Machine Learning

(Hoc máy – IT3190E)

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2023

Contents

Lecture 1: introduction to Machine Learning

- Lecture 2: linear regression
- Lecture 3: classification and kNN
- Lecture 4: random forest
- Lecture 5: neural networks
- Lecture 6: support vector machines
- Lecture 7: clustering with K-means
- Lecture 8: ensemble Learning
- Lecture 10: model assessment & selection
- Lecture 11-12: probabilistic models
- Lecture 13: reinforcement learning

About the course

Period: 16 weeks

□ Lectures: 12-13 weeks □ Project report: 3-4 weeks

Lecture directory: https://users.soict.hust.edu.vn/khoattg/lectures/IT3190E-ML/

https://www.youtube.com/watch?v=ic1wo 8VA1w&list=PLaKukiQCR56ZRh2 cAkweftiZCF2sTa11 &index=1

- Time & location:
- Cyber security: 12:30-15:50, D8-106, Wednesday
- DSAI: 12:30-15:50, B1 lecture hall, Thursday
- Question + advice: khoattq@soict.hust.edu.vn
- Join and discuss somethings with us: http://www.facebook.com/groups/1578056932500777/

Goals of the course

Help students to have a good basic background on Machine Learning (ML).

Identify the main advantages and limitations of the methods/models in ML.

Be able to design & implement an ML-based system and evaluate its performance.

Some technologies/libraries

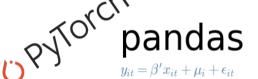


















Capstone Project

- Students work in groups, each consists of 3-5 students.
- Each group choose a problem/topic to be solved, datasets to be used, algorithms in ML.
- Each proposal should be precisely described
- The problem: short description, input, output, data type, future application, ...
- The algorithms or tools, planned to be used
- Data sets to be used
- Project registration: before 10/05/2022
 - Via Google Form (TBA)

Evaluation (đánh giá)

- Attendance and activeness
- Midterm test: Capstone Project
- Final exam
- Online test or Paper-based test
- Overall: Midterm test (40%) + Final exam (60%)

Capstone Project: requirements

- The result will be presented in the ending period of this subject. Every member is required to contribute to his/her project.
- Project report:
- Source code: save your code into one zip file
- **Readme.txt**: describes clearly how to setup, compile, and run your code
- Written report:
 - Introduce the problem to be solved, the data sets were used
- Details about the methods for analyzing data
- Results of different evaluations, new conclusions/findings, ...
- The main components of your code
- The difficulties in this project, and your proposed solution
- ...

Capstone Project: evaluation

- The evaluation of each project will be based on
- The difficulty of the problem of interest
- The appropriateness & quality of the chosen method/solution
- The rigor of the empirical evaluation and assessment on the chosen method/solution
- The quality of the presentation
- The quality of the written report
- Each project will have 15' for slide presentation & demo
- If you use some existing libraries/packages/codes, you have to clearly declare your usage in the written report and slide presentation

Some references

- Lecture slides + Youtube
- Reference books:
- □ T. M. Mitchell. Machine Learning. McGraw-Hill, 1997.
- Trevor Hastie, Robert Tibshirani, Jerome Friedman. The Elements of Statistical Learning. Springer, 2017.
- Ian Goodfellow, Yoshua Bengio, and Aaron Courville. Deep Learning. MIT press, 2016.
- □ E. Alpaydin. Introduction to Machine Learning. The MIT press, 2020.
- Software:
- □ Scikit-Learn (http://scikit-learn.org/)
- Data for experiments:
- UCI repository: http://archive.ics.uci.edu/ml/