

BY DESIGN



NGBD 2020 Symposium Series Next Generation Education Systems 19 May 2020



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Timo Valiharju

Chairman, Educloud Alliance









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de Catalunya

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EduCloud Alliance





Obra Social "la Caixa

Next Generation By Design

VISION

- Global network
- Shift the Market to enable Plug and Play Choice-Based

🗞 EduCloud Alliance 🛛 🗛 🤁 🖓 Strada

● Specific problems → market relevant solutions

NUNIS EÜR BILDUNG

- Targeted collaboration
- Knowledge transfer
- Development and Deployment



LEARNTEC



2 previous editions



1st Edition. 2018 Palau Macaya



2nd Edition. 2019 Palau Macaya











3rd Edition!



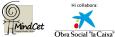


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NGBD Webinar Series Schedule

Webinars	Title	Date
Webinar 1	Comparative Taxonomy Management	Tuesday, 12th May, 2020
Webinar 2	Next Generation Education Systems	Tuesday, 19th May, 2020
Webinar 3	Exploring New Frontiers	Tuesday, 26th May, 2020





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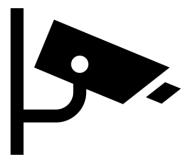


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Next Generation Education Systems

Part						
Welcome	Host: Chairman Timo Väliharju, Educloud Alliance	~ 5 minutes				
Presentations		Panelist 1. Michael Bourque CIO and VP, Information Technology Boston College~ 12-15 minute	es			
		Panelist 2: Janette Vesalainen~ 12-15 minuteProject SpecialistDigiOne project	es			
	Facilitator: Jeff Merriman Chief Technology Officer DXtera Institute, Inc.	Panelist 3: Dr. John Whitmer~ 12-15 minuteSenior Director for Data Science and Analytics~ 12-15 minuteACTNext at ACT, Inc	es			
		Panelist 4: Dr. Marc Alier~ 12-15 minuteAssociate ProfessorUniversitat Politecnica de Catalunya	es.			
		Panelist 5: Christophe Speroni Co-Founder and CPO bettermarks~ 12-15 minute	~ 12-15 minutes			
Discussion	Facilitator: Jeff Merrimam	~ 15 minutes				
Closing	Host: Timo Väliharju, Educloud Alliance	~ 5 minutes				
NEXT GENERATION BY DESIGN DESI						





Sessions are being recorded

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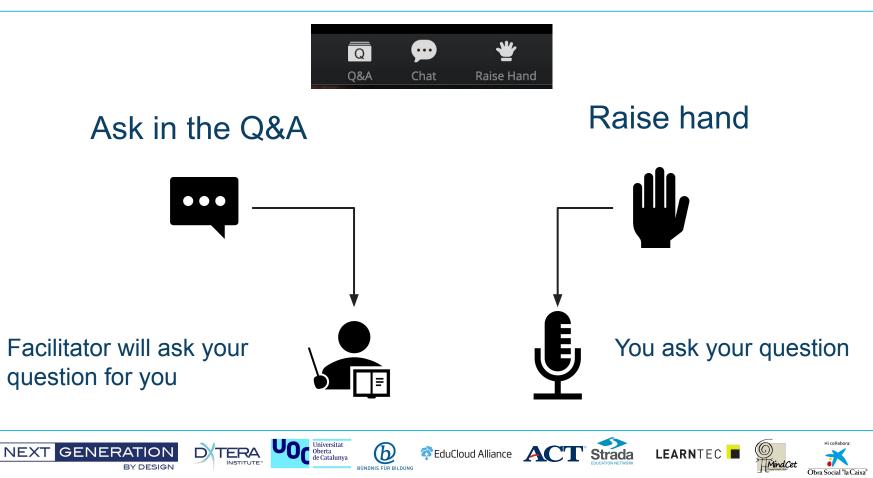
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And after the webinars?



Recordings will be posted



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Summary paper will be published

TeduCloud Alliance









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Jeff Merriman CTO and Co-Founder merriman@DXtera.org

Panel of Innovators



Michael Bourque CIO & VP Information Technology Boston College

NEXT GENERATION

Janette Vesalainen *Project Specialist* DigiOne Project

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Dr. John Whitmer Sr. Director for Data Science and Analytics ACTNext, ACT

ኛ EduCloud Alliance 🛛 🗛 📿 🖵 🖁 Strada

Dr. Marc Alier Associate Professor Universitat Politecnica de Catalunya

LEARNTEC

Christophe Speroni Co-Founder and Chief Product Officer bettermarks





Michael Bourque Chief Information Officer and VP, Information Technology Boston College



Boston College EagleApps System

May 19, 2020





Boston College EagleApps System

- Background and Origins
- Goals
- Capabilities
- Model
- Building Community







EagleApps System Background

- A long journey born out of the Kuali community
- Based on the requirements and vision of 20+ institutions
- Boston College examined market options and decided to complete the system to meet its goals
- Significant investment over past five years
- Strong partnerships with development global partners
- In use at Boston College and expanding functionality





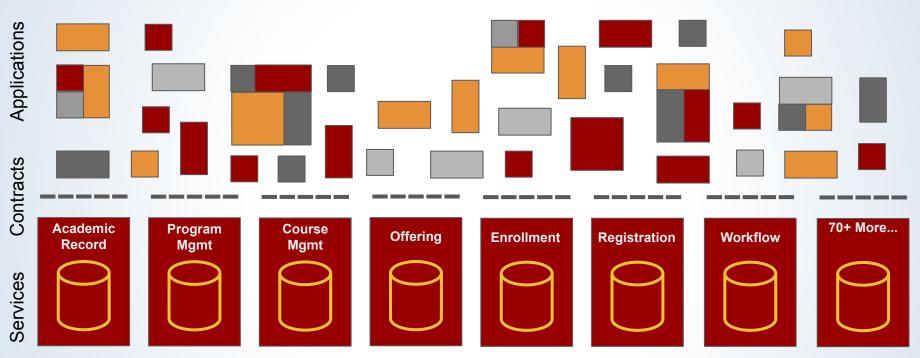
EagleApps System Goals

- Modular system (not monolithic ERP)
- Incremental adoption
- Functionally rich
 - Rules based configuration
 - Extensible
- Technically strong and agile
 - Service based architecture/
 - Designed for integration (APIs)





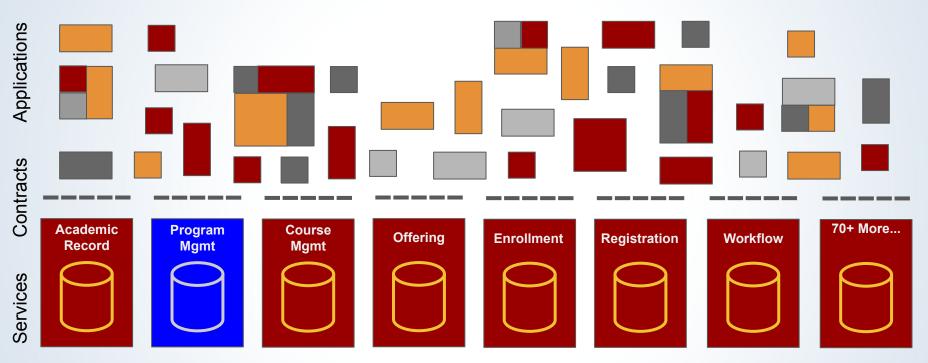
EagleApps System Model/Capabilities







Flexible Service Implementations





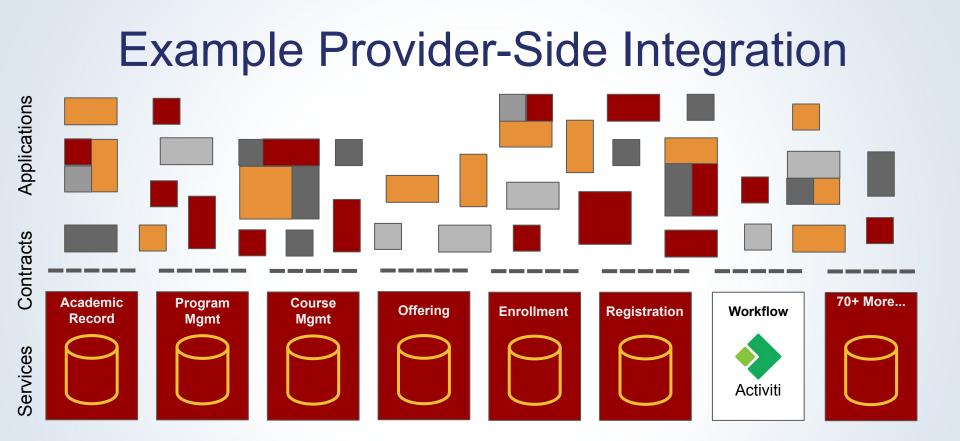


Apps Can Be Replaced and Added

Applications								
Contracts								
S	Academic Record	Program Mgmt	Course Mgmt	Offering	Enrollment	Registration	Workflow	70+ More
Services	\bigcirc	\bigcirc					\bigcirc	











Example Consumer-Side Integration

S Applications	Operational Data Store Data Warehouse						
C Academic							
O O Academic Record	Program Mgmt	Course Mgmt	Offering	Enrollment	Registration	Workflow	70+ More
Services						\bigcirc	





Looking to Engage Interested Institutions

- Institutions Can Engage in a Study to:
 - Develop organizational understanding of EagleApps
 - Understanding how to use EagleApps to implement specific processes. Development of the Future Business Processes as deltas against the EagleApps process
 - Create implementation plan based on requirements and processes
- Building a Community to:
 - Develop implementation planning processes and templates
 - Deploy and configure reference implementation
 - Create and grow EagleApps documentation library
 - Document best practices for EA deployment and use
 - Contribute to long term roadmap/support model





Engagement in the Community

- Monthly Community meetings
- Forums, Gatherings (in-person . . .) and Seminars
- Outputs from the Implementation Studies
- Access to DXtera Solutions
- Advisory Council to provide guidance on:
 - Development Roadmap
 - Maintenance
 - Training & Support
 - Services

- Emerging Models for:
 - Business
 - Licensing
 - Implementation





Boston College EagleApps System

May 19, 2020







Janette Vesalainen

Project Specialist DigiOne project



DigiOne – An education platform connecting users and services

Janette Vesalainen

DigiOne in a nutshell

- DigiOne program aims to create a digital platform and an ecosystem welcoming all actors within the education sector in Finland.
- Currently the preliminary plan is being made with five municipalities, but the aim is to expand nationwide.
- Platform will cover services from early childhood education to upper secondary education.
- At the moment a funding of 11 million euros.
- Estimated time frame for the program is 2019–2023, but development of the platform will continue after that.



Goals

- To support the pedagogical and management change.
- To create more equal opportunities for learning and well-being.
- To create user-friendly services by oneself and also with content/service providers.
- To benefit owners by cutting overlapping work.
- To make sure that the ownership will stay in non-commercial hands.



Current situation

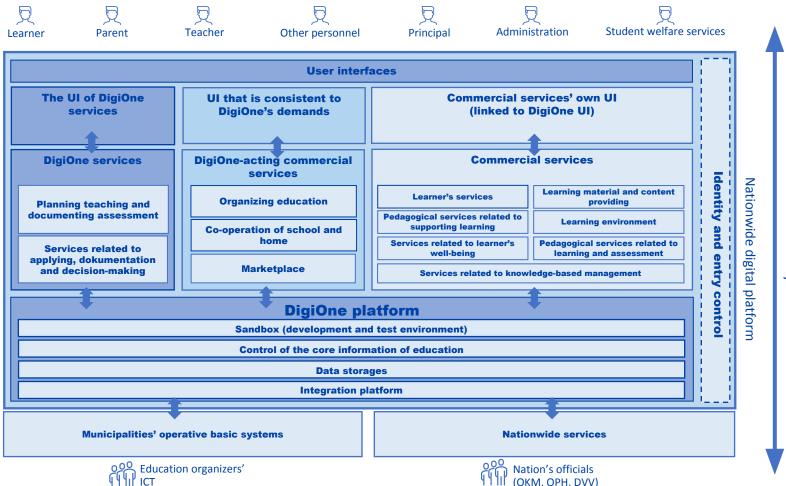


The current systems related to organizing education are scattered and challenging for teachers and learners to use. The systems and services used by administration are in the need of updating.



The systems do not support some of the things required in the new national core curriculums the best possible way. There is improvement needed to be made at least related to continuous assessment, individual support and knowledge-based management.





Nationwide

ecosystem

Empowering Education Ecosystem Connecting Users and Services!



Content providers

Materials and games for studying (non-profit and profit)



Companies, enterprises, public services (Digital service providers)



Students, parents, teachers, administration

DigiOne Value Proposition

Advantages from the ecosystem to all parties

Digital platform owner

Non-profit organizations e.g. other municipalities may join

Cities and education organizations

Cities joining to platform

V

Research and assessment organizations

Commercial and non-profit research institutes





Thank you!

Vantaa.fi/digione/en Twitter: @DigiOneFi



Dr. John Whitmer Senior Director for Data Science and Analytics ACTNext, ACT, Inc.

Predicting Social Emotional Skills & Course Grade

using Learning Analytics

John Whitmer, Ed.D., Sweet San Pedro, Ph.D., Ruitao Liu, Ph.D. Kate E. Walton, Ph.D., John Fritz, Ph.D., Joann L. Moore, Ph.D., and Alejandro Andrade Lotero, Ph.D.



DXtera Next Generation by Design Symposium: Next Gen Ed Systems May 2020



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Learning Analytics

...the measurement, collection, analysis and reporting of data about **learners** and their contexts, for purposes of **understanding and optimizing learning and the environments** in which it occurs.

~Learning and Knowledge Analytics Conference, 2011

At its core, learning analytics (LA) is the collection and analysis of usage data associated with student learning. The purpose of LA is to observe and understand learning behaviors *in order to enable appropriate interventions.*

~Educause Learning Initiative (ELI), 2011





ACT[®]Tessera[®]

Next Generation Social and Emotional Learning Assessment System

Social and Emotional (SE) skills strongly predict academic achievement, career success and lifelong well-being.

Many studies find that these attributes contribute as much or more than academic skills in helping students succeed.

MULTIDIMENSIONAL, COMPREHENSIVE, AND RESEARCH-DRIVEN

Tessera is based on the research-validated and widely adopted Big Five personality factors. The broad, multidimensional Big Five framework encompasses attributes related to successful performance across different ages, contexts, and cultures.

Skill	Description							
Grit	Grit reflects the extent to which a student's actions demonstrate persistence, goal striving, reliability, dependability, and attention to detail at school.							
Teamwork	Teamwork reflects the extent to which a student's actions demonstrate collaboration, empathy, helpfulness, trust, and trustworthiness.							
Resilience	Resilience reflects the extent to which a student's actions demonstrate stress management, emotional regulation, a positive response to setbacks, and poise.							
Curiosity	Curiosity reflects the extent to which a student's actions demonstrate creativity, inquisitiveness, flexibility, open-mindedness, and embracing diversity.							
Leadership	Leadership reflects the extent to which a student's actions demonstrate assertiveness, influence, optimism, and enthusiasm.							



Need for a New Assessment of Socio-Emotional (SE) Skills

- Blackboard LMS Data and learning analytics methods can predict course grade and risk of failing a course, but do not tell us WHY students achieve these predictions(or what they can do about it).
- SE skill assessments also predict course outcomes, but don't reveal how these skills are related to day-to-day student activity.
- What if we could **observe** students' SE skills through their interactions in educational technologies in authentic learning contexts?

Identify students at risk AND understand WHY Design effective interventions





Research Questions

- RQ1: Can we observe student social and emotional skills (SE skills) from behaviors recorded in online learning environments?
- RQ2: How accurate are SE skill predictions compared to student course grade predictions using these behaviors?
- RQ3: How does sequential data mining affect predictive model accuracy and interpretability compared to individual LMS features?
- RQ4: Is there evidence of bias in our predictive models based on student demographic background or educational experience







Data Sources



Student information system (e.g. race/ethnicity, family college history, ACT/SAT scores, college GPA & history)



Scores on Tessera



LMS clickstream data





Research Method Summary

LMS Feature Extraction

- 1. Manually map clickstream data [Action + Title] to categories [110 -> 21]
- 2. Calculate counts, SD, duration for each item [21 -> 63]
- 3. Factor loading using Varimax orthogonal rotation [63 -> 16]
- 4. Sequential data mining ('learning tactics') using Markov transition matrices [64 -> 5]

Statistics Analysis

- Correlational analysis (Spearman Rho) between [#2 #4] and SE skills & student grade
- **2.** Predictive models using gradient boosted machines between [#2 #4] and SE skills & student grade. Create baseline model with SiS only and consolidated models
- **3.** Subpopulation analysis of predictive models [remove criteria of interest, split into groups] use T-test and visual inspection of bias distributions







Research Context

- University of Maryland Baltimore County is public research & teaching university (13,602 FTES Fall 2019)
- Recruited four large-enrollment STEM courses with history of pedagogical innovations by instructors
- Extensive use of LMS for assignments, reading quizzes, test preparation, learning objectives, discussion forum
- Challenging courses with substantial number of students not passing
- Predictive models created with accuracy as of week 4

Course Enrollment & Study Participation

Participation							
Variable	Chemistry	Psychology	Physics	Math			
Enrollment	489	314	319	207			
LMS Entries	700k	450k 375k (2 sections) (2 sections)		99k (2 sections)			
Tessera Participation	83%	55%	70%	50%			
Female	50%	40%	61%	36%			
Race / Ethnicity							
Asian	29%	27%	37%	32%			
Black/African American	17%	22%	19%	25%			
Hispanic/Latino	9%	8%	6%	5%			
White	39%	34%	33%	29%			
Two or More	6%	6%	3%	5%			
Not specified	1%	2%	2%	3%			
First Generation College Student (Y)	24%	25%	27%	31%			
Transfer Student (Y)	11%	15%	32%	43%			





Results RQ1 Correlational Analyses



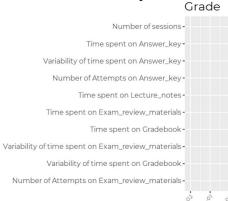


Chem

Grade

- Number of Attempts on Discussion_board-Time spent on Discussion_board-Variability of time spent on Discussion_board-Number of sessions -
 - Time spent on Lecture_class_activities -
 - Time spent on Announcements -
- Variability of time spent on Announcements -
 - Time spent on Gradebook -
- Variability of time spent on Lecture_class_activities -
 - Variability of time spent on Information-

Physics







Correlation Coefficent

313

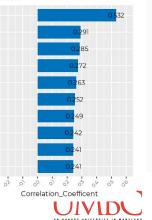
Grade Number of sessions -Time spent on Worksheet -Time spent on Remediation_makeupty of time spent on Exam_review_materials-Variability of time spent on Worksheet -Time spent on Exam_review_materials -Time spent on Discussion_board -Number of Attempts on Discussion_board-Number of Attempts on Answer_key -Time spent on Lecture_notes -

0.233

Psychology Grade

Math

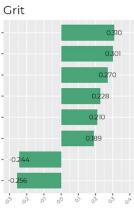




Correlational **Analyses** (Grade by course, p<0.05)

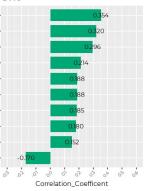


Math



0.335 Variability of time spent on Information-Time spent on General_content -Time spent on Information -Number of Attempts on Information -Number of sessions -Variability of time spent on Remediation_makeup -Time spent on Assignment - -0.244 Number of Attempts on Assignment - -0.256

Psychology Grit



Time spent on Email_messages -Time spent on Help_resources -Time spent on Textbook -Variability of time spent on Email_messages -Variability of time spent on Information -



Number of Attempts on Chapter_readings -



017

0.233

0.233

0303

0.196

0.194

0190

0.212

0.180

0175



Number of sessions -Variability of time spent on Gradebook -Time spent on Reading_guiz -Grade -Variability of time spent on Answer_key -Time spent on Discussion_board -Time spent on Gradebook -Variability of time spent on Reading_guiz -Time spent on Announcements -Time spent on Answer_key -

Chem

Grit

Grade -

Physics

Grit

Variability of time spent on Assignment -

Number of Attempts on Assignment -

Time spent on Lecture class activities -

Number of Attempts on Announcements -

Variability of time spent on Announcements -

Time spent on Assignment -

Time spent on Orientation -

Time spent on Announcements -

Number of sessions -





Results RQ2 Predictive Models





Types of Predictors

1. Background Variables (BV)

Gender, Ethnicity, First Generation Indicator, Transferred Student Indicator. Number of Credits Attempted, Number of Credits Completed, **Current College GPA**

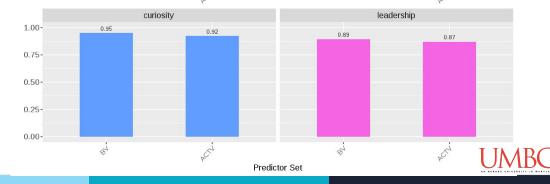
- **2.** All five SE Skill scores (TES)
- **3.** LMS Activity Features (ACTV)
- 4. Learning Tactic Sequence Features (SEQ)

Gradient boosted machines used; linear regression tested and found to be lower accuracy (for every week, for every course), especially at early weeks of the term.



Predictive Model Error by Predictor Set for all SE Skills (Chem)

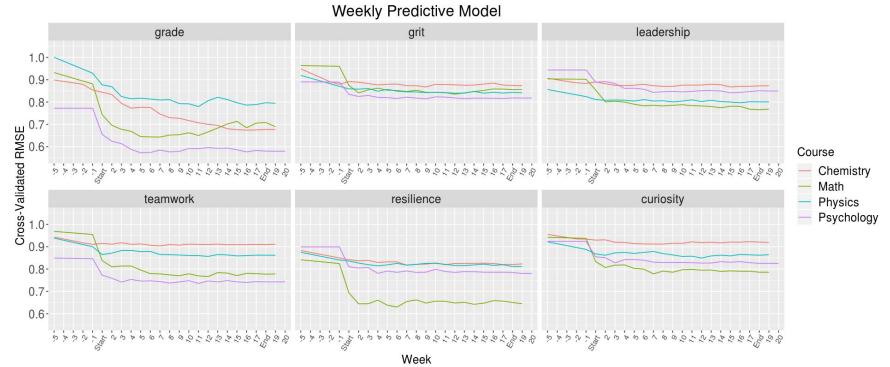
Performance of Predictive Models for CHEMISTRY grit grade 1.00-0.90 0.90 0.88 0.75-0.69 0.70 0.68 0.50-0.25-0.00-AS A R S teamwork resilience 1.00-0.93 Cross-Validated RMSR 0.50-0.25-0.25-0.91 0.87 0.83 0.00-01 S



Note: Accuracy between grade & grit should not be compared. grade scale (0-4) and grit (-3 to 3) means that grit accuracy is better with higher RMSE score



Predictive Model Error by Week by Course







Biggest Surprises

- 1. Consistency of findings despite high variability in LMS usage (both breadth and depth)
- 2. Face validity (& variability) of correlations between LMS activity and SE skills
- **3.** Lower performance & shallow depth of sequential data mining patterns
- **4.** Difficulty of data mapping to join LMS data with course-relevant activity
- Value of faculty feedback to interpret data & findings (not really a surprise)



Image source: Wikimedia commons





Published Reports available (1 of 4 Courses)



ACT Research & Policy | Data Byte | November 2019

The Relationship Between Social and Emotional Skills and Student Behaviors in Predicting Course-Level Student Success

John Whitmer, Sweet San Pedro, Ruitao Liu, Kate E. Walton, Joann L. Moore, & Alejandro Andrade Lotero

Background

Educational technologies, such as learning management systems (LMS), online homework portals, and ePortfolios, are becoming an increasingly important part of student learning experiences. Data collected by these systems provide valuable insights into student learning practices, and learning analytics (LA) researchers have demonstrated that these data can be used to make accurate predictive models of student course grades (Macfayden & Dawson, 2010). However, the psychological constructs that underlie student online behaviors are not well understood. Similarly, the field of social and emotional learning recognizes that social and emotional (SE) skills are associated with academic performance (Poropat, 2009), but the field is somewhat limited in its understanding of the mechanisms by which certain SE skills lead to academic success.

Objective

In this study, we marry the fields of LA and social and emotional learning to determine whether we can observe SE skills from behaviors recorded in online learning environments, whether predictions using LMS mining lachicuges, whether predictions using LMS demographic data, whether and how online behaviors are associated with SE skills and course grades, and whether three is demographic bias in the predictive models.

Method

We evaluated these questions in an undergraduate course making extensive use of the Blackboard LMS.

ACT



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https://tinyurl.com/clicksConstructs





course, and 406 of them also completed the college version of ACT[®] researe[®], an assessment of five SE skills, including Grit, Teamwork, Resilience, Curiosity, and Leadership. For brevity, we present findings for Grit, as well as course grade.

> LNS events, such as accessing the syllabus or taking practice acams, were provided at the individual student-activity level for all interactions with course materials and activities that were recorded in the database. Blackboard extracted this information from the hosted version of the LMS with a total of approximately 700,000 records included in the dataset. To discover meaningful learning behavior patterns, we grouped activity sessions into clusters using a hidden Markor modeling technique.

LMS data were available for 489 students in this

Key Findings

- Some SE skills were systematically observed in student use of the LMS (Figure 1)
- High-level learning tactics could be extracted from activity data using sequential data mining techniques
- Predictions using LMS behavioral data were more accurate than those made using student family and educational background (Figure 2), providing an accurate and powerful basis for actionable interventions
- LMS behavioral data partially mediated the relationship between Grit and course grades
- No evidence of demographic bias was found in predictions





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Thank you!

John Whitmer

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Acknowledgements: Sweet San Pedro, Ph.D, Ruitao Liu, Ph.D. Kate E. Walton, Ph.D., Joann L. Moore, Ph.D., and Alejandro Andrade Lotero, Ph.D.







Dr. Marc Alier Associate Professor Universitat Politecnica de Catalunya



Modding Moodle to improve confidentiality, privacy support.



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Privacy and Confidentiality



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GDPR Support in Moodle now

Do you accept the Yes — Enter the site Terms Of Service? No — Blocked

Wrong answer try again.



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GRIAL Research Group Uriversky of Salamanca





Objection to Terms Of Service is a Right

→ Victims of bullying family violence gender violence extreme paranoia

- → Law enforcement, Secret agents
 - Super heroes, Super Villains and other beings...





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We are re-designing our campuses to accommodate exceptions. Why not Moodle?



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Moodle uses Personal Information mainly in 2 Ways

Administrative purposes -> Necessary to provide the service

Collaborative aspects of Moodle, expose personal data to other students.

-> Not necessary but embedded in the design of Moodle.





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Modding Moodle to improve confidentiality, privacy support.

Course Participants

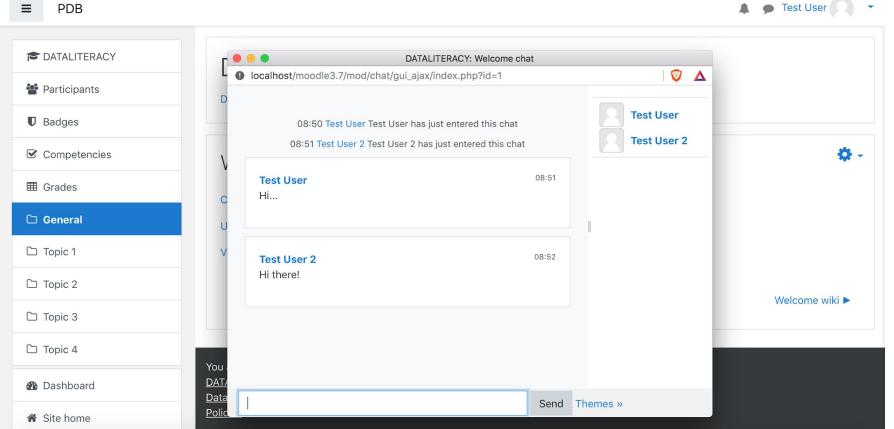
≡ PDB		🌲 🍺 Test User
CATALITERACY	Data literacy	
😤 Participants	Dashboard / My courses / DATALITERACY / Participants	
Badges		
Competencies	Participants	
I Grades	No filters applied	
🗅 General	Search keyword or select filter	
🗅 Topic 1	First name All A B C D E F G H I J K L M N O P Q	R S T U V W X Y Z
🗅 Topic 2	Surname AII A B C D E F G H I J K L M N O P Q	R S T U V W X Y Z
🗅 Topic 3	First name A / Surname Roles Groups	Last access to course
🗅 Topic 4	Test User Student No group:	s 2 secs
🚯 Dashboard		Never
Site home	Test User No roles Group 1	INEVEL



Modding Moodle to improve confidentiality, privacy support.

Chat

🛕 🍺 Test User







Forum							
≡ PDB						🗩 Test U	Jser
 DATALITERACY Participants 	Welcome forum My new discussion			⊠ Si	ubscribe	ed 🌣 Se	ettings 🔻
Badges		•					
Competencies	Display replies in nested form	\$					
I Grades	My new discussion by Test User - Friday, 2 Augus	t 2019, 8:59 am					
🗅 General	Hi there!						
🗅 Topic 1				Permalink	Edit	Delete	Reply
🗅 Topic 2	Re: My new discussion by Test User 2 - Friday, 2	August 2010, 8:50 am					
🗅 Торіс 3	Hi!	August 2019, 6.59 am					
🗅 Topic 4	HI!		Permalink	Show parent	Edit	Delete	Reply
🖚 Dashboard	■ ■ Welcome wiki	Jump to	\$				
Site home							



Modding Moodle to improve confidentiality, privacy support. Wiki history

≡ PDB

🌲 🍺 Test User 🕥 🝷

	Dashboard / N	ly courses / DATAL	ITERACY / General / Welcome wi	ki / History / Page / H	listory
CATALITERACY					Search wikis
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Badges	Welcom	e wiki			
Competencies					
I Grades	View Ed	t Comments	History Map Files		
🗅 General	Page?	August 2019, 8:58 am by	Test User		
🗅 Topic 1	Diff 🝞	Version	User	Modified	
🗅 Topic 2	0	2	Test User	8:58 am	2 August 2019
그 Topic 3					
🗅 Торіс 4		1	Test User 2	8:58 am	2 August 2019
🚯 Dashboard	Compare sele	ected			
爺 Site home	✓ Welcome c	hat	Jump to	\$	Welcome forum ►



Modding Moodle to improve confidentiality, privacy support.

The user profile as viewed by a peer student

Ξ PDB 🛕 🍺 Test User 2 🚺 ★ DATALITERACY Test User 🖻 Message Add to contacts Participants Badges I'm a test user. Competencies I Grades User details Course details 🗅 General Email address Course profiles testuser@example.com Data literacy C Topic 1 Country Roles C Topic 2 Andorra Student City/town C Topic 3 Test City C Topic 4 Web page Miscellaneous https://mytestwebpage.test View all blog entries Dashboard Skype ID Forum posts @testskypeid Status Forum discussions Site home



User enrollment and visibility

An enrolled user with roles assigned can see data about all the participants enrolled to the course.

Even those participants have not accepted the terms of service (yet).



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¹ Modding Moodle to improve confidentiality, privacy support.

How to deal with exceptions?



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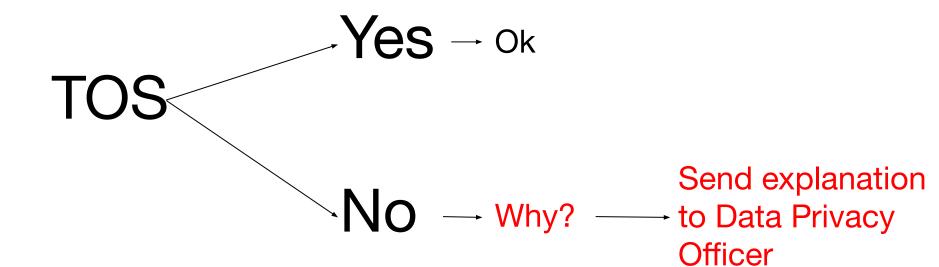






Modding Moodle to improve confidentiality, privacy support.

How to deal with exceptions





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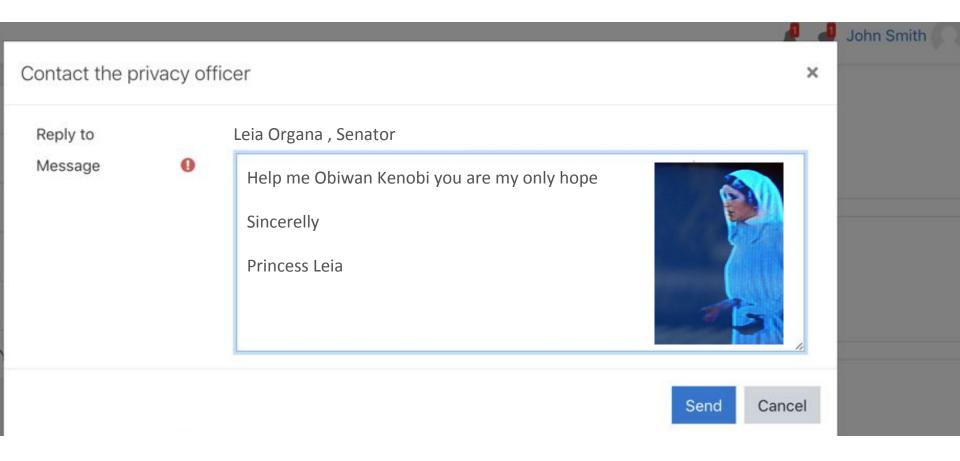


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Solution proposal: "protected user" plugin

> A Protected User is a Moodle user that has special status with regards to data privacy. It is not enrolled in any course.

- > An **Alias** is a Moodle user with fake personal information.
- > The DPO manages the creation and enrollment of Aliases.





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🌲 🌘 Data Privacy Officer 🖉

Moodle

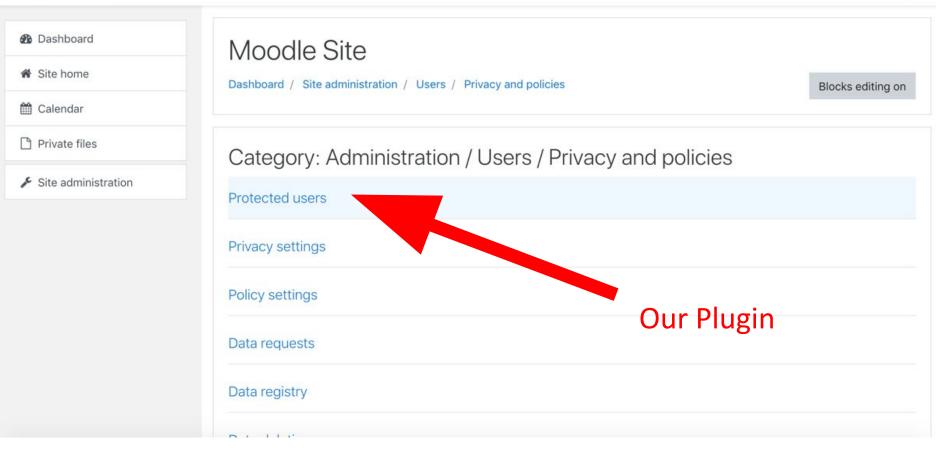
Dashboard / Site administration / Users / Privacy and policies / Data requests

Data requests

Туре	User	Date requested	Status	Message	
Message	John Smith	Tuesday, 17 September 2019, 11:30 am	Pending	Cyberbulling	Actions -

Moodle

Data Privacy Officer





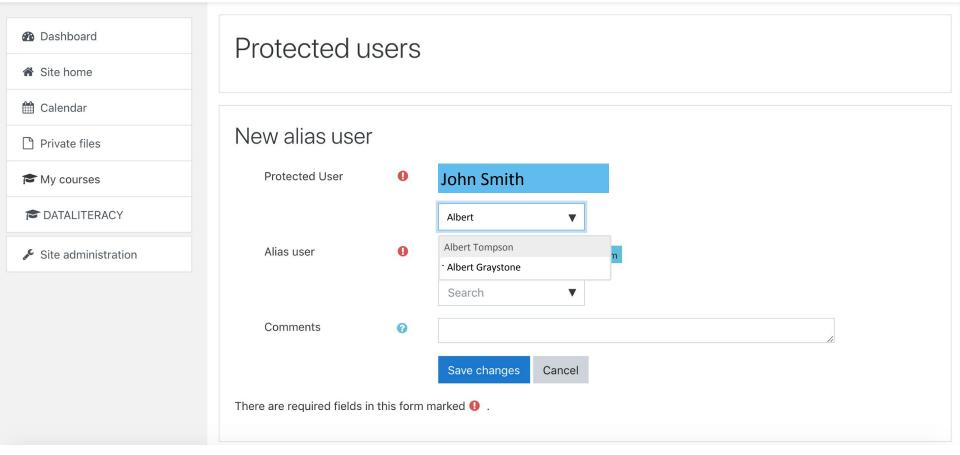


Dashboard / Site administration / Users / Privacy and policies / Protected users



·-/)





				🌲 🍺 Data Privacy Officer
rotected u	iser report:			
John Sr	nith			
Data reque	sts			
Туре	Date requested	Status	Message	
Message	Sunday, 8 September 2019, 5:16 pm	Accepted	Cyberbulling	
			, .	
P				
lias users				New Alia:
Alias	Courses			
Albert Tompson	Curso 1, Curso 2		Enroll Unenroll Eliminar	
			Circle Ci	



> The Protected User can enroll in courses via an Alias preserving the confidentiality.





granludo@gmail.com @granludo



UNIVERSITAT POLITÈCNICA DE CATALUNYA





(This should be Princess Leia's profile page, but we did not have enough CGI Budget)

	Interests	
 Dashboard Site home Calendar 	Music Books Skype ID @testskypeid Status	Reports Browser sessions Grades overview
Private files	Privacy and policies	Login activity
St My courses	Contact the privacy officer Data requests	First access to site Monday, 29 July 2019, 5:21 pm (59 days 3 hours)
CATALITERACY	Export all of my personal data Delete my account Data retention summary	Last access to site Thursday, 26 September 2019, 9:12 pm (1 sec)
	Policies and agreements Alias users	

You are logged in as John Smith (Log out)

	(Jed Leia	i Mind Trick, now you only see)	2	
≡ Moodle				🦺 🏓 John Smith 🔘 🝷
🚳 Dashboard	Jo	hn Smith 📍 Message		
A Site home				
🛗 Calendar				
Private files				
A My courses	Alias users			
T DATALITERACY	Click on the name of the	Alias user to log in as such user.		
	Alias users	Created	Comments	
	Albert Tompson	8 September 2019, 3:49 pm	Cyberbulling	

Moodle

John Smith You are logged in as Albert Tompson

🕐 Dashboard	Data literacy		
Site home	Dashboard / You are logged in as Test2 User		
🛗 Calendar			
Private files			
A My courses	You are logged in as Albert		
CATALITERACY	Tompson		
	Continue		
	[<u>Test User</u>] You are logged in as <u>Test2 User</u> (<u>Log out</u>) <u>Home</u> <u>Data retention summary</u> Policies		





Thank you

Marc Alier @granludo – UPC

Daniel Amo @danielamof – La Salle URL

https://github.com/danielamof/protected_users/



daniel.amo@salle.url.edu @danielamof



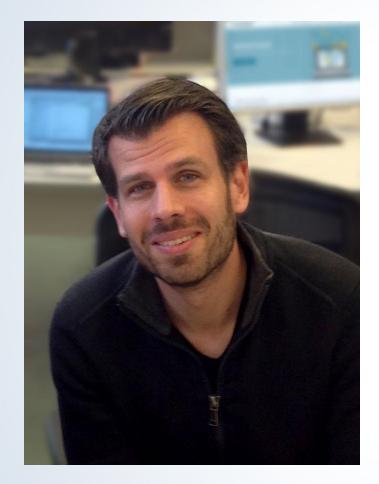
granludo@gmail.com @granludo



UNIVERSITAT POLITÈCNICA DE CATALUNYA







Christophe Speroni Co-Founder and Chief Product Officer bettermarks



Adaptive learning

... and its impact in school lessons



Christophe Speroni CPO & Co-Founder of bettermarks

bettermarl

What does »adaptivity« mean?

Turning mistakes into aha-moments

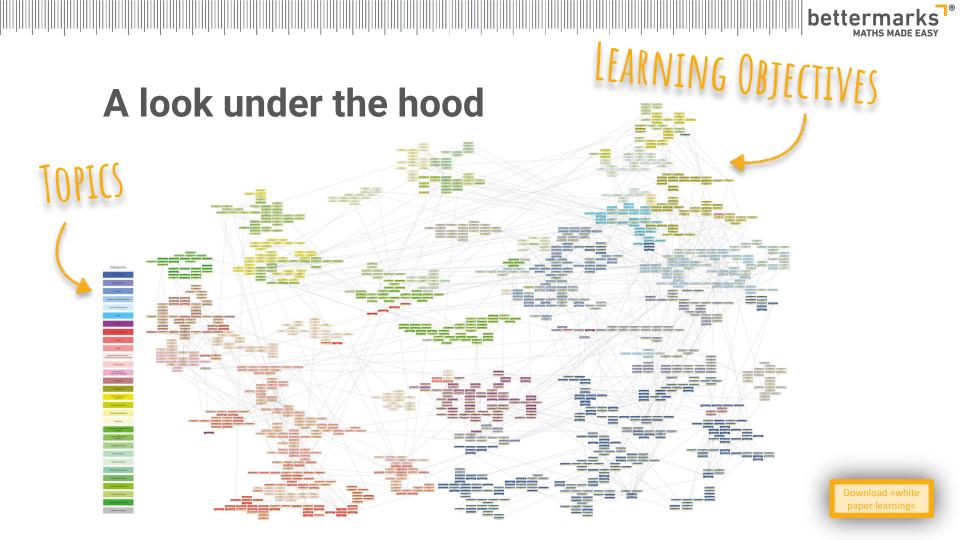
	Pen and Paper	Multiple Choice	
Making mistakes Applying mathematical skills	۲	777	1
Helpful feedback Unterstanding the root of a mistake	~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~	Ş	۲
Detection of knowledge gaps Catching up independently	i		(

bettermarks[®]

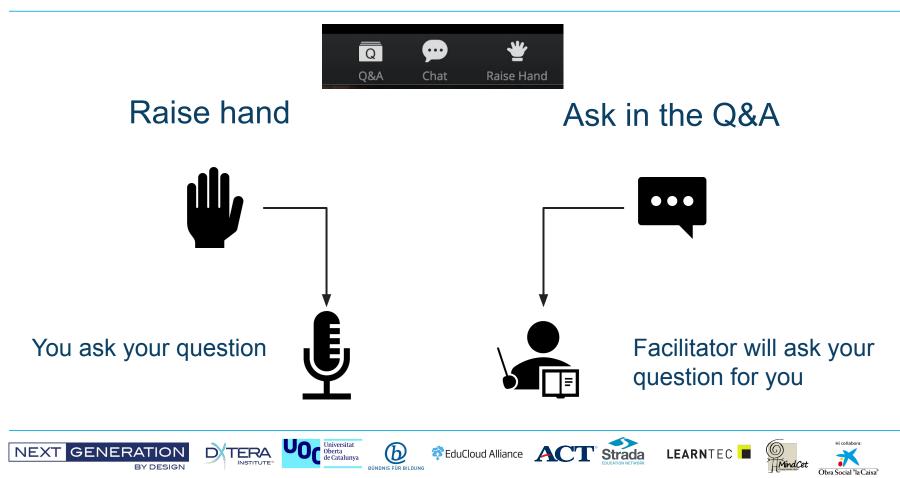
Impact by learning from mistakes

Every mistake with a feedback is an opportunity to understand the root of the mistake and to apply your learning in a second attempt.





Discussion and Questions



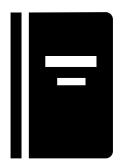
Discussion

NGBD Series Sessions and Next Steps

• All Webinar recordings, papers, community discussions, and details to be posted on the <u>NGBD</u> Site.

🕸 EduCloud Alliance 🛛 🗛 📿 🗂 Strada

- May 26 Exploring New Frontiers
- Summary papers to be published
- Questions or comments to info@dxtera.org



LEARNTEC





BY DESIGN



Thank You