

Subject Description Form

Subject Code	COMP4432		
Subject Title	Machine Learning		
Credit Value	3		
Level	4		
Pre-requisite / Co-requisite / Exclusion	Nil (but students are preferred to have some previous exposure to introductory Artificial Intelligence/Data Analytics concepts and be familiar with basic notions in linear algebra and probability)		
Objectives	<p>The objectives of this subject are to:</p> <ul style="list-style-type: none"> • present the basic principles, concepts and models of modern machine learning; and • introduce recent advances of machine learning technology with impactful applications in pattern recognition, computer vision and other areas. 		
Intended Learning Outcomes	<p>Upon completion of the subject, students will be able to:</p> <p>(a) understand the major concepts of machine learning models and algorithms;</p> <p>(b) develop basic intuitions into the effectiveness of machine learning techniques;</p> <p>(c) gain knowledge of applying machine learning techniques to various cutting-edge applications; and</p> <p>(d) design machine learning solutions to solve new challenging problems in practice.</p>		
Subject Synopsis/ Indicative Syllabus	<table border="1" style="width: 100%;"> <tr> <td style="text-align: left;">Topic</td> </tr> <tr> <td> <p>1. Part I: Machine Learning Fundamentals</p> <ul style="list-style-type: none"> • Linear algebra and probability • Numerical computation and optimization • Learning tasks: Regression, classification, etc. • Performance issues: Cross-validation, Overfitting and curse of dimensionality, Bias-variance dilemma, etc. </td> </tr> </table>	Topic	<p>1. Part I: Machine Learning Fundamentals</p> <ul style="list-style-type: none"> • Linear algebra and probability • Numerical computation and optimization • Learning tasks: Regression, classification, etc. • Performance issues: Cross-validation, Overfitting and curse of dimensionality, Bias-variance dilemma, etc.
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	<p>2. Part II: Models and Techniques</p> <ul style="list-style-type: none"> • Supervised learning: <ul style="list-style-type: none"> ○ Parametric vs non-parametric methods ○ Decision tree based methods ○ Bayesian models ○ Neural networks and support vector machines • Unsupervised learning: <ul style="list-style-type: none"> ○ k-means and hierarchical clustering ○ Spectral clustering and density-based clustering ○ Advanced models: autoencoder, embedding techniques, etc. • Regression and boosting • Feature selection and dimensionality reduction <p>3. Part III: Applications</p> <ul style="list-style-type: none"> • Handwriting recognition challenge, e.g. MNIST • Object detection, recognition and tracking <ul style="list-style-type: none"> ○ Object feature descriptions: Engineering approach vs feature learning approach ○ Object detection examples, e.g. pixel clustering for face detection, etc. ○ Object recognition examples, e.g. face recognition via eigenface features ○ Object tracking examples, e.g. human motion tracking 																																																			
<p>Teaching/ Learning Methodology</p>	<p>39 hours of class activities including lectures on the main concepts and models, together with applicational case studies, tutorials and class/group discussions, laboratory works and student presentations. Additional reading of research papers will be assigned, whenever appropriate.</p>																																																			
<p>Assessment Methods in Alignment with Intended Learning Outcomes</p>	<table border="1" data-bbox="384 1234 1465 1870"> <thead> <tr> <th rowspan="2">Specific assessment methods/tasks</th> <th rowspan="2">% weighting</th> <th colspan="5">Intended subject learning outcomes to be assessed (Please tick as appropriate)</th> </tr> <tr> <th>a</th> <th>b</th> <th>c</th> <th>d</th> <th></th> </tr> </thead> <tbody> <tr> <td>Continuous Assessment</td> <td rowspan="4" style="text-align: center;">55%</td> <td></td> <td></td> <td></td> <td></td> <td></td> </tr> <tr> <td>1. Assignments</td> <td style="text-align: center;">✓</td> <td style="text-align: center;">✓</td> <td style="text-align: center;">✓</td> <td></td> <td></td> </tr> <tr> <td>2. Tests/Quizzes</td> <td style="text-align: center;">✓</td> <td style="text-align: center;">✓</td> <td style="text-align: center;">✓</td> <td></td> <td></td> </tr> <tr> <td>3. Project</td> <td></td> <td></td> <td style="text-align: center;">✓</td> <td style="text-align: center;">✓</td> <td style="text-align: center;">✓</td> </tr> <tr> <td>Examination</td> <td style="text-align: center;">45%</td> <td style="text-align: center;">✓</td> <td style="text-align: center;">✓</td> <td style="text-align: center;">✓</td> <td style="text-align: center;">✓</td> <td></td> </tr> <tr> <td>Total</td> <td style="text-align: center;">100%</td> <td colspan="5"></td> </tr> </tbody> </table> <p>Explanation of the appropriateness of the assessment methods in assessing the intended learning outcomes:</p> <p>Assignment: After-class assessment of the continuous understanding of the concepts, issues, models and applications of machine learning techniques by providing answers to given questions.</p>	Specific assessment methods/tasks	% weighting	Intended subject learning outcomes to be assessed (Please tick as appropriate)					a	b	c	d		Continuous Assessment	55%						1. Assignments	✓	✓	✓			2. Tests/Quizzes	✓	✓	✓			3. Project			✓	✓	✓	Examination	45%	✓	✓	✓	✓		Total	100%					
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	<p>Test/Quiz: In-class assessment of the understanding of the concept, issues, models and applications of machine learning techniques by providing answers to given questions.</p> <p>Project: Assessment of problem solving ability in dealing with practical application problems by written reports and oral presentations.</p> <p>End-of-term Assessment: Assessment of the overall performance by a written examination.</p>	
Student Study Effort Expected	Class contact:	
	<ul style="list-style-type: none"> ▪ Lecture/Tutorial/Lab 	39 Hrs.
	Other student study effort:	
	<ul style="list-style-type: none"> ▪ Self-study 	83 Hrs.
Reading List and References	Reference Books:	
	<ol style="list-style-type: none"> 1. Hastie, T., Tibshirani, R. and Friedman, J., <i>The Elements of Statistical Learning</i>, 2nd Edition, Springer, 2009. 2. Shalev-Shwartz, S. and Ben-David, S., <i>Understanding Machine Learning: From Theory to Algorithms</i>, 2014. 3. Bousquet, O., Boucheron, S. and Lugosi, G., <i>Introduction to Statistical Learning Theory</i>, Advanced Lectures on Machine Learning. 4. Mohri, M., Rostamizadeh, A. and Talwalkar, A., <i>Foundations of Machine Learning</i>. USA, Massachusetts: MIT Press, 2012. 5. Vapnik, V. N., <i>The Nature of Statistical Learning Theory</i>. Springer, 2000. 6. Bishop, Christopher, <i>Pattern Recognition and Machine Learning</i>, Springer, 2006. 7. To be amended and updated at the beginning of the semester. 	