Machine Learning

(Học máy – IT3190E)

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About the course

- Period: 16 weeks
 - □ Lectures: 12-13 weeks
 - Project report: 3-4 weeks
- Lecture directory: https://users.soict.hust.edu.vn/khoattq/lectures/IT3190E-ML/

https://www.youtube.com/watch?v=jc1wo_8VA1w&list=PLaKukjQCR56ZRh2cAkweftiZCF2sTg11_&index=1

- Time & location:
 - 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/

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

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





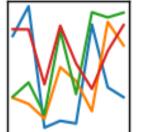


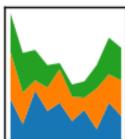




 $\sum_{y_{it}=\beta'x_{it}+\mu_i+\epsilon_{it}} \mathbf{pandas}$







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

- Students work in groups, each consists of 4-6 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 1/04/2024
 - Via Google Form (TBA)

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/