Learning Analytics for 'end-users':

from Human-centred Design to Multimodal Data Storytelling

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Acknowledgement of country

I would like to acknowledge the Traditional Owners
of the land on which I am today,
the Yalukit Willam clan of the Boon Wurrung People, and the Traditional Owners
of the land where you all are at the moment.
I would like us to pay our respects to their Elders past, present and emerging.
We acknowledge and respect their continuing
relationship to the lands upon which we meet.

The power of storytelling



The "yarning circle" Aboriginal pedagogy: www.8ways.online

"We dream in narrative, daydream in narrative, remember, anticipate, hope, despair, believe, doubt, plan, revise, criticize, construct, gossip, learn, hate and love by narrative" Barbara Hardy -(1968, 5)

Storytelling in Journalism

Educators can adjust online classes to fit learning styles

Published April 15, 2020

By Shawna De La Rosa
Contributor, K-12



Flickr; Energy.gov

Dive Brief:

• Educators must balance many learning preferences as students adjust to online learning, which will be a good fit for some and difficult for others, <u>District Administration reports</u>. Some students will struggle



University Guide 2022 Online learning

How has the pandemic changed the way you'll learn?

As students gradually return to campus, many universities will be offering blended learning - mixing face-to-face lectures with the best of digital teaching

The 2022 league table

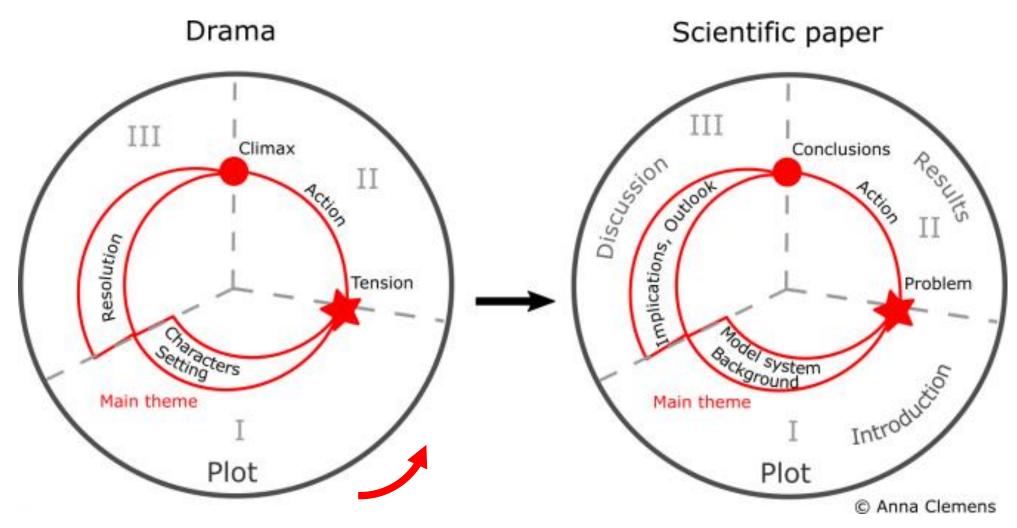




▲ Many universities are shifting large lectures online because they believe it's a better way for students to learn Photograph: Martin DM/Getty Images

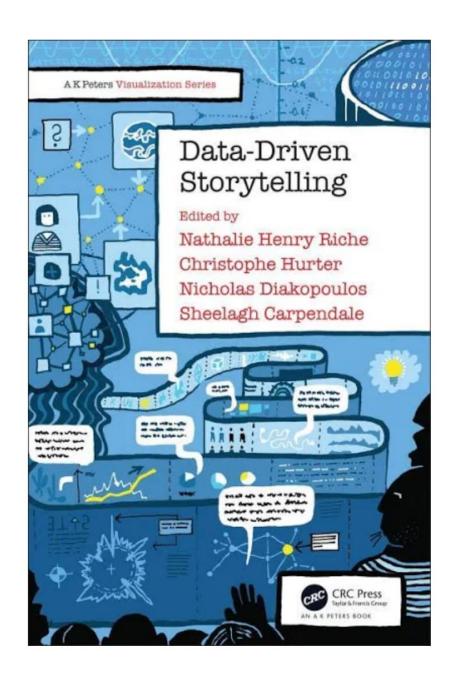
person teaching to motivate you and to meet new people? Some universities are offering a variety of options to suit different learning styles and personal circumstances.

Storytelling in Academia

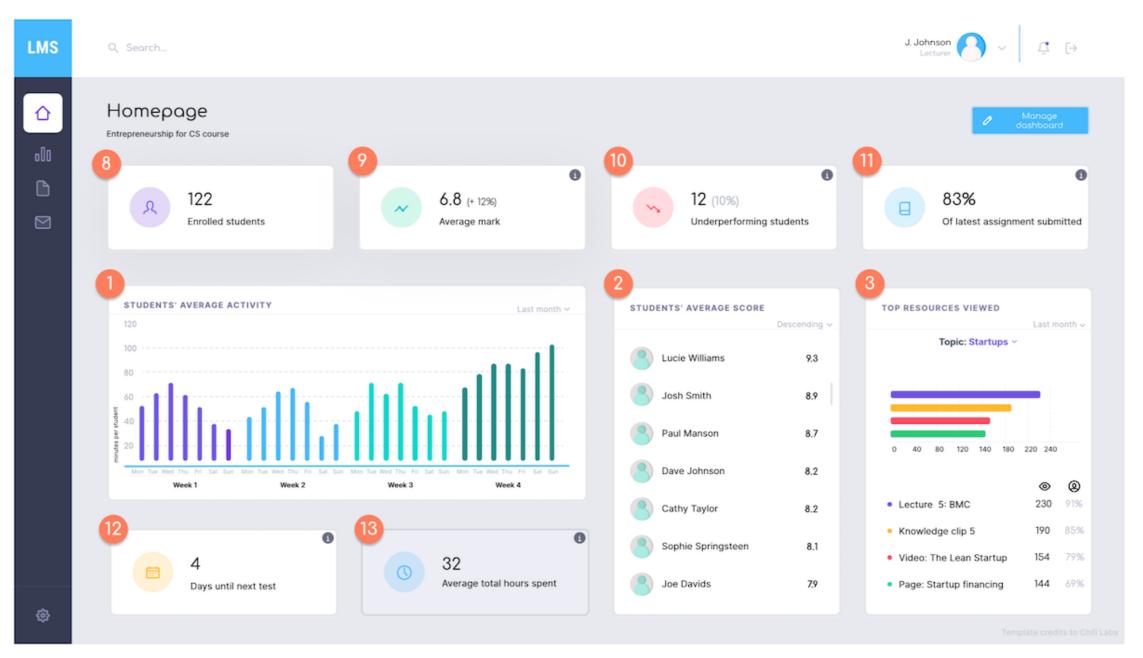


Data-driven Storytelling

"data-driven storytelling is the ability to turn raw data into easy-to-read and easy-to-understand plain stories that help us turn insights into action"







Daryl Zandvliet thesis work. ictinstitute.nl/learning-analytics-research

Dashboards are not delivering their promises

Students find it difficult to interpret/act on data to improve learning (Bodily & Verbert, 2017; Jivet et al., 2018; Matcha et al., 2019; Valle et al., 2021)

....and the same applies to **teachers** (Mangaroska & Giannakos, **2018**).

Bodily, R., & Verbert, K. (2017). Trends and issues in student-facing learning analytics reporting systems research. LAK'17

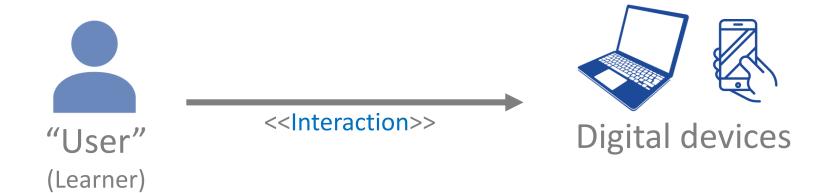
Jivet, I., Scheffel, M., Specht, M., & Drachsler, H. (2018). License to evaluate: Preparing learning analytics dashboards for educational practice. In LAK'18

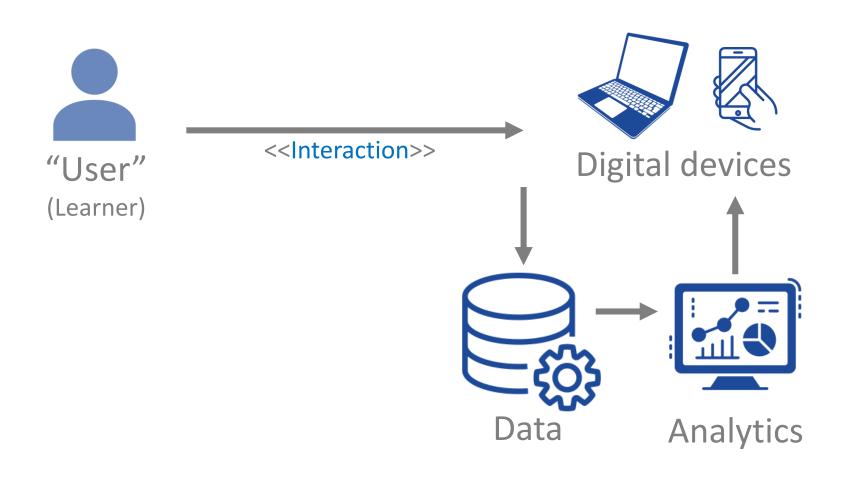
Matcha, W., Gasevic, D., & Pardo, A. (2019). A Systematic Review of Empirical Studies on Learning Analytics Dashboards: A Self-Regulated Learning Perspective. IEE TLT

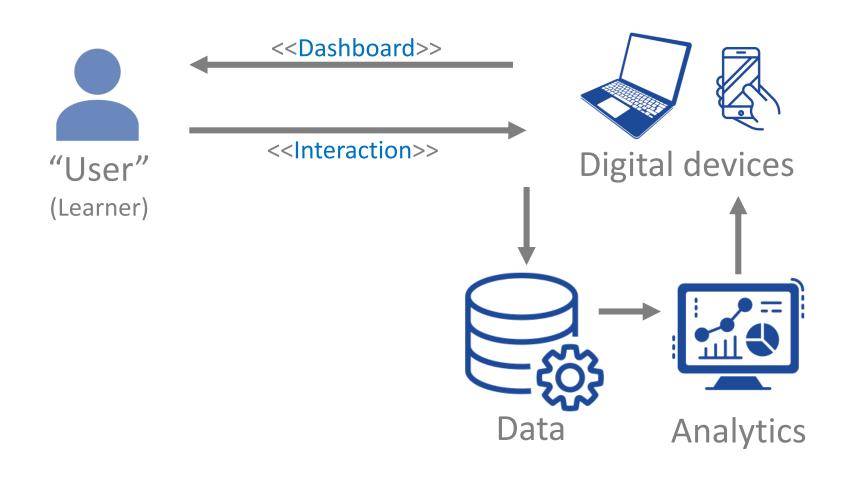
Valle, N., Antonenko, P., Dawson, K., & Huggins-Manley, A. C. (2021). Staying on target: A systematic literature review on learner-facing learning analytics dashboards. BJET

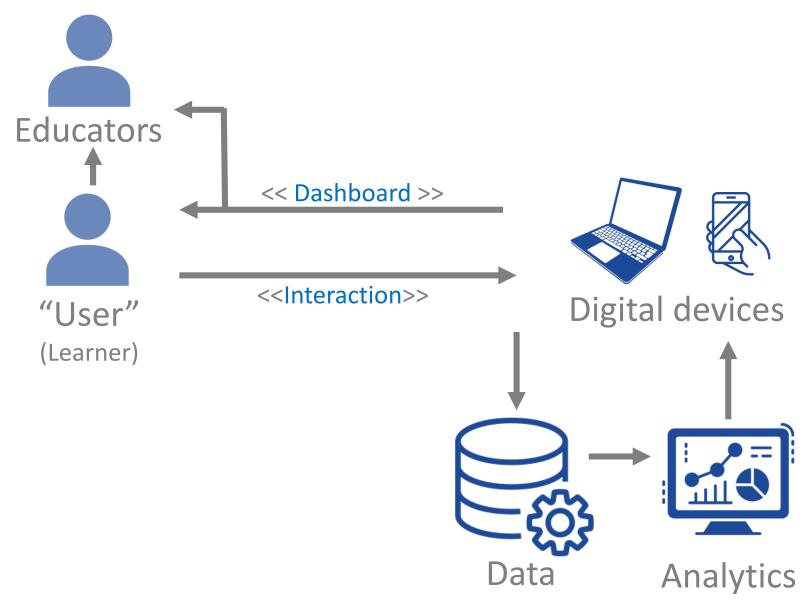
Mangaroska, K., & Giannakos, M. N. (2018). Learning analytics for learning design: A systematic literature review of analytics-driven design to enhance learning. IEEE TLT

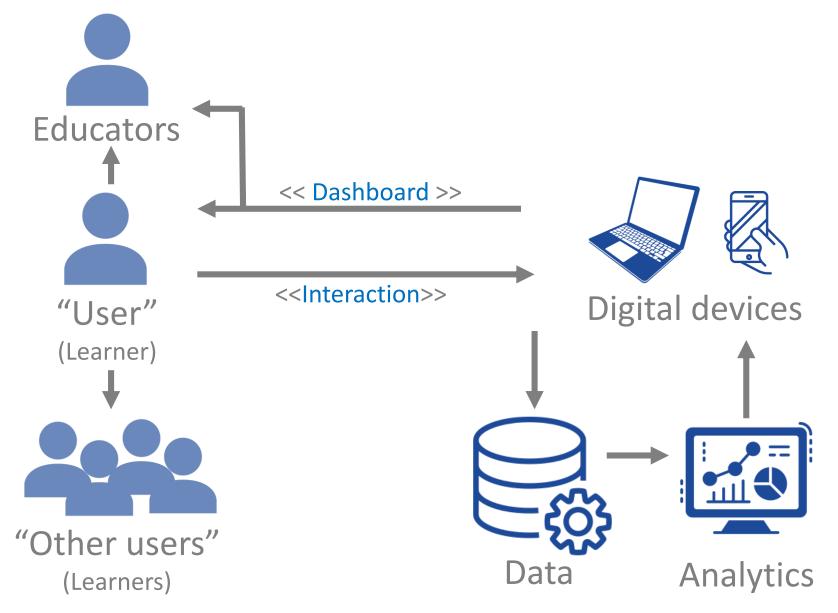








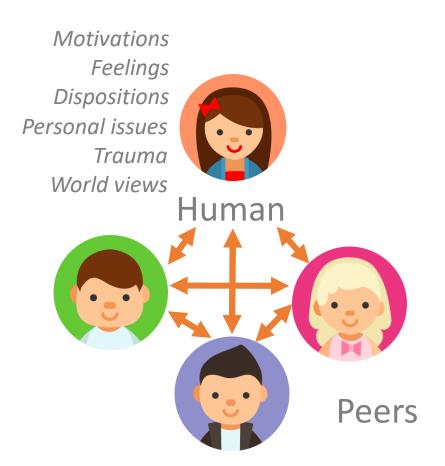








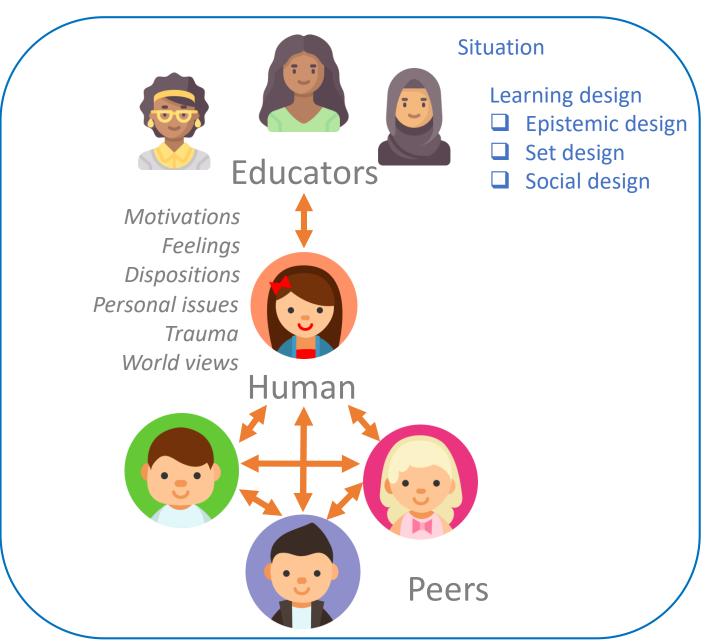




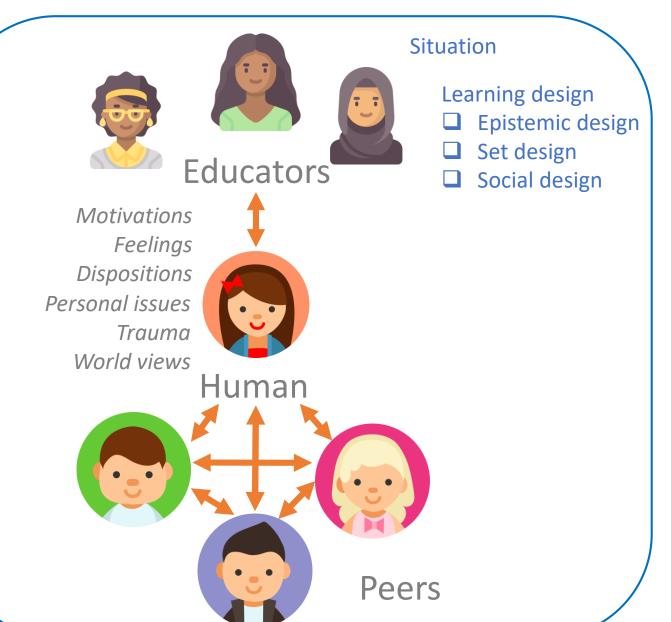






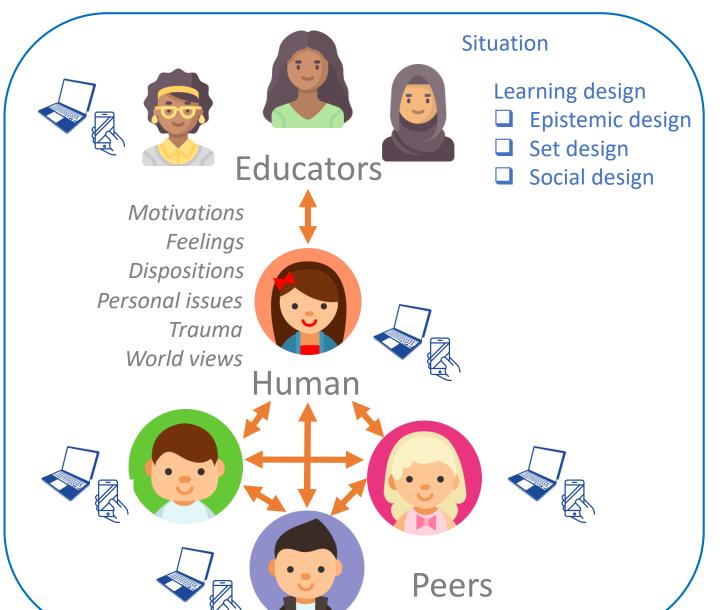






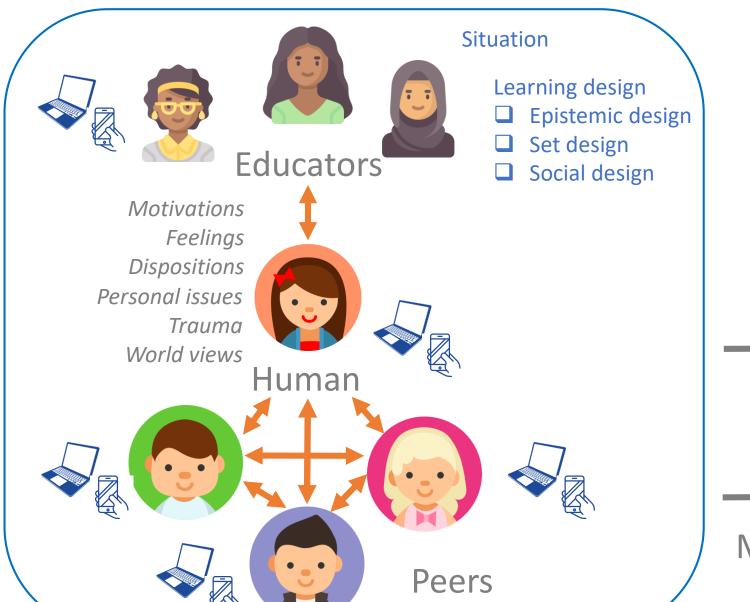
Do students/educators have a problem that can be addressed using their data?

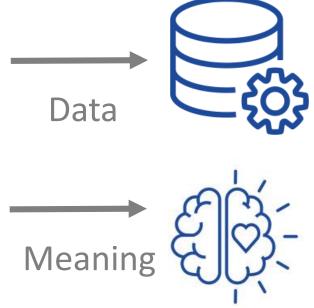




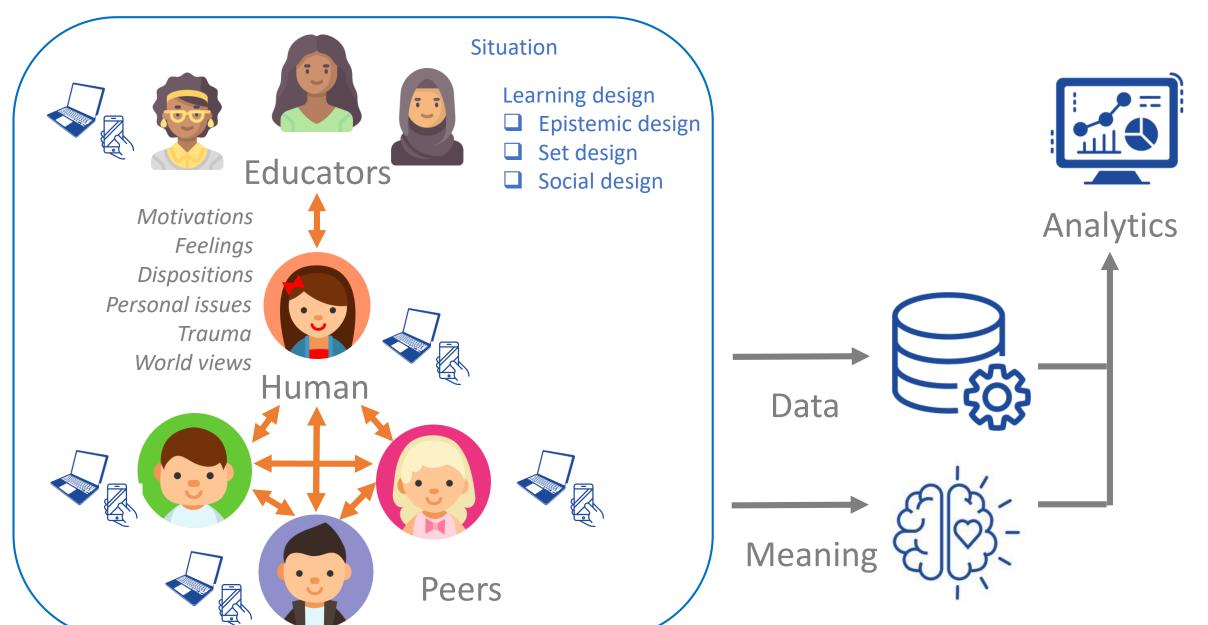




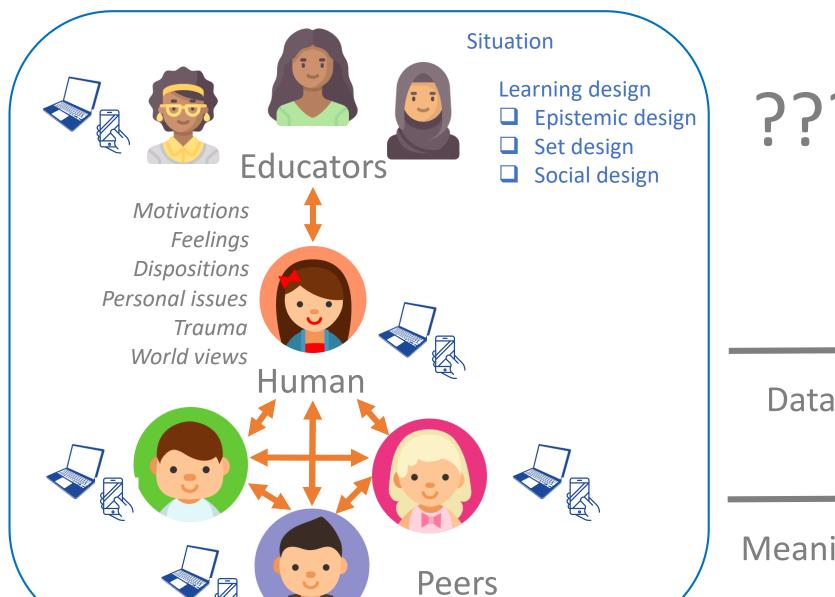


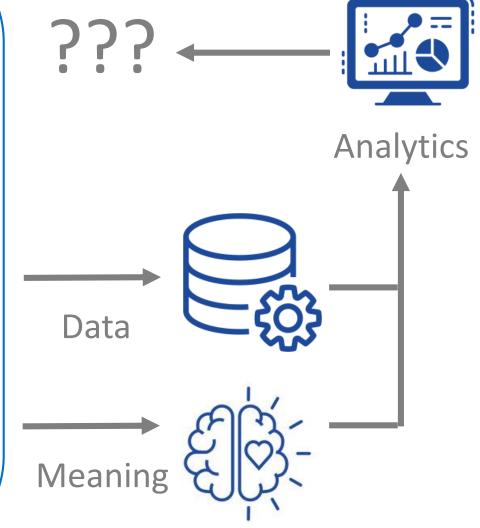


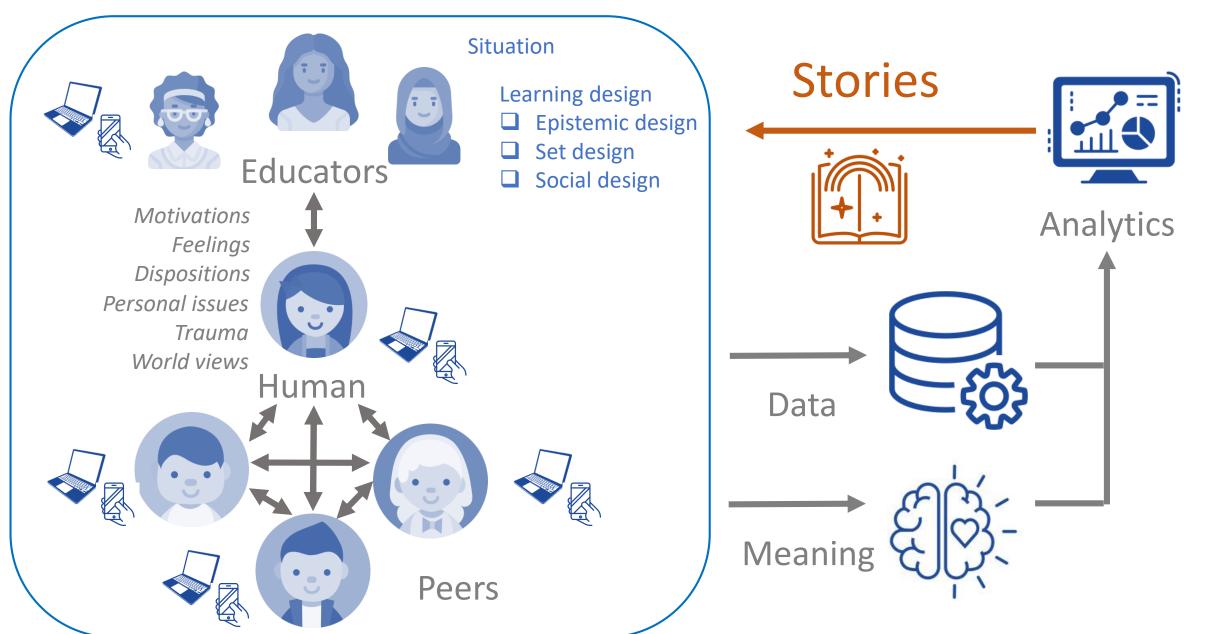


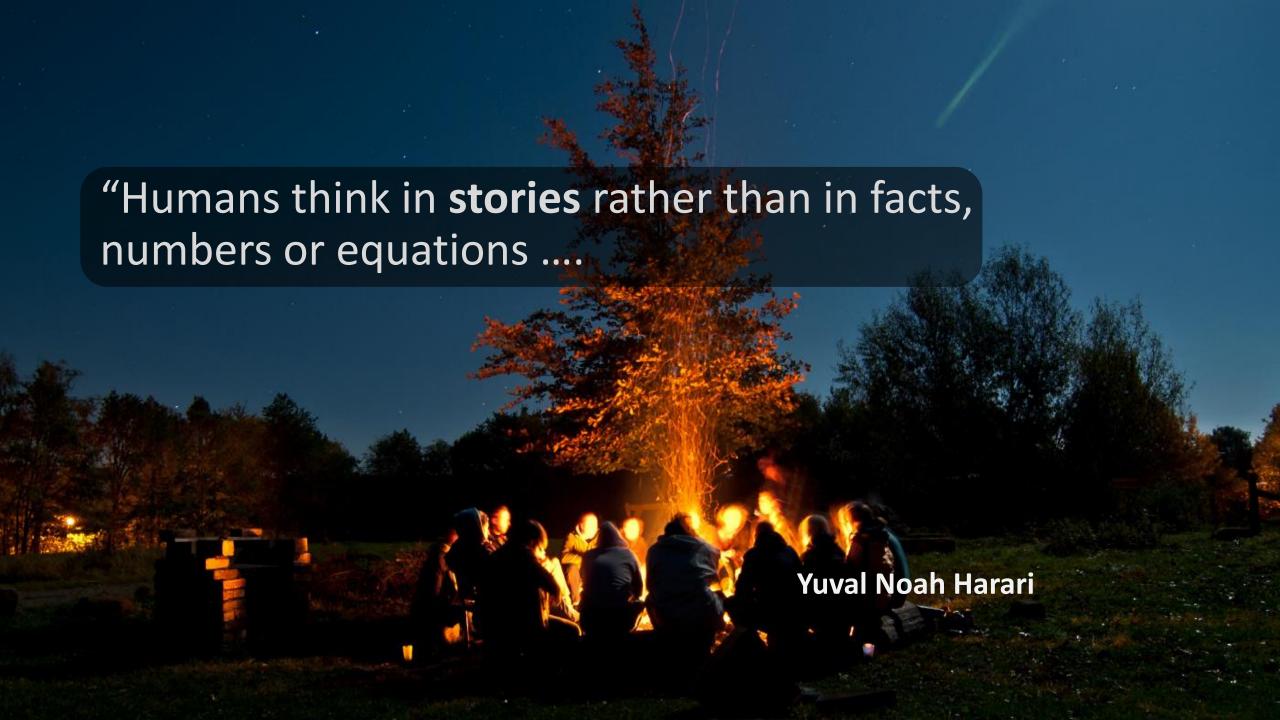












Outline

- ☐ Human centredness
- **□** Multimodality
- ☐ Storytelling

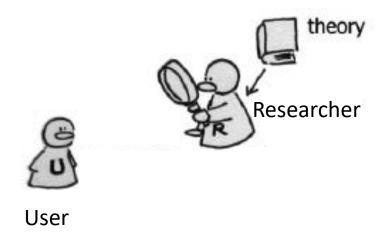
What is human-centredness?

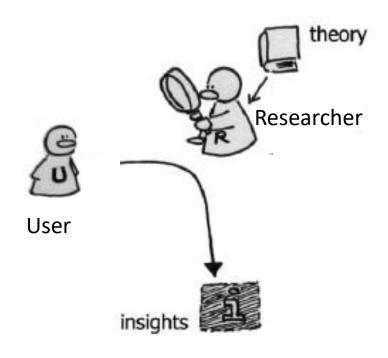
From a learning analytics perspective...

Human centredness has been identified in other fields as a

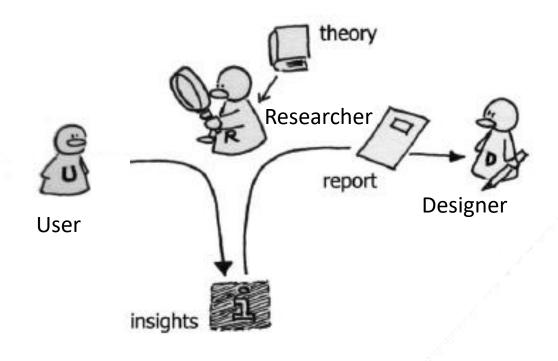
characteristic of systems that have been carefully designed by:

- identifying the critical stakeholders,
- their **relationships**, and
- the **contexts** in which those systems will function





Sanders, E. B. N., & Stappers, P. J. (2008). Co-creation and the new landscapes of design. Co-design, 4(1), 5-18.



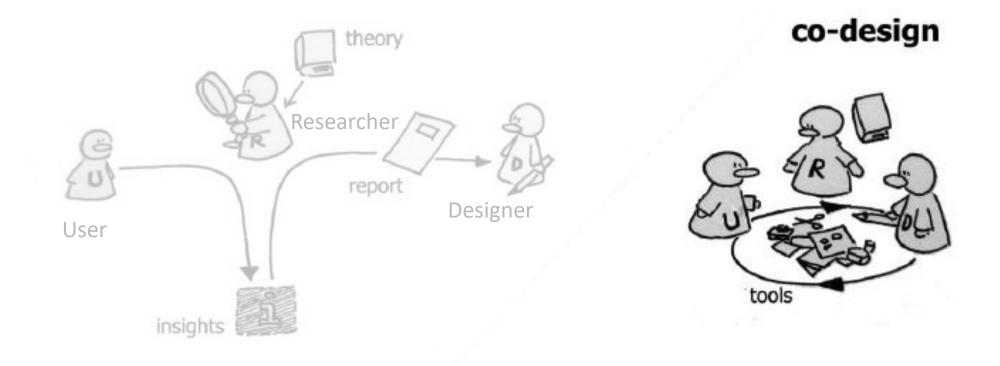
Sanders, E. B. N., & Stappers, P. J. (2008). Co-creation and the new landscapes of design. Co-design, 4(1), 5-18.

An emancipatory perspective for Learning Analytics?

"Learners are not to be seen as passive beneficiaries of a superior control entity. With respect to software adaptations, if Learning Analytics has to play a role, it should be limited to one of awareness and recommendation."

Tchounikine, Pierre. (2019). Learners' agency and CSCL technologies: towards an emancipatory perspective. *International Journal of Computer-Supported Collaborative Learning*, 14(2).

Co-creation (co-design)



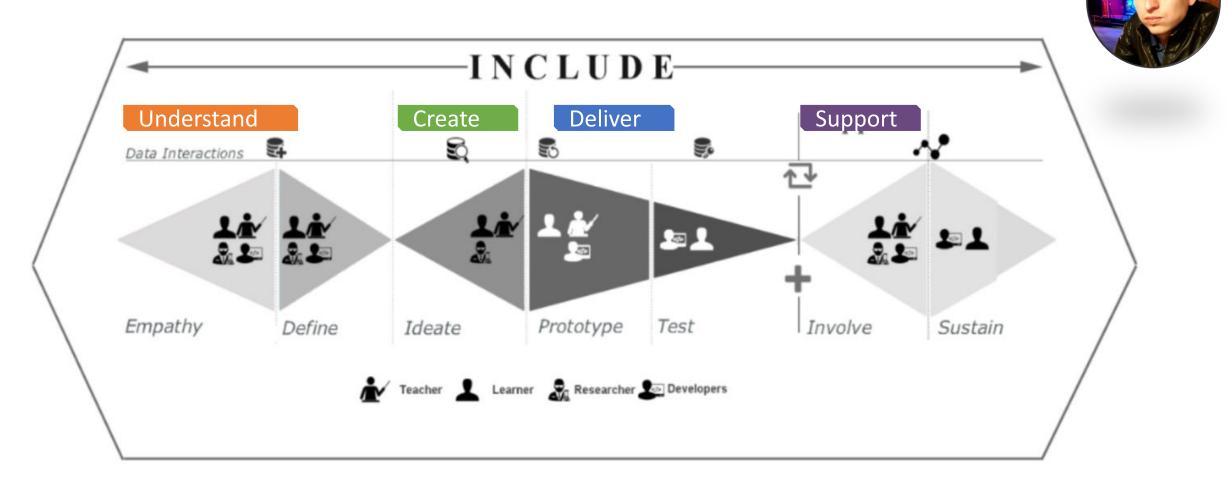
Sanders, E. B. N., & Stappers, P. J. (2008). Co-creation and the new landscapes of design. Co-design, 4(1), 5-18.

"Human-centred design is concerned less with assuring that artifacts work as intended (by their producers, designers, or other cultural authorities) than with enabling many individual or cultural conceptions to unfold into uninterrupted interfaces with technology."

Klaus Krippendorff

Krippendorff, K. 2004, Intrinsic motivation and human-centred design, Theoretic Issues in Ergonomics Science, Vol. 5, No. 1, pp 43-72.

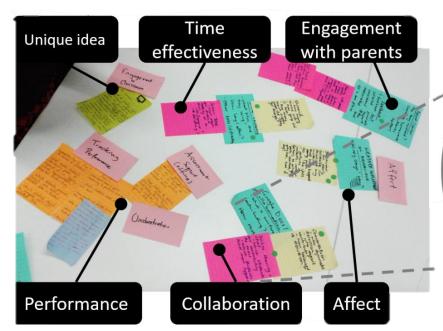
Design Thinking in Learning Analytics



Prieto-Alvarez, C. G., Martinez-Maldonado, R., and Anderson, T. (2018). Co-designing learning analytics tools with learners. In J. M. Lodge, J. C. Horvath, and L. Corrin (Eds.), Learning Analytics in the Classroom (93-110).

Understand

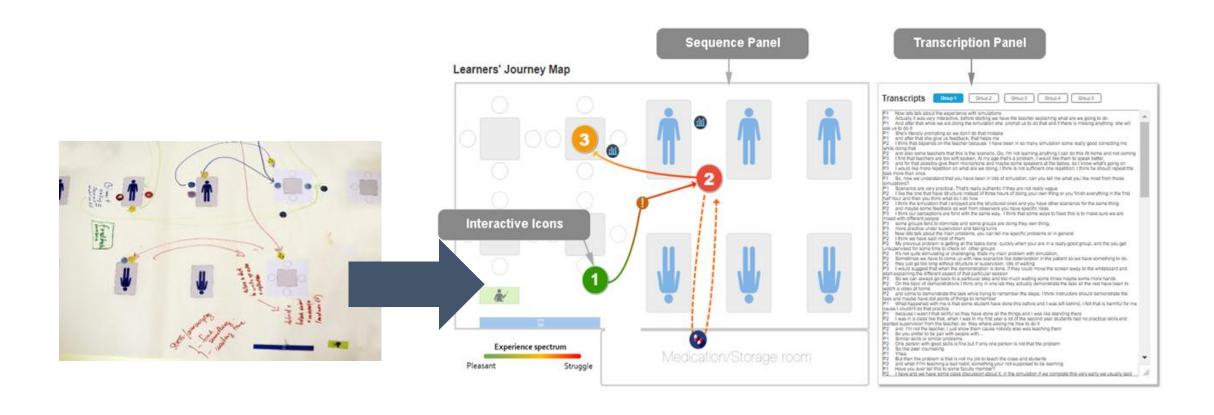
"What aspects of the classroom or the learning activity happening in the classroom you would like to make more visible?"





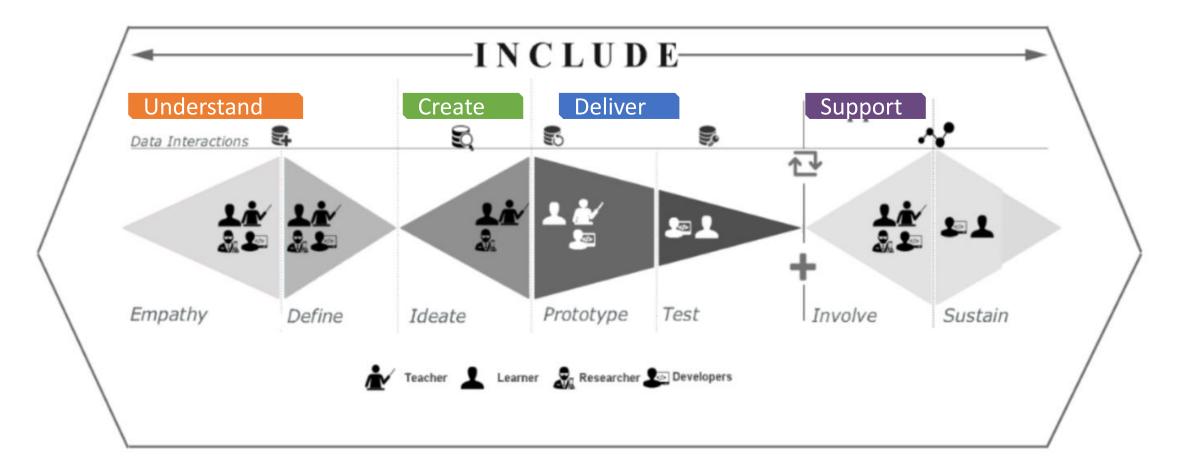
Understand

Learner-data journey mapping



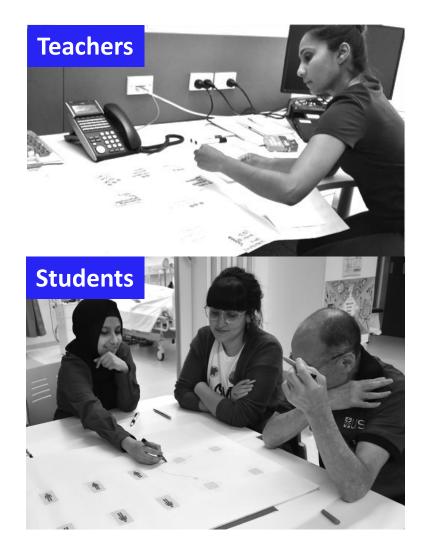
Prieto-Alvarez, C. G., Martinez-Maldonado, R., & Shum, S. B. (2018, December). Mapping learner-data journeys: evolution of a visual co-design tool. In *Proc. OzCHI'18*.

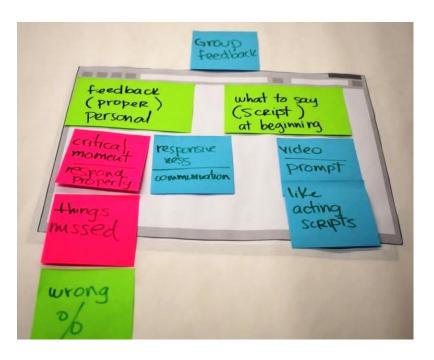
Design Thinking in Learning Analytics



Prieto-Alvarez, C. G., Martinez-Maldonado, R., and Anderson, T. (2018). Co-designing learning analytics tools with learners. In J. M. Lodge, J. C. Horvath, and L. Corrin (Eds.), Learning Analytics in the Classroom (93-110).

Co-design techniques to elicit student and educator perspectives (e.g. "Teacher Superpowers", and "Learning/Data Journey mapping" *)



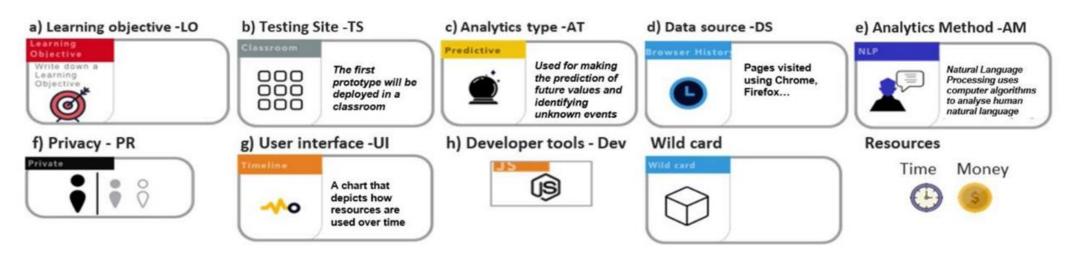


^{*} Prieto-Alvarez, C. G., Anderson, T., Martinez-Maldonado, R., and Buckingham Shum, S. (2018). Mapping Learner/Data Journeys: Evolution of a Visual Co-Design Tool. *Proc. Australian Conference on Human-Computer Interaction*, Melbourne, (ACM, NY), pp.205-214. https://doi.org/10.1145/3292147.3292168

Create

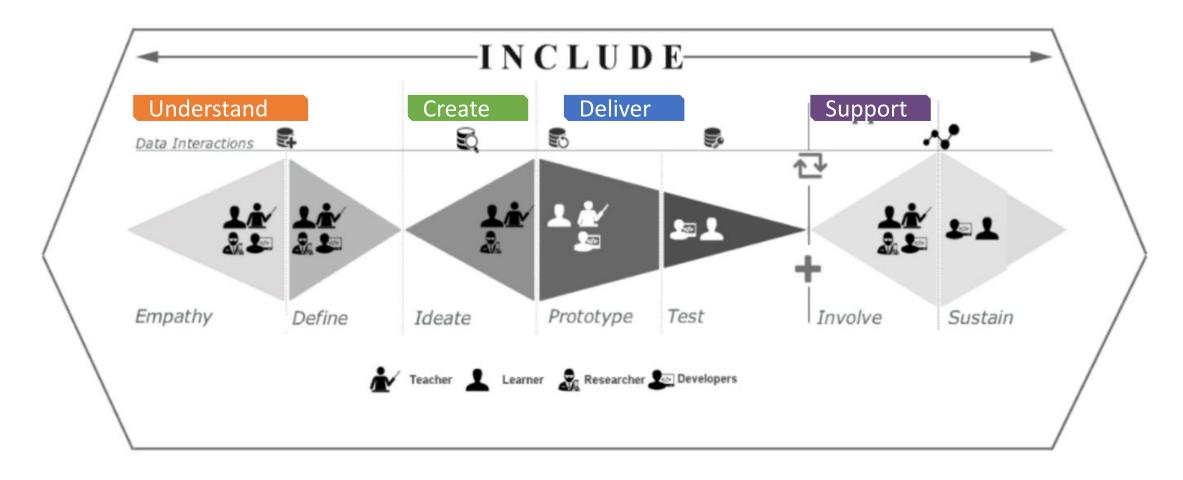
LA-Deck: LA design cards





Alvarez, C. P., Martinez-Maldonado, R., & Shum, S. B. (2020, March). LA-DECK: A card-based learning analytics co-design tool. In *Proc. LAK'20*

Design Thinking (link) in Learning Analytics



Prieto-Alvarez, C. G., Martinez-Maldonado, R., and Anderson, T. (2018). Co-designing learning analytics tools with learners. In J. M. Lodge, J. C. Horvath, and L. Corrin (Eds.), Learning Analytics in the Classroom (93-110).

Deliver

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JOURNAL OF LEARNING ANALYTICS



Volume 6(2), 27-52. http://dx.doi.org/10.18608/jla.2019.62.3

Co-Designing a Real-Time Classroom Orchestration Tool to Support Teacher-Al Complementarity

Kenneth Holstein, ¹ Bruce M. McLaren, ² and Vincent Aleven³



"If I know that I lectured and after I lectured, now my misconceptions decrease significantly. Then I know, okay we're on the right track."

- Teacher 3

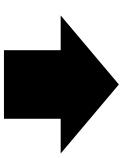
"Yeah, all I care about, it's all about the kids. I don't need rewards and stickers and points for what I am doing. Like, that's my job."



- ... Are you kidding me? This is his heart beating and sweating. Oh my god. IButl part of me thinks. 'It would be nice to know how nervous they are, because I was a very nervous math student back in the day."
- Teacher 9
- "...Could I, like use Ithe dronel to zap my students if they're misbehaving? Nothing that would actually hurt them. a light jolt"



- "I love this because at a snapshot I see who's struggling. I don't need to go from student to student to student.
- Teacher 3
- *There are always students who are shy and just don't raise their hands. [Some] raise their hands when they really don't need help."



Deliver

Testing the same prototypes with

Students

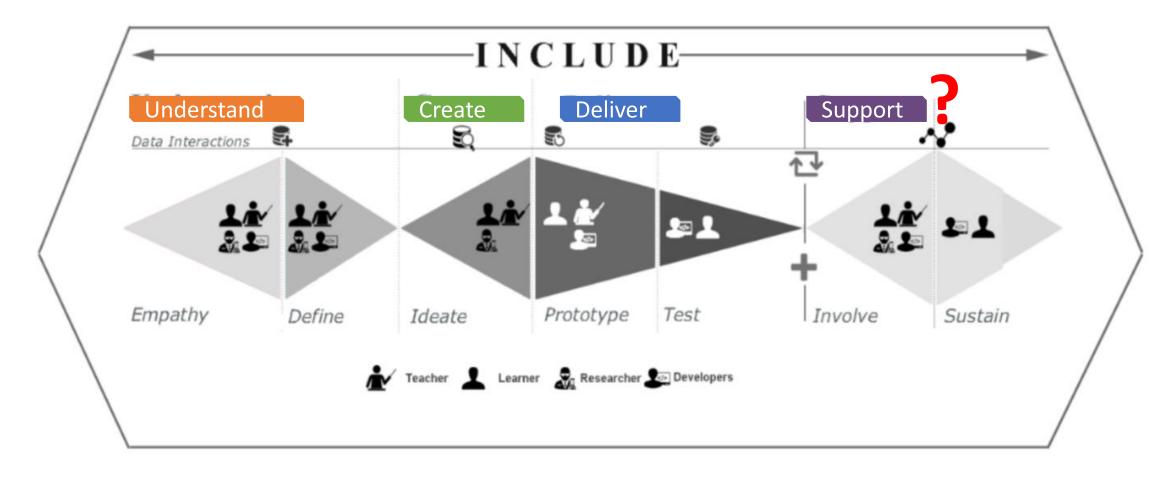


Teachers



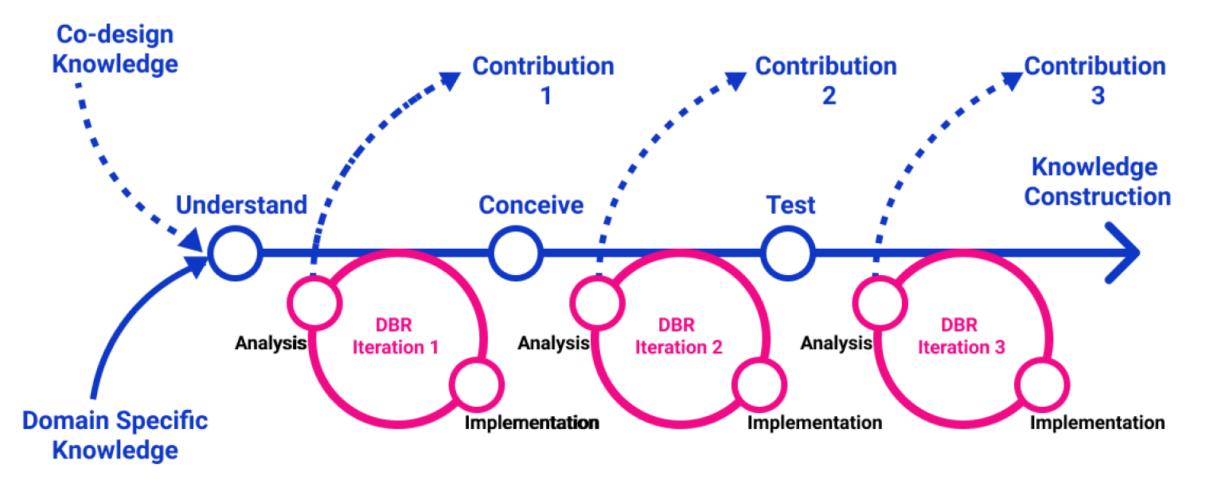


Design Thinking in Learning Analytics



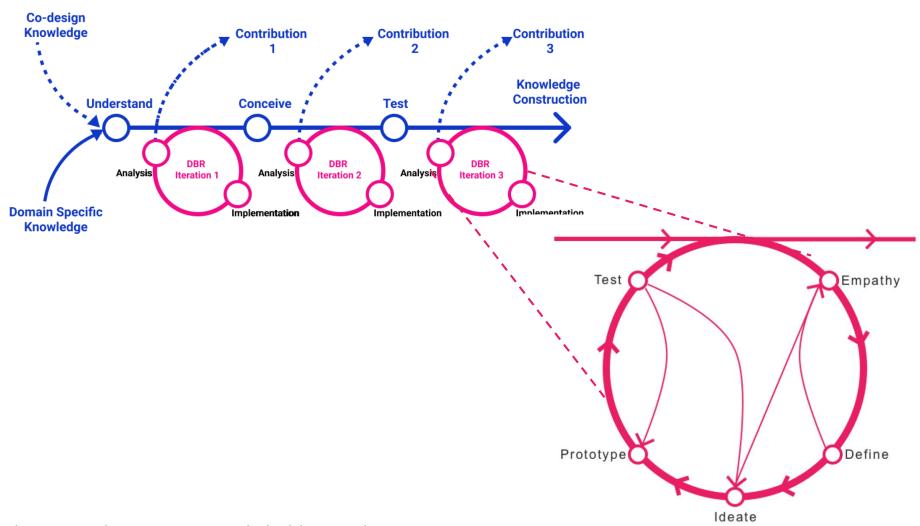
Prieto-Alvarez, C. G., Martinez-Maldonado, R., and Anderson, T. (2018). Co-designing learning analytics tools with learners. In J. M. Lodge, J. C. Horvath, and L. Corrin (Eds.), Learning Analytics in the Classroom (93-110).

What about DBR?



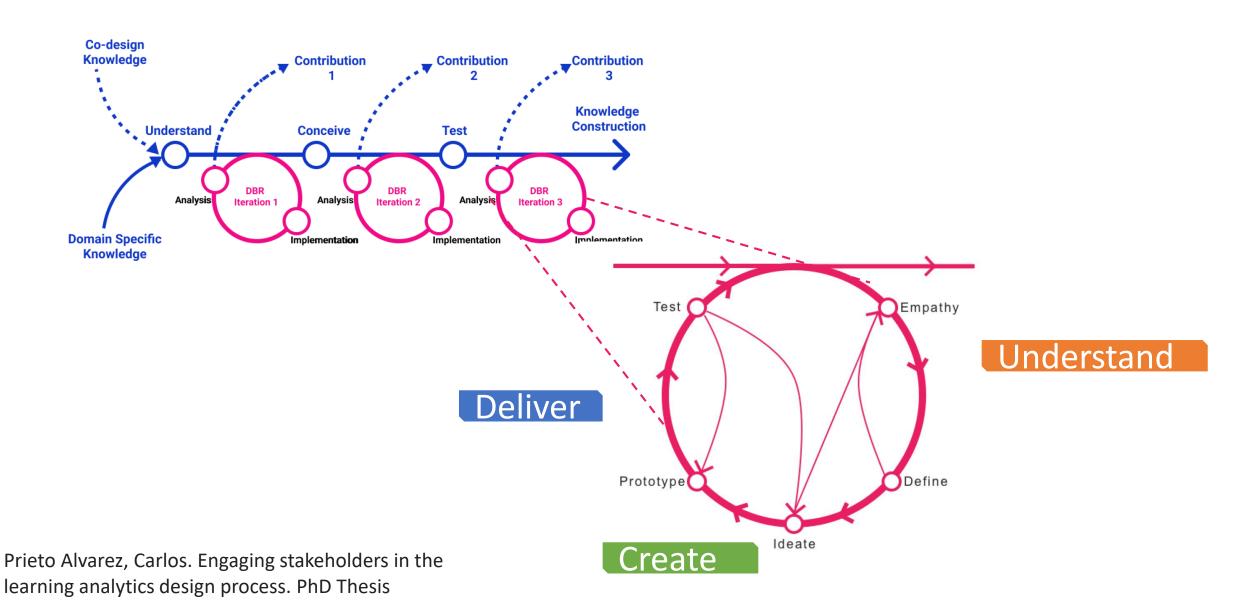
DBR + Design thinking



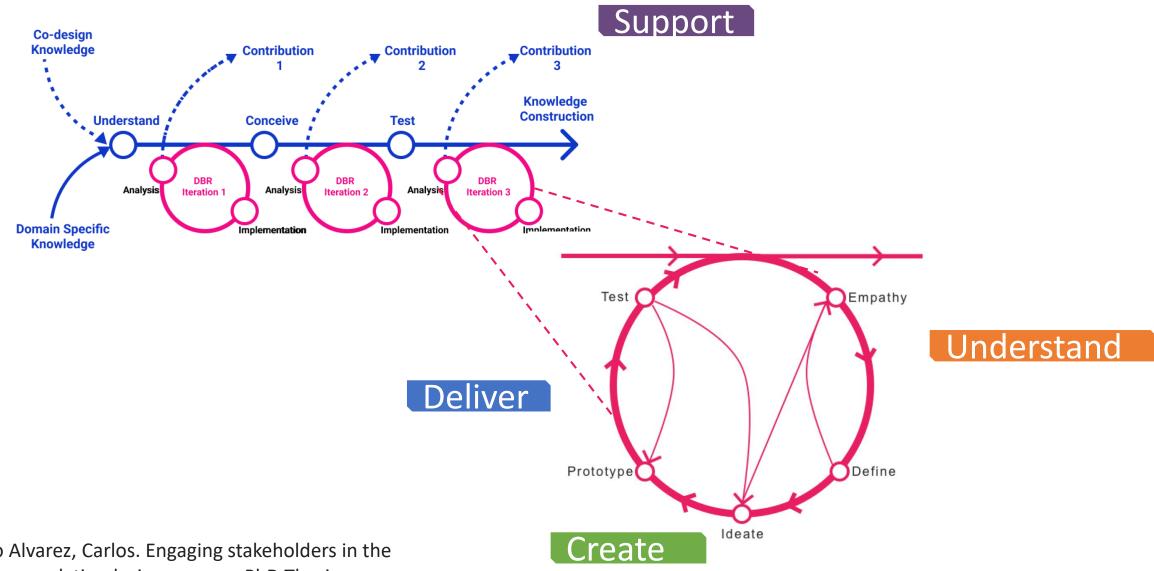


Prieto Alvarez, Carlos. Engaging stakeholders in the learning analytics design process. PhD Thesis

DBR + Design thinking



DBR + Design thinking



Prieto Alvarez, Carlos. Engaging stakeholders in the learning analytics design process. PhD Thesis

Outline

- ☐ Human centredness
- Multimodality
- ☐ Storytelling

The real Learning Analytics situation





Multimodality in Learning Analytics

"Multimodal data is used in recognition of the plurality of ways that students may demonstrate or communicate knowledge, interests and intent"

Worsley et al. 2021

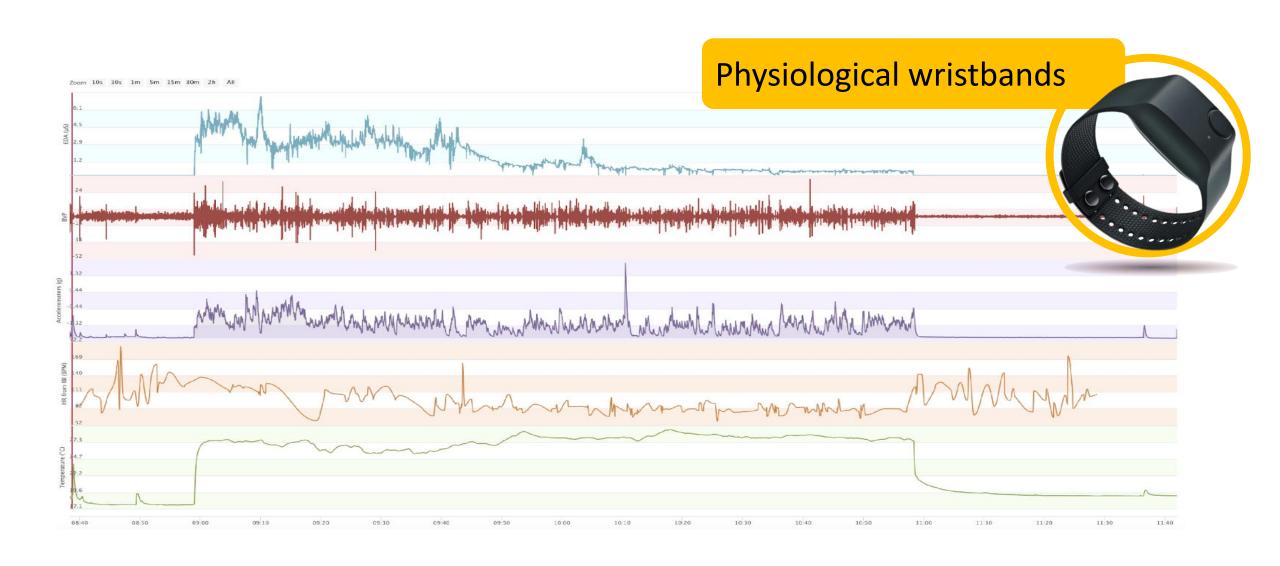


Multimodal learning analytics vision



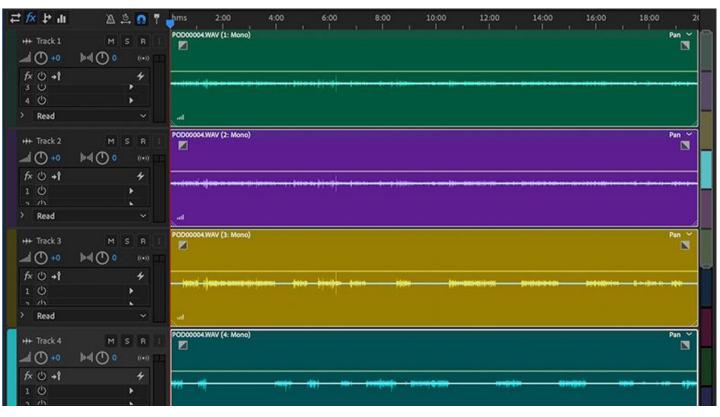
"The goal of MMLA is to support learning experiences that may be collaborative, hands-on, and face-to-face, **de-emphasizing the computer screen** as the primary form or object of interaction"

Resulting multi-channel data streams

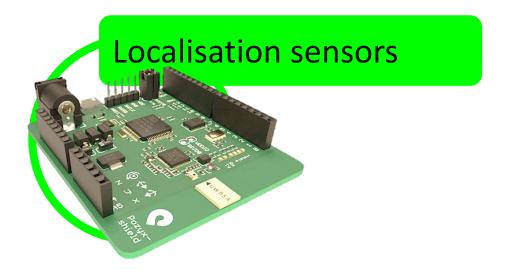


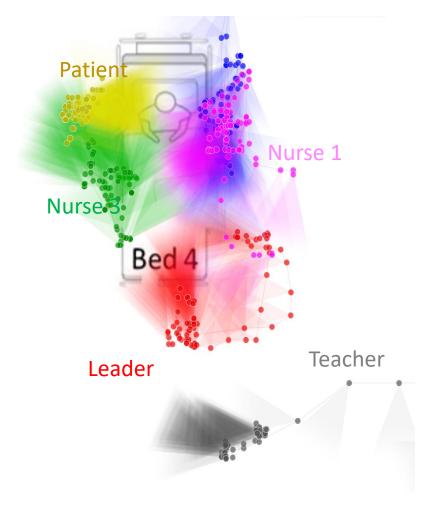
Audio data





Positioning data





2 minute of activity

Action log data

System logs

Observation logs

Patient simulator logs



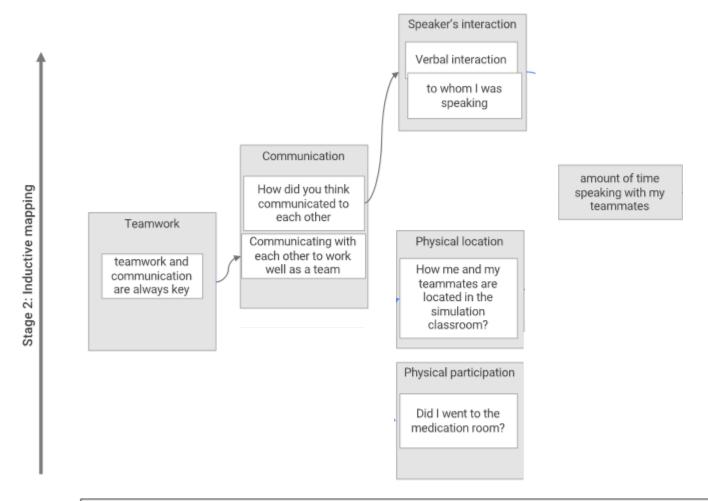
Timestamp	User	Action
00:02	Leader	Connect oximeter
01:00	Nurse 2	Prepare medication
01:20	Nurse 3	Check vital signs
01:55	Leader	Provide medication
02:01	Nurse 2	Connect IV device
02:04	Observer	Log actions
02:32	Nurse 2	Check vital signs

Stage 3: theory-driven deductive mapping

Inductive step with teachers

Categories

Constructs



Sub-constructs

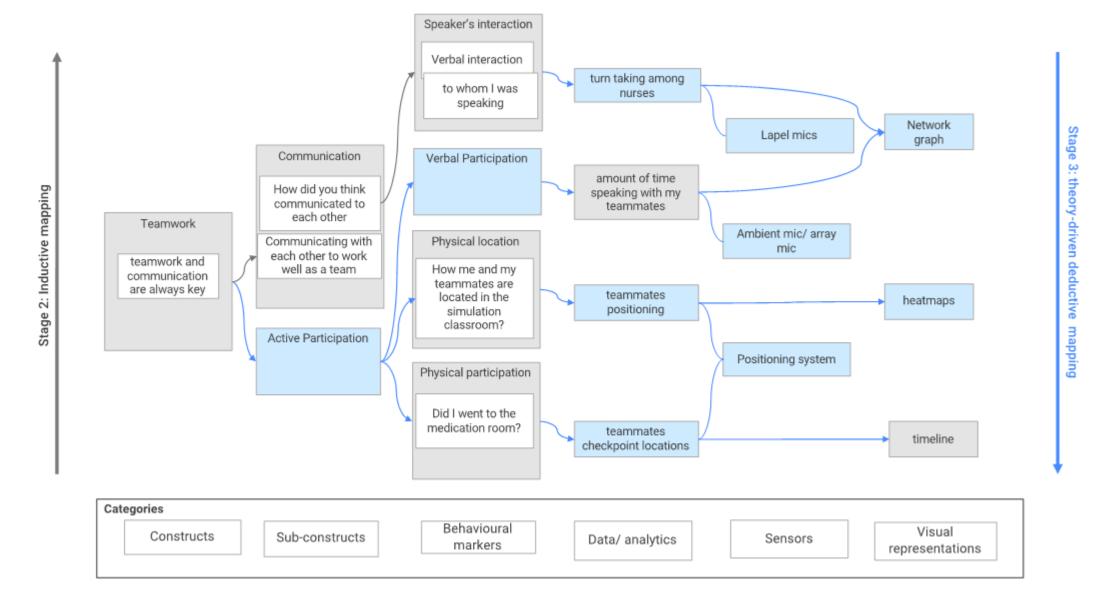
Behavioural

markers

Data/ analytics

Sensors Visual representations

Deductive step: filling the holes from theory

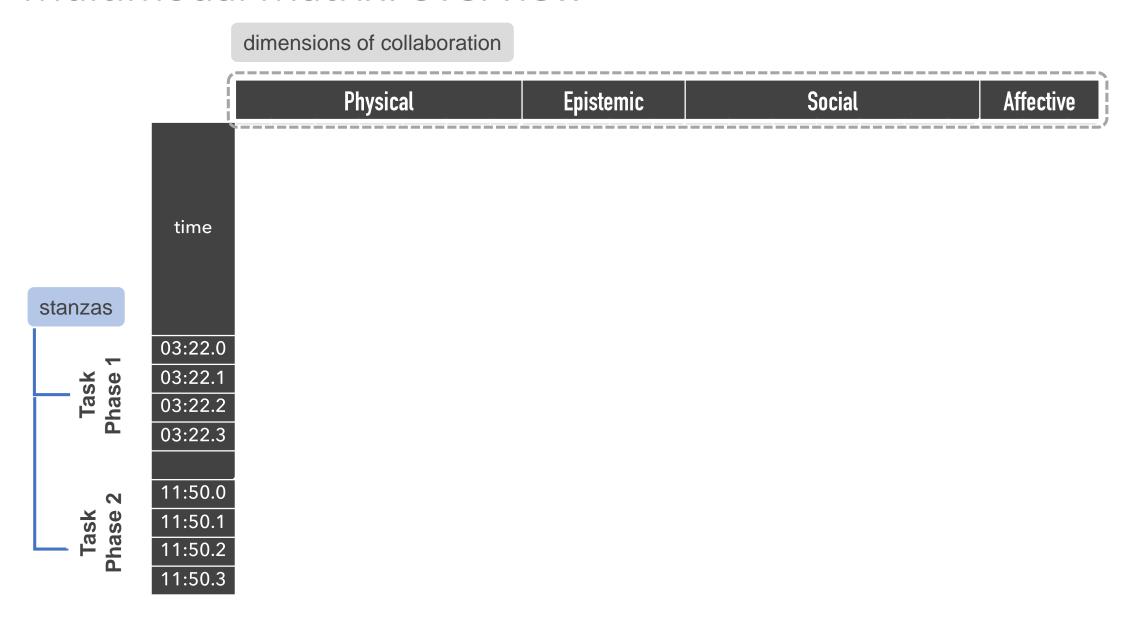


dimensions of collaboration

Bhysical Enistamic Social	\ \Affactiva \ \
r ilysical Lpisicillic Jucial	Allective

dimensions of collaboration

	Physical	Epistemic	Social	Affective
				/
time				
ume				
03:22.0				
03:22.1				
03:22.3				
11:50.0				
11:50.1 11:50.2				
11:50.3				



dimensions of collaboration

	Physical								Epist	emic	Social						Affective			
time	RN1.patient_bed	RN1.next_to_patient	RN1.around_patient	RN1.bed_head	RN1.trolley_area	RN1.pysical_intensity		RN1.ventilations	RN2.ventilations	RN1.compressions	RN1.speaking	RN2.speaking	patient.speaking	RN1.listening	RN2.listening	patient.listening		RN1.EDA_peak	RN2.EDA_peak	

multimodal observations

Task Task Phase 1 Phase 1

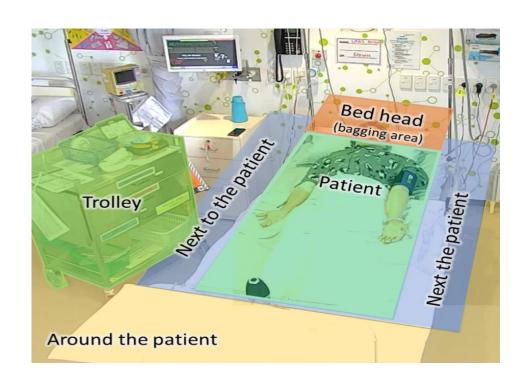
03:22.0 03:22.1 03:22.2 03:22.3 11:50.0 11:50.1 11:50.2

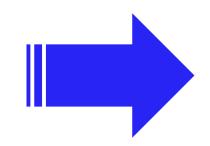
11:50.3

dimensions of collaboration

				P	hysic	al			Epistemic			Social					Af	ectiv	e			
	time	.patient_bed	RN1.next_to_patient	RN1.around_patient	RN1.bed_head	RN1.trolley_area	.pysical_intensity		RN1.ventilations	RN2.ventilations	RN1.compressions		.speaking	RN2.speaking	oatient.speaking	RN1.listening	RN2.listening	patient.listening	RN1.EDA_peak	RN2.EDA_peak		multimodal observations
stanzas		RN1	RN1	RN1	RN1	RN1	RN1		RN1	RNZ	RN1		RN1	RNZ	pati	RN1	RNZ	pati	RN1	RN2		
— [03:22.0	0	1	0	0	0	L		0	0	0		0	0	1	1	1	0	0	0		
	03:22.1	0	1	0	0	0	L		0	0	0		0	0	1	1	1	0	1	0		
Task Phase	03:22.2	0	1	0	0	0	L		0	0	0		0	0	1	1	1	0	0	0		
	03:22.3	0	1	0	0	0	L		0	0	0		0	0	1	1	1	0	0	0		
8	11:50.0	1	0	0	0	0	Н	K	0	1	1		0	0	0	0	0	0	0	0		segment
Task	11:50.1	1	0	0	0	0	Н		0	1	1		0	0	0	0	0	0	1	0		
Task Phase	11:50.2	1	0	0	0	0	Н		0	1	1		0	0	0	0	0	0	1	0		
<u> </u>	11:50.3	1	0	0	0	0	Н		0	1	1		0	0	0	0	0	0	0	0		

Mapping from positional Codes to digital codes in the Multimodal Matrix





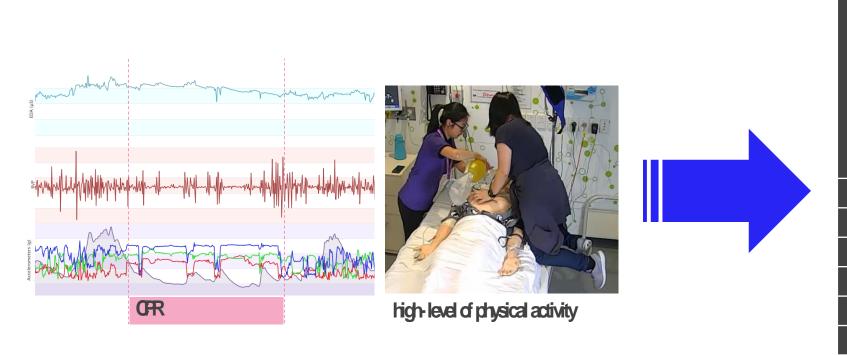
				1,010	,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,
time	RN1.patient_bed	RN1.next_to_patient	RN1.around_patient	RN1.bed_head	RN1.trolley_area
03:22.0	0	1	0	0	0
03:22.1	0	1	0	0	0
03:22.2	0	1	0	0	0
03:22.3	0	1	0	0	0
11:50.0	1	0	0	0	0
11:50.1	1	0	0	0	0
11:50.2	1	0	0	0	0
11:50.3	1	0	0	0	0

Physical

Machine coding

Classifying raw accelerometer data

→ low/medium/high physical activity

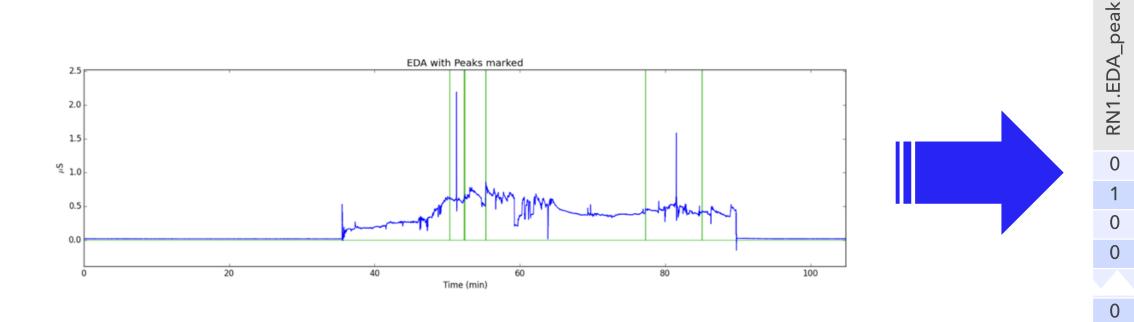


Machine coding

Physical RN1.pysical_intensity time 03:22.0 03:22.1 03:22.2 03:22.3 11:50.0 Н 11:50.1 11:50.2 Н

11:50.3

Thresholding raw EDA traces to focus on what's interesting: EDA peaks + low physical activity



Machine coding

Affective

RN2.EDA_peak

0

0

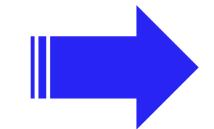
Observers (e.g. researchers or students) use a tablet-based annotation tool to log key actions

Actions Log Actions Start Session Stop Session Anginine student2 Start CPR Stop CPR Ventilation Writing charts Ventilation at: Apr 25, 2019 2:04:18 AM Notes: 3 cycles Writing charts at: Apr 25, 2019 2:01:53 AM Notes: ISBAR chart Delete Anginine at: Apr 25, 2019 1:57:43 AM Notes: one dosis Delete

Human coding

Working towards automatically identifying who is speaking and listening/sec.





Machine coding

		S	ocial			
o o o o RN1.speaking	RN2.speaking	patient.speaking	RN1.listening	RN2.listening	o o o o patient.listening	
0	0	1	1 1 1	1	0	
0	0	1	1	1	0	
0	0 0	1	1	1	0	
0	0	1	1	1	0	
0	0	0	0	0	0	
0	0	0	0	0	0	
0 0 0	0	0	0 0 0	0	0 0 0	
0	0	0	0	0	0	

Outline

- ☐ Human centredness
- Multimodality
- ☐ Storytelling

What is data storytelling?

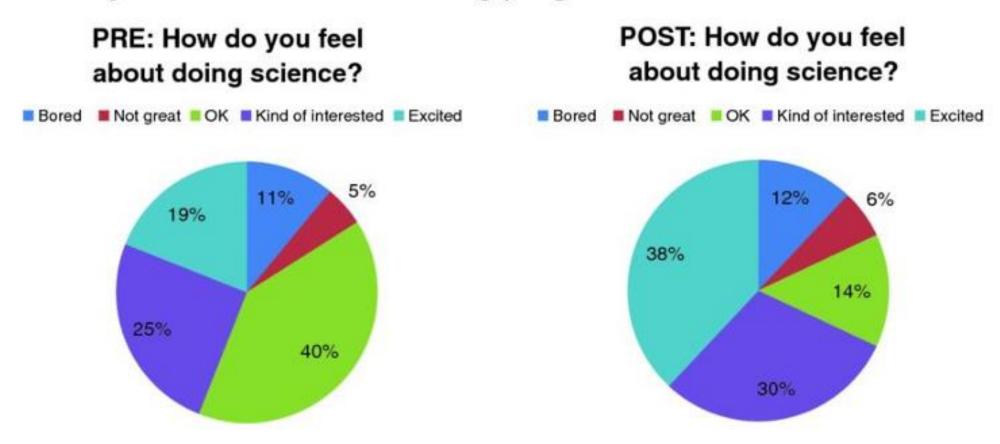
An information compression technique for communicating insights to an audience through the combination of data, visuals, and narrative



Brent Dykes. 2015. Data storytelling: What it is and how it can be used to effectively communicate analysis results. Applied Marketing Analytics. 1, 4, 299-313.

We want to move from this.....

Survey results: summer learning program on science

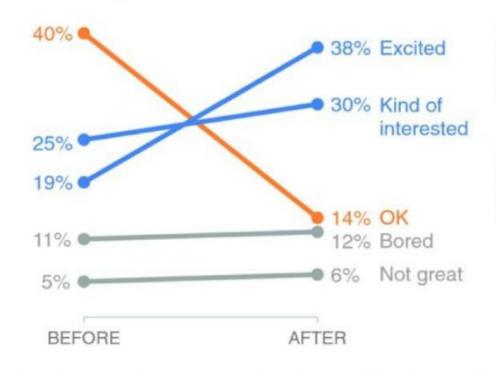


Example from: Knaflic, C. N. (2015). Storytelling with data: A data visualization guide for business professionals. John Wiley & Sons.

...towards data stories like this...

Pilot program was a success

How do you feel about science?



BEFORE program, the majority of children felt just *OK* about science.

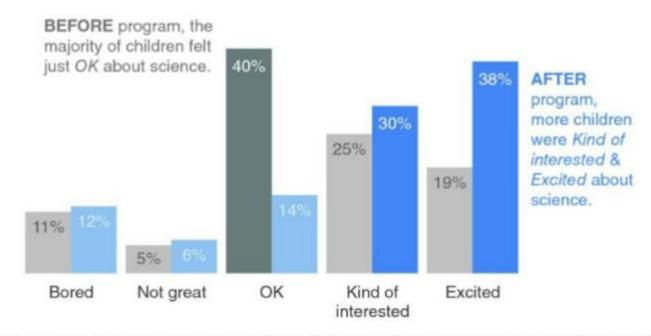
AFTER program, more children were Kind of interested & Excited about science.

Based on survey of 100 students conducted before and after pilot program (100% response rate on both surveys).

... or this one...

Pilot program was a success

How do you feel about science?



Based on survey of 100 students conducted before and after pilot program (100% response rate on both surveys).

... or identifying the indicators that may be most useful to make decisions...

Pilot program was a success

After the pilot program,

68%

of kids expressed interest towards science, compared to 44% going into the program.

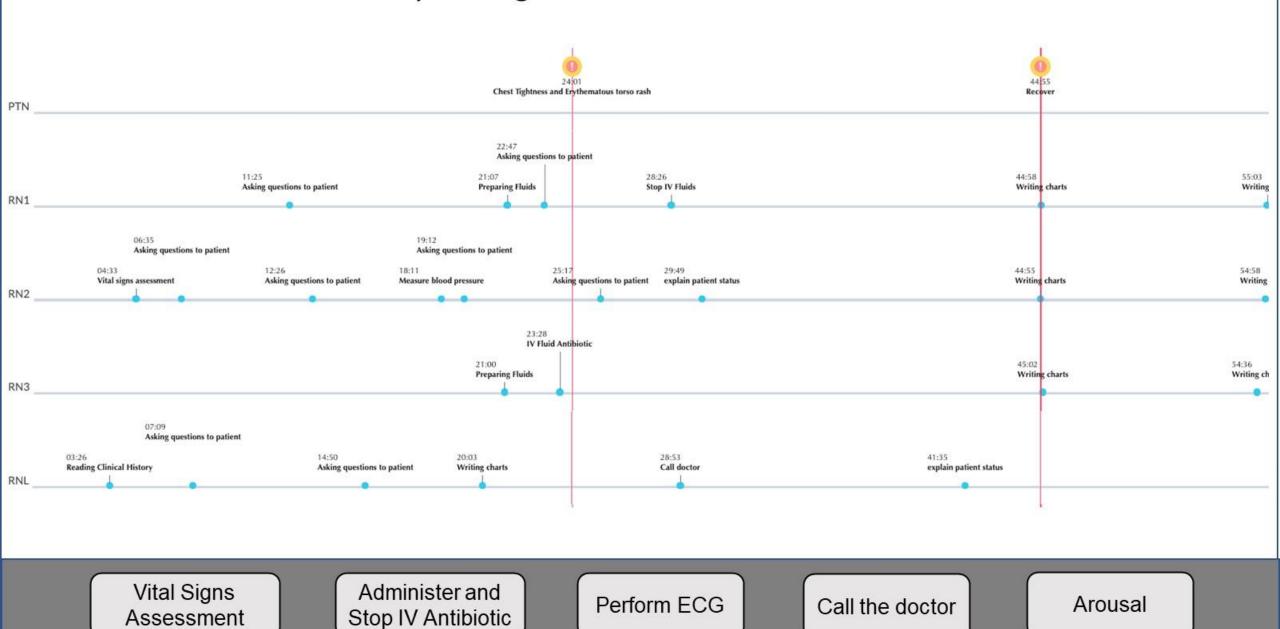
Based on survey of 100 students conducted before and after pilot program (100% response rate on both surveys).

Layered data storytelling prototype

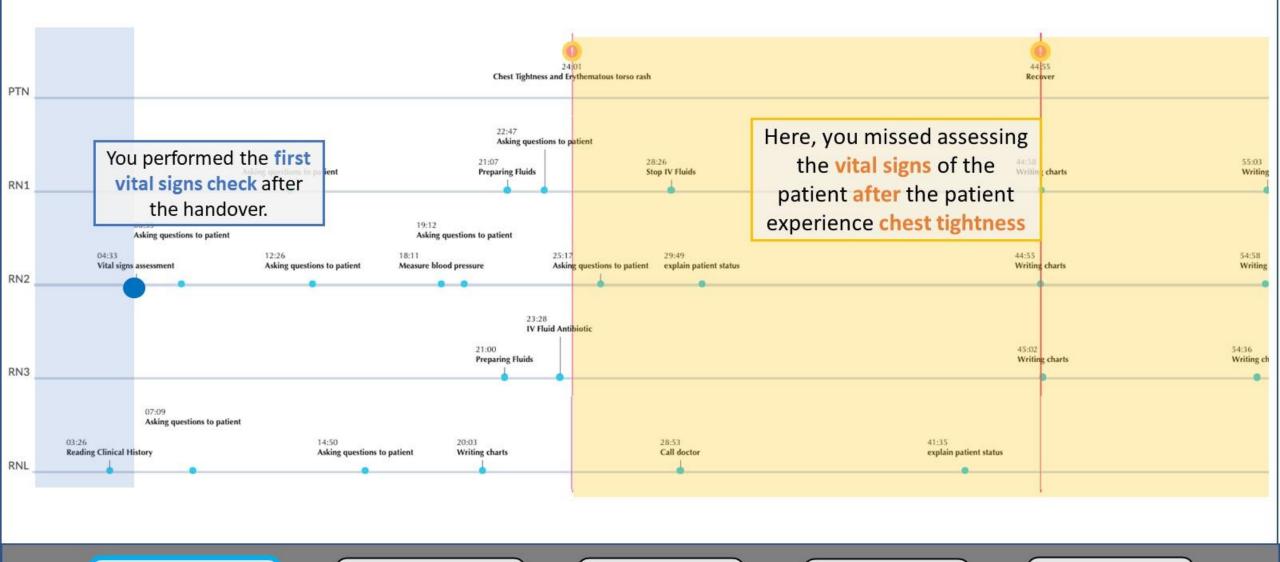
Put to the test with nursing students



Select the critical incidents by clicking on the buttons below



After the patient complains of chest tightness it is very important to assess his/her vital signs



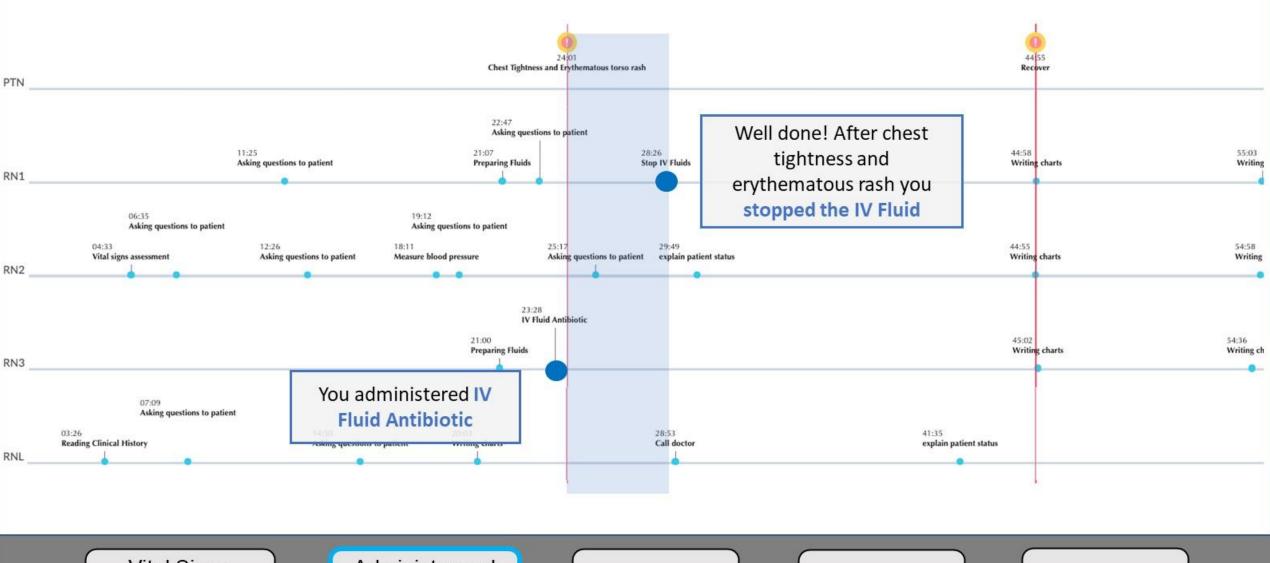
Vital Signs Assessment Administer and Stop IV Antibiotic

Perform ECG

Call the doctor

Arousal

RN1 stopped the medication less than 5 minutes after patient's adverse reaction



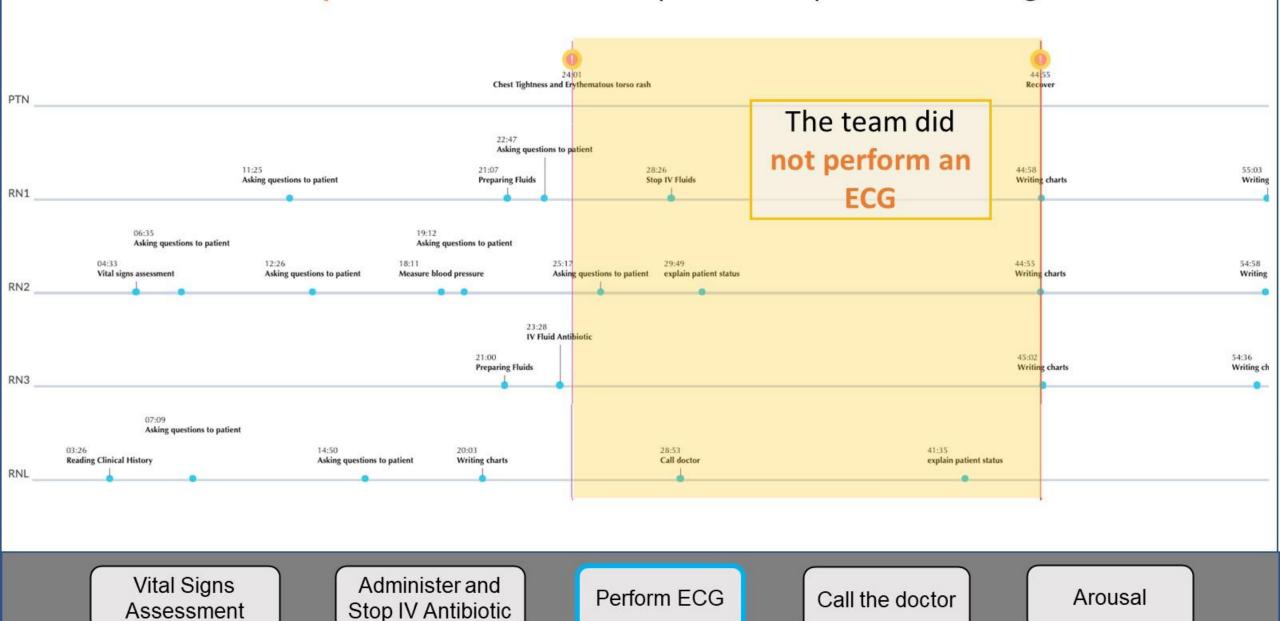
Vital Signs Assessment Administer and Stop IV Antibiotic

Perform ECG

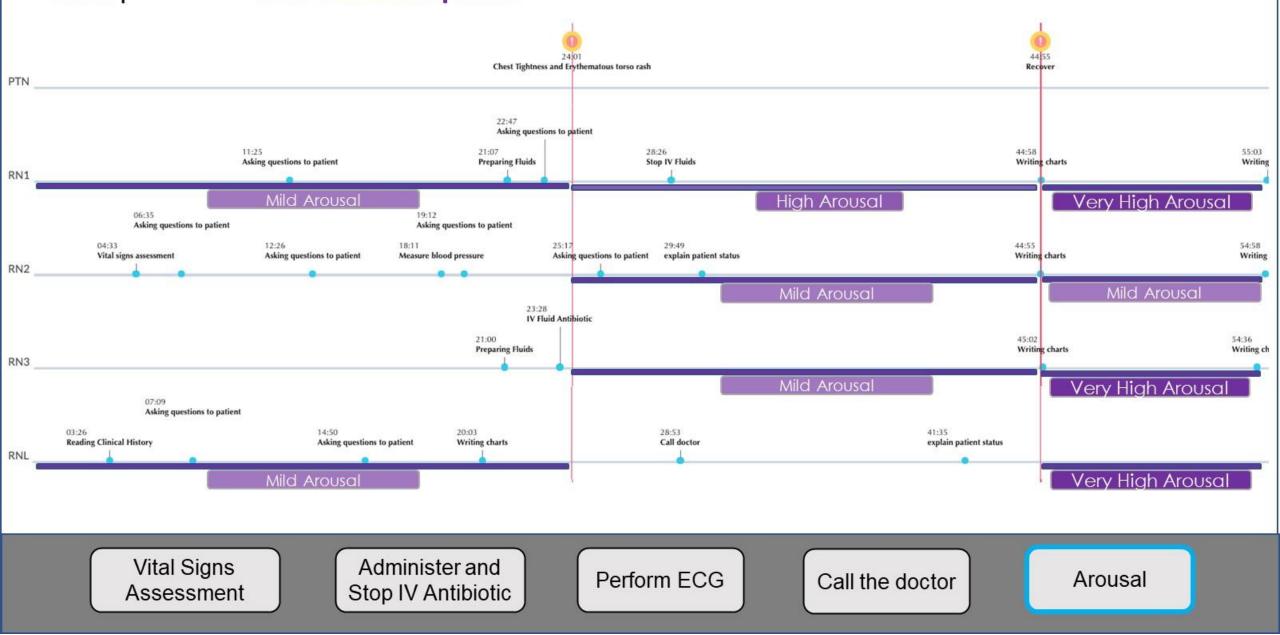
Call the doctor

Arousal

It is recommended to perform an ECG after the patient complains of chest tightness



RN1, RN3 and RNL presented several arousal peaks throughout the simulation RN2 presented a few arousal peaks



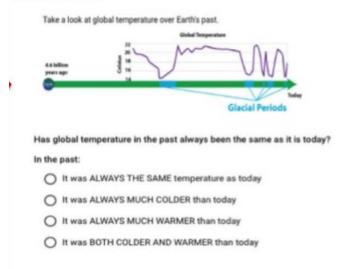
A report as a data storytelling channel

Narrative + Visuals

Hello Ms. Kerrington,

I noticed that most of your students completed Step 1.4 which has the maximum number of attempts feature, so I've analyzed the log data from that step.

Learning goal: Step 1.4 targets MS-ESS3-5 and the stability and change CCC. Students need to be able to recognize the scale of the timeline and interpret the graphical data.



Here is my analysis (by class period) of the log data of your students' responses. Note: You can find the students associated with the workgroup ID by clicking on the "Manage Students" link in the Teacher Tools.

Answered correctly on the first attempt

- Period 1 46% (6/13)
- Period 2 79% (11/14)
- Period 3 75% (9/12)

How many multiple attempts were needed (by workgroup)?

 2 attempts were needed by those who didn't answer correctly on the first attempt.

What was the most common incorrect answer on the first attempt?

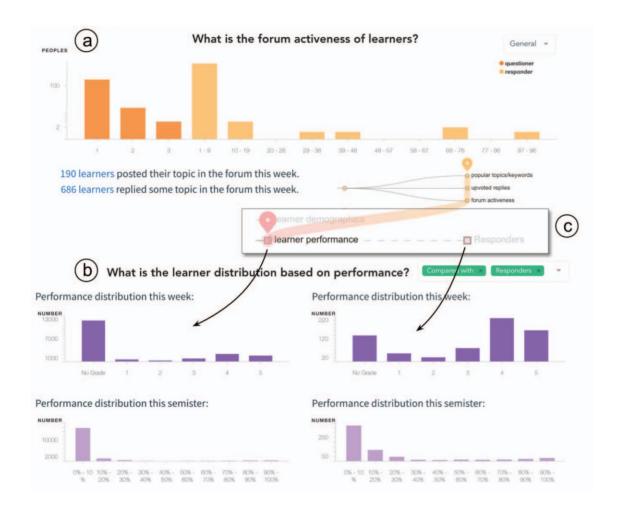
- Most students chose "It was ALWAYS MUCH COLDER than today"
- Students who followed a different pattern, by period:
 - Period 1
 - 397583, 397597 It was ALWAYS MUCH WARMER than today, then answered correctly.
 - Period 2
 - 397640 -It was ALWAYS THE SAME temperature as today, then chose the correct answer
 - Period 3 (all followed primary pattern)

Researcher Insight: This suggests that students' prior knowledge that current global temperatures are the highest they have been in recent history is overriding their analysis of the actual data presented in the timeline of Earth's history.

Wiley, K. J., Dimitriadis, Y., Bradford, A., & Linn, M. C. (2020, March). From theory to action: Developing and evaluating learning analytics for learning design. In *Proceedings of the tenth international conference on learning analytics & knowledge* (pp. 569-578).

A navigation slideshow as a data storytelling channel

Visuals
+
Narrative
+
Navigation



Chen, Q., Li, Z., Pong, T. C., & Qu, H. (2019, April). Designing Narrative Slideshows for Learning Analytics. In 2019 IEEE Pacific Visualization Symposium (Pacific Vis.) (pp. 237-246). IEEE.

Learning Analytics for 'end-users':

from Human-centred Design to Multimodal Data Storytelling

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Take away message 1: Human-centred design

A human-centred approach is critical in the design of any educational technology...



Human

Take away message 1: Human-centred design

A human-centred approach is critical in the design of any educational technology...



Human

...but even more critical to navigate the disruptions and potential risks triggered by the rapid development of AI, Data Analytics and surveillance tools

Take away message 2: Multimodality

A call for embracing multimodality...

...not with the purpose of adding sensors to the learning spaces but to embrace the complexity of learning and human interaction



Take away message 3: Data Storytelling "Humans think in stories rather than in facts, numbers or equations Yuval Noah Harari

Take away message 3: Data Storytelling

"Humans think in **stories** rather than in facts, numbers or equations

.... and the simpler the story, the better"

Yuval Noah Harari

Thanks!