2019 IEEE 43rd Annual Computer Software and Applications Conference

COMPSAC 2019

2019 IEEE 43rd Annual Computer Software and Applications Conference

15–19 July 2019 Milwaukee, Wisconsin

Editors

Vladimir Getov, University of Westminster, UK Jean-Luc Gaudiot, University of California, Irvine, USA Nariyoshi Yamai, Tokyo University of Agriculture & Technology, Japan Stelvio Cimato, University of Milan, Italy Morris Chang, University of South Florida, USA Yuuichi Teranishi, National Institute of Information and Communications Technology, Japan Ji-Jiang Yang, Tsinghua University, China Hong Va Leong, Hong Kong Polytechnic University, Hong Kong Hossian Shahriar, Kennesaw State University, USA Michiharu Takemoto, NTT Labs, Japan Dave Towey, University of Nottingham Ningbo China, China Hiroki Takakura, National Institute of Informatics, Japan Atilla Elci, Aksaray University, Turkey Susumu Takeuchi, NTT Labs, Japan Satish Puri, Marquette University, USA



Los Alamitos, California Washington • Tokyo



Copyright © 2019 by The Institute of Electrical and Electronics Engineers, Inc.

All rights reserved.

Copyright and Reprint Permissions: Abstracting is permitted with credit to the source. Libraries may photocopy beyond the limits of US copyright law, for private use of patrons, those articles in this volume that carry a code at the bottom of the first page, provided that the per-copy fee indicated in the code is paid through the Copyright Clearance Center, 222 Rosewood Drive, Danvers, MA 01923.

Other copying, reprint, or republication requests should be addressed to: IEEE Copyrights Manager, IEEE Service Center, 445 Hoes Lane, P.O. Box 133, Piscataway, NJ 08855-1331.

The papers in this book comprise the proceedings of the meeting mentioned on the cover and title page. They reflect the authors' opinions and, in the interests of timely dissemination, are published as presented and without change. Their inclusion in this publication does not necessarily constitute endorsement by the editors, the IEEE Computer Society, or the Institute of Electrical and Electronics Engineers, Inc.

> BMS Part Number CFP19061-ART ISBN 978-1-7281-2607-4 ISSN 0730-3157

Additional copies may be ordered from:

IEEE Computer Society Customer Service Center 10662 Los Vaqueros Circle P.O. Box 3014 Los Alamitos, CA 90720-1314 Tel: + 1 800 272 6657 Fax: + 1 714 821 4641 http://computer.org/cspress csbooks@computer.org IEEE Service Center 445 Hoes Lane P.O. Box 1331 Piscataway, NJ 08855-1331 Tel: + 1 732 981 0060 Fax: + 1 732 981 9667 http://shop.ieee.org/store/ customer-service@ieee.org IEEE Computer Society Asia/Pacific Office Watanabe Bldg., 1-4-2 Minami-Aoyama Minato-ku, Tokyo 107-0062 JAPAN Tel: + 81 3 3408 3118 Fax: + 81 3 3408 3553 tokyo.ofc@computer.org

Individual paper REPRINTS may be ordered at: <reprints@computer.org>

Editorial production by Lisa O'Conner Cover art production by Hector Torres





IEEE Computer Society Conference Publishing Services (CPS) http://www.computer.org/cps

2019 IEEE 43rd Annual Computer Software and Applications Conference (COMPSAC) COMPSAC 2019

Table of Contents

Message from the Standing Committee Chair	xxiv
Message from the Standing Committee Vice Chair	xxvi
Message from the 2019 General Chairs	xxvii
Message from the Program Chairs	xxviii
Message from COMPSAC 2019 Fast Abstract Track	xxix
COMPSAC 2019 Organizers	xxx

COMPSAC 2019 Symposia

SETA: Software Engineering Technologies & Applications

SETA 1

tatically-Directed Assertion Recommendation for C Programs
Sciences), Renwei Zhang (Huawei Technologies), and Weiliang Yin (Huawei Technologies)
An Empirical Study on API-Misuse Bugs in Open-Source C Programs
Zuxing Gu (Tsinghua University), Jiecheng Wu (Tsinghua University),
Jiaxiang Liu (Shenzhen University), Min Zhou (Tsinghua University),
and Ming Gu (Tsinghua University)
Characterization and Prediction of Popular Projects on GitHub
Junxiao Han (Zhejiang University), Shuiguang Deng (Zhejiang
University), Xin Xia (Monash University), Dongjing Wang (Hangzhou
Dianzi University), and Jianwei Yin (Zhejiang University)

SETA 2

SENSORY: Leveraging Code Statement Sequence Information for Code Snippets Recommendation Lei Ai (Nanjing University of Aeronautics and Astronautics), Zhiqiu Huang (Nanjing University of Aeronautics and Astronautics), Weiwei Li (Nanjing University of Aeronautics and Astronautics), Yu Zhou (Nanjing University of Aeronautics and Astronautics), and Yaoshen Yu (Nanjing University of Aeronautics and Astronautics)	27
Supporting Consistency in the Heterogeneous Design of Safety-Critical Software Andrés Paz (École de Technologie Supérieure, Canada) and Ghizlane El Boussaidi (École de Technologie Supérieure, Canada)	37
A Heuristic Approach to Break Cycles for the Class Integration Test Order Generation Miao Zhang (City University of Hong Kong), Jacky Keung (City University of Hong Kong), Yan Xiao (City University of Hong Kong), Md Alamgir Kabir (City University of Hong Kong), and Shuo Feng (City University of Hong Kong)	47
Assessing the Significant Impact of Concept Drift in Software Defect Prediction Md Alamgir Kabir (City University of Hong Kong), Jacky W. Keung (City University of Hong Kong), Kwabena E. Benniny (Blekinge Institute of Technology), and Miao Zhang (City University of Hong Kong)	53
SETA 3	
In Search of Scientific Agile	59
Learning-Based Anomaly Cause Tracing with Synthetic Analysis of Logs from Multiple Cloud Service Components	66
Using Client-Based Class Cohesion Metrics to Predict Class Maintainability Musaad Alzahrani (Albaha University), Saad Alqithami (Albaha University), and Austin Melton (Kent State University)	72

SETA 4

 ParaAim: Testing Android Applications Parallel at Activity Granularity	1
A Comparative Study of the Effectiveness of Meta-Heuristic Techniques in Pairwise Testing	1
An Adaptive Approach to Recommending Obfuscation Rules for Java Bytecode Obfuscators	7

SETA 5

 AutoPer: Automatic Recommender for Runtime-Permission in Android Applications	70
 Metrics Driven Architectural Analysis using Dependency Graphs for C Language Projects	17
Identifying the Challenges of the Blockchain Community from StackExchange Topics and Trends	23

SETA 6

An Empirical Study on the Spreading of Fault Revealing Test Cases in Prioritized Suites	129
 Supporting Decision Makers in Search-Based Product Line Architecture Design using Clustering Willian Marques Freire (State University of Maringa, Brazil), Carlos Vinícius Bindewald (State University of Maringa, Brazil), Aline M. M. Miotto Amaral (State University of Maringa, Brazil), and Thelma Elita Colanzi (State University of Maringa, Brazil) 	139
Execution Enhanced Static Detection of Android Privacy Leakage Hidden by Dynamic Class Loading . Yufei Yang (Nanjing University), Wenbo Luo (Nanjing University), Yu Pei (The Hong Kong Polytechnic University), Minxue Pan (Nanjing	149

SETA 7

University), and Tian Zhang (Nanjing University)

Deep-AutoCoder: Learning to Complete Code Precisely with Induced Code Tokens	
Xing Hu (Key Laboratory of High Confidence Software Technologies	
(Peking University), Ministry of Education), Rui Men (Key Laboratory	
of High Confidence Software Technologies (Peking University), Ministry	
of Education), Ge Li (Key Laboratory of High Confidence Software	
Technologies (Peking University), Ministry of Education), and Zhi Jin	
(Key Laboratory of High Confidence Software Technologies (Peking	
University), Ministry of Education)	

ConRS: A Requests Scheduling Framework for Increasing Concurrency Degree of Server Programs	169
Biyun Zhu (Institute of Software Chinese Academy of Sciences,	
University of Chinese Academy of Sciences), Ruijie Meng (Institute of	
Software Chinese Academy of Sciences, University of Chinese Academy of	
Sciences), Zhenyu Zhang (Institute of Software Chinese Academy of	
Sciences), and W.K. Chan (City University of Hong Kong)	
LAC: Locating and Applying Consistent and Repetitive Changes	179
Sushma Sakala (University of Nebraska at Omaha), Vamshi Krishna Epuri	
(University of Nebraska at Omaha), Samuel Sungmin Cho (Northern	
Kentucky University), and Myoungkyu Song (University of Nebraska at	
Omaha)	
Dockerfile TF Smell Detection Based on Dynamic and Static Analysis Methods	185
Jiwei Xu (University College Dublin), Yuewen Wu (Chinese Academy of	
Sciences), Zhigang Lu (Chinese Academy of Sciences), and Tao Wang	
(Chinese Academy of Sciences)	

SETA 8

Empirical Analysis of the Growth and Challenges of New Programming Languages	.91
Featured Event Sequence Graphs for Model-Based Incremental Testing of Software Product Lines	.97
Time-Aware and Location-Based Personalized Collaborative Recommendation for IoT Services	:03
An Extended Abstract of "Metamorphic Testing: Testing the Untestable"	:09

Plenary Panel

Agile, Continuous Integration, and DevOps	 1
Carl K. Chang (Iowa State University)	

CELT: Computing Education & Learning Technologies

CELT 1

A Declarative Approach for an Adaptive Framework for Learning in Online Courses Djananjay Pandit (Arizona State University) and Ajay Bansal (Arizona State University)	212
Empowering Engagement through Automatic Formative Assessment Alice Barana (University of Turin), Marina Marchisio (University of Turin), and Sergio Rabellino (University of Turin)	216
VISA: A Supervised Approach to Indexing Video Lectures with Semantic Annotations Luca Cagliero (Politecnico di Torino), Lorenzo Canale (Politecnico di Torino), and Laura Farinetti (Politecnico di Torino)	226

CELT 2

A Mobile System to Increase Efficiency of the Lecturers when Preventing Academic Dishonesty During Written Exams	. 236
Pedro Maroco (Universidade Nova de Lisboa), João Cambeiro (Universidade Nova de Lisboa), and Vasco Amaral (Universidade Nova de Lisboa)	
Study TOUR for Computer Science Students	. 242
(The Hong Kong Polytechnic University), and Grace Ngai (The Hong Kong Polytechnic University)	

NCIW: Networks, Communications, Internet & Web Technologies

NCIW 1

C2P2: Content-Centric Privacy Platform for Privacy-Preserving Monitoring Services	252
Predicting Network Outages Based on Q-Drop in Optical Network	258
Yohei Hasegawa (Waseda University) and Masato Uchida (Waseda	
University)	
HeteroTSDB: An Extensible Time Series Database for Automatically Tiering on Heterogeneous Key-Value	
Stores	264
Yuuki Tsubouchi (SAKURA Internet Inc.), Asato Wakisaka (Hatena Co.,	
Ltd.), Ken Hamada (Hatena Co., Ltd.), Masayuki Matsuki (Hatena Co.,	
Ltd.), Hiroshi Abe (Lepidum Co. Ltd., COCON Inc., Japan Advanced	
Institute of Science and Technology (JAIST)), and Ryosuke Matsumoto	
(SAKURA Internet Inc.)	
FastContainer: A Homeostatic System Architecture High-Speed Adapting Execution Environment Changes Ryosuke Matsumoto (SAKURA Internet Inc.), Uchio Kondo (GMO Pepabo,	270

Inc.), and Kentaro Kuribayashi (GMO Pepabo, Inc.)

NCIW 2

 Yuuichi Teranishi (National Institute of Information and Communications Technology, Japan), Takashi Kimata (National Institute of Information and Communications Technology, Japan), Eiji Kawai (National Institute of Information and Communications Technology, Japan), and Hiroaki Harai (National Institute of Information and Communications Technology, Japan) Modeling Restrained Epidemic Routing on Complex Networks	Hybrid Cellular-DTN for Vehicle Volume Data Collection in Rural Areas	276
of Information and Communications Technology, Japan), Eiji Kawai (National Institute of Information and Communications Technology, Japan), and Hiroaki Harai (National Institute of Information and Communications Technology, Japan) Modeling Restrained Epidemic Routing on Complex Networks	Yuuichi Teranishi (National Institute of Information and	
 (National Institute of Information and Communications Technology, Japan), and Hiroaki Harai (National Institute of Information and Communications Technology, Japan) Modeling Restrained Epidemic Routing on Complex Networks		
 Japan), and Hiroaki Harai (National Institute of Information and Communications Technology, Japan) Modeling Restrained Epidemic Routing on Complex Networks		
Communications Technology, Japan) Modeling Restrained Epidemic Routing on Complex Networks		
Modeling Restrained Epidemic Routing on Complex Networks 285 Natsuko Kawabata (Kwansei Gakuin University), Yasuhiro Yamasaki 285 (Kwansei Gakuin University), and Hiroyuki Ohsaki (Kwansei Gakuin 201 University) 291 Sparse Representation of Network Topology with K-SVD Algorithm 291 Ryotaro Matsuo (Kwansei Gakuin University, Japan), Ryo Nakamura 291		
Natsuko Kawabata (Kwansei Gakuin University), Yasuhiro Yamasaki (Kwansei Gakuin University), and Hiroyuki Ohsaki (Kwansei Gakuin University) Sparse Representation of Network Topology with K-SVD Algorithm	Communications Technology, Japan)	
Natsuko Kawabata (Kwansei Gakuin University), Yasuhiro Yamasaki (Kwansei Gakuin University), and Hiroyuki Ohsaki (Kwansei Gakuin University) Sparse Representation of Network Topology with K-SVD Algorithm	Modeling Restrained Epidemic Routing on Complex Networks	285
(Kwansei Gakuin University), and Hiroyuki Ohsaki (Kwansei Gakuin University) Sparse Representation of Network Topology with K-SVD Algorithm		
University) Sparse Representation of Network Topology with K-SVD Algorithm		
Sparse Representation of Network Topology with K-SVD Algorithm		
Ryotaro Matsuo (Kwansei Gakuin University, Japan), Ryo Nakamura	University)	
	Sparse Representation of Network Topology with K-SVD Algorithm	291
	Rvotaro Matsuo (Kwansei Gakuin University, Japan), Rvo Nakamura	
(Kwunser Gakuin Oniversity, Japan), and Hiroyaki Onsaki (Kwunser		
Gakuin University, Japan)	Gakun University, Japan)	

EATA: Emerging Advances in Technology & Applications

Message from EATA Symposium Chairs	299
Ali Hurson (Missouri University of Science & Technology), Hiroyuki	
Sato (University of Tokyo), Toyokazu Akiyama (Kyoto Sangyo	
University), and Dan Lin (University of Missouri)	

EATA 1

DirectFlow: A Robust Method for Ocular Torsion Measurement	300
Bruno Kozen Stahl (School of Technology, PUCRS), Leonardo Pavanatto	
Soares (School of Technology, PUCRS), Vicenzo Abichequer Sangalli	
(School of Technology, PUČRS), Pedro Costa Klein (School of	
Technology, PUCRS), Rafael Neujahr Copstein (School of Technology,	
PUCRS), and Márcio Sarroglia Pinho (School of Technology, PUCRS)	
Pothole Detection in Asphalt: An Automated Approach to Threshold Computation Based on the Haar	
Wavelet Transform	306
Ricardo Silveira Rodrigues (UFSM, Brazil), Marcia Pasin (UFSM,	
Brazil), Alice Kozakevicius (UFSM, Brazil), and Vinicius Monego (UFSM,	

Brazil)

EATA 2

An Ontology Enh	hanced User Profiling Algorithm Based on Application Feedback	
Xin Dong (Be	eijing University of Technology), Tong Li (Beijing	
University of	Technology), and Zhiming Ding (Beijing University of	
Technology)		

A Dataflow Application Deployment Strategy for Hierarchical Networks	326
Shintaro Ishihara (Kyoto Sangyo University), Satoshi Tanita (Kyoto	
Sangyo University), and Toyokazu Akiyama (Kyoto Sangyo University)	

DSAT: Data Sciences, Analytics & Technologies

DSAT 1

How the Academics Qualification Influence the Students Learning Development Edna Dias Canedo (UnB), Rhandy Rafhael De Carvalho (UnB), Heloise Acco Tives Leão (CEULP/ULBRA), Pedro Henrique Teixeira Costa (UnB), and Marcio Vinicius Okimoto (UnB)	336
 Alchemy: Stochastic Feature Regeneration for Malicious Network Traffic Classification	346
Purchasing Behavior Analysis Based on Customer's Data Portrait Model Jing Sun (North China University of Technology), Huiqun Zhao (North China University of Technology), Sanwen Mu (North China University of Technology), and Zimu Li (North China University of Technology)	352
Recommender Systems Based on Autoencoder and Differential Privacy Jiahui Ren (East China University of Science and Technology), Xian Xu (East China University of Science and Technology), Zhihuan Yao (East China University of Science and Technology), and Huiqun Yu (East China University of Science and Technology)	358
DSAT 3	
Matrix Factorization Model with Dual Preferences for Rating Prediction Yuan Li (Peking University, China) and Kedian Mu (Peking University, China)	364
 View-Adaptive Weighted Deep Transfer Learning for Distributed Time-Series Classification Sreyasee Das Bhattacharjee (UNC Charlotte), William J. Tolone (UNC Charlotte), Ashish Mahabal (California Institute of Technology), Mohammed Elshambakey (IRI, SRTA-City), Isaac Cho (UNC Charlotte), and George Djorgovski (California Institute of Technology) 	373
Multi-Source Heterogeneous Core Data Acquisition Method in Edge Computing Nodes Hong Xia (Xi'an University of Posts and Telecommunications), Mingdao Zhao (Xi'an University of Posts and Telecommunications), Yanping Chen (Xi'an University of Posts and Telecommunications), Zhongmin Wang (Xi'an University of Posts and Telecommunication), Zhong Yu (Xi'an University of Posts and Telecommunication), and Jingwei Yang (California State University Sacramento)	382

DSAT 4

Chinese Social Media Entity Linking Based on Effective Context with Topic Semantics Chengfang Ma (Institute of Information Engineering, CAS, School of Cyber Security, University of CAS), Ying Sha (Institute of Information Engineering, CAS, School of Cyber Security, University of CAS; Huazhong Agricultural University), Jianlong Tan (Institute of Information Engineering, CAS, School of Cyber Security, University of CAS), Li Guo (Institute of Information Engineering, CAS, School of Cyber Security, University of CAS), and Huailiang Peng (Institute of Information Engineering, CAS, School of Cyber Security, University of CAS), Li Guo (Anstitute of CAS), and Huailiang Peng (Institute of Information Engineering, CAS, School of Cyber Security, University of CAS)	386
Identification of Cybersecurity Specific Content Using the Doc2Vec Language Model Otgonpurev Mendsaikhan (Nagoya University), Hirokazu Hasegawa (Nagoya University), Yukiko Yamaguchi (Nagoya University), and Hajime Shimada (Nagoya University)	396
Semantic Data-Driven Microservices Ivan Luiz Salvadori (Federal University of Santa Catarina), Alexis Huf (Federal University of Santa Catarina), and Frank Siqueira (Federal University of Santa Catarina)	402
Big Data Analytics in Telecommunication using State-of-the-art Big Data Framework in a Distributed Computing Environment: A Case Study Mohit Ved (Centre for Development of Advanced Computing) and Rizwanahmed B (Indian National Centre for Ocean Information Services)	411

DSAT 5

Keyword-Based Semi-Supervised Text Classification	417
TypoWriter: A Tool to Prevent Typosquatting	423
Ishtiyaque Ahmad (Bangladesh University of Engineering and	
Technology), Md Anwar Parvez (Bangladesh University of Engineering and	
Technology), and Anindya Iqbal (Bangladesh University of Engineering	
and Technology)	
Vehicle Travel Time Estimation by Sparse Trajectories	433
Mingyang Jiang (Shanghai Jiao Tong University) and Tianqi Zhao	
(Tsinghua University)	

DSAT 6

Xu: An Automated Query Expansion and Optimization Tool Morgan Gallant (Queen's University, Canada), Haruna Isah (Queen's University, Canada), Farhana Zulkernine (Queen's University), and Shahzad Khan (Gnowit Inc.)	443
Parallel Discovery of Trajectory Companions from Heterogeneous Streaming Data	453
Yongyi Xian (Concordia University), Chuanfei Xu (Concordia	
University), Sameh Elnikety (Microsoft Research, USA), and Yan Liu	
(Concordia University)	

ASYS: Autonomous Systems

ASYS 1

Cooperative UAVs Gas Monitoring using Distributed Consensus Daniele Facinelli (University of Trento, Italy), Matteo Larcher (University of Trento, Italy), Davide Brunelli (University of Trento, Italy), and Daniele Fontanelli (University of Trento, Italy)	463
Increasing Self-Adaptation in a Hybrid Decision-Making and Planning System with Reinforcement	
Learning	469
Christopher-Eyk Hrabia (Technische Universität Berlin, DAI-Lab), Patrick Marvin Lehmann (Technische Universität Berlin, DAI-Lab), and	
Sahin Albayrak (Technische Universität Berlin, DAI-Lab)	
AILiveSim: An Extensible Virtual Environment for Training Autonomous Vehicles Jérôme Leudet (AILiveSim Oy, Finland), François Christophe (University	479
of Helsinki, Finland), Tommi Mikkonen (University of Helsinki,	
Finland), and Tomi Männistö (University of Helsinki, Finland)	
ASYS 2	
Learning Distributed Cooperative Policies for Security Games via Deep Reinforcement Learning	489

Hassam Ullah Sheikh (University of Central Florida), Mina Razghandi (University of Central Florida), and Ladislau Boloni (University of Central Florida)	59
The SAMBA Approach for Self-Adaptive Model-Based Online Testing of Services Orchestrations	∂ 5
 Visual Tracking with Autoencoder-Based Maximum A Posteriori Data Fusion	31

ITiP: IT in Practice

ITiP 1

Enhanced Detection of Crisis-Related Microblogs by Spatiotemporal Feedback Loops	. 507
Christian Meurisch (TU Darmstadt), Zain Hamza (TU Darmstadt), Bekir	
Bayrak (TU Darmstadt), and Max Mühlhäuser (TU Darmstadt)	
AssistantGraph: An Approach for Reusable and Composable Data-Driven Assistant Components	. 513
Christian Meurisch (TU Darmstadt), Bekir Bayrak (TU Darmstadt), and	

Max Mühlhäuser (TU Darmstadt)

Integrating Static Code Analysis Toolchains	23
Matthias Kern (FZI Research Center for Information Technology), Ferhat	
Erata (Yale University), Markus Iser (Karlsruhe Institute of	
Technology), Carsten Sinz (Karlsruhe Institute of Technology),	
Frederic Loiret (KTH Royal Institute of Technology), Stefan Otten (FZI	
Research Center for Information Technology), and Eric Sax (FZI	
Research Center for Information Technology)	
Producing Green Computing Images to Optimize Power Consumption in OLED-Based Displays	29
Sorath Asnani (Politecnico di Torino), Maria Giulia Canu (Politecnico	
di Torino), and Bartolomeo Montrucchio (Politecnico di Torino)	

ITiP 3

LIPs: A Protocol for Leadership Incentives for Heterogeneous and Dynamic Platoons	535
Brian Ledbetter (Tennessee Tech University), Samuel Wehunt (Tennessee	
Tech University), Mohammad Ashiqur Rahman (Florida International	
University), and Mohammad Hossein Manshaei (Florida International	
University)	
Risk Assessment Methods for Cloud Computing Platforms	545
Tim Weil (Alcohol Monitoring Systems)	
Employer Branding in the IT Industry: An Employer view	548
Amir Dabirian (KTH Royal Institute of Technology)	

ITiP 4

A Game-Theoretic Analysis of Pricing Competition between Aggregators in V2G Systems	549
Information Exposure (IEX): A New Class in the Bugs Framework (BF) Irena Bojanova (NIST), Yaacov Yesha (UMBC and NIST), Paul E. Black (NIST), and Yan Wu (BGSU)	559
DroidPatrol: A Static Analysis Plugin For Secure Mobile Software Development	565
Open Source Fog Architecture for Industrial IoT Automation Based on Industrial Protocols Mohammad Ghazi Vakili (Politecnico di Torino), Claudio Demartini (Politecnico di Torino), Mauro Guerrera (Politecnico di Torino), and Bartolomeo Montrucchio (Politecnico di Torino)	570

CAP: Computer Architecture & Platforms

CAP 1

An Actor-Based Design Platform for System of Systems	. 579
Marjan Sirjani (Mälardalen University), Giorgio Forcina (Mälardalen	
University), Ali Jafari (Reykjavik University), Stephan Baumgart	
(Volvo Construction Equipment AB), Ehsan Khamespanah (University of	
Tehran, Reykjavik University), and Ali Sedaghatbaf (Mälardalen	
University)	
Detecting Malicious Attacks Exploiting Hardware Vulnerabilities Using Performance Counters	. 588
Congmiao Li (University of California, Irvine) and Jean-Luc Gaudiot	
(University of California, Irvine)	

CAP 2

Research on Index Mechanism of HBase Based on Coprocessor for Sensor Data Feng Ye (Hohai Univercity), Songjie Zhu (Hohai University), Yuansheng Lou (Hohai University), Zihao Liu (Jiangsu University of Science and Technology), Yong Chen (Nanjing Longyuan Micro-Electronic Company), and Qian Huang (Hohai University)	598
Testing Performance-Isolation in Multi-core Systems	604
Jakob Danielsson (Mälardalen University), Tiberiu Seceleanu (ABB AB),	
Marcus Jägemar (Ericsson AB), Moris Behnam (Mälardalen University),	
and Mikael Sjödin (Mälardalen University)	

HCSC: Human Computing & Social Computing

HCSC 1

Optimal Hand Sign Selection Using Information Theory for Custom Sign-Based Communication	10
Fouch-Based Ontology Browsing on Tablets and Surfaces 6	16
Fulvio Corno (Politecnico di Torino), Luigi De Russis (Politecnico di	
Torino), and Luisa Barrera León (Politecnico di Torino)	
Investigating Differences in Gaze and Typing Behavior Across Age Groups and Writing Genres	22
Jun Wang (The Hong Kong Polytechnic University), Eugene Yujun Fu (The	
Hong Kong Polytechnic University), Grace Ngai (The Hong Kong	
Polytechnic University), and Hong Va Leong (The Hong Kong Polytechnic	
University)	

HCSC 2

Application of Reconstructed Phase Space in Autism Intervention	630
Piyush Saxena (Direct Supply), Devansh Saxena (Marquette University),	
Xiao Nie (Direct Supply), Aaron Helmers (Direct Supply), Nithin	
Ramachandran (Direct Supply), Alana McVey (Marquette University), Amy	
VanHecke (Marquette University), and Sheikh Ahamed (Marquette	
University)	
Using Gamification to Motivate Occupants to Energy Efficiency in a Social Setting of a Building	
Automation System	638
Joana Páris (Universidade Nova de Lisboa), João Cambeiro (Universidade	
Nova de Lisboa), Vasco Amaral (Universidade Nova de Lisboa), and	
Armanda Rodrigues (Universidade Nova de Lisboa)	
A Tale of the Social-Side of ASD	644
Shameem Ahmed (Western Washington University), Md. Forhad Hossain	
(Missouri State University), Kurt Price (Western Washington	
University), Cody Pranger (Western Washington University), Md. Monsur	
Hossain (We Work), and Moushumi Sharmin (Western Washington	
University)	

HCSC 3

Analyzing Happiness: Investigation on Happy Moments using a Bag-of-Words Approach and Related	
Ethical Discussions	. 653
Riddhiman Adib (Marquette University), Eyad Aldawod (Marquette	
University), Nathan Lang (Marquette University), Nina Lasswell	
(Marquette University), and Shion Guha (Marquette University)	
Long-Term Monitoring of NIRS Signals for Mental Health Assessment	. 663
Labiblais Rahman (Nihon University, Japan) and Katsunori Oyama (Nihon	
University, Japan)	

MOWU: Mobile, Wearable & Ubiquitous Computing

MOWU 1

Barrier Detection Using Sensor Data from Multiple Modes of Transportation with Data Augmentation	667
An Energy Efficient Pedestrian Heading Estimation Algorithm using Smartphones Yankan Yang (Inner Mongolia University), Baoqi Huang (Inner Mongolia University), and Runze Yang (Inner Mongolia University)	676

MOWU 2

Towards An Effective and Efficient Machine-Learning-Based Framework for Supporting Event Detection	
in Complex Environments	. 685
Alfredo Cuzzocrea (University of Trieste), Enzo Mumolo (University of	
Trieste), and Marco Tessarotto (University of Trieste)	

SCH: Smart & Connected Health

SCH 1

Improving Classification of Breast Cancer by Utilizing the Image Pyramids of Whole-Slide Imaging and Multi-scale Convolutional Neural Networks Li Tong (Georgia Institute of Technology and Emory University), Ying Sha (Georgia Institute of Technology), and May D Wang (Georgia Institute of Technology and Emory University)	. 696
Compliance Checking of Open Source EHR Applications for HIPAA and ONC Security and Privacy	
Requirements Maryam Farhadi (Kennesaw State University), Hisham Haddad (Kennesaw State University), and Hossain Shahriar (Kennesaw State University)	. 704
Computer Vision Based Systems for Human Pupillary Behavior Evaluation: A Systematic Review of the	
Literature Cleyton Rafael Gomes Silva (Federal University of Goias), Cristhiane Gonçalves (Federal University of Goias), Joyce Siqueira (Federal University of Goias), Fabrizzio A. A. De Melo Nunes Soares (Federal University of Goias), Rodrigo Albernaz Bezerra (Federal University of Goias), Hedenir Monteiro Pinheiro (Federal University of Goias), Ronaldo Martins Da Costa (Federal University of Goias), Eduardo Nery Rossi Camillo (Goias Eye Bank Hospital), and Augusto Paranhos Junior (Goias Eye Bank Hospital)	. 714

SCH 2

The Causes Analysis of Ischemic Stroke Transformation into Hemorrhagic Stroke using PLS (partial Least Square)-GA and Swarm Algorithm	720
Chiĥhsiong Shih (Tunghai University), Cheng-Chung Chu William (Tunghai University), and You-Wei Chang (Tunghai University)	
Fully Automatic Intervertebral Disc Segmentation Using Multimodal 3D U-Net	730
Chuanbo Wang (University of Wisconsin-Milwaukee), Ye Guo (University	
of Wisconsin-Milwaukee), Wei Chen (Army Medical University), and Zeyun Yu (University of Wisconsin-Milwaukee)	
Tu (University of Wisconsin-Mitwaukee)	
Comparing Health Outcomes in San Francisco and Boston Metro Areas	.740
Swapna Gokhale (University of Connecticut)	

SEPT: Security, Privacy & Trust in Computing

SEPT 1

Dynamic Data Publishing with Differential Privacy via Reinforcement Learning Ruichao Gao (Inner Mongolia University) and Xuebin Ma (Inner Mongolia University)	
Enforcing Optimal Moving Target Defense Policies	
Jianjun Zheng (Texas Tech University) and Akbar Siami Namin (Texas	
Tech University)	
Automatic Detection of NoSQL Injection Using Supervised Learning	
Md Rafid Ul Islam (Bangladesh University of Engineering & Technology),	
Md. Saiful Islam (Bangladesh University of Engineering & Technology),	
Zakaria Ahmed (Bangladesh University of Engineering & Technology),	
Anindya Iqbal (Bangladesh University of Engineering & Technology), and	
Rifat Shahriyar (Bangladesh University of Engineering & Technology)	

SEPT 2

Exploration into Gray Area: Efficient Labeling for Malicious Domain Name Detection	
Naoki Fukushi (Waseda University), Daiki Chiba (NTT Secure Platform	
Laboratories), Mitsuaki Akiyama (NTT Secure Platform Laboratories),	
and Masato Uchida (Waseda University)	
Precise and Robust Detection of Advertising Fraud	
Fumihiro Kanei (NTT Secure Platform Laboratories, Japan), Daiki Chiba	
(NTT Secure Platform Laboratories, Japan), Kunio Hato (NTT Secure	
Platform Laboratories, Japan), and Mitsuaki Akiyama (NTT Secure	
Platform Laboratories, Japan)	
Raising the Bar Really High: An MTD Approach to Protect Data in Embedded Browsers	
Fadi Mohsen (University of Michigan-Flint) and Haadi Jafaarian	
(University of Colorado Denver)	

SEPT 3

Efficient SVM Based Packer Identification with Binary Diffing Measures	795
Yeongcheol Kim (Chungnam National University, South Korea), Joon-Young	
Paik (Tianjin Polytechnic University, China), Seokwoo Choi (National	
Security Research Institute, South Korea), and Eun-Sun Cho (Chungnam	
National University)	
CTRL-ALT-LED: Leaking Data from Air-Gapped Computers Via Keyboard LEDs	301
Mordechai Guri (Ben-Gurion University of the Negev), Boris Zadov	
(Ben-Gurion University of the Negev), Dima Bykhovsky (Ben-Gurion	
University of the Negev; Shamoon College of Engineering, Beer-Sheva,	
Israel), and Yuval Elovici (Ben-Gurion University of the Negev)	

Zeyad Al-Odat (North Dakota State University) and Samee Khan (North	
Dakota State University)	

SEPT 4

CSKES: A Context-Based Secure Keyless Entry System	817
Privacy Is The Best Policy: A Framework for BLE Beacon Privacy Management	823
Emmanuel Bello-Ogunu (The University of North Carolina at Charlotte),	
Mohamed Shehab (The University of North Carolina at Charlotte), and	
Nazmus Sakib Miazi (The University of North Carolina at Charlotte)	
Safety and Security Co-Analyses: A Systematic Literature Review	833
Elena Lisova (Malardalen University), Irfan Sljivo (Malardalen	
University), and Aida Causevic (Malardalen University)	

SISA: Smart IoT Systems & Applications

SISA 1

 Sensor Networks and Data Management in Healthcare: Emerging Technologies and New Challenges	334
Multi-Breath: Separate Respiration Monitoring for Multiple Persons with UWB Radar	340
Yanni Yang (The Hong Kong Polytechnic University), Jiannong Cao (The	
Hong Kong Polytechnic University), Xiulong Liu (The Hong Kong	
Polytechnic University), and Xuefeng Liu (Beihang University)	
sEmoD: A Personalized Emotion Detection Using a Smart Holistic Embedded IoT System	350
Akm Jahangir Alam Majumder (University of South Carolina Upstate),	
Tanner M. Mcwhorter (Miami University), Yezhou Ni (Miami University),	
Hanging Nie (Miami University), Jacob Iarve (Miami University), and	

SISA 2

Donald R Ucci (Miami University)

RIVER-MAC: A Receiver-Initiated Asynchronously Duty-Cycled MAC Protocol for the Internet of Things 860 Mathew L. Wymore (Iowa State University) and Daji Qiao (Iowa State University)

GeneSIS: Continuous Orchestration and Deployment of Smart IoT Systems	370
Nicolas Ferry (SINTEF ICT), Phu Nguyen (SINTEF ICT), Hui Song	
(SINTEF), Pierre-Emmanuel Novac (University Côte d'Azur, CNRS),	
Stéphane Lavirotte (University Côte d'Azur, CNRS), Jean-Yves Tigli	
(University Côte d'Azur, CNRS), and Arnor Solberg (TellUAS)	
Degree Distribution of Wireless Networks for Mobile IoT Applications	376
Renato Ferrero (Politecnico di Torino) and Filippo Gandino	

(Politecnico di Torino)

Student Research Symposium

SRS 1

Experimental Comparison of Pure Flooding and Its Variants on Raspberry Pi in Small Scale Ad Hoc Networks Sangwoo Jung (CNU) and Ki-Il Kim (CNU)	882
Threshold-Driven Class Decomposition Mohammed Hamdi (Oakland University), Rashmi Pethe (Oakland University), Annapoorani Sevugan Chetty (Oakland University), and Dae-Kyoo Kim (Oakland University)	884
Optimized Division of Exploration Areas in Multi-robot Systems Considering Static and Dynamic Charging Stations Robison Cris Brito (Federal University of Technology (UTFPR)), Nicollas Saque (Federal University of Paraná (UFPR)), Diego Addan Gonçalves (Federal University of Paraná (UFPR)), Fabio Favarim (Federal University of Technology (UTFPR)), and Eduardo Todt (Federal University of Paraná (UFPR))	888
Analysis of the Evolution of the Influence of Central Nodes in a Twitter Social Network Minami Uehara (University of Tsukuba, Japan) and Sho Tsugawa (University of Tsukuba, Japan)	892

SRS 2

Spam Domain Detection Method Using Active DNS Data and E-Mail Reception Log	896
Kenya Dan (Tokyo University of Agriculture and Technology), Naoya	
Kitagawa (Tokyo University of Agriculture and Technology), Shuji	
Sakuraba (Internet Initiative Japan Inc., Japan), and Nariyoshi Yamai	
(Tokyo University of Agriculture and Technology)	

A Traffic Distribution System Among Multiple Terminals Using MPTCP in Multihomed Network Environment.... 900

Ryuji Asakura (Tokyo University of Agriculture and Technology), Reido Horigome (Tokyo University of Agriculture and Technology), Nariyoshi Yamai (Tokyo University of Agriculture and Technology), Naoya Kitagawa (Tokyo University of Agriculture and Technology), and Satoshi Ohzahata (The University of Electro-Communications)

A Modified Smart Contract Execution Enviroment for Safe Function Calls	.904
Sooyeon Lee (Chungnam National University, Republic of Korea) and	
Eun-Sun Cho (Chungnam National University, Republic of Korea)	
Parallelization of Plane Sweep Based Voronoi Construction with Compiler Directives	908
Anmol Paudel (Marquette University), Jie Yang (Marquette University),	

Fast Abstracts

and Satish Puri (Marquette University)

Fast Abstract 1

Toward an Optimal Anomaly Detection Pattern in Wireless Sensor Networks	912
Application and Research of Image-Based Modeling and 3D Printing Technology in Intangible Cultural Heritage Quanzhou Marionette Protection	914
Logistic Regression and Random Forest for Effective Imbalanced Classification	Э16
 Improving Prediction Accuracy for Logistic Regression on Imbalanced Datasets	∂ 18
IoT Malware Analysis	920
Blockchain Development Platform Comparison	922
Protecting Data in Android External Data Storage	∂ 24
Achievements Visualization in Programming Education	926
Virtualization for Flexibility and Network-Aware on 5G Mobile Devices	928

Fast Abstract 2

CloneTM: A Code Clone Detection Tool Based on Latent Dirichlet Allocation	930
VisioTM: A Tool for Visualizing Source Code Based on Topic Modeling) 32
A Protocol for Preventing Transaction Commitment without Recipient's Authorization on Blockchain) 34
Rogue Wireless AP Detection using Delay Fluctuation in Backbone Network) 36
Trust-Oriented Live Video Distribution Architecture	938
Detecting No-Sleep Bugs Using Sequential Reference Counts	940
A Linear Regression Approach to Modeling Software Characteristics for Classifying Similar Software	942
Load Balancing Algorithm for Multiple UAVs Relayed Tactical Ad Hoc Networks	Э44
Garbage Weight Estimation System	946

Fast Abstract 3

Topic Shift Detection in Online Discussions using Structural Context	948
A Clicked-URL Feature for Transactional Query Identification Yingcheng Sun (Case Western Reserve University) and Kenneth Loparo (Case Western Reserve University)	950
Context Aware Image Annotation in Active Learning with Batch Mode Yingcheng Sun (Case Western Reserve University) and Kenneth Loparo (Case Western Reserve University)	952

Information Extraction from Free Text in Clinical Trials with Knowledge-Based Distant Supervision	954
Basic Concept of Emergency Optical Network Planning Using Multiagent-Based Flexible and Autonomous	
	956
Satoru Izumi (Tohoku University), Masaki Shiraiwa (National Institute	
of Information and Communications Technology), Goshi Sato (National	
Institute of Information and Communications Technology), Sugang Xu	
(National Institute of Information and Communications Technology), and	
Takuo Suganuma (Tohoku University)	
Basic Design of Network Control Method Based on Disaster Risk of OpenFlow C/M-Plane Satoru Izumi (Tohoku University), Hiroyuki Takahira (Tohoku	958
University), Kosuke Gotani (Tohoku University), Misumi Hata (Tohoku	
University), Luis Guillen (Tohoku University), Toru Abe (Tohoku	
University), and Takuo Suganuma (Tohoku University)	
Predicting Opioid Use Disorder (OUD) Using A Random Forest Adway Wadekar (Saint John's High School)	960

Empowering Engagement through Automatic Formative Assessment

Alice Barana Department of Mathematics University of Turin Torino, Italy 0000-0001-9947-5580 Marina Marchisio Department of Mathematics University of Turin Torino, Italy 0000-0003-1007-5404 Sergio Rabellino Department of Computer Science University of Turin Torino, Italy 0000-0002-1757-2000

Abstract—Engagement is one of the most powerful driving forces that moves learners forward in a learning experience. The effects of engagement are related with enjoyment in learning, good results and prevention to drop-outs. Technologies, and in particular automatic formative assessment, can positively influence engagement, as well as providing data to measure it. This paper discusses how computer-based activities with automatic formative assessment and interactive feedback can promote engagement in Mathematics at school level. After discussing a theoretical framework for defining and measuring engagement, the potentialities of activities with automatic formative assessment and interactive feedback designed according to a specific model are illustrated. The activities have been experimented with 299 students of grade 8 in Italy. Through questionnaires administered before and at the end of the experimentation and through data from the online platform, the effects of the online activities on student engagement is evaluated. The activities resulted particularly effective to increase the engagement level in students that had low attitudes toward Mathematics at the beginning of the project.

Keywords—Automatic Formative Assessment, Engagement, Interactive Feedback, Mathematics Education, Online Learning.

I. INTRODUCTION

Engagement is one of the most powerful driving forces that moves learners forward in a learning experience. When students are engaged in a task, they tend to keep focused, to enact deep learning strategies and self-regulation, to achieve good results and even to get satisfaction and pleasure for their activity. More generally, encouraging and controlling engagement is an effective communication technique: the evolution of audience attention is a key factor that has to be taken into consideration when scheduling the duration not only of lessons, but also of presentations and performances. Going back to education, increasing engagement can be useful because engaged learners may enter a virtuous circle: when they obtain good results because of their work, their self-efficacy beliefs are intensified, and this keeps them engaged and makes them continue succeeding. On the other side, disengaged students who have difficulties in achieving good results may be trapped in a negative rather than positive loop, which hinders them from success [1]. Students coming from poor families or low socioeconomic status might find it harder to be engaged in learning activities than students coming from medium-high social class, due to the little support they may find in their family, or to the greater difference they may perceive between their school and home environments [2]. These might be the root cause for dropouts and early school leaving, and that is why supporting didactic projects, aimed at enhancing learning engagement especially for students with challenging backgrounds, is often a key strategy pursued by policymakers and institutions, interested in improving the quality of education over their territory.

In Mathematics, which is often considered a "hard science", engagement is related to the development of strong aspirations for carrying on advanced studies in this field [1]. Since the development of Mathematical understanding is a crucial access key for workplace in the modern society, much attention to student engagement in Mathematics should be paid by teachers and educators from the very early school years.

Even more than for traditional schooling, engagement is a key point especially for online learning. In particular, it has been shown to be a strong predictor of MOOC retention [3]: studying solutions aimed at keeping users engaged is crucial to increase the completion rates, which are often very low due to the weak motivations of the enrolled students to complete the courses [4]. Technologies often support the engagement process: learning materials provided through gamification, simulation or interaction seem to be more effective than static resources in maintaining the users involved.

Technology also supports the teaching process: in virtual environments, unlike in traditional classrooms, the large amount of data registered and made available by the systems can provide useful information that help researchers understand the processes activated during learning situations, evaluate the effectiveness of teaching strategies and support teacher decisions.

This paper discusses how computer-based activities with automatic formative assessment and interactive feedback can promote engagement in Mathematics at school level. After discussing a theoretical framework for defining and measuring engagement, the potential of activities with automatic assessment designed according to a specific model are illustrated. Then, a didactic experiment conducted at grade 8 is presented. The methods for data collection and analysis are described in detail and the results are shown and discussed in the light of the theoretical framework.

II. THEORETICAL FRAMEWORK

A. What is "engagement"?

Engagement is highly studied in the educational research field, and it is possible to find many different definitions and characterizations of this construct in literature. Some authors associate engagement with the level of attention [5] or with motivation [6]; others conceptualize it in terms of visible students' behavior which should reflect the way they engage with learning materials [7], in terms of intensity and quality of students' involvement in learning activities [8], or in terms of effort and investment students expend in the learning task [9]. In all these researches, active participation is a central theme for understanding engagement. In this paper we accept the definition given by Ng, Barlett and Elliott, who refer to engagement as "students' dynamic participation and coparticipation in recognition of opportunity and purpose in completing a specific learning task" [1]. Peculiarity of this definition is the characterization of engagement as an interactive and purposive process; it allows to examine how it may change over time and vary according to situations and contexts. When students participate eagerly to a specific learning activity, they deploy appropriate strategies, regulate processes and monitor their actions. They feel happy, spend time and effort on the task, and show high levels of focus and concentration. However, these conditions may fluctuate: sometimes, students can fail to plan their actions, feel worried or not so willing to making efforts, or they can become distracted. Thus, engagement is not a mere personal property, but it is a set of actions undertaken by a person in a specific context where interactions with other people, artifacts and tools occur. Engagement has a fluctuant nature, this means that it depends on specific situations and it may change over time; it has a focal object, it is situational and malleable: it can be modified by changing task design, support or rules project [1]. The malleable nature of engagement means that it is possible to create repeated episodes eliciting engagement and so, in time, contribute to these students establishing positive stable beliefs and behaviors [10].

Despite the number of definitions, researches on engagement agree on the fact that it is a multidimensional construct [11]. The number and nature of the components may vary, but the main trend recognizes the three main components of student engagement identified by Fredricks, Blumenfeld and Paris: behavioral, emotional and cognitive [12]. *Behavioral engagement* concerns involvement in learning activities and it includes behaviors such as effort, persistence, attention, concentration, and completion of work [12]. Other definitions of behavioral engagement entail positive conduct, such as following rules, or participating to school-related activities [1]. When the focus is on homework, effort expenditure and timely completion are indicators of behavioral engagement. However, strict adherence to norms is not a good indicator for high-order thinking, enjoyment and interest: students could just keep quiet and pretend to pay attention, their level of interest being indeed very low [13].

Emotional engagement is understood as students' affective reactions in a classroom, which can vary from interest to boredom, from happiness to sadness, from satisfaction to anxiety. Interest and value for learning are important indicators of emotional engagement [12]. Emotional engagement is linked to several outcomes, such as improved persistence, learning achievements, but also liking school subjects and positive attitudes towards school [1].

Cognitive engagement is the mental investment people make in learning; it involves the use of deep strategies, self-regulation, openness to problem solving [12]. A high level of cognitive engagement can be detected when students enter into an interactive dialogue to generate new knowledge [7]. Students with high levels of cognitive engagement are less likely to give up their learning and more likely to keep engaged with school [1].

There are other dimensions that are considered by scholars in addition to these three. For instance, Appleton, Christenson and Furlong considered a four-component model having psychological engagement as fourth dimension [14]. Patrick, Ryan and Kaplan found that the social component is a strong factor of engagement [15]. Social engagement can be defined as the extent to which students collaborate and share responsibility in order to complete a task.

Ng, Barlett and Elliott make a distinction between these components, labelled as "indicators" of engagement, and the so called "facilitators" of engagement, which can be cognitive and social [1]. Among cognitive facilitators, the most acknowledged are:

- self-efficacy, that is the child's perceived ability to successfully complete a task within a specific domain or setting. Students who have developed a high sense of self-efficacy are more confident in their capacities and are more likely to get involved in tasks;
- achievement goals, that are students' perceived purposes to learn. There are two main types of achievement goals: mastery and performance goals. Mastery goals represent a focus on learning for the sake of improvement and understanding, whereas performance goals reflect students' attention to achievement; the former are more desirable than the latter, as they are linked to highest outcomes;
- autonomy, which refers to the choices that students can make freely during the learning process, and it is at the

core of the self-determinant theory for promoting engagement [6];

• interest, which involves both cognitive and emotional components and promotes and supports learning motivations.

Social facilitators refer to social conditions, interactions, and relationships that promote engagement; especially for adolescents, the influence that peers, teachers, family and the environment have on them is critical for their choices, behaviors and emotions [16].

This framework allows to classify research models based on engagement: in facilitators-focused models engagement is a desired outcome, while in outcome-focused models engagement is a mediator. In literature, the indicators of engagement are operationalized and translated into variables with the aim of measuring the level of engagement [11].

Low levels of engagement and low achievements are often related. In a study conducted by OECD after PISA 2012 survey it emerges that low performers in Mathematics, that are students who scored less than 2 in a 6-level scale, are less interested in Mathematics than better-performing students. They report low levels of perseverance, that is associated with work ethic. It seems that they do not devote enough time to homework, and that their effort in school related activities is not very productive, as it does not result in significative outcomes. Moreover, they tend to skip classes and school days and they show little sense of belonging at school [17]. Students with disadvantaged background are more likely low performers than top ones: socioeconomic status has a great influence on school achievements as well as on attitudes and beliefs towards Mathematics [18].

B. Engagement and learning technologies

Students engagement in technology enhanced learning environments includes any interaction of the learner with instructors, peers or learning content through the use of digital technology; this can happen face to face or remotely, and the courses involved may be entirely online, blended, or face to face [19]. The potential of technologies can open up new ways in the research about engagement, contributing both to the measurement of the indicators and to the creation of facilitators of engagement.

When the focus is the detection of engagement, technologies offer many sources of data such as logs, registration of dialogues and answers that can be usefully integrated in the research [8]. Many authors agree that the mere number of logs is not a reliable indicator of behavioral engagement, if considered alone: the amount of time and actions spent on activities may vary largely among students according to their cognitive needs or to external factors. However, the data provided by automated systems can be combined in order to generate meaningful information about user experience [19].

Digital technologies are acknowledged to be powerful cognitive and social facilitators of engagement for several reasons: they enhance the possibilities to activate learning by doing or active learning strategies, which enable students to intellectually engage in the task [20], [21]; they increase the chances of interactions among peers and with the instructor [8];

they can facilitate self-regulation and adaptive learning through formative assessment [22], [23]; asynchronous activity enable learners to study at their own pace and to reflect on the learning process [24], [25]; in Mathematics and other STEM, they let you analyze real-world problems, thus making the subject interesting and relevant [10], [26].

C. Engagement and automatic formative assessment

If online learning can provide new tools for engaging students, its effect when combined with formative assessment should be promising. In fact, formative assessment strategies as prompting discussions, providing feedback that move learners forward, activating students and peers as protagonists of their learning and have, as their main consequence, that of acting on student engagement [27], [28].

Our research group proposed a model of automatic formative assessment for developing learning activities using an automatic assessment system particularly suitable for Mathematics; the model has been experimented using Moebius Assessment (formerly known as Maple TA) [29], a system based on the mathematical engine of Maple, which allows the definitions of mathematical formulas, graphics and algorithms running behind questions. In particular, it is possible to define grading codes aimed at evaluating mathematical answers for their mathematical meaning, and to create worksheets with several possibilities of interaction, step-by-step guided resolutions, and allow students to enter graphs or symbolic formulas [30], [31]. Fig. 1 shows one example of question created through Moebius Assessment: the geometrical figure was created through Maple's mathematical engine, the formula of the area is accepted if written in a mathematical correct form, even if different from the provided solution. Moreover, a preview of the graph of the function entered is provided to students before grading the answer, in order to enable them to self-assess their answer.

The model was shown to bring enhancement both to teaching and learning [23], [32], acting on competence and on self-regulation [33], [34]. According to this model, learning activities created with an adequate automatic assessment system should have the following features:

- Availability. Assignments are always available to students, who can attempt them at their own pace, without limitation in data, time and number of attempts.
- Algorithm-based questions and answers. Random values, parameters or formulas make questions, and the relative answers randomly change at every attempt. This can be realized through the implementation of algorithms running behind the questions. By algorithmic variables, different representational registers (words, numbers, symbols, tables, graphics, schemas) can be shown in questions and feedback.
- Open answers. The multiple choice modality is avoided whenever possible, to make room for open answers, where students are asked to respond in one of the different registers listed above.

- Immediate feedback. Results are computed in a few moments and they are shown to the students while they are still focused on the task.
- Interactive feedback. Right after answering one question the system can show whether it is correct and go through a step-by-step guided resolution which interactively shows a possible process for solving the task. Students who fail to answer autonomously to the main question are asked questions about prerequisites or simpler tasks. At each step, if they give the wrong answer, the correct one is shown to be used in the following step. They can gradually acquire the background and the processes that enable them to solve the initial problem. They earn partial credits for the correctness of their answer in the step-by-step process.
- Contextualization. Whenever possible, questions refer to real-world issues which can be relevant to students as well as for the discipline.

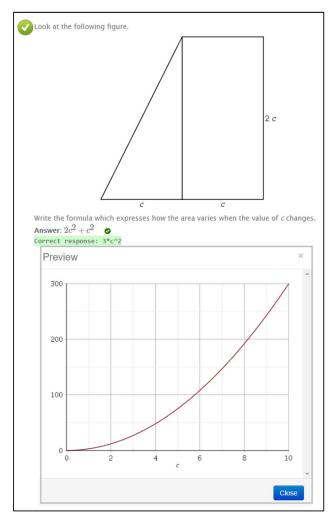


Fig. 1. Example of question created through Moebius Assessment

Fig. 2 shows one example of question created through Moebius Assessment according to the model previously described. Students can follow the step-by-step solution to learn how to apply the Pythagorean theorem to a real-world relevant situation. When students fail one answer to one of the subquestions, the correct solution is given and it can be used in the following steps, so that students can individuate mistakes and correct them before proceeding. Numeric answers are accepted within a tolerance and formulas are matched with their mathematical meanings.

These types of activities should be cognitive facilitators of engagement: when students fail one answer, the interactive feedback activates them through the solving process making it possible to individuate the exact source of the mistake, thus acting on self-efficacy. Students can try the question again and find different parameters, so they have to autonomously repeat the solving process. Multiple attempts before showing the correct solution act on autonomy as well. The immediate feedback helps them focus on mistakes as a source of knowledge and to set mastery goals instead of performance goals. Lastly, real-world settings act on the student's interest for the subject, connecting abstract Mathematics to the real world [35], [36].

III. RESEARCH HYPOTHESIS AND QUESTION

We can reasonably assume that students who, at the beginning of a learning path, show interest towards Mathematics and value their learning, who have high levels of self-efficacy, perseverance and openness to problem solving, can be easily engaged in learning activities of any kind. Therefore, we will focus on students at risk of disengagement in order to understand whether an experimental didactic intervention based on formative automatic assessment help students of 8th grade to reengage in Mathematics.

IV. RESEARCH METHOD

In order to investigate the research question, a didactic experimentation has been designed and realized by our research group in the city of Turin (Italy) in 2017/2018 school-year. The experimentation involved 299 8th grade students attending 6 different lower secondary schools in different areas of Turin. In particular, about half of the students belonged to low socio-economical class, while the other half to middle-high social class. The sample was composed by randomly selecting 13 classes from the 6 schools, which were entirely included in the project with their Mathematics teachers.

All the students filled in an initial questionnaire in November 2017, which aimed at investigating the initial level of students' attitude toward school and, more specifically, toward Mathematics. The questionnaire was administered online; it was composed of 35 Likert-scale questions inspired to PISA 2012 student questionnaire [37], with items about intrinsic and extrinsic motivation toward Mathematics, perceived control of success, Mathematics work ethic, Mathematics behavior, perseverance and openness to problem solving. Table 1 shows the items in the questionnaire. Intrinsic and extrinsic motivation are related to the emotional component of attitude toward Mathematics, as they entail the extent to which students are interested in and value Mathematics; work ethic, perseverance and Mathematics behavior are usually closely related and they

express the initial behavioral component of student attitude; the perceived control of success and openness to problem solving are indicators of cognitive engagement. All items were on a 4-points Likert scale (strongly disagree/disagree/agree/strongly agree) except for the items on perseverance and openness to problem solving, where, through a 5-point Likert scale, a neutral position was allowed.

C The new TV					
Marco wants to buy a new TV. In his livingroom he has a space 150 cm long where to place it. He needs to know what is the maximum					
size of the screen that he can buy.					
The size of the screens is given by the lenght of the diagonal line measured in inches.					
In meters, 1 inch = 2.54 cm					
Correct response: 2.54 cm (browse the Internet to find the data!)					
The size of modern TVs is 16:9. It means that the ratio between base and height is 16:9.					
In formulas, if b is the base and h is the height,					
<i>h</i> = 9/16*b ₫ 🖸 🖉					
Correct response: b*9/16					
VThe TV of maximum dimensions can have the base 150					
cm 📀					
Correct response: 150 cm long and the height 266					
cm					
Correct response: 84.38±0.01 cm long.					
The diagonal line will be 172.1 cm					
Correct response: 172.1±0.1 cm long.					
Thus, Marco can buy, at most, a 67					
Correct response: 67 inches TV.					

Fig. 2. Example of activity with automatic formative assessment and interactive feedback.

TABLE I.	ITEMS OF THE INITIAL Q	UESTIONNAIRE
----------	------------------------	--------------

Subscale	Code	Items	
	QII	I like lectures about Mathematics.	
Intrinsic	QI2	I can't wait for Mathematics lessons	
motivation	QI3	I do Mathematics because I like it	
	QI4	I am interested in the things that I learn in Mathematics	
Extrinsic QI5		Making an effort in mathematics is worthy because it will help me in the job that I want to do later on	
motivation	QI6	Mathematics is an important subject for me because I need it for what I want to study later on	

Subscale	Code	Items		
	QI7	If I put enough effort, I can succeed in Mathematics		
Perceived control of success	QI8	It is completely my choice whether or not I do well in Mathematics		
	Q19	Family demands or other problems prevent me from spending a lot of time for my Mathematics work		
	QI10	If lessons were different, I would try harder in Mathematics		
	QI11	Whether I study or not, I am bad at Mathematics		
	QI12	I finish my homework in time for mathematics class		
	QI13	I work hard on my Mathematics homework		
	QI14	I am prepared for my Mathematics exams		
Mathematics	QI15	I study hard for mathematics quizzes		
work ethic	QI16	I keep studying until I understand Mathematics material		
	QI17	I pay attention in Mathematics class		
	QI18	I avoid distractions when I am studying mathematics		
	QI19	I keep my Mathematics work well organized		
	QI20	I talk about mathematics problems with my friend		
	QI21	I help my friends with Mathematics		
Mathematics	QI22	I do Mathematics as an extracurricular activity		
behavior	QI23	I do Mathematics more than 2 hours a day outside of school		
	QI24	I play chess		
	QI25	I program computers		
	QI26	When confronted with a problem, I give up easily		
	QI27	I put off difficult problems		
Perseverance	QI28	I remain interested in the tasks that I start		
	QI29	I continue working on tasks until everything is perfect		
	QI30	When confronted with a problem, I do more than what is expected from me		
	QI31	I can handle a lot of information		
On one oco to	QI32	I am quick at understanding things		
Openness to problem	QI33	I seek explanations for things		
solving	QI34	I can easily link facts together		
	QI35	I like to solve complex problems		

Together with the questionnaire, they also completed a paper-based learning test with 10 multiple choice or open-ended items.

From December 2017 to June 2018 the classes with their teachers had access to an online course in a Moodle platform, populated with activities designed by the research group according to the model of formative automatic assessment. The online course covered the whole program for 8th grade Mathematics, ranging from negative numbers to solid geometry,

from linear functions to equations. Materials and methodologies were shared with teachers through periodic focus groups in order to enable them to use the materials autonomously, in class through the Interactive White Board or in a computer lab, or to ask students to work on them as homework. A group of lessons in each class were held under the supervision of research group members.

Fig. 3 shows one example of activity with automatic formative assessment on solid geometry. The computation of the volume of a parallelepiped is applied to a real-world problem; students who fail to find the correct solution autonomously can use the interactive feedback in support of their reasoning. Partial grade is allowed for answers in the step-by-step process. Students, teachers and researchers had access to the students' answers that could be used as prompts for discussion or research.

At the end of the school year students were asked to complete a second paper-based learning test with questions about the topics covered in the online course and used by all the classes. Moreover, they were asked to fill a second online questionnaire conceived to evaluate the impact of automatic assessment activities on student engagement. In deep, questions focused on emotional aspects, such as interest, enjoyment and value for the learning activities, and on cognitive aspects, like cognitive and metacognitive processes enacted by the learning materials. The 15 items are reported in Table III; they are all 5point Likert scale.

TABLE II. ITEMS OF THE FINAL QUESTIONNAIRE

Subscale	Code	Items			
Emotional engagement	QF1	Having the materials available in every moment is useful			
	QF2	The proposed problems are interesting			
	QF3	The tests are useful to practice			
	QF4	It is useful to visualize the correct answer after submitting a response.			
	QF5	The online assignments are a valid help for studying			
	QF6	Online assignments made me appreciate the topics studied more			
	QF7	Using the platform from home helped me identi the topics on which we worked in class			
	QF8	The online tests helped me better understand the topics studied			
	QF9	The online tests helped me understand if I understood the topics studied			
	QF10	The immediate feedback helped me understand how the task should be solved			
Cognitive Engagement	QF11	Problems with step-by-step resolution helped me understand the solving process			
	QF12	Online assignments helped me autonomously sol Mathematics exercises			
	QF13	Online assignments helped me become more confident about my capabilities			
	QF14	Online assignments helped me acknowledge my preparation			
	QF15	When I gave an incorrect answer, I used to try the exercise again			

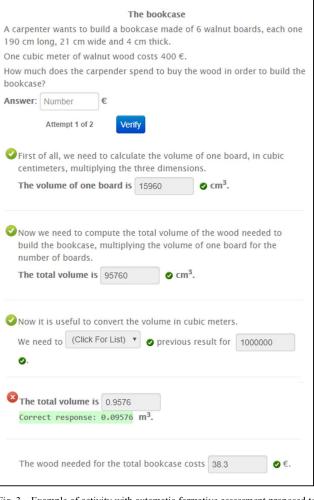


Fig. 3. Example of activity with automatic formative assessment proposed to the 8th grade students.

In order to evaluate the impact of the learning materials on student engagement, an initial profile of all the students has been depicted through their answers to the initial questions. Answers to Q19, Q110, Q111, Q126 and Q127 were reversed so that higher answers correspond to higher attitudes. Through factor analysis, answers were classified in three main sets, corresponding to emotional, behavioral and cognitive components of initial student attitudes; continual and categorical variables were defined as indicators of the initial level of student engagement in the three components.

Similarly, questions of the final questionnaire were split in two subscales, one related to emotional engagement and one related to cognitive engagement; two continuous variables were created as indicator of the final level of emotional and cognitive engagement. As behavioral indicator, the number of logs to any course activity, the number of submissions of automatically assessed assignments and the average rate of submission per assignment were collected and taken into consideration. The final level of engagement in each subscale was compared with the initial one; analyses were conducted using SPSS 25. The reliability of all the questionnaires and the subscales were checked through Cronbach Alpha.

Students' socio-economic factor was determined using data from national surveys; the sample has been split in two categories: students from low social class and students from medium or high social class. The division broadly coincides with the division in schools: 4 out of 6 schools considered for the experimentation were mostly attended by students with low socio-economic status. The two schools attended by students from medium-high class participated to the project with more classes.

The learning data are not considered for this study. They are object of study in other papers, where they are compared with the results of a control group [34], [38].

V. DATA ANALYSIS AND RESULTS

The initial questionnaire was answered by 278 out of 299 students (93%). Students who did not answer to the questionnaire were excluded from the sample. The reliability of the survey was checked through the Cronbach Alpha, which resulted sufficiently high (0.82). An initial factor analysis lead to the elimination of 5 variables: QI9, QI10, QI12, QI24 and QI25. As of the last two eliminated variables (playing chess and programming computers), they are probably not common actions for 8th grade students, so they didn't contribute effectively to the detection of Mathematics behavior. QI9 and OI10 ("Family demands or other problems prevent me from putting a lot of time into my Mathematics work" and "If lessons were different, I would try harder in Mathematics") probably concerned external factors compromising student success more than interior control of their actions. Although the effect of the teacher and the family environment may be important factors for learning and developing competences, they are usually not such as to impede school work and the achievement of success [17]. Regarding completing homework before classes (QI12), it has been noticed that the mere compliance with rules does not necessarily imply engagement: the homework can be finished just to avoid punishments (at grade 8 many teachers are usually very strict in demanding that homework is done on time) but this does not necessarily mean that behind the homework there is effort, and it could also be copied from classmates. The Cronbach Alpha computed on the remaining 30 items increased to 0.85. Factor analysis also evidenced a distinction between intrinsic and extrinsic motivation, which again is not in disagreement with literature: in the OECD volume "Low performing students" only intrinsic motivation is considered in analyzing the relation between drive and achievements [17]. Students who are intrinsically motivated engage in tasks because they simply enjoy them, and they are more likely to set mastery goals; on the other hand, students who are extrinsically motivated tend to set performance goals, that are less stable than mastery ones [39].

Through factor analysis, three standardized variables were created as linear function of students' answers to the remaining items: intrinsic motivation, which represent the emotional component of initial level of student engagement (EE_i); behavioral engagement (BE_i), composed by items on Mathematics work ethic from QI13 to QI19, items on Mathematics behavior from QI19 to QI23, and by all items of

perseverance; cognitive engagement (CE_i), to which items QI7, QI8 and QI11 on perceived control of success and all items on openness to problem solving contribute.

The three categorical variables were built on the basis of the sum of students' answers to questions in the subscales. The variable had values 1, 2 or 3, meaning low, moderate high and high attitudes. In particular, for emotional engagement (EE_cat), the value of the sum of the answers to the four items could range from 4 to 16; a low level was defined for values equals or below 8; a medium-high level was defined for values ranging from 9 to 12 and high level for values higher than 12. For cognitive engagement (CE_cat) the cut-off values were 21 and 29 in a range from 8 to 37; for behavioral engagement (BE_cat) the cut-off values were 37 and 53 in a range from 16 to 69. Table II shows the distribution of students for each of the three variables. It is noticeable that students in the lowest levels of engagement are the minority.

TABLE III. DISTRIBUTION OF STUDENTS FOR LEVEL OF INITIAL ATTITUD	Е
--	---

	Percentage of students		
Initial engagement level	Emotional Engagement (EE_cat)	Cognitive Engagement (CE_cat)	Behavioral Engagement (BE_cat)
Low level	8.3%	20.9%	15.0%
Moderate high level	23.4%	52.2%	72.6%
High level	68.3%	27.0%	11.6%

The final questionnaire was answered by 75% of students; all of them had previously completed the initial questionnaire. The decrease in the number of students completing the survey is probably due to the fact that, at the end of the school-year, teachers had more difficulties and less time to verify that students filled it. However, Cronbach Alpha is very high (0.90), showing a high reliability of the items. In order to exclude the hypothesis that students who did not answer to the final questionnaire were concentrated in the lowest levels of initial engagement, a Chi-squared test was run to verify the incidence of having or not having answered to the final questionnaire on the distributions of the initial levels of emotional, cognitive and behavioral engagement. None of the tests gave significant results (p=0.25 for emotional engagement; p=0.33 for cognitive engagement; p=0.66 for behavioral engagement), meaning that answerers and non-answerers were equally distributed in terms of initial engagement.

As a preliminary analysis, variance analysis (ANOVA) was conducted over the students' answers on the single final questionnaire items of the two subscales, considering as independent variables the corresponding level of initial engagement, emotional or cognitive. For almost all the items, the initial level of engagement did not explain students' answers (p>0.1); for some items the trend was even decreasing, meaning that students with low levels of engagement showed a higher interest for this kind of activities than their classmates; only items QF2 and QF6 registered a significative, though weak, dependence on the initial emotional engagement level. Table IV reports some examples of students' answers to the items of the cognitive subscales, analyzed for levels of initial engagement.

Initial cognitive	Means of students' answers		
engagement level	QF14	QF10	QF13
Low level	3.66	4.07	3.59
Moderate high level	3.53	3.96	3.53
High level	3.81	4.30	3.66
p-value	0.31	0.16	0.75

TABLE IV. EXAMPLES OF STUDENTS' ANSWERS TO THE FINAL QUESTIONNAIRE FOR LEVEL OF COGNITIVE ENGAGEMENT

These results are promising for our research question; however, as one can reasonably suppose, students in the highest initial levels of engagement tended to give higher answers than students from lowest levels: deeper analyses are needed in order to show the impact of the adopted methodologies on initially demotivated students.

Factor analysis, repeated on the final questionnaire, left the original schema unchanged: thus, two standardized variables, one expressing the cognitive engagement and the other expressing emotional engagement, were created as output of the factor analysis; they embed the students' answers to the items considered in Table II. Moreover, ANOVA analysis was conducted over these variables, considering the corresponding initial engagement levels (EE_cat and CE_cat) as independent variables. Results were once again not significant, showing initially highly engaged students a little more engaged in online activities than the others, making it difficult to draw further conclusions.

In order to investigate whether any effect occurred on changes in the factors of student engagement, the difference between the level of emotional engagement with online activities observed with the final questionnaire and the initial level of emotional engagement was computed; the same was done for cognitive engagement. The variable expressing the difference in emotional engagement (EE_diff) has mean -0.062 and standard deviation 1.276; the variable expressing the difference in cognitive engagement (CE_diff) has mean -0.022 and standard deviation 1.329.

Through ANOVA, the dependence of EE_diff from the initial categorical level of emotional engagement (EE_cat) was tested. Results are reported in Table V: students with initial low levels of emotional engagement improved their level by 1.580, which is more than one standard deviation. The difference decreases as the initial engagement level increases. ANOVA test shows significant relations among the variables (p<0.0001); Eta test shows that this relation is moderate, explaining the 18% of the variance (Squared eta: 0.183, p<0.0001). The effect size, restricting the sample to students with initially low levels of engagement, is d=1.15, which is a noticeable value.

A similar analysis was conducted for cognitive engagement, by studying, through ANOVA, the dependence of CE_diff on the CE_cat. Results are reported in Table V: students with initial low levels of emotional engagement improved their level by 1.1810, which is similar to standard deviation. The difference decreases as the initial engagement level increases, reaching -1.076 for initially highly engaged students. ANOVA test shows significant relations among the variables (p<0.0001); Eta test shows that this relation is strong, explaining the 34% of the variance (Squared eta: 0.342, p<0.0001). The effect size, considering only students having low initial levels of engagement, is d=1.08, which is a very high value as well.

Initial level of emotional/cognitive engagement	EE_diff	CE_diff
Low level	1.5796	1.1810
Moderate high level	0.2747	0.1510
High level	-0.3393	-1.0764
p-value	< 0.0001	< 0.0001
Squared eta	0.183	0.342

TABLE V. DIFFERENCES BETWEEN INITIAL AND FINAL LEVEL OF ENGAGEMENT, PER LEVEL OF INITIAL ENGAGEMENT

The behavioral variables considered at the end of the didactic experience are: the number of logs registered in the platform, the number of submitted assignments on the platform and the rate of submission per assignment. Considering data from the informatic systems brings the advantage that they were collected for the whole 100% of students, so there are no missing data. Table VI shows means and standard deviations of the three variables.

TABLE VI. DATA FROM THE PLATFORM USAGE

Variables	Mean	<i>S.D</i> .
Number of logs	96.51	72.48
Number of assignment submissions	18.65	17.15
Rate of submission per assignment	1.66	0.87

From the literature we already know that logs are related to behavioral engagement, but they can be influenced by other factors, so they are not reliable indicators [19]. As a matter of fact, in our analysis 28% of variances of the number of log and 26% of variances of the number of submissions is explained by the class teacher: probably the way teachers asked students to do the online activities and the way they checked the homework impacted on student work. These variables turn out to be weakly associated with the initial level of behavioral engagement, as shown in Table VII: students with a low level of behavioral engagement tended to work less on the platform than their classmates. For the number of submissions, the relation is statistically significative; for the number of logs, however, there is not statistical evidence.

The situation changes when considering the average rate of resubmission per assignments. Recalling that the assignments have unlimited attempts, that numbers and situations change at every attempt and that mistakes are explained through interactive feedback, when students try questions again it means that they are eager to autonomously solve the problem, that they understood the solution and want to challenge themselves once again: thus, the task managed to engage students. This variable seems not to be related to the initial behavioral engagement level, as shown in Table VII (p=0.21). Even students with initially low levels of behavioral engagement could be engaged by activities with automatic formative assessment.

 TABLE VII.
 Average data from the platorm, per level of initial behavioral engagement

Initial level of behavioral engagement	Number of logs	Number of submissions	Rate of submissions per assignment
Low level	78.52	14.52	1.75
Moderate high level	97.33	18.45	1.60
High level	116.06	25.56	1.89
p-value	0.079	0.020	0.210

Lastly, we focused on the socio-economic status of the students, with the purpose of verifying that the online activities experimented were useful also to students with challenging backgrounds. Through ANOVA tests, we noticed that the two groups registered similar values in the difference between final and initial emotional engagement (p=0.273); from a cognitive point of view, engagement level grew more for students from a lower social class than those from a higher one (p=0.041); the same trend was registered in the rate of submission per assignment (p=0.027). Results are displayed in Table VIII. These results are extremely important since the sociocultural origin is a strong predictor of scholastic success [18]. Engaging students from low socio-economic status is challenging, but important to prevent drop-outs.

TABLE VIII. IMPACT ON ENGAGEMENT ON STUDENTS OF DIFFERENT SOCIO-ECONOMIC STATUS

Socio-economic status	EE_diff	CE_diff	Rate of submissions per assignment
Low	-0.204	0.252	1.80
Medium-high	0.024	-0.188	1.54
p-value	0.273	0.041	0.027

VI. DISCUSSION AND CONCLUSION

The data analysis showed that the online activities, designed accordingly to the model of automatic formative assessment with interactive feedback developed by the research group, successfully contributed to the engagement of students that initially showed little involvement in school. The theoretical framework on engagement helped distinguish the components and define variables to measure student engagement. We showed that students who, at the beginning of the school year, were little emotionally engaged with Mathematics, that is, they had little interest toward things learned and felt little enjoyment when learning Mathematics, could be engaged with this kind of online activities. We found out that students who are initially highly engaged with Mathematics remained engaged with this kind of activities, however the most important effects are perceived on initially little engaged students, whose engagement could increase.

The online activities managed to catch student attention thanks to the use of the computer, that is still an unusual practice in the majority of 8th grade classes in Italy, and thanks to the realworld settings, which help students associate a meaning to abstract concepts. The possibility to individuate, self-correct and understand mistakes offered by the automatic formative assessment is not possible with a traditional paper textbook: students acknowledge that this kind of work help them improve and value it. Moreover, the interactive feedback opens a dialogue between students and the system and encourages them to understand solving processes. There is a big difference between reading an example of correct resolution from a book or a file, or hearing it from the teacher at the blackboard, and following an interactive path where you are asked to insert answers to sub-tasks at each step. The latter is undoubtedly more engaging, under a cognitive perspective. Thus, even those students who, at the beginning of the school year, did not show much behavioral engagement towards Mathematics, that is, they were used to carry out little mathematical activity outside school and not to persevere in mathematical tasks, using the automatic assessment changed their attitude and made more attempts to the assignments. Seeing their scores increasing, their emotional engagement increased as well, and so did their cognitive engagement and their learning results, starting that positive loop that traditional instruction often fails to activate [1]. The impact of the activities on students with challenging backgrounds is of considerable importance, too. It is shown that school disengagement is related to disaffection, bullying, early school leaving and criminality. Increasing students' engagement in such environments is an outstanding goal [15].

The activities monitored in this project were not occasional practices, but they were regularly used over the school-year. Some teachers decided to exclusively use the online platform for assigning homework, thus abandoning the textbook. The final questionnaire was administered after 6 months of regular online activities: we can suppose that the effects on engagement acquired stability and could influence student beliefs towards Mathematics. If a similar methodology could become part of the everyday teacher practice as it was for the classes included in the experimentation, there would be decisive enhancements in the culture of Mathematics. It is fundamental to connect schools and universities with similar projects, that offer essential materials and data for advances in the research, but also valuable learning experiences for students. Many teachers who participated to the project enrolled to a teacher training course about these methodologies the following year, with the purpose of continuing to use these kinds of activities and becoming autonomous in the preparation of learning materials [40].

The University of Turin is experimenting similar educational models using the same technological asset for learning other disciplines, such as Science, Latin and Foreign Languages. Interdisciplinary research collaborations are essential to develop innovative materials for learning and new insights for the research.

REFERENCES

 C. Ng, B. Bartlett, and S. N. Elliott, Empowering engagement: creating learning opportunities for students from challenging backgrounds. New York, NY: Springer Science+Business Media, 2018.

- [2] J.J. Appleton, S.L. Christenson, D. Kim, and A.L. Reschly, "Measuring cognitive and psychological engagement: Validation of the Student Engagement Instrument," in Journal of School Psychology, vol. 44, no. 5, pp. 427–445, 2006.
- [3] Y. Xiong, H. Li, M.L. Kornhaber, H.K. Suen, B. Pursel, and D.D. Goins, "Examining the Relations among Student Motivation, Engagement, and Retention in a MOOC: A Structural Equation Modeling Approach," in Global Education Review, vol. 2, no. 3, pp. 23–33, 2015
- [4] K. Jordan, "Initial trends in enrolment and completion of massive open online courses," The International Review of Research in Open and Distributed Learning, vol. 15, no. 1, 2014.
- [5] J.A.C. Hattie, Visible Learning: A Synthesis of Over 800 Meta-Analyses Relating to Achievement. Routledge, 2009.
- [6] E.A. Skinner and J.R. Pitzer, "Developmental Dynamics of Student Engagement, Coping, and Everyday Resilience," in Handbook of Research on Student Engagement, S.L. Christenson, A.L. Reschly, and C. Wylie, Eds. Boston, MA: Springer US, 2012, pp. 21–44.
- [7] M.T.H. Chi and R. Wylie, "The ICAP Framework: Linking Cognitive Engagement to Active Learning Outcomes," in Educational Psychologist, vol. 49, no. 4, pp. 219–243, 2014.
- [8] Y.F. Yang, "Engaging students in an online situated language learning environment," in Computer Assisted Language Learning, vol. 24, no. 2, pp. 181–198, 2011.
- [9] H.M. Marks, "Student Engagement in Instructional Activity: Patterns in the Elementary, Middle, and High School Years," in American Educational Research Journal, vol. 37, no. 1, pp. 153–184, 2000.
- [10] R. Pierce, K. Stacey, and A. Barkatsa, "A scale for monitoring students' attitudes to learning mathematics with technology," in Computers & Education, vol. 48, no. 2, pp. 285–300, 2007.
 [11] J.A. Fredricks and W. McColskey, "The Measurement of Student
- [11] J.A. Fredricks and W. McColskey, "The Measurement of Student Engagement: A Comparative Analysis of Various Methods and Student Self-report Instruments," in Handbook of Research on Student Engagement, S.L. Christenson, A.L. Reschly, and C. Wylie, Eds. Boston, MA: Springer US, 2012, pp. 763–782.
- [12] J.A. Fredricks, P.C. Blumenfeld, and A.H. Paris, "School Engagement: Potential of the Concept, State of the Evidence," in Review of Educational Research, vol. 74, no. 1, pp. 59–109, 2004.
- [13] V. Dagley, "Making the invisible visible: a methodological and a substantive issue," in Educational Action Research, vol. 12, no. 4, pp. 613–630, 2004.
- [14] J.J. Appleton, S.L. Christenson, and M.J. Furlong, "Student engagement with school: Critical conceptual and methodological issues of the construct," in Psychology in the Schools, vol. 45, no. 5, pp. 369– 386, 2008.
- [15] H. Patrick, A. M. Ryan, and A. Kaplan, "Early adolescents' perceptions of the classroom social environment, motivational beliefs, and engagement.," in Journal of Educational Psychology, vol. 99, no. 1, pp. 83–98, 2007.
- [16] T.A. Kindermann, "Effects of Naturally Existing Peer Groups on Changes in Academic Engagement in a Cohort of Sixth Graders," in Child Development, vol. 78, no. 4, pp. 1186–1203, 2007.
- [17] OECD, Low-performing students: why they fall behind and how to help them succeed. Paris: OECD Publishing, 2016.
- [18] OECD, PISA 2012 results: Ready to learn, vol. 3. Paris: OECD, 2013.
- [19] C.R. Henrie, L.R. Halverson, and C.R. Graham, "Measuring student engagement in technology-mediated learning: A review," in Computers & Education, vol. 90, pp. 36–53, 2015.
- [20] F. Gossen, D. Kuhn, T. Margaria, and A.L. Lamprecht, "Computational Thinking: Learning by Doing with the Cinco Adventure Game Tool," in Proceedings of 2018 IEEE 42nd Annual Computer Software and Applications Conference, 2018, pp. 990–999.
- [21] M. Prince, "Does Active Learning Work? A Review of the Research," in Journal of Engineering Education, vol. 93, no. 3, pp. 223–231, 2004.
- [22] D. Nicol and C. Milligan, "Rethinking technology-supported assessment practices in relation to the seven principles of good feedback practice," in Innovative Assessment in Higher Education, C. Bryan. and K. Clegg., Eds, Taylor and Francis Group, 2006, pp. 1–14.
- [23] A. Barana, M. Fioravera, M. Marchisio, and S. Rabellino, "Adaptive Teaching Supported by ICTs to Reduce the School Failure in the

Project 'Scuola Dei Compiti,'" in Proceedings of 2017 IEEE 41st Annual Computer Software and Applications Conference, 2017, pp. 432–437.

- [24] A. Barana, A. Bogino, M. Fioravera, M. Marchisio, and S. Rabellino, "Open Platform of self-paced MOOCs for the continual improvement of Academic Guidance and Knowledge Strengthening in Tertiary Education," in Journal of e-Learning and Knowledge Society, vol 13, no 3, pp. 109–119, 2017.
- [25] A. Barana, A. Bogino, M. Fioravera, F. Floris, M. Marchisio, L. Operti and S. Rabellino, "Self-Paced Approach in Synergistic Model for Supporting and Testing Students," in Proceedings of 2017 IEEE 41st Annual Computer Software and Applications Conference, 2017, pp. 407–412.
- [26] A. Barana, M. Fioravera, and M. Marchisio, "Developing problem solving competences through the resolution of contextualized problems with an Advanced Computing Environment," in Proceedings of the 3rd International Conference on Higher Education Advances, 2017, pp. 1015-1023.
- [27] P. Black and D. Wiliam, "Developing the theory of formative assessment," in Educational Assessment, Evaluation and Accountability, vol. 21, no. 1, pp. 5–31, 2009.
- [28] A. Barana, S. Boffo, F. Gagliardi, R. Garuti, M. Marchisio, "Empowering Engagement in a Technology Enhanced Learning Environment", in Proceedings of 4th conference on Smart Learning Ecosystems and Regional Development, in press.
- [29] Moebius Assessment, available at https://www.digitaled.com/products/assessment/, last accessed 02-Jan-2019.
- [30] A. Barana, M. Marchisio, and S. Rabellino, "Automated Assessment in Mathematics," in Proceedings of 2015 IEEE 39th Annual Computer Software and Applications Conference, 2015, pp. 670–671.
- [31] A. Barana, L. Di Caro, M. Fioravera, F. Floris, M. Marchisio, and S. Rabellino, "Sharing system of learning resources for adaptive strategies of scholastic remedial intervention," in Proceedings of the 4th International Conference on Higher Education Advances, 2018, pp. 1495-1503.
- [32] A. Barana and M. Marchisio, "Ten Good Reasons to Adopt an Automated Formative Assessment Model for Learning and Teaching Mathematics and Scientific Disciplines," Procedia - Social and Behavioral Sciences, vol. 228, pp. 608–613, 2016.
- [33] A. Barana, A. Conte, M. Fioravera, M. Marchisio, and S. Rabellino, "A Model of Formative Automatic Assessment and Interactive Feedback for STEM," in Proceedings of 2018 IEEE 42nd Annual Computer Software and Applications Conference, 2018, pp. 1016–1025.
- [34] A. Barana, M. Marchisio, and M. Sacchet, "Advantages of Using Automatic Formative Assessment for Learning Mathematics," in Proceedings of 2018 TEA Conference, in press.
- [35] A. Brancaccio, M. Marchisio, C. Palumbo, C. Pardini, A. Patrucco, and R. Zich, "Problem Posing and Solving: Strategic Italian key action to enhance teaching and learning Mathematics and Informatics in the high school," in Proceedings of 2015 IEEE 39th Annual Computer Software and Applications Conference, 2015, pp. 845–850.
- [36] A. Barana, M. Marchisio, e R. Miori, "MATE-BOOSTER: Design of an e-Learning Course to Boost Mathematical Competence", in Proceedings of the 11th International Conference on Computer Supported Education (CSEDU 2019), 2019, vol. 1, pp. 280–291.
- [37] OECD, PISA 2012 Technical Report. PISA, OECD Publishing, 2014.
- [38] A. Barana, S. Boffo, F. Gagliardi, and M. Marchisio, "Problem Posing & Solving: a Digital Way to Learn Mathematics," In press.
- [39] J.A. Middleton and P.A. Spanias, "Motivation for Achievement in Mathematics: Findings, Generalizations, and Criticisms of the Research," in Journal for Research in Mathematics Education, vol. 30, no. 1, p. 65-88, 1999.
- [40] A. Barana, M. Fioravera, and M. Marchisio, "Teacher training: a model for introducing innovative digital methodologies for learning Mathematics," in Proceedings of the 3rd International Conference on Higher Education Advances, 2017, pp. 608-616.

Author Index

A. A. De Melo Nunes Soares, Fabrizzio
Abe, Hiroshi
Abe, Naoto
Abe, Toru
Abichequer Sangalli, Vicenzo
Acco Tives Leão, Heloise
Adib, Riddhiman
Agu, Emmanuel
Ahamed, Sheikh
Ahamed, Sheikh Iqbal
Ahmad, Ishtiyaque
Ahmed, Shameem
Ahmed, Zakaria
Ai, Lei
Akiyama, Mitsuaki
Akiyama, Toyokazu
Alahi, Irfan
Albayrak, Sahin
Albernaz Bezerra, Rodrigo
Aldawod, Eyad
Al-Odat, Zeyad
Alqithami, Saad
Alves, Everton L. G
Alzahrani, Musaad
Amaral, Aline M. M. Miotto
Amaral, Vasco
Amrizal, Muhammad Alfian
Anu, Han
Asakura, Ryuji
Asnani, Sorath
B, Rizwanahmed
Bansal, Ajay
Barana, Alice
Barrera León, Luisa
Baumgart, Stephan
Bayrak, Bekir
Behnam, Moris

Fukazawa, Yoshiaki117
Fukuoka, Tomoyuki117
Fukushi, Naoki
Gallant, Morgan 443
Gandino, Filippo
Gao, Chao
Gao, Hongcan 107
Gao, Ruichao
Gaudiot, Jean-Luc
George, Olawunmi 690
Ghazi Vakili, Mohammad570
Gokhale, Swapna 417, 740
Gonçalves, Cristhiane
Gonçalves, Diego Addan888
Gotani, Kosuke958
Gu, Ming 11
Gu, Zuxing11
Guerrera, Mauro
Guha, Shion 653
Guillen, Luis
Guo, Chenkai107
Guo, Li
Guo, Ye
Guri, Mordechai801
Haddad, Hisham704
Ham, Jae-Hyun944
Hamada, Ken
Hamdi, Mohammed
Hamza, Zain
Han, Junxiao21
Hanayama, Kaisei
Harai, Hiroaki
Hasan, Mgm Mehedi 549
Hasegawa, Hirokazu 396, 936
Hasegawa, Yohei
Hata, Misumi
Hato, Kunio
Helmers, Aaron 630

Kattamreddy, Thulasidhar Reddy94	6
Kawabata, Natsuko	5
Kawai, Eiji 270	6
Kawakami, Tomoya93	8
Kern, Matthias	3
Keung, Jacky 4'	7
Keung, Jacky W 5	3
Khamespanah, Ehsan	9
Khan, Mohammed Salman930, 932	2
Khan, Samee	1
Khan, Shahzad	3
Kim, Dae-Kyoo	0
Kim, Ki-Il	4
Kim, Yeongcheol79	5
Kimata, Takashi	6
Kitagawa, Naoya	0
Kohama, Munetaka 11'	7
Kondo, Uchio 270	0
Konishi, Hiroshi	7
Kozakevicius, Alice	6
Kozen Stahl, Bruno 30	0
Kumagai, Atsutoshi 340	6
Kurauchi, Yuki	7
Kuribayashi, Kentaro	0
Kusumoto, Shinji920	6
Lang, Nathan	3
Larcher, Matteo 46.	3
Lasswell, Nina	3
Lavirotte, Stéphane	0
Leal, Lucas	5
Ledbetter, Brian	5
Lee, Sooyeon	4
Lehmann, Patrick Marvin 46	9
Leong, Hong Va62	2
Leong, H. V	2
Leudet, Jérôme 479	9
Li, Congmiao	8
Li, Ge	9

Li, Tong	Matsu
Li, Weiwei	Matsu
Li, Yuan	Matsu
Li, Zhuolin	McVe
Li, Zimu	Mcwh
Liang, Bin 66	Mede
Lim, Hyun-Il	Melto
Lin, Dan 299	Men,
Lisova, Elena	Mend
Liu, Jiaxiang	Meng
Liu, Xiulong	Meuri
Liu, Xuefeng	Miazi
Liu, Yan	Mikko
Liu, Zihao 598	Moha
Lo, Dan	Mohs
Loiret, Frederic	Mone
Loparo, Kenneth	Monte
Lou, Yuansheng	Montr
Lounis, Karim	Mu, K
Lu, Zhigang 185	Mu, S
Luo, Hanwu	Mühll
Luo, Wenbo 149	Mulla
Ma, Chengfang	Mumo
Ma, Xiaoxing	Musta
Ma, Xuebin	Nakar
Machado, Patrícia D. L	Nakao
Mahabal, Ashish	Nery
Majumder, Akm Jahangir Alam 850	Neuja
Männistö, Tomi	Ngai,
Manshaei, Mohammad Hossein 535, 549	Nguy
Mao, Hongyan 203	Nguy
Marchisio, Marina216	Ni, Yo
Maroco, Pedro	Nie, F
Marques Freire, Willian	Nie, X
Martins, Eliane	Nova
Martins Da Costa, Ronaldo 714	Odone
Matsuki, Masayuki	Ohsak
Matsumoto, Ryosuke	Ohzał

Matsumoto, Satoru	····· !	938
Matsumoto, Shinsuke	9	926
Matsuo, Ryotaro	•••••	291
McVey, Alana	(630
Mcwhorter, Tanner M		850
Medeiros, Henry		501
Melton, Austin		. 72
Men, Rui	••••	159
Mendsaikhan, Otgonpurev		396
Meng, Ruijie	••••	169
Meurisch, Christian	507, 3	513
Miazi, Nazmus Sakib		823
Mikkonen, Tommi	•••••	479
Mohammad, Salim Ali Khan	•••••	. 91
Mohsen, Fadi	•••••	786
Monego, Vinicius		306
Monteiro Pinheiro, Hedenir	•••••	714
Montrucchio, Bartolomeo	529, :	570
Mu, Kedian		364
Mu, Sanwen		352
Mühlhäuser, Max	507, :	513
Mullangi, Sai	9	946
Mumolo, Enzo	(685
Mustafa, Nasser M		834
Nakamura, Ryo		291
Nakao, Akihiro		346
Nery Rossi Camillo, Eduardo	•••••	714
Neujahr Copstein, Rafael		300
Ngai, Grace	242,	622
Nguyen, Kien		928
Nguyen, Phu		870
Ni, Yezhou		850
Nie, Hanqing		850
Nie, Xiao	(630
Novac, Pierre-Emmanuel		870
Odone, Francesca		501
Ohsaki, Hiroyuki	285,	291
Ohzahata, Satoshi		900

Okada, Kazuya 346	
Okimoto, Marcio Vinicius	
Otten, Stefan	
Oyama, Katsunori	
Öztürk, Dilek	
Paik, Joon-Young 795	
Pan, Minxue	
Pan, Xiubao	
Panda, Subhrakanta91	
Pandit, Djananjay 212	
Paranhos Junior, Augusto	
Páris, Joana	
Parvez, Md Anwar	
Pasin, Marcia	
Paudel, Anmol 908	
Pavanatto Soares, Leonardo	
Paz, Andrés	
Pei, Yu 149	
Peng, Huailiang	
Peng, Yanru	
Pethe, Rashmi	
Pike, Matthew	
Prampolini, Enrico 501	
Pranger, Cody	
Price, Kurt	
Puri, Satish	
Qian, Kai 565, 914	
Qian, Ying	
Qiao, Daji	
Qin, Bo	
Rabellino, Sergio	
Rafhael De Carvalho, Rhandy	
Rahman, Labiblais	
Rahman, Mohammad 565	
Rahman, Mohammad Ashiqur 535, 549	
Ramachandran, Nithin	
Ramadugu, Shanthan946	
Razghandi, Mina	

Reddivari, Sandeep 930,	932
Ren, Jiahui	358
Reznichenko, Yevgeniy	501
Rodrigues, Armanda	638
Roh, Bongsoo	944
Roushan, Tanvir	690
Saad, Walid	549
Sakala, Sushma	179
Sakhare, Priyanka Bharat	940
Sakuraba, Shuji	896
Salvadori, Ivan Luiz	402
Saque, Nicollas	888
Sarroglia Pinho, Márcio	300
Sato, Goshi	956
Sato, Hiroyuki	299
Sax, Eric	523
Saxena, Devansh	630
Saxena, Piyush	630
Seceleanu, Tiberiu	.604
Sedaghatbaf, Ali	579
Segura, Sergio	209
Sekiya, Hiroo	928
Sekiya, Yuji	346
Seshimo, Hitoshi	667
Severin, Karl	417
Sha, Ying 386,	696
Shahriar, Hossain 565, 704, 918, 920, 922,	924
Shahriyar, Rifat 191,	760
Shao, Rumeng	203
Sharmin, Moushumi	.644
Shehab, Mohamed	.823
Sheikh, Hassam Ullah	489
Shen, Beijun	. 97
Shen, Zhishu	252
Shi, Wenchang	66
Shih, Chihhsiong	720
Shimada, Hajime	936
Shiraiwa, Masaki	956

Siami Namin, Akbar 753	3
Siddique, Abubakar 501	l
Silva, Cleyton Rafael Gomes 714	ł
Silveira Rodrigues, Ricardo	5
Sinz, Carsten	3
Siqueira, Frank 402	2
Siqueira, Joyce 714	ł
Sirjani, Marjan 579)
Sjödin, Mikael604	ł
Sljivo, Irfan833	3
Soderstrom, Ola	5
Solberg, Arnor)
Song, Hui)
Song, Myoungkyu179)
Suganuma, Takuo 912, 956, 958	3
Suksomboon, Kalika 252	2
Sun, Jing	2
Sun, Yingcheng	ł
Tagami, Atsushi	2
Takahashi, Kenji 340	5
Takahashi, Tokio610)
Takahira, Hiroyuki	3
Talukder, Md Arabin Islam 565	5
Tamaki, Junji117	7
Tan, Jianlong 386	5
Tanita, Satoshi	5
Tao, Lixin	3
Teixeira Costa, Pedro Henrique	5
Teranishi, Yuuichi	3
Tessarotto, Marco 685	5
Tigli, Jean-Yves)
Tiwari, Devansh117	7
Todt, Eduardo 888	3
Tolone, William J	3
Tong, Li 696	5
Torres, Wesley N. M 129)
Towey, Dave	ł
Tsubouchi, Yuuki 264	ł

Tsugawa, Sho	892
Tuglular, Tugkan	197
Ucci, Donald R	850
Uchida, Masato	3, 610, 770
Ueda, Kazuaki	
Uehara, Minami	892
Ul Islam, Md Rafid	760
Valepe, Sathvik Vamshi	91
VanHecke, Amy	630
Ved, Mohit	411
Vinícius Bindewald, Carlos	139
Wadekar, Adway	960
Wakabayasi, Shin'ichi	934
Wakisaka, Asato	264
Wang, Chuanbo	730
Wang, Cong	1
Wang, Dongjing	21
Wang, Juan	817
Wang, Jun	622
Wang, May D	696
Wang, Qingshun	916
Wang, Tao	
Wang, Zhongmin	382
Ward, Robert	59
Washizaki, Hironori	117
Wehunt, Samuel	535
Weil, Tim	545
William, Cheng-Chung Chu	720
Wu, Fan	. 565, 924
Wu, Jiecheng	11
Wu, Yan	559
Wu, Yanfeng	107
Wu, Yuewen	185
Wymore, Mathew L.	860
Xia, Hong	382
Xia, Xin	21
Xian, Yongyi	453
Xiao, Yan	47

Xu, Chuanfei
Xu, Jing 107
Xu, Jiwei 185
Xu, Sihan 107
Xu, Sugang
Xu, Xian
Yamaguchi, Yukiko
Yamai, Nariyoshi
Yamasaki, Yasuhiro
Yamauchi, Ryosuke
Yang, Jie 908
Yang, Jingwei
Yang, Runze
Yang, Yankan
Yang, Yanni
Yang, Yufei 149
Yao, Junfeng
Yao, Zhihuan
Ye, Feng
Ye, Shasha 916
Yesha, Yaacov 559
Yin, Jianwei 21
Yin, Weiliang 1
Yoo, Wook-Sung946
Yoshihisa, Tomoki
Yu, Huiqun
Yu, Ping
Yu, Yaoshen
Yu, Zeyun
Yu, Zhong
Yuan, Yue
Zadov, Boris
Zhang, Hao918, 924
Zhang, Miao 47, 53
Zhang, Renwei 1
Zhang, Tian 149
Zhang, Zhenyu 169
Zhang, Ziwei

Zhao, Huiqun
Zhao, Mingdao 382
Zhao, Tianqi433
Zheng, Jianjun752
Zhe-Tao, Li928
Zhou, Min 1
Zhou, Yu27
Zhou, Zhi Quan 209
Zhu, Biyun 169
Zhu, Songjie
Zulkernine, Farhana
Zulkernine, Mohammad817