Subject code:	Subject name: Advanced Information Technology		
I.7(3)			
Study load: 4 ECTS	Load of contact hours: 36	Study semester: Autumn	Assessment: Credit / No credit
Objectives:	The aim of the course - to give an idea of the modern, advanced information technologies.		
Course outline: Learning Outcomes:	 Topics covered: Overview of approaches to solving the problems of classification and regression. Statistical methods. Machine learning. Basics of statistical information processing. Linear regression. The choice of dimension feature space. Logistic regression. Methods of time series analysis. Fuzzy uncertainty, fuzzy inference. Review of ensembles of classification models. Methods of decision trees. Contact lessons will be divided into two parts: lectures&practicals and information technology workshops with individual tasks. 		
	At the end of the cours advanced information to for solving regression a The student should be algorithms that are ade environment.	e, the student should ha technologies as an instru- and classification proble able to choose effective quate to the given cond	we a holistic view of umental environment ems. models, methods and itions and application
Assessment Methods:	Assessment of student knowledge is carried out according to the results of solving two independent tasks according to the statistics provided by the teacher. The first is the construction of multidimensional dependence, the second is the classification problem. Software implementation tools are selected by the student on their own.		
Teacher(s):	Mikhail Matveev		
Prerequisite subject(s):	Preliminary knowledge mathematical statistics required.	e of mathematics, proba , computational method	bility theory and ls, and programming is
Compulsory Literature:	Statistical analysis of d V.P. Bogatova, T.Ya. I 124 p.	lata in R. / A.G. Bukhov Biryuchinskaya – Voror	vets, P.V. Moskalev, nezh, VGAU. 2010

	Paklin N.B. Business Intelligence: From Data to Knowledge: A		
	Study Guide. 2nd ed. corrected / N.B. Paklin. V.I. Oreshkov - St.		
	Petersburg: Peter. 2013 704 p.		
Renlacement	Rafolovich V Data mining or Data mining for the employed		
Literature	Practical course /AT Rafolovich - "I-trade" 2014		
	Fractical course. /AT. Kalolovich - 1-trade , 2014.		
Participation	Lower limit of lectures attendance is 80%, each task must be		
requirements:	presented by end of the course.		
Independent work:	1. The task of building dependency		
	2. The task of classification		
Grading criteria scale			
or the minimal level	Failed < 50 points		
necessary for passing	Passed ≥ 50 points		
the subject:			
	Points distribution:		
	Ongoing assessment:		
	The task of building dependencies: 50 points		
	Classification Problem: 50 points		
Information about			
the course:	Room on at		
	,,		
1) Date 1	Lecture 1. Overview of approaches to solving classification and		
	regression problems. Statistical methods. Machine learning.		
	regression problems. Statistical methods. Machine learning.		
	regression problems. Statistical methods. Machine learning.		
	<i>regression problems. Statistical methods. Machine learning.</i> The unity and difference of classification and regression problems.		
	<i>regression problems. Statistical methods. Machine learning.</i> The unity and difference of classification and regression problems. Classification of uncertainties. Problem solving in the face of		
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	Statement of the classification problem. The concept of logit conversion. Approaches to the assessment of logistic regression		
	parameters.		
	Practical 4		
	Knowledge consolidation on examples.		
5) Date 5	Lecture 5. Time series analysis methods.		
	Time series as a tool for describing a dynamic process. Definition of		
	stationary and non-stationary time series. Models of autoregression		
	and moving average. Violation of the Gauss-Markov conditions.		
	Method of moments for estimating autoregressive parameters.		
	Models for predicting the behavior of a series.		
	Practical 5		
	Knowledge consolidation on examples.		
6) Date 6	Lecture 6. Fuzzy uncertainty, fuzzy inference.		
	Fuzzy sets as a method of formalizing fuzzy uncertainty. The basics		
	of fuzzy logic. Fuzzy production systems. Fuzzy neural production		
	networks like ANFIS.		
	Practical 6		
	Knowledge consolidation on examples.		
7) Date 7	Lecture 7. Methods of decision trees.		
	Classification using decision trees. Algorithms for constructing		
	decision trees, measures of partition quality. An example of a		
	partition using the Shannon entropy. The concept of random forest.		
	Practical 7		
	Knowledge consolidation on examples.		
8) Date 8	Lecture & Browse ensembles of classification models		
0) Date 0	Lecture 6. Drowse ensembles of clussification models.		
	Theoretical substantiation of the capabilities of the ensemble of		
	"weak" algorithms. Types of ensembles: homogeneous,		
	heterogeneous; component training, ensemble training. The		
	formation of the output of the ensemble. The methods of formation		
	of the ensemble: running, boosting, stacking.		
	Practical 8		
	Knowledge consolidation on examples.		
9) Date 9	Assessment		