

INFO 7390 Advances in Data Sciences and Architecture

Course Information

Course Title: Advances in Data Sciences and Architecture

Course Number: INFO 7390 Term and Year: Spring 2022

Credit Hour: 4

Course Format: Online

Instructor Information

Full Name: Ramkumar Hariharan

Email Address: r.hariharan@northeastern.edu

Course Prerequisites

Graduate level INFO 6105 Minimum Grade of B

Course Description

Covers a wide range of skills and responsibilities that are necessary for managing complex business performance and operational data. Such data tend to be fragmented, poorly organized, and often flawed. Offers students an opportunity to learn how a more up-to-date mapping of complex data works and to be alerted to the care and attention they must give to such a task as well as the implications of the results. Covers best practices for managing all aspects of the data transformation life cycle, covering broad areas such as requirements gathering, meta-model design, data integration and transformation, as well as implementation and ongoing operations. Discusses tools for mapping fragmented data into business intelligence solutions that guide successful strategies.

Standard Learning Outcomes

Learning outcomes common to all College of Engineering Graduate programs:

- 1. An ability to identify, formulate, and solve complex engineering problems.
- 2. An ability to explain and apply engineering design principles, as appropriate to the program's educational objectives.
- 3. An ability to produce solutions that meet specified end-user needs with consideration of public health, safety, and welfare, as well as global, cultural, social, environmental, and economic factors.

The Information Systems Program accepts students of different engineering backgrounds with minimum programming skills and produces first class Information Systems engineers that operate at the intersection of real-world complexity, software development, and IT management. Graduating students will be able to construct end-to-end advanced software applications that meet business needs.

Specific Learning Outcomes for the Information Systems program:

1. Create a strong technical foundation through diverse, high-level courses

- 2. Built crucial interpersonal skills needed to succeed in any industry
- 3. Foster a deep level of applied learning through project based case studies

Course Outcomes and Assesment Standards

Are you looking to acquire practical data science skills in areas like machine and deep learning? Maybe, you are looking to learn some fundamentals, at the same time applying them on new problems involving big data? Or you are thinking about how best you can hone your existing skills towards becoming a professional data scientist? If you are then, check out INFO 7390. Since we have a very different teaching philosophy, individuals from a variety of backgrounds and strengths will find this course useful. This course will be taught via not just online videos but also occasional live coding videos!

The goal of this course is to foray into machine learning by start doing a project from day one of the course. We will cover basic and advanced machine learning with focus on deep learning topics.

Students enrolled in this course can expect to learn both theory and practice of modern data science in the context of big data, in a very interactive and intuitive manner. After completing this course, participants will be able to build effective ML pipelines for solving industry focused problems.

What will you learn in this course?

- 1. We start off by doing a pre-defined machine learning project selected by the instructor or by you.
- 2. Students are expected (with guidance from instructor) to gather data, preprocess it, build deep learning model and evaluate performance.
- 3. We focus on aspects of big data including choice of algorithms, compute environment, etc....
- 4. Understand the central principles of machine learning, recommender systems, and transfer learning from a conceptual, mathematical, and programming point of view.
- 5. Understand the why and how of data pre-processing for machine learning.
- 6. How to do all of the above in a cloud computing framework (Google Colab or the like)
- 7. Machine learning best practices for the industry as it's currently being practiced.
- 8. Where to go from here to further your data science career goals.

Course Syllabus

All these topics will be discussed from a project-centric point of view.

Course introduction — project expectations, presentation, writing up an academic paper based on this project, etc.

Unsupervised learning — importance, clustering, K-means clustering, feature importance in the context of clustering.

Introduction to artificial neural networks and deep learning — why are they popular, examples, their relation to linear algebra. Different neural network architectures.

Basics of linear algebra and differential calculus. Neural networks as successive transformations of the input vector.

Convolutional neural networks — theory and practice. Using excel to understand convolutions. Transfer learning with convolutional neural networks. How to use variable learning rates for transfer learning. Image augmentation techniques and an understanding of some common pretrained deep neural network models (Vgg 19, Resnets, etc.).

Components of neural networks — forward of activations, error calculation, back- propagation of the error gradients, weights updating. Activation functions - logistic, tanh, ReLU, and softmax. Loss functions — binary and categorical cross-entropy.

Introduction to Keras, tensor flow and pytorch deep learning libraries. Stochastic gradient descent, learning rate, and the loss function landscape for deep neural networks. Saddle Points.

Convolutional neural network — theory and practice. Using excel to understand convolutions. Transfer learning with convolutional neural networks. How to use variable learning rates for transfer learning.

Recommender systems: theory and practice. Collaborative filtering, content-based filtering and hybrid approaches. Live coding with the surprise package.

Basics of Natural Language Processing (NLP). Data preprocessing (removing stop words), tokenization, numericalization, language models and classification models. Concept of embeddings and their importance.

How is the course going to be delivered?

This course is organized as a series of live streaming and online videos.

How are you going to be graded?

There will be one assignment. Your performance on the assignments will contribute 50% towards your final grade.

You will be asked to submit code that goes all the way from data pre-processing to final results. Machine learning model performance plots are very important. Scores on your project will contribute 50% towards your final grade for this course.

Grade Scale

| 96-100% | Α | 87-90.9% | B+ | 77-79.9% | C+ | 69.9% or below F |
|----------|----|----------|----|----------|----|---------------------|
| | | 84-86.9% | В | 74-76.9% | С | |
| 91-95.9% | A- | 80-83.9% | B- | 70-73.9% | C- | |

Attendance Policy

This is an online course and you can take it at your schedule.

Late Work Policy

Students must submit assignments by the deadline in the time zone noted in the syllabus.

Students must communicate with the faculty prior to the deadline if they anticipate work will be submitted late.

Work submitted late without prior communication with faculty will not be graded.

Course reviews by previous INFO 6105, INFO 7390 and CSYE 7245 students

"Ram's INFO 7390 is one of my favorite courses at NEU" — Spring 2020 student "Ram is one of the friendliest professors I have come across..." — Spring 2020 student

"Ram can teach machine learning to my grandma and she will completely understand it" — Spring 2019 student

"I thoroughly enjoyed the course. would 100% recommend your course to anyone interested in starting out with Data Science" — Summer 2019 student

"I found the course to be very interesting as its design is very simple and understandable" — Summer 2019 student

"Used techniques from your course for my data science internship. Thank you" — Spring 2019 student

How to ask for help and other benefits

Ram and TA's will be available by email throughout the duration of this course and will gladly help out students.

Perks: for active data science job seekers, Ram will be happy to leverage his professional network to pass along CVs of students! This has resulted in some of his previous students landing jobs, or sometimes getting interviews from companies!

Text Books

These are a few suggestions. Please remember that we made this course from scratch and we will not follow any single textbook!

Practical treatment

 Introduction to Machine Learning with Python: A Guide for Data Scientists, Andreas C. Müller and Sarah Guido

- Python Machine Learning: Machine Learning and Deep Learning with Python, scikit- learn, and TensorFlow, 2nd Edition, Sebastian Raschka and Vahid Mirjalili
- Theoretical treatment
- An Introduction to Statistical Learning: With Applications in R, Daniela Witten, Gareth James, Robert Tibshirani, and Trevor Hastie (Legally free e-book here: https://www-bcf.usc.edu/~gareth/ISL/)
- The Elements of Statistical Learning, Jerome H. Friedman, Robert Tibshirani, and Trevor Hastie (legally free e-book here: https://web.stanford.edu/~hastie/ElemStatLearn/)
- Pattern Recognition and Machine Learning, Christopher Bishop

End-of-Course Evaluation Surveys

Your feedback regarding your educational experience in this class is very important to the College of Professional Studies. Your comments will make a difference in the future planning and presentation of our curriculum.

At the end of this course, please take the time to complete the evaluation survey at https://neu.evaluationkit.com. Your survey responses are completely anonymous and confidential. For courses 6 weeks in length or shorter, surveys will be open one week prior to the end of the courses; for courses greater than 6 weeks in length, surveys will be open for two weeks. An email will be sent to your HuskyMail account notifying you when surveys are available.

Academic Integrity

A commitment to the principles of academic integrity is essential to the mission of Northeastern University. The promotion of independent and original scholarship ensures that students derive the most from their educational experience and their pursuit of knowledge. Academic dishonesty violates the most fundamental values of an intellectual community and undermines the achievements of the entire University.

As members of the academic community, students must become familiar with their rights and responsibilities. In each course, they are responsible for knowing the requirements and restrictions regarding research and writing, examinations of whatever kind, collaborative work, the use of study aids, the appropriateness of assistance, and other issues. Students are responsible for learning the conventions of documentation and acknowledgment of sources in their fields. Northeastern University expects students to complete all examinations, tests, papers, creative projects, and assignments of any kind according to the highest ethical standards, as set forth either explicitly or implicitly in this Code or by the direction of instructors.

Go to http://www.northeastern.edu/osccr/academic-integrity-policy/ to access the full academic integrity policy.

Student Accommodations

Northeastern University and the Disability Resource Center (DRC) are committed to providing disability services that enable students who qualify under Section 504 of the Rehabilitation Act and the Americans with Disabilities Act Amendments Act (ADAAA) to participate fully in the activities of the university. To receive

accommodations through the DRC, students must provide appropriate documentation that demonstrates a current substantially limiting disability.

For more information, visit http://www.northeastern.edu/drc/getting-started-with-the-drc/.

Library Services

The Northeastern University Library is at the hub of campus intellectual life. Resources include over 900,000 print volumes, 206,500 e-books, and 70,225 electronic journals.

For more information and for Education specific resources, visit http://subjectguides.lib.neu.edu/edresearch.

24/7 Blackboard Technical Help

For immediate technical support for Blackboard, call 617-373-4357 or emailhelp@northeastern.edu

Within Blackboard, open a support case via the red support button on the right side of the screen, click Create Case

myNortheastern, e-mail, and basic technical support
Visit the Information Technology Services (ITS) Support Portal

Email: help@northeastern.edu

ITS Customer Service Desk: 617-373-4357

Diversity and Inclusion

Northeastern University is committed to equal opportunity, affirmative action, diversity and social justice while building a climate of inclusion on and beyond campus. In the classroom, member of the University community work to cultivate an inclusive environment that denounces discrimination through innovation, collaboration and an awareness of global perspectives on social justice.

Please visit http://www.northeastern.edu/oidi/ for complete information on Diversity and Inclusion

TITLE IX

Title IX of the Education Amendments of 1972 protects individuals from sex or gender-based discrimination, including discrimination based on gender-identity, in educational programs and activities that receive federal financial assistance.

Northeastern's Title IX Policy prohibits Prohibited Offenses, which are defined as sexual harassment, sexual assault, relationship or domestic violence, and stalking. The Title IX Policy applies to the entire community, including male, female, transgender students, faculty and staff.

In case of an emergency, please call 911.

Please visit <u>www.northeastern.edu/titleix</u> for a complete list of reporting options and resources both on- and off-campus.