

1. Course number and name:

COE 594-xx– Big Data and Analytics (3-0-3)

2. Credits and contact hours

3 – 0 – 3

3. Prerequisites and/or co-requisites

Prerequisites: consent of instructor

4. Name and Contact Information of Instructor:

Instructor: Dr. Imran Zualkernan
Office: EB1-260
Phone: Ext. 2892
Email: izualkernan@aus.edu
Office Hours: TBA

5. Course Description (Catalog Information)

Covers the end-to-end process for big data analytics including Extract-Transform-and-Load (ETL), descriptive analytics, supervised and un-supervised learning methods, deep learning, and big data storage and clustering eco-systems. Includes model evaluation techniques.

6. Textbook, title, author and year:

Primary:

Readings, excerpts from book chapters, and notes.

Secondary:

Ian Goodfellow and Yoshua Bengio and Aaron Courville, *Deep Learning*, MIT Press, 2016.

Tan, Steinbach, Kumar, *Introduction to data mining*, Pearson New International Edition. First edition, Essex Pearson, 2014.

Hadley Wickham, Garrett Grolemund, *R for Data Science: Import, Tidy, Transform, Visualize, and Model Data*, O'Reilly Media, 2016.

François Chollet with J. J. Allaire, *Deep Learning with R*, Manning Publications, January 2018.

Foster Provost and Tom Fawcett, *Data Science for Business: What you need to know about data mining and data-analytic thinking*, O'Reilly Media; 1 edition (July 27, 2013).

Thomas H. Davenport and Jeanne G. Harris, *Competing on Analytics: The New Science of Winning*, Harvard Business Review Press; 1 edition (March 6, 2007).

7. Learning Outcomes

This course requires student to demonstrate the following:

1. Identify and describe the various big data analytic processes.
2. Explain the components and alternatives of a big data eco-system.
3. Conduct complex data manipulation and visualization tasks on large data sets.
4. Design ETL techniques for complex and large data sets.
5. Apply and evaluate supervised learning methods large data sets
6. Apply and evaluate unsupervised learning methods for large data sets
7. Build, evaluate and optimize deep learning models for big data

8. Teaching and Learning Methodologies:

Different teaching and learning methods will be adopted in class to help students achieve the course's learning outcomes and to encourage student participation, creativity, and interaction with each other. Methods include formal lectures, class discussion, case studies, and a term project.

9. Course Topics and Schedule:

| Topic | Weeks |
|-----------------------------------------------------------------------|-----------|
| Introduction to Big Data and Analytics – Case Studies | 1 |
| Introduction to Data Analysis and Visualization | 1 |
| Big Data Storage - Hadoop, Hive, RDFS and Spark | 2 |
| Extract Transform and Load (ETL) | 2 |
| Supervised Learning (Logistic regression, SVM, Bayes, Decision Trees) | 2 |
| Unsupervised Learning (K-Means, Gaussian Mixtures) | 1 |
| Evaluating Supervised and Unsupervised Learning Models | 1 |
| Deep Learning (CNN, RNN) | 2 |
| Explanation tools for Deep Learning Models | 1 |
| Optimization for Deep Learning Models | 1 |
| Evaluation | 2 |
| Total: | 16 |

10. Schedule of Laboratory and other Non-Lecture Sessions:

None

11. Out-of-Class Assignments with Due Dates:

Each homework will consist of a hands-on application of the techniques learned to data.

| Assignment | Due Date (tentative) |
|------------|----------------------|
| HW-1 | Week 3 |
| HW-2 | Week 5 |
| HW-3 | Week 7 |

| | |
|------|---------|
| HW-4 | Week 9 |
| HW-5 | Week 11 |

12. Student Evaluation:

| Assessment | Weight | Due Date (tentative) |
|-------------------------------------|--------|------------------------|
| Homework Assignments | 50% | Once every two weeks |
| Final Project Report & Presentation | 20% | Evaluation Week |
| Final Examination | 30% | Final Examination Week |

13. Contribution of Course to Program Outcomes

This course contributes to the accomplishment of the following program outcomes:

| Program Outcome | Extent of Contribution |
|-----------------------------------------------------------------------------------------------------------------------------------|------------------------|
| 1. Perform research emphasizing creativity, independent learning and scientific methods in a chosen area of computer engineering. | ● |
| 2. Apply advanced mathematics and engineering knowledge in identifying, formulating and solving engineering problems. | ● |
| 3. Select and use techniques, skills and modern tools necessary for research or professional practice. | ● |
| 4. Communicate effectively. | ○ |
| 5. Recognize the need for, and engage in, lifelong learning. | ● |
| 6. Attend to professional and ethical responsibilities. | ○ |

Extent of contribution: ● high; ● medium; ○ low