



@DrBartRienties

Professor of Learning Analytics



Learning analytics at the Open University and the UK: reviewing 6 years of implementation at scale

Webinar 11 June 2020



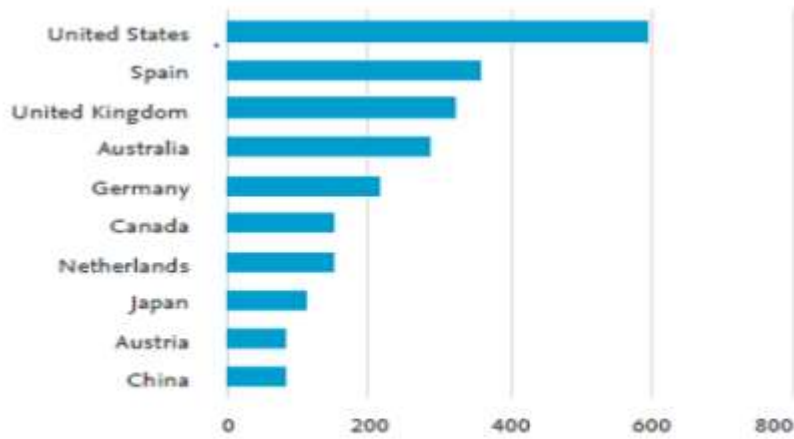
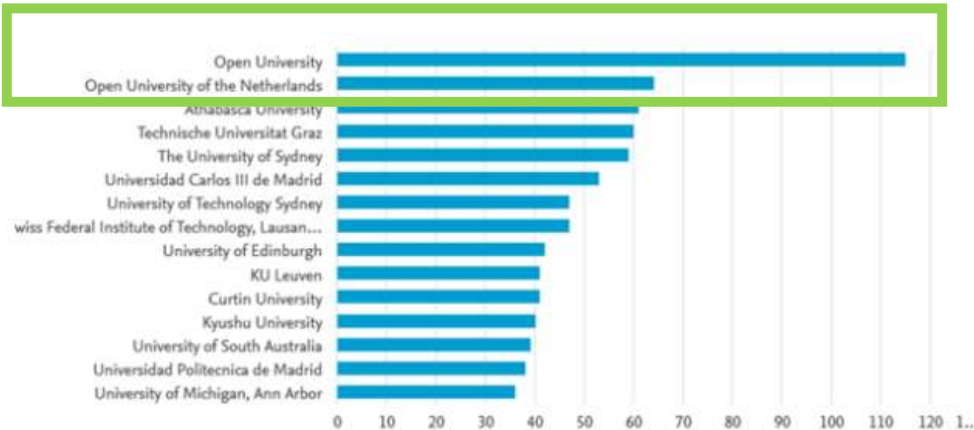


Figure 7: Learning Analytics Affiliation by County - ALL



Author / Institution Country	Cited	%	H	Subject Areas
Pardo, Abelardo AUS U of Sydney	45	1.6	21	Learning Analytics, Learning Technologies, Technology Enhanced Learning, Learning Design, Educational Technology
Draschler, Hendrick NLD- The Open University	40	1.5	28	Educational Technologies, Learning Analytics, Recommender Systems, Medical Education, Self-Regulated Learning
Dawson, Shane AUS U of South Australia	33	1.2	35	Learning Analytics, Social Network Analysis, Technology Enhanced learning, Learning Design Education
Rienties, Bart GBR The Open University	32	1.2	27	Learning Analytics, Learning Design, Social Network Analysis, Computer-Supported Collaborative Learning
Kinshuk USA U of North Texas	31	1.1	46	Online learning, Mobile learning, Ubiquitous Learning, Cognitive Profiling, Adaptivity Information
Kloos, Carlos D ESP U of Carlos III Madrid	31	1.1	35	Digital Education, Technology-enhanced, Learning, Educational Technology, MOOCs, eLearning
Munz-Merino, Pedro ESP U of Carlos III Madrid	31	1.1	21	Educational Data Mining, Learning Analytics, Gamification, Educational Technology
Gasevic, Dragan AUS Monash University	30	1.1	51	Learning Analytics, Self-regulated learning, Technology, Enhanced Learning, Collaborative Learning, Learning Technologies
Ebner, Martin AUT Graz University	28	1.0	35	E-learning, Open Educational Resources, Learning Analytics MOOC TEL
Ferguson, Rebecca GBR The Open University	28	1.0	24	Education, Learning Analytics, MOOCs, Distance Learning, Online learning
Verbert, Katrien BEL KU Leuven	26	1.0	31	HCI, Visualization user interfaces for recommender systems, Technology Enhanced Learning, Digital Humanities
Scheffel, Maren UK The Open University	25	1.0	14	Learning Analytics, Evaluation Self-Regulated Learning, Learning Design
Ogata, Hiroaki JPN Kyoto University	24	.9	30	Educational Data Science, Learning Analytics, Mobile and Ubiquitous

Web of Science



Results Analysis

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Showing 2,400 records for TOPIC: ("learning analytics")

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Web of Science Categories

Publication Years

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Organizations-Enhanced

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Visualization **Treemap**

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Sort by **Record count**

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Select records to view, or exclude. Choose "View records" to view the selected records only or "Exclude records" to view the unselected records only.

Select	Field: Organizations-Enhanced	Record Count	% of 2,400	Bar Chart
<input type="checkbox"/>	OPEN UNIVERSITY UK	57	2.375 %	

Wakelam, E., Jefferies, A., Davey, N., & Sun, Y. (2020). The potential for student performance prediction in small cohorts with minimal available attributes. *British Journal of Educational Technology*, 51(2), 347-370. doi: 10.1111/bjet.12836



“In the UK the Open University (OU) is a **world leader** in the collection, intelligent analysis and use of large scale student analytics. It provides academic staff with systematic and high quality actionable analytics for student, academic and institutional benefit (Rienties, Nguyen, Holmes, Reedy, 2017). Rienties and Toetenel’s, 2016 study (Rienties & Toetenel, 2016) identifies the importance of the linkage between LA outcomes, student satisfaction, retention and module learning design. These analytics are often provided through **dashboards tailored for each of academics and students** (Schwendimann et al., 2017).

The OU’s **world-class Analytics4Action initiative** (Rienties, Boroowa, Cross, Farrington-Flint et al., 2016) supports the university-wide approach to LA. In particular, the initiative provided valuable insights into the identification of students and modules where interventions would be beneficial, analysing over 90 large-scale modules over a two-year period...

The deployment of LA establishes the **need and opportunity for student and module interventions** (Clow, 2012). The study concludes that the faster the feedback loop to students, the more effective the outcomes. This is often an iterative process allowing institutions to understand and address systematic issues.

Legal, ethical and moral considerations in the deployment of LA and interventions are key challenges to institutions. They include informed consent, transparency to students, the right to challenge the accuracy of data and resulting analyses and prior consent to intervention processes and their execution (Slade & Tait, 2019)”

Title/Abstract matches "learning analytics"

Displaying results 1 to 20 of 265.
Refine search New search 1 2 3 4 5 6 7 8 9 10 11 Next
Order the results: by year (most recent first) Reorder

Export results as ASCII Citation Export RSS 1.0 Atom RSS 2.0 SURL

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1. Nguyen, Quan; Rienties, Bart and Richardson, John T. E. (2020). Learning analytics to uncover inequality in behavioural engagement and academic attainment in a distance learning setting. *Assessment & Evaluation in Higher Education*, 45(4) pp. 594-606. 
2. Rodríguez León, Lucy Jane (2020). *Literacy Experiences: An Exploration Of Young Children's Orientations, Identities, And Affective Relations With Text*. PhD thesis. The Open University.  +1 more...
3. Nguyen, Quan (2020). *Unravelling the Temporal Process of Learning Design and Student Engagement in Distance Education using Learning Analytics*. PhD thesis. The Open University.  +1 more...

<http://oro.open.ac.uk/>

The Open University



Leading global distance learning, delivering high-quality education to anyone, anywhere, anytime



Largest
University
in Europe

38% of part-time
undergraduates
taught by OU in UK

173,927 formal
students



No formal
entry
requirements



33%
enter with one
A-level or less

55%

of students are
'disadvantaged'

66% ²⁵
of new
undergraduates
are 25+

60%

FTSE 100 have
sponsored staff on OU
courses in 2017/8

1 in 8

Open University students
has a disability (23,630)



3 in 4

Students are
already in work



1,300

employers use
OU learning
solutions to
develop
workforce



A special thanks to Vaclav Bayer, Avinash Boroowa, Shi-Min Chua, Simon Cross, Doug Clow, Chris Edwards, Rebecca Ferguson, Mark Gaved, Christothea Herodotou, Martin Hlosta, Wayne Holmes, Garron Hillaire, Simon Knight, Nai Li, Vicky Marsh, Kevin Mayles, Jenna Mittelmeier, Vicky Murphy, Mark Nichols, Quan Nguyen, Tom Olney, Lynda Prescott, John Richardson, Saman Rizvi, Jekaterina Rogaten, Matt Schencks, Mike Sharples, Dirk Tempelaar, Belinda Tynan, Lisette Toetenel, Thomas Ullmann, Denise Whitelock, Zdenek Zdrahal, and others...





What have I learned in six years at the OU

Change is slow, but can be enhanced with:

1. Clear senior management support
2. Bottom-up support from teachers and researchers who are willing to take a risk
3. Evidence-based research can gradually change perspectives and narratives
4. You quickly forget about the small/medium/large successes and fail to realise that you are making a real impact
5. Large-scale innovation takes substantial time and effort
6. It is all about people...



Magic of learning design (does not come easy)



Learning Design: European Approaches

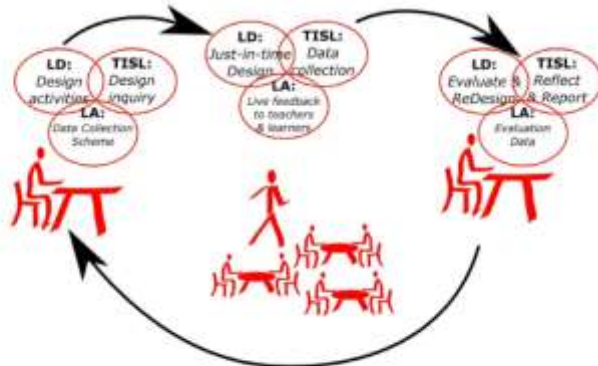
Barbara Wasson¹ · Paul A. Kirschner²

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Abstract

Research on instructional and learning design is “booming” in Europe, although there has been a move from a focus on content and the way to present it in a formal educational context (i.e., instruction), to a focus on complex learning, learning environments including the workplace, and access to learner data available in these environments. We even see the term “learning experience design” (Neelen and Kirschner 2020) to describe the field. Furthermore, there is an effort to empower teachers (and even students) as designers of learning (including environments and new pedagogies), and to support their reflection on their own practice as part of their professional development (Hansen and Wasson 2016;

Fig. 7 Teacher-led design inquiry of learning and innovation cycle (Wasson et al. 2016)



“Research on **the relationship between learning design and learning analytics** has also been a focus in European research in recent years. For example, in their research at **the Open University UK**, Toetenel and Rienties combine learning design and learning analytics where learning design provides context to empirical data about OU courses enabling the learning analytics to give insight into learning design decisions. **This research is important as it attempts to close the virtuous cycle between learning design to improve courses and enhancing the quality of learning, something that has been lacking in the research literature.** For example, they study the impact of learning design on pedagogical decision-making and on future course design, and the relationship between learning design and student behaviour and outcomes (Toetenel and Rienties 2016; Rienties and Toetenel 2016; Rienties et al. 2015).”

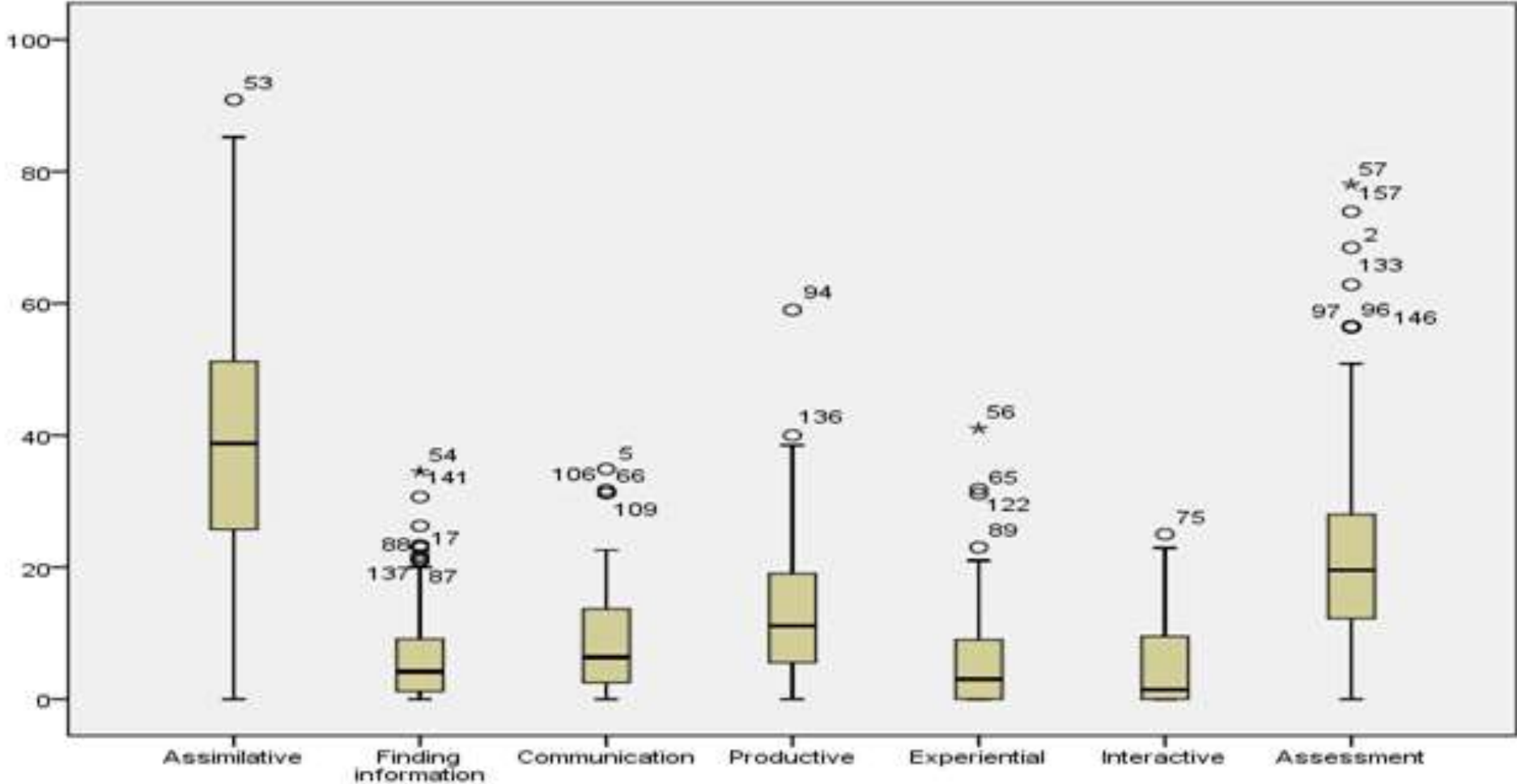


Open University Learning Design Initiative (OULDI)

	Assimilative	Finding and handling information	Communication	Productive	Experiential	Interactive/ Adaptive	Assessment
Type of activity	Attending to information	Searching for and processing information	Discussing module related content with at least one other person (student or tutor)	Actively constructing an artefact	Applying learning in a real-world setting	Applying learning in a simulated setting	All forms of assessment, whether continuous, end of module, or formative (assessment for learning)
Examples of activity	Read, Watch, Listen, Think about, Access, Observe, Review, Study	List, Analyse, Collate, Plot, Find, Discover, Access, Use, Gather, Order, Classify, Select, Assess, Manipulate	Communicate, Debate, Discuss, Argue, Share, Report, Collaborate, Present, Describe, Question	Create, Build, Make, Design, Construct, Contribute, Complete, Produce, Write, Draw, Refine, Compose, Synthesise, Remix	Practice, Apply, Mimic, Experience, Explore, Investigate, Perform, Engage	Explore, Experiment, Trial, Improve, Model, Simulate	Write, Present, Report, Demonstrate, Critique

Conole, G. (2012). *Designing for Learning in an Open World*. Dordrecht: Springer.

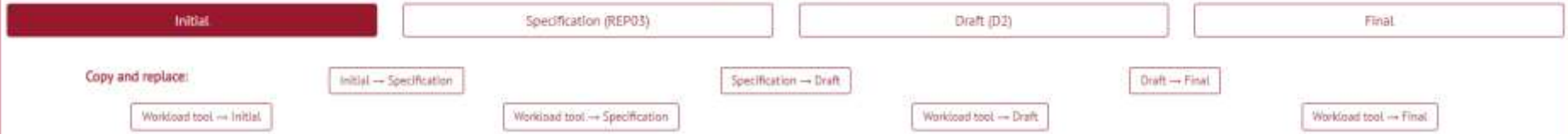
Rienties, B., Toeteneel, L., (2016). The impact of learning design on student behaviour, satisfaction and performance: a cross-institutional comparison across 151 modules. *Computers in Human Behavior*, 60 (2016), 333-341



Toeteneel, L., Rienties, B. (2016). Analysing 157 Learning Designs using Learning Analytic approaches as a means to evaluate the impact of pedagogical decision-making. *British Journal of Educational Technology*, 47(5), 981–992.

Hours spent undertaking each type of activity

Design stages

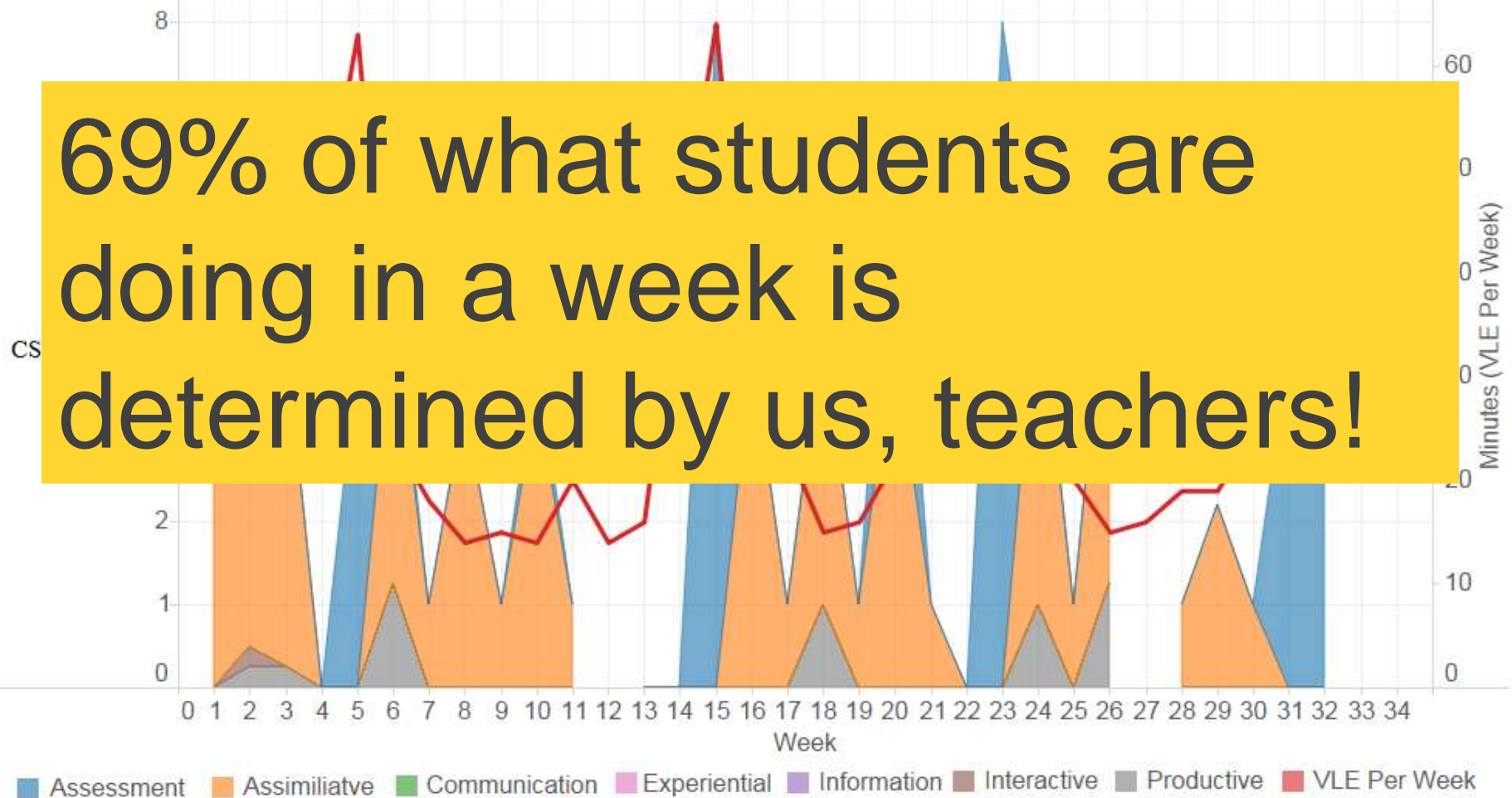


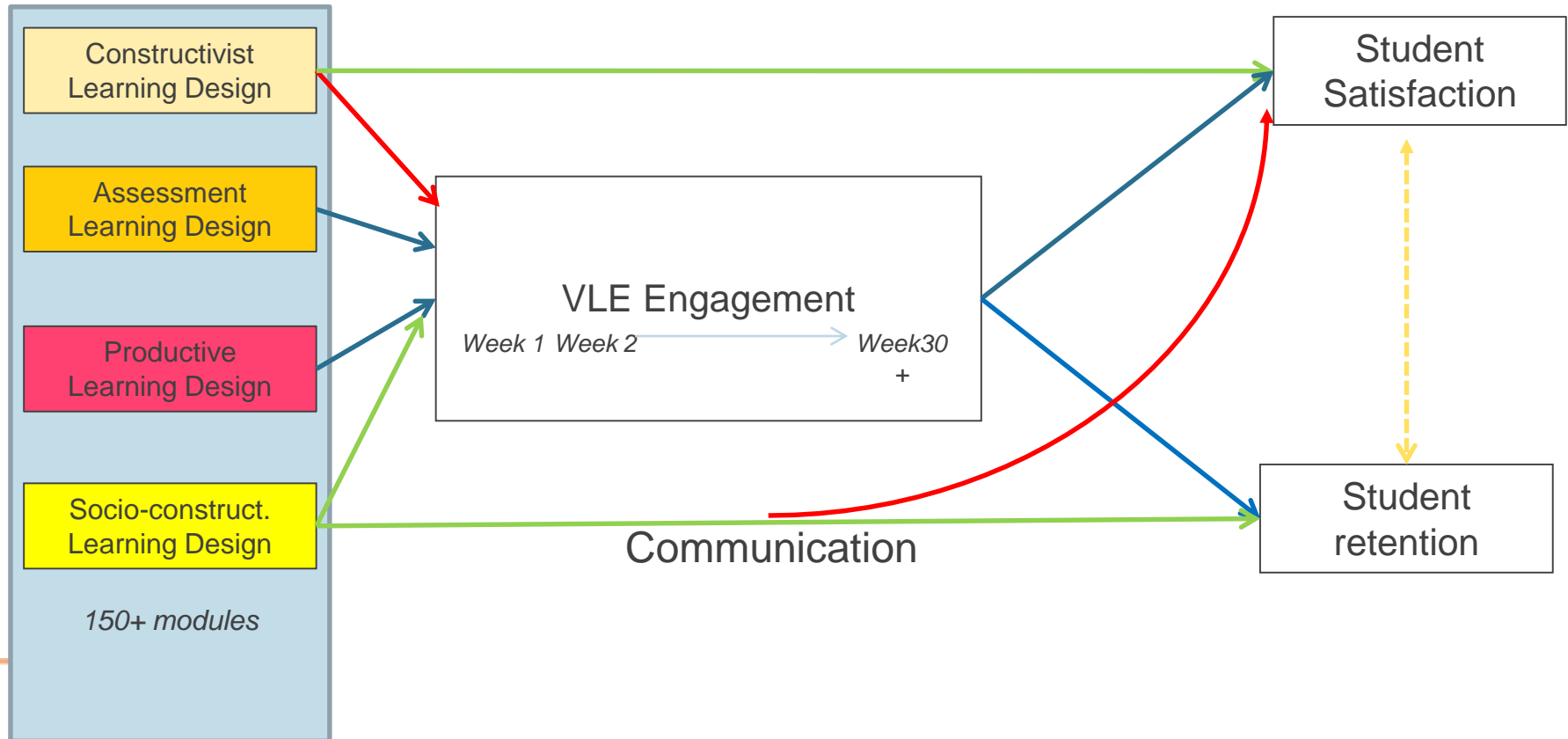
Week	Assimilative		Finding and handling information		Communication		Productive		Experiential		Interactive / Adaptive		Assessment		Total hours	
	Value	Icon	Value	Icon	Value	Icon	Value	Icon	Value	Icon	Value	Icon	Value	Icon	Avg: 12.16, StDv: 6.28	Hide guides
Week 1	10	✂	1.5	✂	1	✂	0.6	✂	0	✂	0	✂	0.2	✂	13.30	
Week 2	6.1	✂	0	✂	0.5	✂	0.6	✂	2	✂	0	✂	6.2	✂	15.40	
Week 3	6.1	✂	0	✂	0	✂	2.2	✂	2.85	✂	0	✂	3.5	✂	14.65	
Week 4	0	✂	0	✂	0	✂	0	✂	0	✂	0	✂	0	✂	0	
Week 5	5.8	✂	0	✂	0	✂	0	✂	19.1	✂	0	✂	10.9	✂	35.85	
Week 6	13.5	✂	0	✂	0	✂	3.55	✂	4.3	✂	0	✂	1.8	✂	23.15	
Week 7	7.25	✂	0.4	✂	0	✂	1	✂	0.7	✂	0	✂	3.3	✂	11.65	
Week 8	5.79	✂	0	✂	0	✂	0	✂	0	✂	0	✂	9.3	✂	15.09	
Week 9	10.5	✂	0	✂	0	✂	3	✂	0.1	✂	0	✂	2.5	✂	16.15	
Week 10	6.31	✂	0	✂	0.5	✂	0.55	✂	0.7	✂	0	✂	2.65	✂	10.51	
Week 11	7.46	✂	4	✂	0	✂	2.1	✂	0	✂	0	✂	3.2	✂	16.76	
Week 12	5.69	✂	0	✂	0	✂	1.3	✂	0.35	✂	0.5	✂	1.8	✂	9.64	
Week 13	7.43	✂	0.65	✂	0	✂	2.8	✂	0.6	✂	0	✂	1.6	✂	11.08	

Merging big data sets

- Learning design data (>300 modules mapped)
- VLE data
 - >140 modules aggregated individual data weekly
 - >37 modules individual fine-grained data daily
- Student feedback data (>140)
- Academic Performance (>140)
- Predictive analytics data (>40)
- Data sets merged and cleaned
 - 111,256 students undertook these modules

69% of what students are doing in a week is determined by us, teachers!





Rienties, B., Toeteneel, L., (2016). The impact of learning design on student behaviour, satisfaction and performance: a cross-institutional comparison across 151 modules. *Computers in Human Behavior*, 60 (2016), 333-341

Nguyen, Q., Rienties, B., Toeteneel, L., Ferguson, R., Whitelock, D. (2017). Examining the designs of computer-based assessment and its impact on student engagement, satisfaction, and pass rates. *Computers in Human Behavior*. DOI: 10.1016/j.chb.2017.03.028.

Learning Design

About

Blog

Processes

Tools

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[▶ Online Learning Design Tool](#)
[▶ The ICEBERG Model](#)
[▶ Qualification Design Tools](#)
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Analytics and Evaluation

Scholarship and Research

Partnerships

Training and Support

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Activity Planner Categories

The Activity Planner categories represent different types of activities that students can engage with in the course of a module, each providing a distinct set of pedagogical benefits. Here you will find some explanations of each category and the type of activities that fall within it.



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Assimilative

Attending to information

Students study and think about theories and concepts encountered in materials and resources, case studies, etc.

Often the first part of a learning cycle where students receive and begin to make sense of new information, before they then apply or test their new knowledge, or go on to reflect, review and communicate their understanding.

Read, Watch, Listen, Think about, Access, Observe, Review, Consider, Study



Finding and Handling information

Searching for and processing information

Students are actively and critically engaged in gathering and manipulating information.

Predictive analytics and professional development



• Predictions

Student ID	Name	Total PT	Staff/Part PT	Risk	Next Prediction (Week 20/21/22)	Score	Completion
XXXXXXXX	Frank Hoad	100000	100000	Green	Green	Green	75.0%
XXXXXXXX	Helen Jones	100000	100000	Red	Red	Red	80.0%
XXXXXXXX	Alan Day	100000	100000	Green	Green	Green	85.0%
XXXXXXXX	Wendy Chen	100000	100000	Yellow	Yellow	Yellow	70.0%
XXXXXXXX	Charles Wilson	100000	100000	Green	Green	Green	80.0%
XXXXXXXX	Angela Adams	100000	100000	Red	Red	Red	90.0%
XXXXXXXX	Christopher Brown	100000	100000	Green	Green	Green	80.0%
XXXXXXXX	Carol Taylor	100000	100000	Yellow	Yellow	Yellow	85.0%
XXXXXXXX	Clare Miller	100000	100000	Yellow	Green	Green	50.0%
XXXXXXXX	Pauline Smith	100000	100000	Red	Red	Red	90.0%
XXXXXXXX	Emily Wilson	100000	100000	Red	Red	Red	85.0%
XXXXXXXX	Thomas Clark	100000	100000	Yellow	Green	Green	80.0%
XXXXXXXX	Oliver Lee	100000	100000	Green	Green	Green	85.0%
XXXXXXXX	Alice White	100000	100000	Red	Green	Green	90.0%
XXXXXXXX	Ben King	100000	100000	Green	Green	Green	70.0%
XXXXXXXX	Wendy Jones	100000	100000	Yellow	Green	Green	90.0%
XXXXXXXX	Carl Roberts	100000	100000	Green	Green	Green	85.0%
XXXXXXXX	Neil Green	100000	100000	Green	Red	Green	80.0%
XXXXXXXX	Esther Henry	100000	100000	Green	Green	Green	85.0%
XXXXXXXX	Oliver King	100000	100000	Yellow	Green	Green	80.0%
XXXXXXXX	Robert Young	100000	100000	Yellow	Red	Green	50.0%
XXXXXXXX	Wendy Clark	100000	100000	Green	Green	Green	85.0%
XXXXXXXX	Andrew Powell	100000	100000	Green	Green	Green	85.0%
XXXXXXXX	Jamie Brown	100000	100000	Green	Green	Green	85.0%
XXXXXXXX	Ben Adams	100000	100000	Green	Green	Green	85.0%

Showing 1 to 25 of 25 entries.

Kuzilek, J., Hlosta, M., Herrmannova, D., Zdrahal, Z., & Wolff, A. (2015). OU Analyse: analysing at-risk students at The Open University LACE Learning Analytics Review (Vol. LAK15-1). Milton Keynes: Open University.

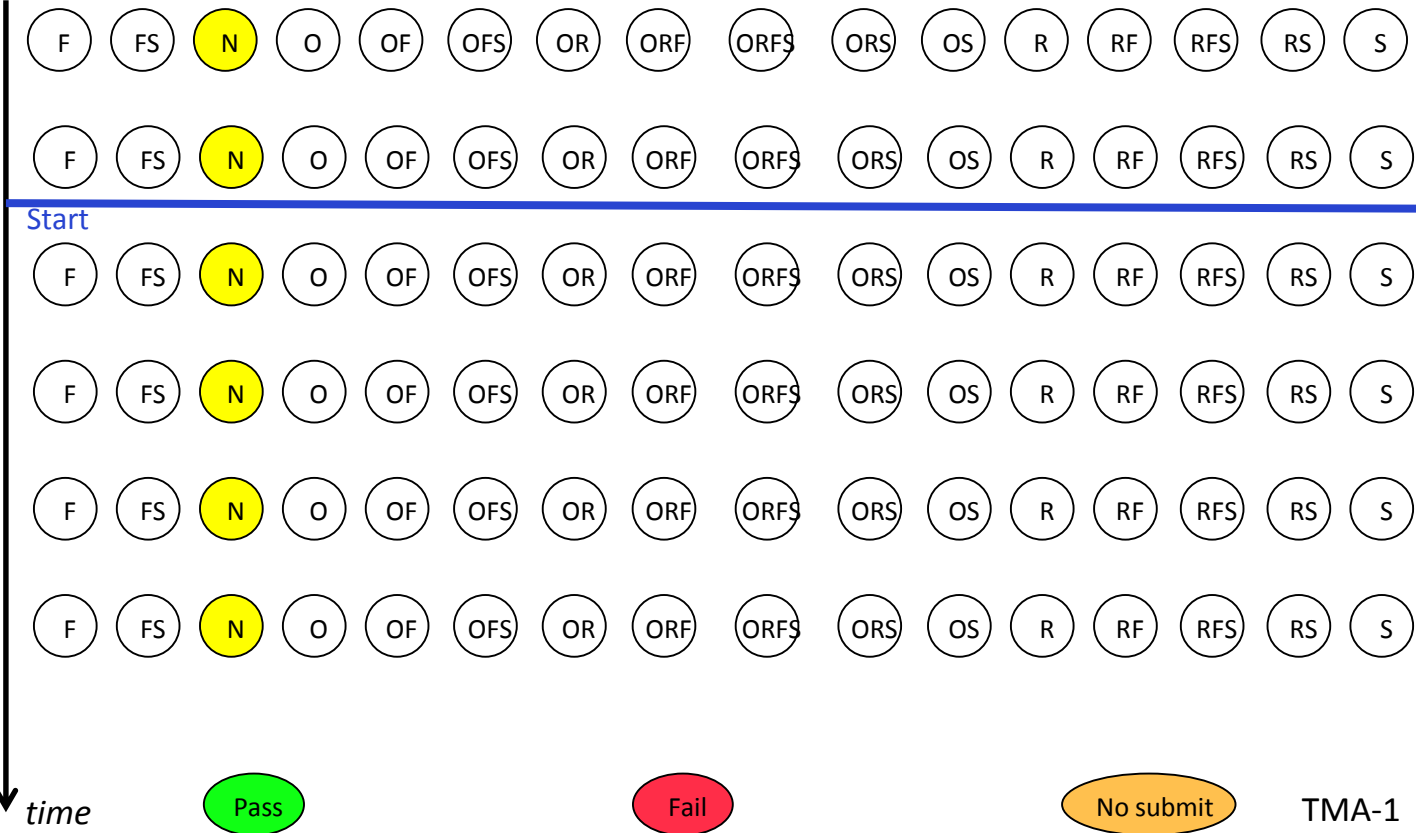
Kuzilek, J., Hlosta, M., & Zdrahal, Z. (2017). Open University Learning Analytics dataset. Scientific Data, 4, 170171. doi: 10.1038/sdata.2017.171

Wolff, A., Zdrahal, Z., Herrmannova, D., Kuzilek, J., & Hlosta, M. (2014). Developing predictive models for early detection of at-risk students on distance learning modules, Workshop: Machine Learning and Learning Analytics Paper presented at the Learning Analytics and Knowledge (2014), Indianapolis.

Activity space

Start

VLE opens



OU Analyse demo

<http://analyse.kmi.open.ac.uk>



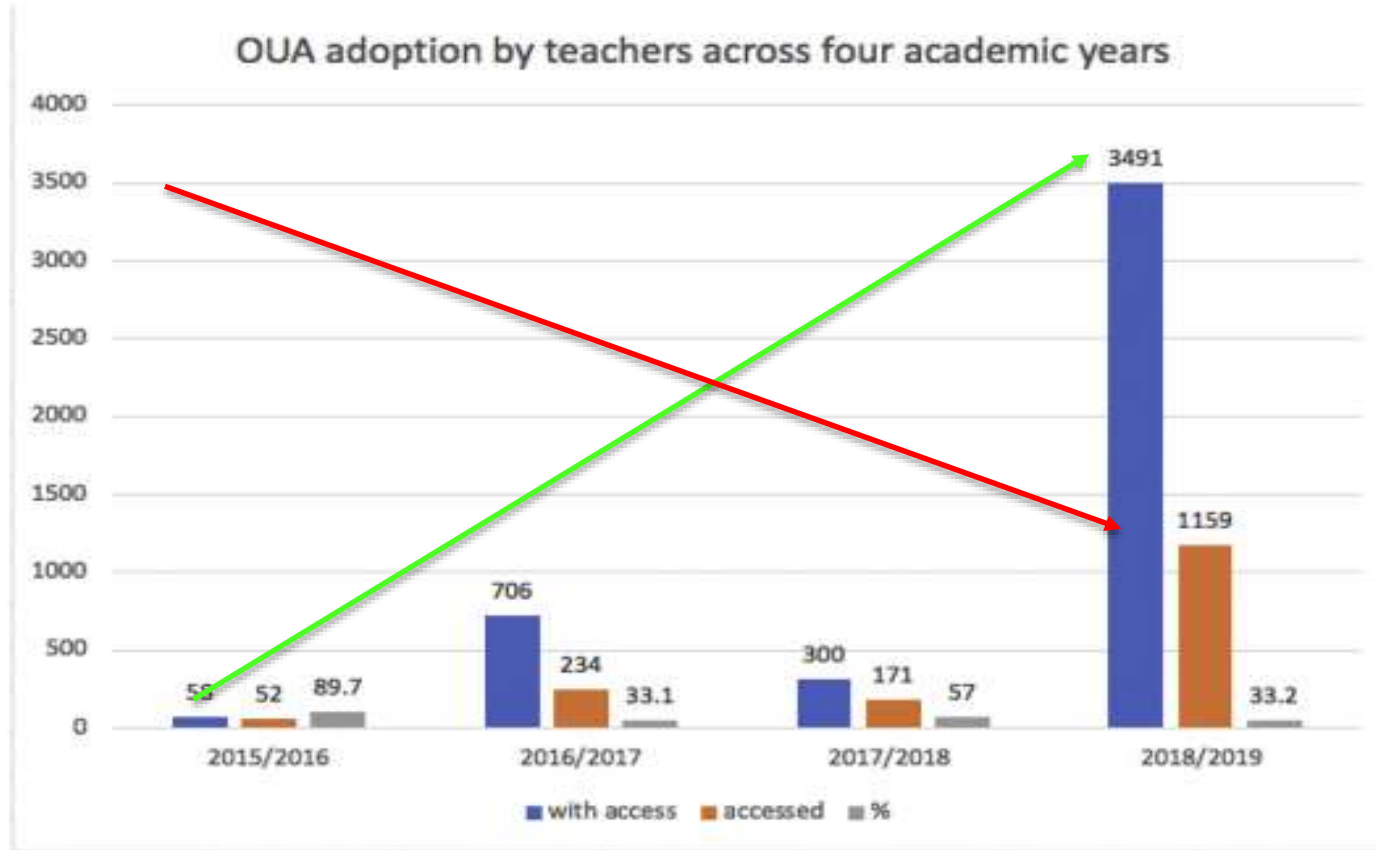


Fig. 2. OUA adoption by teachers during the last 4 academic years.

2015/16

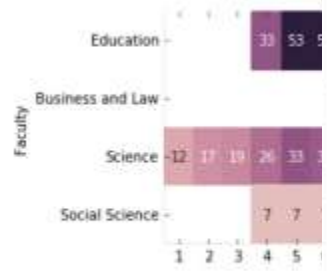
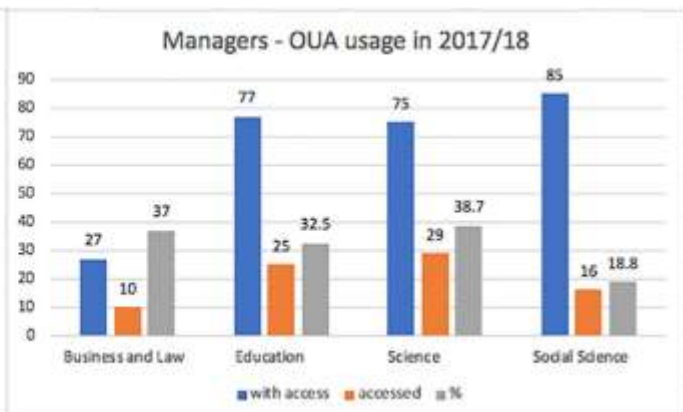
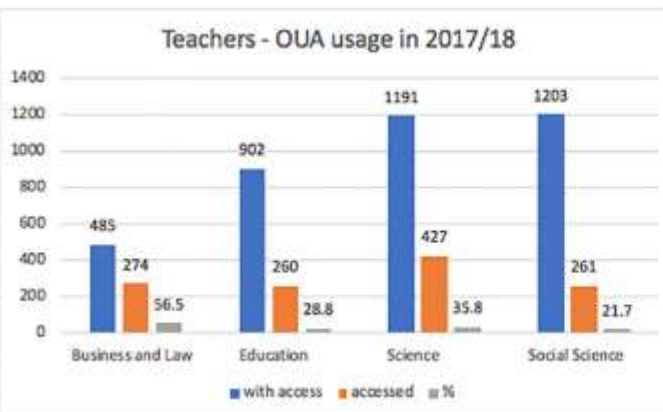


Fig. 3. Percentage



2016/17

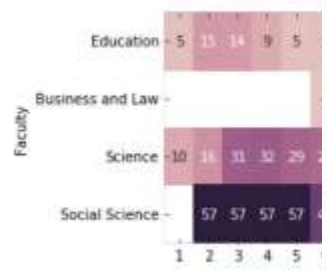
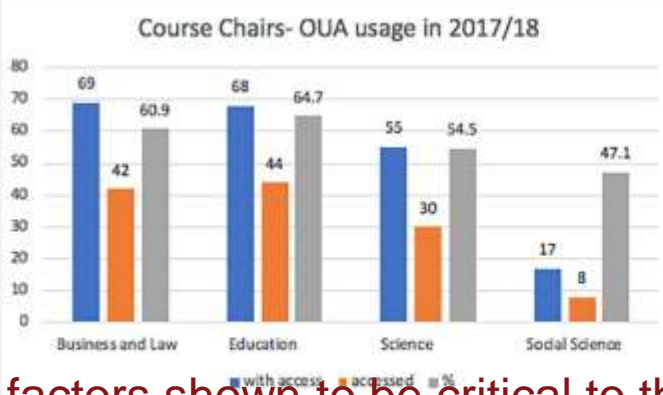


Fig. 4. Percentage

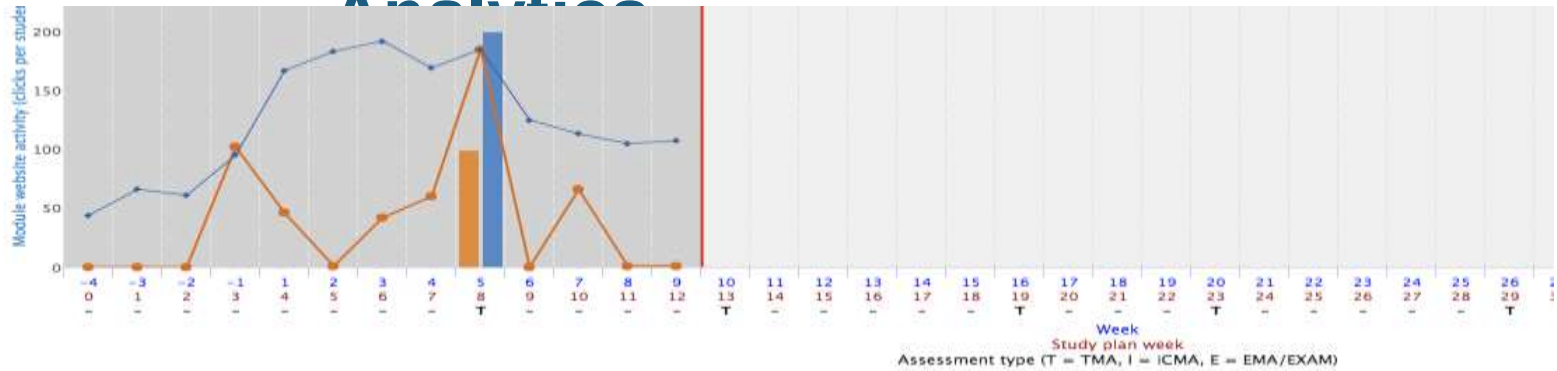


- Amongst the factors shown to be critical to the scalable PLA implementation were: Faculty's engagement with OUA, teachers as "champions", evidence generation and dissemination, digital literacy, and conceptions about teaching online.

Fig. 7. Staff usage of OUA in 2018/19 across four faculties.



Student Facing Analysis



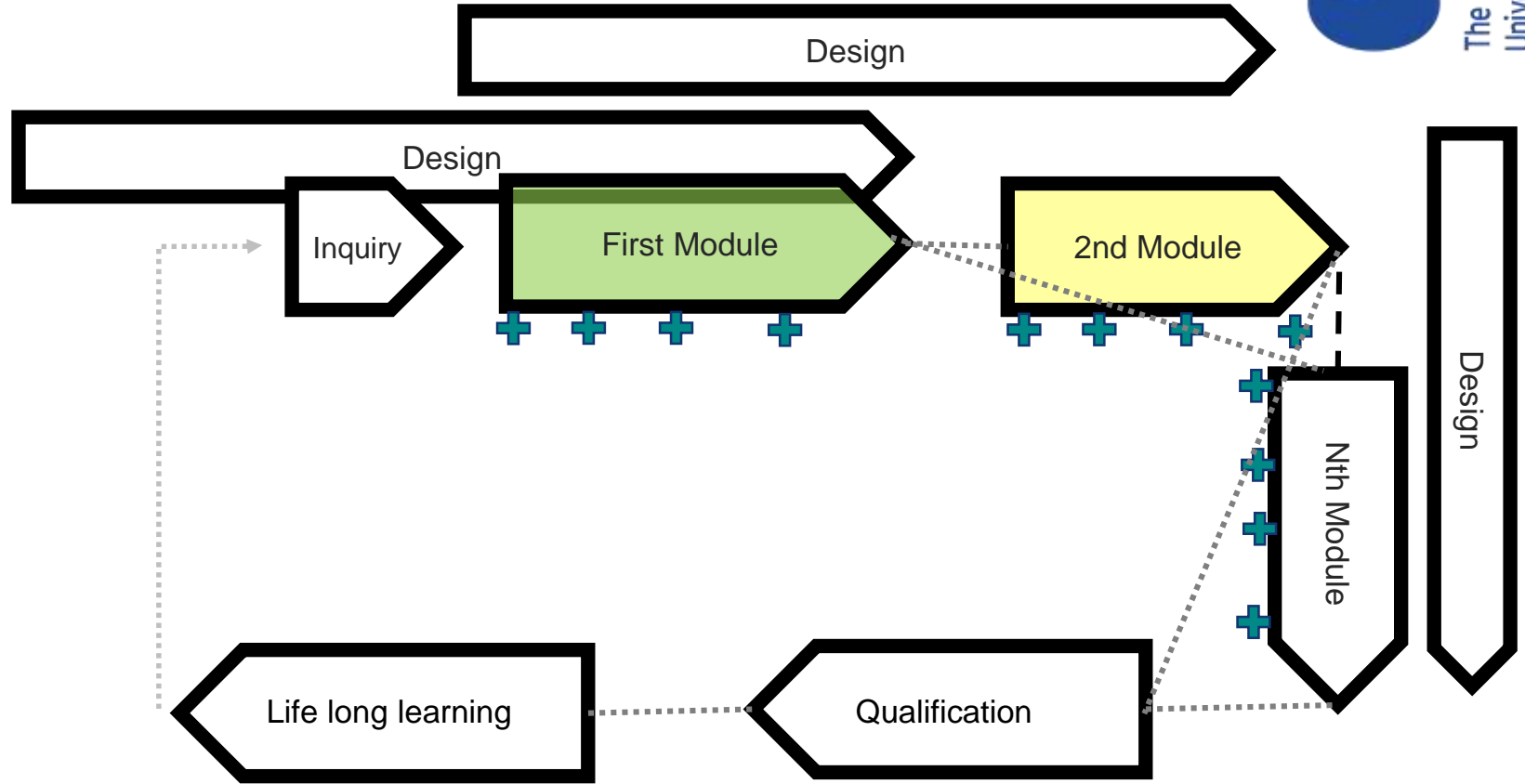
Study recommender

This link will enable you to navigate to the study material **Block 1 Part 6: Wireless communications and mobile computing** on the module website.

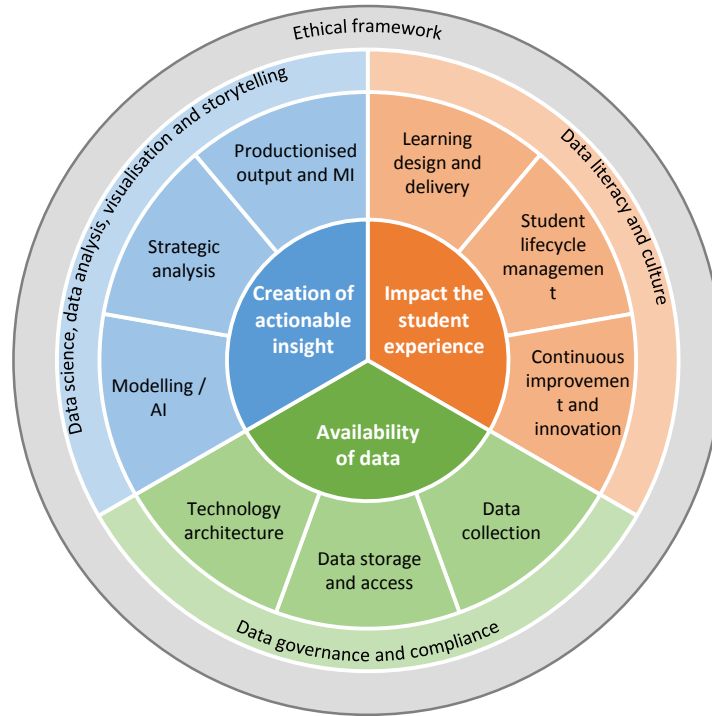
- Visit Block 1 Part 4: Geography is history
- Visit Block 1 Part 6: Wireless communications and mobile computing
- Consider participating in Sense surgery
- Visit Sense Programming Guide: Sessions 3'5 (online version)

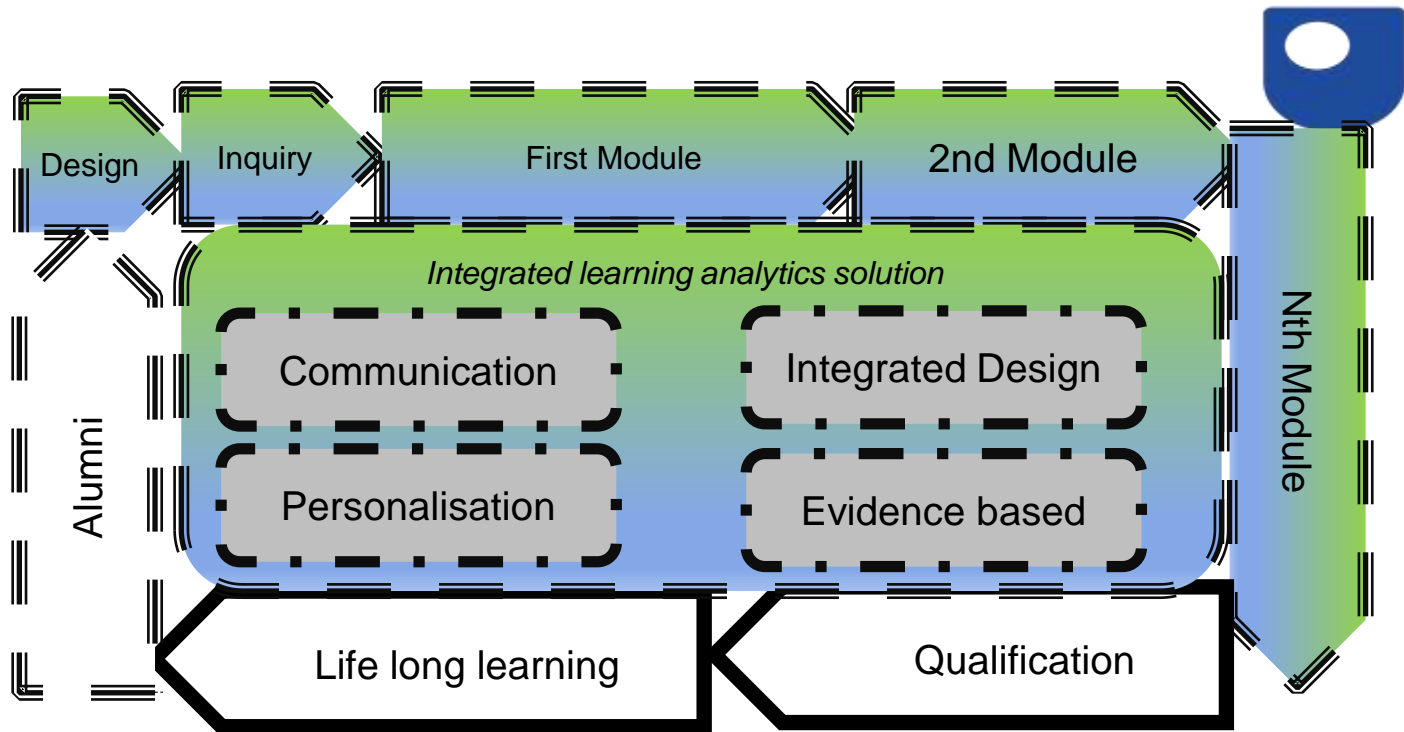
What do practitioners want and where should distance learning institutions be going?





ORGANISATIONAL CAPABILITIES





Nth Module

Integrated learning analytics solution

Communication

Integrated Design

Personalisation

Evidence based

Alumni

Life long learning

Qualification

Design

Inquiry

First Module

2nd Module



What have I learned in six years at the OU

Change is slow, but can be enhanced with:

1. Clear senior management support
2. Bottom-up support from teachers and researchers who are willing to take a risk
3. Evidence-based research can gradually change perspectives and narratives
4. You quickly forget about the small/medium/large successes and fail to realise that you are making a real impact
5. Large-scale innovation takes substantial time and effort
6. It is all about people...

Further reflections

1. What about the ethics?
2. What about professional development?
3. Are we optimising the record player?





iet



Yes I donate
ORGAN DONATION

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