





Data Sciences Symposium

7th & 8th March, University of Toulon City Center - Free Entrance

Prog. com. : H. Glotin & A. Paiement, LIS - DYNI - PSD - CNRS - U. Toulon V. Kornilov, M. Komarov & S. Maltseva, HSE Moscow

7th March, amphi FA001

11h00 Présentation du Pôle INPS : *Hervé Glotin*, Université de Toulon

- 11h10 Conférence invitée INPS: Information Spaces for Big Data Problems: Transforming Sequential Processing into Parallel *Peter Golubtsov*, Dr., Pr., Dept of Mathematics, Faculty of Physics, Lomonosov Moscow State University
- 11h40 Table ronde : des Pôles interdisciplinaires de recherche à l'université de Toulon, pourquoi et avec qui?
- 12h00 Pause déjeuner salle CO.315 3^e étage du plot Coudon

7	Jeudi mars 2019 Co-org. <i>H. Glotin, A. Paiement, V. Kornilov, M. Komarov, S. Maltseva,</i> RU-FR BigData Session INPS - Amphithéâtre FA.110 Method for Big data	8	Vendredi mars 2019 Co-org. <i>H. Glotin, A. Paiement, V. Kornilov, M. Komarov, S. Maltseva,</i> RU-FR BigData Session INPS - Amphithéâtre FA.110
13:15	DataMall: a new platform for trading with big data <u>Vitalii Veliamidov</u> , Engineer, Nat. Center of Cognitive Tech., ITMO University, St Petersbourg, RU	9:00	Deep Learning Bioacoustics & Deep Learning Jan Schlüter, Dr., Researcher, LIS Lab & OFAI, CNRS, UTLN, Toulon, FR & Vienna, Austria
13:35	All-Russian Big Data competition among students Irina Deeva, Engineer, Nat. Center of Cognitive Tech., ITMO University, St Petersbourg, RU		Forecasting
13:55	High Performance Computing for mono channel source localisation Julie Patris, Dr, Pr., Dept of Physics, CNRS, LIS, Aix Marseille University	9:45	Decisions for Large Retail Network - Machine learning for tornado prediction <i>Fuad Aleskerov</i> , Dr., Pr, Head of Dept of Mathematics, Faculty of Economic Sciences, Nat. Research Univ. Higher Sch. of Economics, Moscow, RU
14:15	Long term survey of the Fukushima Nuclear Exclusion Zone : Systems & challenges Daisuke Shimotoku, PhD Student, Faculty of Computer Science, Univ. Tokyo, JP	10:30	Deep learning methods for time series forecasting <i>Petr Gladilin</i> , Researcher, eScience Research Institute, ITMO University, St Petersbourg, RU
14:30	The pain and gain of maintaining the ML models zoo Natalia Khapaeva, Senior Architect, Mobile TeleSystems PJSC, RU	10:50	Forecasting purchase categories by transactional data: a comparative study of classification methods
	Learning from Big data		Egor Shikov, Engineer, Nat. Center of Cognitive Tech., ITMO University, St Petersbourg, RU
14:40	Big Data in climate: models fitting, analysis & storage of models' results <i>Nikolay Nikitin</i> , Engineer, Nat. Center of Cognitive Tech., ITMO University, St Petersbourg, RU	11:20	On the Ethics of Big Data Olga Tsukanova, Dr Pr HSE Moscow, with Tatyana Telysheva
15:00	Logic-based boundaries of some Modular Neural Networks for Big Data <i>Denis Fedyanin</i> , Research Fellow, Int. Lab. for Logic, Linguistics & Formal Philosophy, Nat. Research Univ. Higher Sch. of Economics, Moscow, RU	11:45	Discussion <i>Svetlana Maltseva</i> , Dr., Pr., Head of Sch. of Business Informatics, Fac. of Business & Managt., Nat. Research Univ. Higher Sch. of Economics, Moscow, RU
15:20	Large scale marine mammal monitoring: Learning Orca calls	12:30	Pause déjeuner - salle C0.315 - 3° étage du plot Coudon
	Hervé Glotin, Dr., Pr., Dept of Computer Science, CNRS, UTLN, Toulon, FR	14:05	End to End SpliNet Learning Randall Balestriero, PhD Student, Faculty of Computer Science, Rice Univ., Houston, USA
15:35	On compression, learning & searching regularity in big data <u>Tatiana Makhalova</u> , PhD Student, Faculty of Computer Science, Nat. Research Univ. Higher Sch. of Economics, Moscow, RU	14:45	City-scale event prediction using convolutional neural networks & adaptive geogrids Ksenia Mukhina, Engineer, Nat.Center of Cognitive Tech., ITMO University, St Petersbourg, RU
	Deep Learning	15:15	Discussion
16:00	Manifold learning from Stereo Auto-Encoder Maxence Ferrari, PhD Student, Faculty of Computer Science, CNRS LAMFA LIS, FR	16:00	End
16:20	Convolutional neural networks for automated search of anomalies in metocean fields <i>Pavel Vychuzhanin</i> , Engineer, Nat.Center of Cognitive Tech., ITMO University, St Petersbourg, RU		
16:45	Time series in Astrophysics & Deep Learning Adeline Paiement, Dr., Pr., Dept of Computer Science, CNRS, UTLN, Toulon, FR		

ABSTRACTS

Fuad Aleskerov	Dr., prof., Head of Department of Mathematics,	National Research University	Decisions for Large Retail Network	
	Faculty of Economic Sciences	Higher School of Economics		
collected for more The problem was to evaluate up-sell po New methods to so	ill network with 1.5 mln customers, more than 400 the than one year for each customer. o find a segmentation of customers on the basis of the tential as well as find churning customers. Ive these problems as well as general view on the pro oplied for one of the European countries and large Eur	busand goods, and data Fir basket consumption to blems are discussed.	https://www.hse.ru/en/org/persons/140159Professor, Member of Academia Europaea, HonoraryScience and Education of RussiaMember of Editorial Board of 14 journals10 books, more than 200 articles, more than 100 in perjournals and volumesCopyright certificates, patents - 6Invited speaker for more than 100 conferences and wLogic-based boundaries of some Modular Neural	eer-reviewed
Senis reayanni		Control Sciences	Networks for BigData	
of people. Such div a unique NN for eac one result by specia distributed NN has architecture has so	cognition of faces, where one can take predefined NN rision a large NN into smaller Networks could be very of the big part of BigData. And then we can just integrate al NN without teaching NN for entire dataset for the very an ability to replace local NN when we need it. It gives me limits and properties. We will focus on its bounda and show connections to dynamic epistemic logic and	useful for BigData. We can host results of these local NN into ery beginning each time. These s us high flexibility. Such ries based on using some	Applied Mathematics and Game Theory.	
Peter Golubtsov	Dr., prof. Department of Mathematics, Faculty of Physics	Lomonosov Moscow State University	Information Spaces for Big Data Problems: Transforming Sequential Processing into Parallel	
because it avoids ac by introducing a sp properties of the co efficiency of the inf	sequential updating of information is important for "b secumulating and storing large data sets. It is shown the ecial intermediate form of information representation prresponding information space, such approach allows ormation update. It also leads to various parallelization procedure, which are ideally suited for distributed da	at processing can be simplified Thanks to the rich algebraic unifying and increasing the n options for the inherently	Doctor of Science, Professor of National Research Unit School of Economics and Lomonosov Moscow State U Main areas of research: mathematical foundations of theory, information processes in Big Data systems, inf effects in dynamic stochastic games with applications and optimal management of natural resources.	niversity. information ormation

Natalia Khapaeva	Senior Architect	Mobile TeleSystems PJSC	The pain and gain of maintaining the ML models zoo	75 1438312
When we talk about	the Data Science models, the most recent topics are	algorithm selection, feature	https://www.linkedin.com/in/nmkh/	
engineering, etc. Thus, the «fog of war» covers almost half of the model lifecycle. However, deploying,			Senior Big Data Architect at MTS Group, Data Governar	nce, Data
monitoring and quality assessment stages of the model lifecycle can be the critical bottleneck of the			Quality and DS Products Operation projects.	
whole project.				
This talk covers susta	ainable deployment and monitoring in a web product	tion environment.		
Svetlana Maltseva	Dr., prof., Head of School of Business Informatics,	National Research University		
	Faculty of Business and Management	Higher School of Economics		
			https://www.hse.ru/en/org/persons/67823	
Tatiana Makhalova	PhD Student, Faculty of Computer Science	National Research University	On compression, learning and searching regularity	
		Higher School of Economics	in data	
	an umbrella term that encompasses a big variety of r	nethods, processes, and	PhD Student, Faculty of Computer Science, National Re	esearch
-	getting insight from data.		University Higher School of Economics	
-	erstood differently and results in different approache			
completely different objectives. More than that, being developed independently the fields of DS are				
		-		
constantly moving a	part. It seems unlikely that well-adapted to solve dist	inct tasks, the methods from		
constantly moving a one field can contrib	part. It seems unlikely that well-adapted to solve dist oute to the development of the other fields. We consi	inct tasks, the methods from ider the most common fields in		
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	compare the efficiency of different classes of machi s, long-short term memory networks and convoluti			
Ksenia Mukhina	Engineer, National Center of Cognitive Technologies	ITMO University	City-scale event prediction using convolutional neural networks and adaptive geogrids	
Active development of modern cities requires not only efficient monitoring systems but furthermore forecasting systems that can predict future state of the urban area with high accuracy. In this work, we present a method for urban area prediction based on geospatial activity of users on Instagram. We propose three different deep learning architectures that are able to solve a target problem and show that convolutional neural network based on three-dimensional convolution layers provides the best results with an accuracy of 99%.			Ksenia Mukhina is a Lecturer at High-Performance Computing Department and a Research Assistant at eScience Research Institute in ITMO University. Ksenia works in "GSS infrastructure" research team where she focuses on social networks analysis. She an author of more than 20 scientific papers.	
Vitalii Veliamidov	Engineer, National Center of Cognitive Technologies	ITMO University	DataMall: a new platform for trading with big data	
data but don't know own business or kn dedicated to bring been developed wi • safety of da • flexible acc • different m "pay-as-yo • easy-to-use desired qua	connected processes in modern business world lead w how to or can not extract value for himself and the ows how to help others but lacks appropriate data. them all together and solve the problem of data access th 4 main aspects in mind: ata with adjustable degree of control for their owner ess to elastic computing power provided by clouds; conetization policies of data access - wholesales for u-go" policy for small and medium business; e web hub that serves as a showcase for customers antities and buy it.	hose who wants to improve his DataMall is a new platform cess and data selling. Its design has er; large business and to help find desired data in	Vitalii Veliamidov has graduated from St. Petersburg S University with M.D. (dep. of Information and Analytic His professional interests include distributed data stor processing systems and now he is working on the Data platform at National Center of Cognitive Technologies, University	al Systems). age and Moll
Nikolay Nikitin	Engineer, National Center of Cognitive Technologies	ITMO University	Big Data in climate modelling: models fitting, analysis and storraging of models' results	
The Big Data concept becomes extremely important for metocean data analysis and storage. The experience of semantic data storage's development in a frame of long-term Arctic ocean simulation project is presented. Several data engineering solutions that allow providing real-time data processing are concerned. Also, the TerraXT software for the dynamic environmental data visualisation is described.			Nikolay Nikitin is an Engineer of National Center for Cognitive Technologies, Teaching assistant and PhD Student at High-Performance Computing Department of ITMO University.	

Pavel Vychuzhanin	Engineer, National Center of Cognitive Technologies	ITMO University	CNN for automated search of anomalies in metocean fields	
modeling as a patter methods can be imp currents and ice con	pposes to consider the problem of detecting anoma on recognition tasks. As practical examples, it will be lemented for fields with different nature of anomali centration in the Arctic region. The advantage of the lds in a fully automatic mode, without expert super	shown how the proposed ies, in particular, for fields of sea e presented approach is the	Pavel Vychuzhanin is an Engineer of National Center f Technologies, Master's Student at High-Performance Department of ITMO University. Pavel works in "Natu Simulation" research group where he focuses on data methods in geophysical modeling.	Computing ral Systems
Irina Deeva	Engineer, National Center of Cognitive Technologies	ITMO University	All-Russian Big Data competition among students: results and experience	
Competitive data analysis as a separate unit in the world of machine learning and big data. The experience of the All-Russian competitions in Big Data Analytics. How to cover machine learning and computing infrastructure. Results, statistics and trends among students in the field of big data.			Irina Deeva is an Engineer of National Center for Cognitive Technologies and PhD Student at High-Performance Computing Department of ITMO University. Irina works in "Natural Systems Simulation" research group where she focuses on multiscale and surrogate modeling	
Jan Schlüter	Postdoctoral research fellow	LIS lab, University of Toulon	Deep Learning for Acoustic Bird Detection and Species Identification	
is vital to understand help to monitor biod Using artificial neura well as identify the s present the technica	ety of life on Earth, is constantly reduced by human d and control it. We explored how well current meth liversity from audio recordings, supporting or compl I networks, we were able to automatically detect bi pecies to some extent, reaching top results in two s I solutions in detail, discuss shortcomings, and finish ogy to underwater recordings of whales.	nods of artificial intelligence may ementing human observations. rd calls in field recordings, as cientific competitions. I will		
Julie Patris	Dr., Pr.	Aix-Marseille University, LIS lab	High performance computing for mono channel source localization	
(see Marques et al. 2 hydrophones (Kuper installing one single Several attempts hav hydrophone (McDor usually permit only r multiple arrivals of a Here, we use acoust	whale density estimation through passive acoustics 2013). Most commonly, the localization process require man et al. 2004), which is a more technically comple- sensor. we been made towards recovering the position of th hald et al. 1999, Kuperman et al. 2001, Bonnel et al. range (and sometimes depth) estimation, and are us short sound are separable. ic modeling methods (SPECFEM, Tromp et al. 2008 s.org/cig/software/specfem3d/) to understand the	uires an array of 4 synchronized ex and expensive option than e singer with only one 2014), however, these methods ually possible only when	Dept. Physics, Aix Marseille University	

information inside th	cific context. Thanks to this precise time-domain more e signal, including modifications produced by enviror			
	ariation, etc.). Subsequently we applied a low-cost ir iprocity principle to reconstruct the whale's position			
	ed data confirm that the technique is well suited to l			
	e Southeast Pacific 2 blue whale song type (describe			
	lities of this method for determining the range of an			
	two simulations in a 3D 10 km-wide box lead to the			
	n 100 m) in 98 % of cases, of the depth in 100 % of c			
-	nuth, thanks to our 3D modelling, with a resolution g			
cases.	,	, ,		
We also present the f	irst results of this method applied to Blue Whales in	Northern Chile, where a		
coupled visual/acous	tic study was performed in austral summer 2017.			
Randall Balestriero	PhD candidate	Rice University	End to end SpliNet learning	
We propose to tackle	the problem of end-to-end learning for raw wavefor	rm signals by introducing	Computer Science Dept., Rice University, Houston, USA	
learnable continuous	time-frequency atoms. The derivation of these filter	s is achieved by defining a		
functional space with	a given smoothness order and boundary conditions.	. From this space, we derive		
the parametric analytical filters. Their differentiability property allows gradient-based optimization. As				
such, one can utilize any Deep Neural Network (DNN) with these filters. This enables us to tackle in a				
front-end fashion a large scale bird detection task based on the freefield1010 dataset known to contain				
key challenges, such as the dimensionality of the inputs data (>100,000) and the presence of additional				
noises: multiple sour				
Adeline Paiement	Dr., Dept. Computer Science	University of Toulon, LIS lab	Deep learning for astronomy image analysis	
This talk will present	our recent and ongoing work on astronomy image ar	nalysis. We will examine the	Computer Science Dept., Université de Toulon, LIS lab	
	pose on the design of new computer vision based ar			
particularly consider	deep learning methods, which have been so far pred	lominantly developed for		
natural images, and we will illustrate the potential for this paradigm in enhancing the interpretation and				
exploitation of astronomy images.				