Fau	
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UNIVERSITY

## **COURSE CHANGE REQUEST Graduate Programs**

Department Computer and Electrical Eng. and Comp. Sci.

College Engineering and Computer Science

UGPC Approval	_
UFS Approval	_
SCNS Submittal	_
Confirmed	_
Banner Posted	_
Catalog	

Current Course	
Prefix and Number	CAP 5615

**Current Course Title** 

Introduction to Neural Networks

Syllabus must be attached for **ANY** changes to current course details. See <u>Guidelines</u>. Please consult and list departments that may be affected by the changes; attach documentation.

Change description to:	
Change prerequisites/minimum grades to:	
COP 3530 Data Structures and Algorithm Analysis	
Change corequisites to:	
Change registration controls to:	
Please list existing and new pre/corequisites, specify AND or OR and include minimum passing grade.	
Terminate course? Effective Term/Year for Termination:	
@fau.edu / 561-297-3452	
Date  Date  Date	

Email this form and syllabus to <a href="https://www.uGPC@fau.edu">UGPC@fau.edu</a> one week before the UGPC meeting.

GRADUATE COLLEGE

	1. Course title/number, number of credit hours			
	Introduction to Neural Networks -	- CAP5615	3 credit hours	
	2. Course prerequisites, corequis	ites, and where	e the course fits in the program of study	
	Prerequisite: COP 3530 (Data Stru	ctures and Algo	rithm Analysis)	
	3. Course logistics			
	Term:			
	Class location and time:			
	4. Instructor contact information	1		
47	Instructor's name Office address Office Hours	Xingquan (Hill Engineering E	) Zhu ast (EE-96) Bldg. <b>,</b> Room 509	
	Contact telephone number Email address	561-297-3452 xzhu3@fau.ed	<u>du</u>	
	5. TA contact information			
	TA's name Office address Office Hours Contact telephone number Email address	N/A		
	6. Course description			
	mechanisms. It also teaches stud solve real-world problems, such basic architectures of supervised	ents hands-on as digital chard and unsupervis	eural networks, neural network architectures, and learning experiences in designing/implementing neural networks to acter classification or face recognition. Students will learn sed neural networks and how neural networks can be used a will also cover deep learning networks, and deep learning	
	7. Course objectives/student lea	rning outcome	s/program outcomes	
CDAD	Course objectives	I .	damental concepts of artificial neural networks, classificat s networks, and advanced learning framework, such as de	
	E COLLEGE	2. Develop ab	ilities to analyze artificial neural networks. e basic understanding of Back Propagation for weight updat	ting
	5 2019	in neural netw		,
1.	Student learning outcomes	COLLEGE	in the areas of software design and development, data	



need	& relationship to ABET objectives structures, and operating systems 2. An ability to plan and execute an engineering design to m	neet an identified
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### 8. Course evaluation method

1.	Homework	45%
2.	Mid-Term	15%
3.	Term Project	10%
4.	Final Exam	15%
5.	Student Presentation	10%
6.	Participation	5%

### 9. Course grading scale

### Grading Scale:

 v
 90% and above
 A (including A, A-)

 v
 70-89%
 B (including B+, B, and B-)

 v
 60-69%
 C (including C+, C, and C-)

 v
 40-59%
 D (including D+, C, and D-)

v 39% and below or cheating F

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

Makeup tests are given only if there is solid evidence of a medical or otherwise serious emergency that prevented the student of participating in the exam. Makeup exam should be administered and proctored by department personnel unless there are other pre-approved arrangements

Late work is subject to late penalty.

*Incomplete grades* are against the policy of the department. Unless there is solid evidence of medical or otherwise serious emergency situation and the student is currently passing the class, incomplete grades will not be given.

### 11. Special course requirements

All homework assignments and all lab work in this course must be **INDIVIDUAL** effort. Please take the time to read the documentation. You are responsible for the information outlined in it. Please see the instructor, any teaching assistant, or Engineering Student Services tutoring for assistance. Check the Where to Find **Help** Section on Blackboard.

### 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. Attendance Policy Statement

Students are expected to attend all of their scheduled University classes and to satisfy all academic objectives as outlined by the instructor. The effect of absences upon grades is determined by the instructor, and the University reserves the right to deal at any time with individual cases of non-attendance.

Students are responsible for arranging to make up work missed because of legitimate class absence, such as illness, family emergencies, military obligation, court-imposed legal obligations or participation in University approved activities. Examples of University-approved reasons for absences include participating on an athletic or scholastic team, musical and theatrical performances and debate activities. It is the student's responsibility to give the instructor notice prior to any anticipated absences and within a reasonable amount of time after an unanticipated absence, ordinarily by the next scheduled class meeting. Instructors must allow each student who is absent for a University-approved reason the opportunity to make up work missed 3 without any reduction in the student's final course grade as a direct result of such absence.

### 14. Disability policy statement

In compliance with the Americans with Disabilities Act Amendments Act (ADAAA), students who require reasonable accommodations due to a disability to properly execute coursework must register with Student Accessibility Services (SAS) and follow all SAS procedures. SAS has offices across three of FAU's campuses – Boca Raton, Davie and Jupiter – however disability services are available for students on all campuses. For more information, please visit the SAS website at www.fau.edu/sas/.

### 15. Counseling and Psychological Services (CAPS) Center

Life as a university student can be challenging physically, mentally and emotionally. Students who find stress negatively affecting their ability to achieve academic or personal goals may wish to consider utilizing FAU's Counseling and Psychological Services (CAPS) Center. CAPS provides FAU students a range of services – individual counseling, support meetings, and psychiatric services, to name a few – offered to help improve and maintain emotional well-being. For more information, go to <a href="https://www.fau.edu/counseling/">https://www.fau.edu/counseling/</a>

### 16. 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 <a href="https://www.fau.edu/regulations/chapter4/4.001">www.fau.edu/regulations/chapter4/4.001</a> Code of Academic Integrity.pdf

#### 17. Texts/reading

### Textbook:

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

#### 18. Supplementary/recommended readings

### Reference books:

- 1. <u>Neural Networks for Pattern Recognition</u>, Christopher M. Bishop, Clarendon Press, 1996 (Online version available)
- Pattern Recognition and Machine Learning Christopher M. Bishop, Springer, October, 2007, (Online version available)

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

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Weekly Schedule	Topics	
Week 1	Introduction, R programming	
Week 2	Introduction to pattern recognition and machin	e learning
Week 3	Decision tree learning	[homework 1]
Week 4	R for decision tree learning	
Week 5	Bayes learning	
Week 6	R for Bayes learning	[homework 2]
Week 7	Single perceptron learning	
Week 8	Multi-layer feedforward neural networks	[homework 3]
Week 9	Multi-layer feedforward neural networks for fa	ace recognition
		[mid-term]
Week 10	Radial basis function networks	[homework 4]
Week 11	Support vector machines	
Week 12	Intro to deep learning	[term project]
Week 13	Convolutional neural networks	
Week 14	Student presentation	
Week 15	Student presentation and review	