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IV week V week VI week	Unstructured da restricted Boltzn (LSTM), gated r Unstructured da representation v	ta processing with con nann machine, and re ecurrent units (GRU) ta representation mo	ecursive networks with long- and other neuron types in d	short term memory leep neural networks.	
IV week V week VI week	restricted Boltzn (LSTM), gated r Unstructured da representation v	nann machine, and re ecurrent units (GRU) ta representation mo	ecursive networks with long- and other neuron types in d	short term memory leep neural networks.	
V week VI week	Unstructured da representation v	ta representation mo			
VI week		Unstructured data representation models (bag-of-words, LDA, topic models, vector representation with word2vec, doc2vec and others).			
		Classification, clustering and concept mining from unstructured data based on their meaning of the content.			
VII week	Sentiment analy	sis aiming to determi	ine in the text expressed fee	lings and attitude.	
	document.		ing of text to only the major		
VIII week		Optimization of unstructured data mining with nature inspired algorithms.			
IX week	Named-entity re	cognition and entity r	elation modeling		
X week					
XI week					
XII week					
XIII week					
XIV week					
XV week Methods of education					
	II.				
 lectures, project assignment 					
project assignment.					
Students' load					
N	Veekly		In Semest	ter	
		•	Lectures: 60		
		•	Individual work: 240		

Literature:

- Sarkar, D., 2016. Text Analytics with Python: A Practical Real-World Approach to Gaining Actionable Insights from Your Data. Apress.
- Bird, S., Klein, E. and Loper, E., 2009. Natural language processing with Python: analyzing text with the natural language toolkit. " O'Reilly Media, Inc.".
- Reese, R.M., 2015. Natural language processing with Java. Packt Publishing Ltd.
- Hearty, J., 2016. Advanced Machine Learning with Python. Packt Publishing Ltd.
- Goodfellow, I., Bengio, Y. and Courville, A., 2016. Deep learning. MIT press.
- Géron, A., 2017. Hands-on machine learning with Scikit-Learn and TensorFlow: concepts, tools, and techniques to build intelligent systems. O'Reilly Media.

Learning outcomes (complied with the outcomes for the study programme):

Knowledge and understanding:

On completion of this course the student will be able to

- demonstrate knowledge and understanding of the process of preparation and building of the mining system for unstructured data
- demonstrate knowledge and usage of methods of unstructured data mining on real cases
- demonstrate the ability to evaluate and interpret the results of the mining process for unstructured data

Transferable/Key skills and other attributes:

Use of tools and technologies: use of software tools and technologies for data mining of unstructured data in one of the programming languages.

Forms of tests and evaluation:

- completed project 50%
- oral examination 50%

Name and surname of teacher and associate: Sašo Karakatič

Particularities needed to be emphasized for the subject:

Note (if needed):