

Machine Learning

Introduction. ML History

Aleksandr Petiushko

ML Research



Content

① Introduction

Content

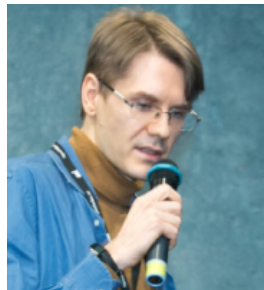
- ➊ Introduction
- ➋ Course logistics and syllabus

Content

- ① Introduction
- ② Course logistics and syllabus
- ③ Historical reference

About the lecturer¹

- Aleksandr Petiushko, PhD in theoretical CS (2016)
- Lecturer in Lomonosov MSU / MIPT for Machine Learning, Computer Vision, Deep Learning Theory, Python for an ML Researcher since 2019
- Former Huawei Chief Scientist (Scientific Expert), AIRI Director of Key Research Programs (Leading Scientific Researcher)
- Currently at Nuro, leading the ML Research



¹Homepage: <https://petiushko.info/>

Intro

Time to introduce yourselves: what are your hobbies, motivation in ML, etc.: please go into “**Module 1 Students Introduction**” thread

Sofia Plagiarism Policy

- It covers parts “*sourced from AI*”
 - ▶ Please read the “**Sofia Plagiarism Policy**” thread
 - ▶ **First offense:** students need to rewrite assignment
 - ▶ **Second offense:** students fail the course
 - ▶ **Third offense:** students are to be withdrawn from their program

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 - ▶ It can produce very different information in comparison to the source used to ask question (e.g., book chapter)

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- Only the answers with some non-trivial arguments that contradict the initial post will be considered as graded ones

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- Preliminary grading scale:

Grade	Percent accumulated
A	90-100 %
B	75-89 %
C	60-74 %

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 - ▶ **Math:** for research and design of ML algorithms
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- Hopefully we could touch on both a little

- Course page: <https://github.com/fatheral/sofia-ml-2024>
- The professor's lectures will be uploaded there

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Natural Intelligence (human)

- Able to perceive the information, analyze it, make decisions based on this analysis

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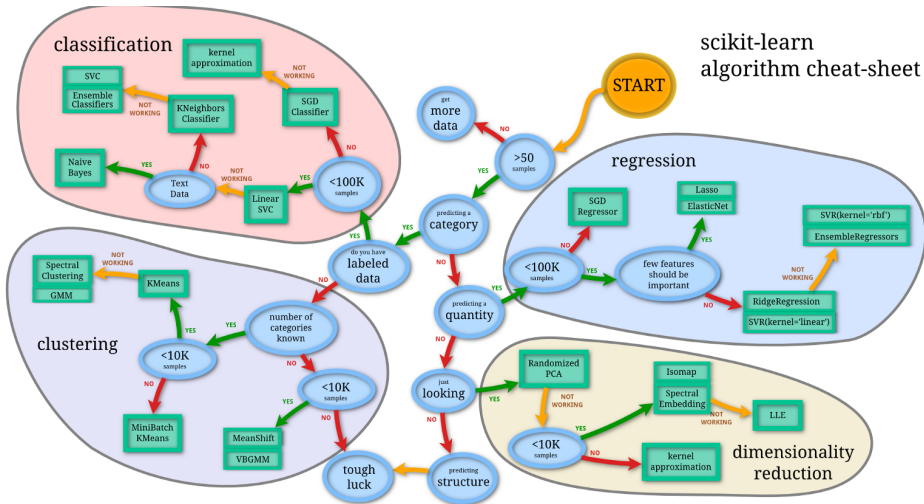
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Artificial Intelligence

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- (**Weak**) Algorithm which is able to be trained using the input data in order to do tasks afterward — instead of human

Scikit-Learn² Roadmap



²https://scikit-learn.org/stable/tutorial/machine_learning_map/

(Tentative) future content

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Practice part

- Data processing and analysis by Python
 - Scikit-Learn, Numpy, ...

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In 1997 Tom M. Mitchell introduced more formal definition of a machine learning algorithm.

Formal definition

A **computer program** is said **to learn** from examples E for some set of problems T and a quality metric P if its performance on problems from T , as measured by P , is improved by using examples E .

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- **1906**: Andrey Andreyevich Markov develops the apparatus of Markov chains, which in **1913** he uses to study the text “Eugene Onegin”. Markov chains are used to generate and recognize signals.

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- **1997:** The Deep Blue computer beat world chess champion Garry Kasparov.

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- **2022:** OpenAI, a (not so) non-profit research company, provided the breakthrough in LLMs: ChatGPT.

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Definitions

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- Reinforced
 - Action generation based on interaction with the environment

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- 4 ML History is intriguing!

Thank you!