

COURSE DESCRIPTION FORM

INSTITUTION FAST – National University

**PROGRAM (S) TO
BE**

EVALUATED

A. Course Description

(Fill out the following table for each course in your computer science curriculum. A filled out form should not be more than 2-3 pages.)

Course Code	CS5102
Course Title	Deep Learning
Credit Hours	3
Prerequisites by Course(s) and Topics	Undergraduate machine learning or artificial intelligence Undergraduate programming/Python
Assessment Instruments with Weights (homework, quizzes, midterms, final, programming assignments, lab work, etc.)	<ul style="list-style-type: none">• 1 Mid-term exam: 15% each• Final exam: 40%• 3 Projects: 30%
Course Coordinator	
URL (if any)	piazza.com/nu.edu.pk/fall2018/cs5102/home
Current Catalog Description	
Textbook (or Laboratory Manual for Laboratory Courses)	- Deep Learning by Ian Goodfellow, Yoshua Bengio - Stanford deep learning for visual recognition http://cs231n.stanford.edu/2017/syllabus.html
Reference Material	- Neural Networks and Deep Learning by Michael Nielsen (Dec 2014). - Pattern Recognition and Machine Learning, Christopher M Bishop, Springer 2006. - Deep learning for medical imaging @ Purdue https://docs.google.com/document/d/1zEL-nu_To7Olc3cD-dg5iADvWrErAQSJD8n-1CLrGGA/edit#heading=h.ml4r2vcdkioV

Course Goals	<ul style="list-style-type: none"> – Understand the fundamentals of neural networks – Understand deep learning with CNNs – Apply deep learning to real problems, especially in the context of vision and language processing
Topics Covered in the Course, with Number of Lectures on Each Topic (assume 15-week instruction and one-hour lectures)	<p>Learning overview:</p> <p>Introduction and motivation, biological neural network and its history, learning vs programming, learning components, supervised learning, classification vs regression</p> <p>Basics of Neural Networks:</p> <p>Perceptron, Multilayer Perceptron, Gradient Descent in MLPs, Activation functions, Back-propagation</p> <p>Convolutional Neural Networks:</p> <p>Motivation for CNNs, basics of convolution, distinguishing features of CNNs, What makes CNNs tick, deep learning with CNNs</p> <p>Deep learning - Applications:</p> <p>Applying CNNs: transfer learning, CNNs in computer vision, Recurrent NNs, Inception and GoogleNet, LSTMs and deep learning for NLP, Generative Adversarial Networks</p> <p>Practical issues: Overview of Python, GPU, Cloud-based GPU solutions, useful libraries</p>
Laboratory Projects/Experiments Done in the Course	<p>Project 1: Implementation of NN backpropagation(MATLAB)</p> <p>Project 2: Implementation of a CNN(cats vs dogs problem) in Keras Training, testing and visualization Hyperparameter tuning: optimizer, layers, parameters etc</p> <p>Project 3: Implementation/modification of DL model for:</p> <ol style="list-style-type: none"> GAN for generating skin cancer data <i>or</i> LSTM model for describing the scene of a movie



Class Time Spent on (in credit hours)	Theory	Problem Analysis	Solution Design	Social and Ethical Issues
	0.9	0.9	0.9	0.3
Oral and Written Communications	Every student is required to submit at least 3 written reports of typically 5-10 pages. These reports are NOT graded for oral and verbal proficiency beyond what is expected of a technical report.			

Instructor Name M Usman Sadiq

Instructor Signature _____

Date _____