FLORIDA CTLANTIC UNIVERSITY Graduate Programs—NEW COURSE PROPOSAL ¹ DEPARTMENT: DEPT. OF COMPUTER & ELECTRICAL ENGINEERING AND COMPUTER & ELECTRICAL ENGINEERING AND COMPUTER & ELECTRICAL ENGINEERING AND COMPUTER & ELECTRICAL						UGPC APPROVAL UFS APPROVAL SCNS SUBMITTAL CONFIRMED BANNER POSTED CATALOG S AND COMPUTER SCIENCE	
Recommended Course Identification: PrefixCAP Course Number6619 La (to obtain a course number, contact <u>nmaldonado@fau.edu</u>) Complete Course Title: Deep Learning				Lab Code (I	. or C)	EFFECTIVE DATE (first term course will be offered) FALL 2017	
CREDITS ² : 3	TEXTBOOK INFORMATION: Deep Learning, Ian Goodfellow, Yoshua Bengio, and Aaron Courville, MIT Press, 2016. Available online at http://www.deeplearningbook.org.						
COURSE DESCRIPTION, NO MORE THAN THREE LINES: This course teaches students basic concepts of deep learning, with applications in computer science, engineering, business and other areas. The class will cover major topics including math preliminaries, machine learning basics, deep forward networks, convolution networks, autoencoders, representation learning networks, their implementations and applications.							
PREREQUISITES *: COP3530 Data Structures and Algorithm Analysis or equivalent		Corequisites*:			REGISTRATION CONTROLS (MAJOR, COLLEGE, LEVEL)*: GRADUATES IN COMPUTER ENGINEERING, COMPUTER SCIENCE, AND ELECTRICAL ENGINEERING.		
* PREREQUISITES, COREQUISITES AND REGISTRATION CONTROLS WILL BE ENFORCED FOR ALL COURSE SECTIONS. MINIMUM QUALIFICATIONS NEEDED TO TEACH THIS COURSE:							
MEMBER OF THE GRADUATE FACULTY OF FAU AND HAS A TERMINAL DEGREE IN THE SUBJECT AREA (OR A CLOSELY RELATED FIELD) Faculty contact, email and complete phone number: Taghi Khoshgoftaar, khoshgof@fau.edu Xingquan Zhu, xzhu3@fau.edu Hanqi Zhuang, zhuang@fau.edu 561-297-3452/561-297-3413							
Approved by: Department Chair: College Curriculum College Dean UGPC Chair: Graduate College D UFS President: Provost:	Chair:	hely		Date: 	22/16 124/16 y 12010	 Syllabus must be attached; see guidelines for requirements: www.fau.edu/provost/files/course syllabus.2011.pdf Review Provost Memorandum: Definition of a Credit Hour www.fau.edu/provost/files/Definition_ Credit Hour_Memo_2012.pdf Consent from affected departments (attach if necessary) 	

materials may be viewed on the UGPC website prior to the meeting.

FAUnewcrseGrad—Revised November 2014

1. Course title/number, number of credit hours					
Deep Learning – CAP 6619		3 credit hours			
2. Course prerequisites, corequisites, and where the course fits in the program of study					
Prerequisites: COP3530 Data S	tructures and Algorithm A	nalysis or equivalent			
3. Course logistics					
<i>Term:</i> Fall 2017					
Class location and time: TBD					
4. Instructor contact informat	tion				
Instructor's name Office address Office Hours Contact telephone number Email address	Dr. Taghi Khoshgoftaar, Dr. Xingquan Zhu and Dr. Hanqi Zhuang Engineering East (EE-96) Bldg. TBD 561-297-3452/561-297-3413				
		13(@1a0.edu, 210a119(@1a0.edu			
5. TA contact information					
TA's name Office address Office Hours Contact telephone number Email address	N/A N/A N/A N/A				
6. Course description					
This course teaches students basic concepts of deep learning, with applications in computer science, engineering, business and other areas. The class will cover major topics including math preliminaries, machine learning basics, deep forward networks, convolution networks, autoencoders, representation learning networks, their implementations and applications. Note: Deep learning is a field that studies deep artificial neural networks composed of many hidden layers. It is made possible due to available big datasets and fast computing power. Many favorable results in applications have been obtained, especially for applications where the cost function is complex, the datasets are huge, and labels may not be completely available.					
7. Course objectives/student learning outcomes/program outcomes					
Course objectives	The goal of this class is for hands-on experiences on should be able to underst algorithmic and impleme popular deep learning mo	or students to gain theoretical foundation and deep learning. At the end of the class, students cand the fundamentals of deep learning, entation details and should be able to apply odels to study their research problems.			

8. Course evaluation method	
4 Home Works (each home work is worth 10%) -	40%
Test -	30%
Project -	30%
9. Course grading scale	
Grading Scale:	

90 and above: "A", 85-89: "A-", 76-84: "B+", 70-75: "B", 66-74 : "C+", 60-65: "C", 50-59: "D", 49 and below: "F."

10. Policy on makeup tests, late work, and incompletes

Makeups are possible, and are given only if there is solid evidence of medical or otherwise family/personal emergency issues that prevent the student from participating in the exam. Makeup exam should be administered and proctored by department personnel unless there are other pre-approved arrangements

Late work is not acceptable.

A grade of incomplete will be assigned only in the case of solid evidence of medical or otherwise serious emergency situation.

11. Special course requirements

N/A

12. Classroom etiquette policy

University policy requires that in order to enhance and maintain a productive atmosphere for education, personal communication devices, such as cellular phones and laptops, are to be disabled in class sessions.

13. Disability policy statement

In compliance with the Americans with Disabilities Act (ADA), students who require special accommodations due to a disability to properly execute coursework must register with the FAU Students Accessibility Services (SAS) located in Boca Raton, Davie, and Jupiter campuses and follow all SAS procedures <u>http://www.fau.edu/sas.</u>

14. Honor code policy

Students at Florida Atlantic University are expected to maintain the highest ethical standards. Academic dishonesty is considered a serious breach of these ethical standards, because it interferes with the university mission to provide a high quality education in which no student enjoys unfair advantage over any other. Academic dishonesty is also destructive of the university community, which is grounded in a system of mutual trust and place high value on personal integrity and individual responsibility. Harsh penalties are associated with academic dishonesty. See University Regulation 4.001 at www.fau.edu/regulations/chapter4/4.001 Code of Academic Integrity.pdf

15. Required texts/reading

Deep Learning, Ian Goodfellow, Yoshua Bengio, and Aaron Courville, MIT Press, 2016. Available online at

http://www.deeplearningbook.org.

16. Supplementary/recommended readings

Research papers and forums, see explanations.

17. Course topical outline, including dates for exams/quizzes, papers, completion of reading

Weekly course topics (tentative)

Weekly schedule	Торіс	
Week 1	Introduction/Overview	
Week 2	Linear Algebra Review:	HW1
	vectors, matrices, tensors; linear dependence and	
	span; normed space; eigen-decomposition; SVD; PCA;	
Week 3	Optimization Basics:	
	Gradient-based optimization; constrained optimization	
Week 4	Probability Review:	HW2
	Probability; marginal probability; conditional	
	probability; Bayes' rule; intro to information theory;	
	estimators, bias and variance	
Week 5-6	Machine Learning Overview:	
	Learning algorithms; capacity, overfittng and	
	underfitting; validation sets; Bayesian statistics;	
	supervised learning algorithms; unsupervised learning	
	algorithms; stochastic gradient descent	
Week 7-8	Deep Forward Networks:	Hw3
	Gradient-based learning; hidden units;	
	Architecture design; differential algorithms; examples	
Week 9-10	Convolution Neural Networks:	Test, Project
	The convolution operation;	Announcement
	motivation; pooling; variants of convolution function;	
	structured output; combine CNN with DFN	
Week 11	Recurrent Neural Networks:	HW4
	Unfolding computational graphs; recurrent neural	
	networks; other variations	
Week 12-13	Autoencoder learning:	
	Undercomplete autoencoders; regularized	
	autoencoders; representational power, layer size and	
	depth; stochastic encoders and decoders; learning	
	manifold with autoencoders	
Week 14-15	Representation Learning Networks: unsupervised deep	
	learning; transfer learning and domain adaptation;	
	semi-supervised learning; distributed representation	
Week 16	Project Report	

Project: The goal of the term project is to practice the algorithms and techniques learned in class. Students will work on the project in the second half of the class. Each student will select a topic related to the material taught in class. A list of tentative topics such as image classification will be provided in class. Students are expected to conduct research in that topic, implement and validate the algorithms, and collect experimental results. At the end of the project, the students will submit a report and the programming source code for evaluation.

Research papers and forums (updates will be given from time to time):

- <u>Representation Learning: A Review and New Perspectives</u>, Yoshua Bengio, Aaron Courville, Pascal Vincent, Arxiv, 2012.
- The monograph or review paper <u>Learning Deep Architectures for AI</u> (Foundations & Trends in Machine Learning, 2009).
- Deep Machine Learning A New Frontier in Artificial Intelligence Research a <u>survey paper</u> by Itamar Arel, Derek C. Rose, and Thomas P. Karnowski.
- Graves, A. (2012). *Supervised sequence labelling with recurrent neural networks*(Vol. 385). Springer.
- Schmidhuber, J. (2014). Deep Learning in Neural Networks: An Overview. 75 pages, 850+ references, <u>http://arxiv.org/abs/1404.7828</u>, PDF & LATEX source & complete public BIBTEX file under <u>http://www.idsia.ch/~juergen/deep-learning-overview.html</u>.
- LeCun, Yann, Yoshua Bengio, and Geoffrey Hinton. <u>"Deep learning."</u> Nature 521, no. 7553 (2015): 436-444.
- <u>http://www.deeplearning.net/</u>
- <u>http://www.deeplearningbook.org/</u>