

## Subject Description Form

<b>Subject Code</b>	AMA4680
<b>Subject Title</b>	Statistical Machine Learning
<b>Credit Value</b>	3
<b>Level</b>	4
<b>Pre-requisite/</b>	<p>Applied Linear Models for Finance Analytics (AMA2602) or            Applied Linear Models (AMA3602) or            Statistics for Data Science (AMA3631) or equivalent  <b>and</b>            Mathematical Methods for Data Science (AMA3001/AMA3701) or            Further Mathematical Methods for Finance (AMA3723) or            Further Mathematical Methods (AMA3724) or equivalent  <b>and</b>            Programming Fundamentals and Applications (COMP1012) or            Principles of Programming (AMA2222/AMA2222A) or equivalent</p>
<b>Exclusion</b>	Statistical Machine Learning (AMA3800)
<b>Objectives</b>	To provide a basic introduction to machine learning. To present fundamental concepts and algorithms for selected topics of machine learning, to provide the students with the necessary background for the application of machine learning to real problems, and to provide a starting point for students who are interested in pursuing research in machine learning or related fields.
<b>Intended Learning Outcomes</b>	<p>Upon completion of the subject, students will be able to:</p> <ol style="list-style-type: none"> <li>a) Demonstrate mastery of the principles of machine learning</li> <li>b) Develop quantitative skills of machine learning and interpret the outcomes of machine learning algorithms.</li> <li>c) Identify, define, and formulate problems of machine learning in real applications and generate workable solutions to problems.</li> </ol>
<b>Subject Synopsis/ Indicative Syllabus</b>	<p><u>Supervised Learning, Regression and Model Selection:</u>            Generalized linear models; ridge regression, model selection, dimension reduction, principal component analysis, lasso; reproducing kernels, kernel ridge regression, regularization; stochastic gradient descent, online learning.</p> <p><u>Supervised Learning, Classification:</u>            Naive Bayes; decision trees; <math>k</math>-nearest-neighbor classifier; logistic regression; support vector machines, kernelized support vector machines; score-based classifiers, the receiver operating characteristic curve, AUC scores, imbalanced data.</p> <p><u>Unsupervised Learning, Clustering:</u></p>

	<p><i>K</i>-means; agglomerative hierarchical clustering.</p> <p><u>Other Selected Topics:</u>  Cross validation; artificial neural networks, back-propagation, introduction to deep learning algorithms; introduction to software packages for machine learning; random forests.</p>				
<b>Teaching/Learning Methodology</b>	The subject will mainly be delivered through lectures and tutorials in computer lab. The theoretical background and the real applications of learning algorithms are both emphasized.				
<b>Assessment Methods in Alignment with Intended Learning Outcomes</b>	Specific assessment methods/tasks	% weighting	Intended subject learning outcomes to be assessed (Please tick as appropriate)		
			a	b	c
	1. Assignments/Projects	15%	✓	✓	✓
	2. Midterm Test	25%	✓	✓	✓
	3. Final Exam	60%	✓	✓	✓
Total	100%				
	<p>Explanation of the appropriateness of the assessment methods in assessing the intended learning outcomes:  This subject focuses on the machine learning algorithms. Some of the algorithms concern important mathematical and statistical background. So it is important and appropriate to set the assignments and a midterm test for continuous assessment, and a final exam.  This subject emphasizes the practice of statistical machine learning. The projects are appropriate for assessing the related intended learning outcomes in the continuous assessment.</p>				
<b>Student Study Effort Required</b>	Class contact:				
	Lecture		26 Hrs.		
	Tutorial		13 Hrs.		
	Other student study effort:				
	Assignments/Projects		58 Hrs.		
	Self-study		30 Hrs.		
	Total student study effort		127 Hrs.		
<b>Reading List and References</b>	<p><u>Textbooks:</u>  Han, J., Kamber, M., and Pei, J.      Data Mining: Concepts and Techniques, 3rd Edition.      Morgan Kaufmann, 2011</p>				

James, G., Witten, D., Hastie, T., and Tibshirani, R.	An Introduction to Statistical Learning	Springer 2013
<u>References:</u>		
Tan, P.N., Steinbach, M., and Kumar, V.	Introduction to Data Mining	Pearson 2006
Hastie, T., Tibshirani, R, and Friedman, J.	The Elements of Statistical Learning	Springer 2009
Kelleher, J.D., Namee M.B., D'Arcy, A.	Fundamentals of Machine Learning for Predictive Data Analysis	The MIT Press 2015
Steinwart, I., Christmann A.	Support Vector Machines	Springer 2008
Goodfellow I., Bengio Y., Courville A.	Deep Learning	The MIT Press 2016