

- elements to analyze:
  - how does activation function affect the model's accuracy? Experiment with sigmoid and two other activation functions. The activation function in an output layer should be chosen accordingly to the problem;
  - how does the number of hidden layers and number of neurons in hidden layers impact the model's accuracy? Analyze different architectures;
  - how does the loss function affect the model's accuracy? Consider two different loss functions for both classification and regression.
- the application should plot training and test error
- visualization of a training set and classification/regression result (as a background)

Assessment: source code, report

# Project I - multilayer perceptron - classification

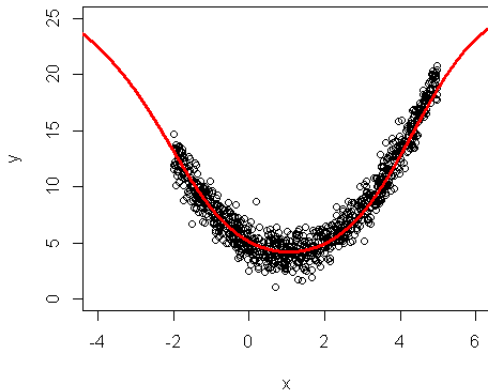
	A	B	C	D
1	x,y,cls			
2	-0.00292545510455966,0.722530109807849,1			
3	0.504570618271828,-0.789261351339519,2			
4	0.894131738692522,-0.720432524569333,1			
5	0.220206981524825,0.12407802278176,2			
6	0.267815329600126,0.925336269196123,1			
7	-0.27340721199289,0.479693677276373,1			
8	-0.287346473895013,-0.972735119983554,2			
9	0.467718373052776,0.81061793724075,1			
10	0.992102677622122,0.191424742640704,1			





# Project I - multilayer perceptron - regression

	A	B
1	x,y	
2	-5,-1253	
3	-4.99,-1247.368296	
4	-4.98,-1241.753168	
5	-4.97,-1236.154592	
6	-4.96,-1230.572544	
7	-4.95,-1225.007	
8	-4.94,-1219.457936	
9	-4.93,-1213.925328	
10	-4.92,-1208.409152	
11	-4.91,-1202.909384	
12	-4.9,-1197.426	
13	-4.89,-1191.958976	
14	-4.88,-1186.508288	
15	-4.87,-1181.073912	



# Project I - multilayer perceptron

- The second part of the project consists in fitting the implemented model to the deep learning *Hello world* dataset - MNIST

Assessment: source code, extended report (or supplement to the report)

# Project I - multilayer perceptron

## Useful resources:

- <https://www.deeplearningbook.org>
- <http://neuralnetworksanddeeplearning.com/>
- <https://www.coursera.org/specializations/deep-learning>
- <https://drive.google.com/drive/folders/0B5DSlxnH-fzR1hxd3VYOUhua2c>

# Project II - convolutional neural networks

Topic: **Image classification with convolutional neural networks**

Dataset: CIFAR-10

<https://www.kaggle.com/c/cifar-10/>

# Project II - convolutional neural networks

- you can utilize code from external sources (books, articles, blogs) provided that:
  - reference is cited in the report
  - some modifications to the original solution are applied
- violation of the above or any other kind of plagiarism result in a failing grade
- application of pre-trained models (AlexNet, VGG, etc.) is permitted (and even recommended as a part of the experiment)
- full discretion in terms of packages use
- crucial elements taken into account while assessment:
  - report quality
  - achieved accuracy registered on Kaggle
  - the originality of the solution

# Project II - convolutional neural networks

- take care of reproducibility by initializing a random number generator with a constant seed.
- to obtain statistically significant results, each experiment should be repeated multiple times.

Assessment: source code, report, presentation

# Project II - convolutional neural networks

## Useful resources:

- Google Colab or GPU to speed up training
- ensembling (soft voting or majority/hard voting)
- data augmentation
- <https://benchmarks.ai/cifar-10>
- <https://www.robots.ox.ac.uk/vgg/practicals/cnn/index.html>
- <https://adeshpande3.github.io/adeshpande3.github.io/A-Beginner's-Guide-To-Understanding-Convolutional-Neural-Networks/>
- <https://medium.com/kaggle-blog/profiling-top-kagglers-bestfitting-currently-1-in-the-world-58cc0e187b>

# Project III - recurrent neural networks

Topic: **Speech commands classification with recurrent neural networks**

Dataset: Speech Commands Dataset

<https://www.kaggle.com/c/tensorflow-speech-recognition-challenge/data>



# Project III - recurrent neural networks

- test and compare different network architectures (at least one of them should be Long short-term memory (LSTM))
- investigate influence of parameters change on the obtained results
- present confusion matrix (with appropriate discussion)
- in case of accuracy or efficiency problem a subset of classes can be selected and tested (e.g. only “yes” and “no” commands)
- please pay special attention on “silence” and “unknown” classes - test different approaches (e.g. separate network for their recognition)

Assessment: source code, report

# Project III - recurrent neural networks

## Useful resources:

- <https://www.kaggle.com/davids1992/speech-representation-and-data-exploration>
- <https://towardsdatascience.com/recognizing-speech-commands-using-recurrent-neural-networks-with-attention-c2b2ba17c837>
- <https://www.coursera.org/lecture/nlp-sequence-models/recurrent-neural-network-model-ftkzt>
- <https://pathmind.com/wiki/lstm>

# Project IV - generative adversarial networks

Topic: **Image generation with generative adversarial networks**

Dataset: 10% sample of the LSUN Bedrooms Dataset

[https://www.kaggle.com/jhoward/lsun\\_bedroom](https://www.kaggle.com/jhoward/lsun_bedroom)

You only need the data from the *sample* directory ( $\sim 4.7$ GB in total).

# Project IV - generative adversarial networks

- test and compare different network architectures (at least one of them should converge to generate satisfactory images)
- potential architectures (can use different ones): vanilla GAN with CNNs, DC-GAN, Wasserstein GAN, WGAN-GP, SN-GAN, StyleGAN1, StyleGAN2 (with or without data augmentations)
- calculate the Fréchet Inception Distance (FID) for your generated images and compare it to results from literature
- assess your results qualitatively
- investigate the influence of hyperparameters on obtained results

# Project IV - generative adversarial networks

- discuss sets of hyperparameters which help in overcoming training collapse and mode collapse
- select two of your generated images together with their latent vectors; interpolate linearly between the two latent vectors to generate 8 additional latent vectors; use these 8 vectors to generate images from your model; present the 10 generated images (8 newly generated and 2 generated previously) and discuss the importance of the results
- discuss any additional findings

Assessment: source code, report

# Project IV - generative adversarial networks

## Useful resources:

- Original GAN paper: <https://arxiv.org/pdf/1406.2661.pdf>
- DC-GAN paper: <https://arxiv.org/pdf/1511.06434.pdf>
- StyleGAN2 with data augmentations: <https://arxiv.org/pdf/2006.06676.pdf>
- FastAI on GANs (skip to relevant part): <https://course18.fast.ai/lessons/lesson12.html>
- Machine Learning Mastery (general overview):  
<https://machinelearningmastery.com/what-are-generative-adversarial-networks-gans/>
- DeepLearning.AI (for those interested; ability to access content for free):  
<https://deeplearning.ai/program/generative-adversarial-networks-gans-specialization/>