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# PROGRAM ROADMAP

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Berlin, 26.04.2021



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# AGENDA

- Intro Round
- The DSR Roadmap
- Get You Ready for Data Science
- How to Set Up a Data Science Project
- An Example of a Data Science Project
- Cheat Sheets
- Some Meetup groups (in Berlin)

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**INTRO ROUND**

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# INTRO

- About me:
  - A data enthusiast - 7 years in the field
  - Did DSR to transition from BI Analyst to Data Scientist
  - <https://www.linkedin.com/in/iskriyanavasileva/>
- About this class:
  - You can ask questions at any time
  - If you'd like me to change something during the lecture – let me know
  - If you think of something after the lecture – write to me on LinkedIn or Slack

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# INTRO

- What about you?

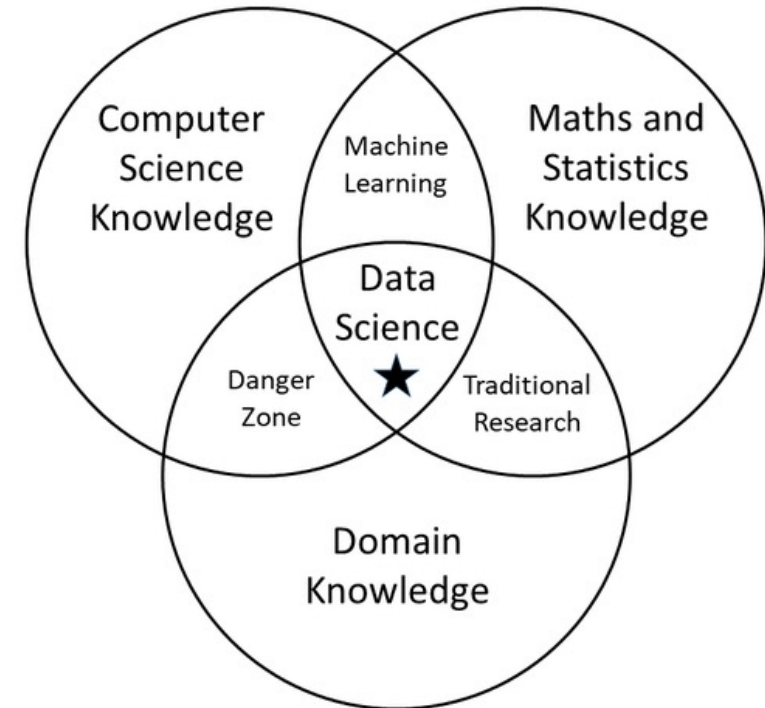
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# THE DSR ROADMAP

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# THE DSR ROADMAP

- Theoretical & Technical Fundamentals
- Machine Learning Fundamentals
- Mini Competition
- Deep Learning (or the really “cool stuff”)
- Practical Data Science
- Soft Skills
- The Final Project



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# DSR ROADMAP: THEORETICAL & TECHNICAL FUNDAMENTALS

- Python
  - Git & Bash
  - NumPy
  - Pandas
  - SQL
  - Visualisation
  - Statistics, Probability
  - Docker & Databases
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# DSR ROADMAP: THEORETICAL & TECHNICAL FUNDAMENTALS



**Purpose – the tools enabling you to work as a data scientist**



**Tips - do not underestimate them!**

Even if you know, use these lecture to refresh your skills & strengthen them

Prepare questions you encountered during your preparation for the bootcamp, but could not find the answer to them. It can be useful for everyone including your teachers



**Useful literature**

[Data Wrangling with Python](#) (Jacqueline Kazil, Katharine Jarmul)

[Python for Data Analysis](#) (Wes Mckinney)

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# DSR ROADMAP: MACHINE LEARNING FUNDAMENTALS

- DS Fundamentals
- ML fundamentals
- Trees
- Time Series



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# DSR ROADMAP: MACHINE LEARNING FUNDAMENTALS



## Purpose

**DS Fundamentals** – helps you structure a data science project from a data perspective. You get to do steps such as exploratory data analysis (EDA), data cleaning & preparation (fill in NULL values, oversampling etc.), feature engineering, model development & training, model testing, results delivery

**ML Fundamentals** – here you go through the “classics” of ML such as linear regression, logistic regression, evaluation metrics of classification etc.

**Trees & Time Series** – dedicated lectures only for these topics



## Tips – absolute must-haves

Some HRs may have questions with ready answers to filter out candidates already during the screening interview

If you get the chance, go through the notebooks in advance & identify questions or potential points you have to optimise

The best way to internalise these concepts is by coding – take a look at projects on Kaggle, towardsdatascience.com and re-do them. Even if you copy code, type it! Speaking from experience – it really does make a difference, as it makes you aware of things you would not notice if only copy-paste-ing!



## Useful literature

[The Elements of Statistical Learning](#) (Jerome H. Friedman, Robert Tibshirani, & Trevor Hastie)

[Data Analysis and Data Mining: An Introduction](#) (Adelchi Azzalini & Bruno Scarpa)

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# DSR ROADMAP: MINI COMPETITION

- 3 days of real data science work 🥰💻💻



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# DSR ROADMAP: MINI COMPETITION



**Purpose – during the competition you get to experience:**

**How to work in a team** – both technically (repository) & conceptually (task distribution, tackling dependencies, data strategy)

**The different stages of a data science project** – data exploration, data cleaning & preparation, feature engineering, model choice, model training, model testing & results delivery

**Working on a realistic project**



**Tips**

Enjoy it!

If you have the time after that, develop a clean solution on your own. It can be a good template for future reference

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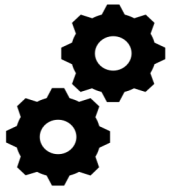
# DSR ROADMAP: DEEP LEARNING

- Backpropagation
  - Sequences
  - Deep Learning Basics
  - Natural Language Processing (NLP)
  - Computer Vision
  - Transfer Learning
  - Reinforcement Learning
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# DSR ROADMAP: DEEP LEARNING



## Purpose – this is where the real fun starts! Most probably this is why you chose this bootcamp

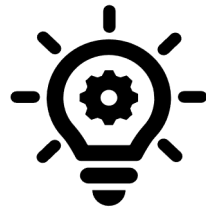
**Backpropagation** – central in deep learning – a mathematical technique for quickly calculating derivatives (the gradients).

**Deep Learning Basics** – shows you the “deep” in deep learning - layers of increasingly meaningful representations. The number of layers define the depth of the model.

**Computer Vision** – image classification using convolutional layers, data pipelining (processing, augmentation, shuffling, batching), regularisation, transfer learning or how to use pre-trained models, sound processing as spectrograms and as time series, variational autoencoders, GANs, briefly about image semantics

**Natural Language Processing (NLP)** – how to do a sentiment analysis with bag-of-words, non-DL approach to it (for e.g. using Gaussian NB), simple DL approach (for e.g. building Dense-Dropout layers using Sequential), what are word embeddings, LSTM, briefly about text generation, transfer learning and attention-based models

**Reinforcement Learning** - an agent receives information about its environment and learns to choose actions that will maximize some reward. For instance, a neural network that “looks” at a video- game screen and outputs game actions in order to maximize its score can be trained via reinforcement learning.



## Tips

If you can, prepare before the lectures. Do an online course in deep learning. For ex. [Coursera's Deep Learning](#)

Definitely pay attention in these lectures – for some of you, it can be a lot to take in. Take some time after the lectures to revisit what was done during the day

[PyTorch vs. Tensorflow before](#) & [PyTorch vs. Tensorflow now](#)

Try to set a GPU on your own (see literature section). If you decide to do a project with deep learning, you will most probably need a GPU.



## Useful literature

[Deep Learning](#) (Ian Goodfellow, Yoshua Bengio, Aaron Courville) – THE BIBLE of deep learning

[Deep Learning with Python](#) (François Chollet)

[Set a GPU on AWS](#)

[Set a GPU on Google Cloud](#)

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# DSR ROADMAP: PRACTICAL DATA SCIENCE

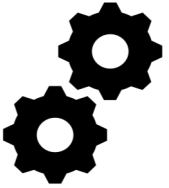
- Debugging
- Serverless
- Practical Aspects of Data Science
- Practical Aspects of Machine Learning





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# DSR ROADMAP: PRACTICAL DATA SCIENCE



**Purpose – practical advice, real-life project approach, deployment**



## **Tips**

Try not to skip these lectures. They showcase what you have to deal with once you start working as a data scientist.

Debugging is extremely useful. In this class you will also get to know PyCharm.

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# DSR ROADMAP: SOFT SKILLS

- Communication
- Career Support



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# DSR ROADMAP: SOFT SKILLS



**Purpose – what to expect at interviews, strategy & effective communication. Both teachers are extremely capable & know their stuff.**



## **Tips**

Prepare questions for the Career Support Class

Please, do respond on time for the preparation of the communication class.  
Kevin prepares accordingly

For the communication class, pick a topic / situation that was challenging for you. You will get very useful feedback & actionable advise.

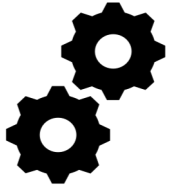
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# DSR ROADMAP: THE FINAL PROJECT



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# DSR ROADMAP: THE FINAL PROJECT



## **Purpose**

”Probieren geht über Studieren” (“The proof of the pudding is in the eating.”) - practice all you have learned so far.

Data science is about doing and showcasing your skills with meaningful projects:

- you found a topic, for which data science offers a solution / optimisation / invention
- you dealt with the issue of finding or creating data
- you delivered some kind of result

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# DSR ROADMAP: THE FINAL PROJECT - HOW TO START RESEARCH ON FINDING ONE



## **Disclaimer – no golden formula for this**

One way is to pick a topic, that addresses a challenge in an industry, in which you later want to work

- Batch 25 “Sound of failure” reduce industrial downtime by diagnosing the failure of machines using their acoustic footprint;
- Batch 24 “CAN DEEP LEARNING RELIABLY PREDICT THE CONSUMPTION AND GENERATION OF RENEWABLE ENERGY?”

Address an issue that can solve an every-day problem of many

- Batch 25 “Check your pose” an app that detects your pose and provides you with feedback on the proper posture
- Batch 22 “Deep Food” – an app that recognises ingredients from your fridge and gives you recipe suggestions

Take a look at existing research and see if you could improve it or it could inspire you to build upon it

- [Paperswithcode](#) – it will provide you with trending machine learning research and the code to it

Take a look at business magazines (for e.g. The Economist, Harvard Business Review) and see what are popular data topics

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# DSR ROADMAP: THE FINAL PROJECT



## Tips

Data availability is crucial – you can either find a good data set, generate it and / or use transfer learning. Either way – make sure you start early enough with gathering the data

The most complex model is not always the best one. Start with simple & quick solutions to test your approach

Your results may not always be what you'd expect them to be. However, the way to them is also interesting. Share this!

Idea for a presentation structure:

- Why did you choose this project – motivation, use cases, background info
- Data used and interesting aspects, challenges and how did you tackle them
- Deep dive into the models used – what models did you try, which one is the final one and why, final model performance
- Next steps
- Demo / Recording

If you need a GPU, try to set up one before that

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**GET YOU READY FOR DS**



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# GET YOU READY FOR DS

- [Shell](#)
  - A computer program which exposes an operating system's services to a human user or other program. It is named a shell because it is the outermost layer around the operating system. You can steer Anaconda, Python, Git etc. with it.
  - You access it via a [terminal emulator](#) – ex. for macOS - Terminal, iTerms. [List of Terminals](#)
  - command-line interface shells use specific scripting language - bash / zsh
- [Anaconda](#) – provides you with Python 🐍 and with some useful application around them such as PyCharm, Jupyter Notebook & Lab
- [Git](#) – a version control tool. “Version control is a system that records changes to a file or set of files over time so that you can recall specific versions later.”
- Integrated Development Environment (IDE) – [PyCharm](#), [Spyder](#)
- [Jupyter Lab / Jupyter Notebook](#) – Interactive data science environment – more visual than IDE, therefore it is used for educational & presentation purposes

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# GET YOU READY FOR DS

- Useful:
  - [How to Set Up a Data Science Project](#)
  - Terminal for Mac users – [iTerms \(features\)](#)
  - [Terminal modification](#), if you'd like it to be more colourful
  - [Bash vs. zsh](#)
  - [Difference between conda and pip](#)
  - [Git in a nutshell](#)
  - [Fundamentals of computing & programming](#)

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# HOW TO SET UP A DATA SCIENCE PROJECT

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# HOW TO SET UP A DATA SCIENCE PROJECT

Make sure you have [Git](#), [Anaconda](#) & [PyCharm](#) installed and ideally have a [GitHub account](#).

Let's set a Data Science Project from scratch by following the instructions here: <https://github.com/Iskriyana/dsr-teaching-setup>



- Maintain a learning structure - all of the notebooks can be used for future reference while you are preparing for an interview or when you are working on a future task
- My approach
  - Environment per lecture – most of the teachers already have it in their prep instructions
  - Folder / Repo per lecture
  - Notes per lecture
  - Bookmarks per lecture

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A DS PROJECT

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# A DS PROJECT

<https://github.com/Iskriyana/nlp-product-sentiment-classification>

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# CHEAT SHEETS

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# CHEAT SHEETS

- [Anaconda:](#)
- [ohmyzsh:](#)
- More:
  - <https://github.com/Iskriyana/dsr-teaching-setup/tree/main/cheatsheets>
  - <https://github.com/ADGEfficiency/programming-resources/tree/master/cheat-sheets>



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# MEETUPS

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# MEETUPS IN BERLIN

- [Data Science Retreat](#)
- [Berlin DataTalks Club](#) & [their slack](#)
- [Berlin Machine Learning Group](#)
- [meetup.ai](#)
- [Deep Learning Würzburg](#)
- [PyData](#)
- [Google Developer Group](#)
- [Berlin Computer Vision Group](#)
- [Advanced Machine Studying Group](#)
- [Women Who Code Berlin](#)
- [PyLadies Berlin](#)
- [Women Techmakers Berlin](#)

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THANK YOU & HAVE FUN!

