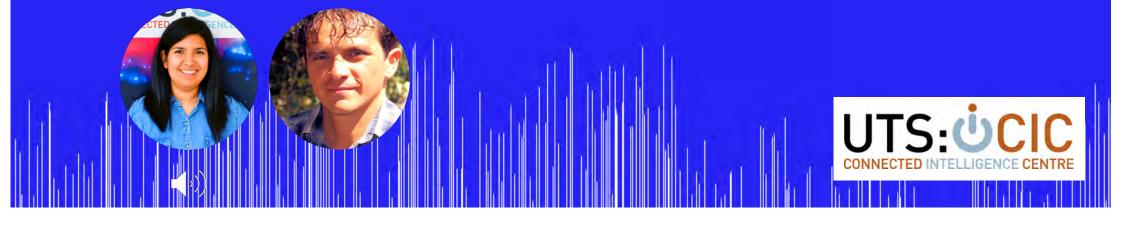


University of Sydney, School of Computer Science • 28<sup>th</sup> May 2019 Invited Talk, Computer-Human Adapted Interaction Research Group

# Towards Collaboration Translucence: Giving Meaning to Multimodal Group Data

Vanessa Echeverria, Roberto Martinez-Maldonado & Simon Buckingham Shum Connected Intelligence Centre, University of Technology Sydney, AUS @vanechev @RobertoResearch @sbuckshum



CHI 2019, May 4-9, 2019, Glasgow, Scotland, UK

CHI 2019 Paper

# Extended version of CHI 2019 talk

#### Towards Collaboration Translucence: Giving Meaning to Multimodal Group Data

#### Vanessa Echeverria<sup>1,2</sup>, Roberto Martinez-Maldonado<sup>1</sup> and Simon Buckingham Shum<sup>1</sup> <sup>1</sup>University of Technology Sydney, Ultimo, NSW, 2007, Australia. <sup>2</sup>Escuela Superior Politécnica del Litoral, ESPOL, Campus Gustavo Galindo Km 30.5 Vía Perimetral, Guayaquil, Ecuador. Vanessa.EcheverriaBarzola, Roberto.Martinez-Maldonado, Simon.BuckinghamShum}@uts.edu.au

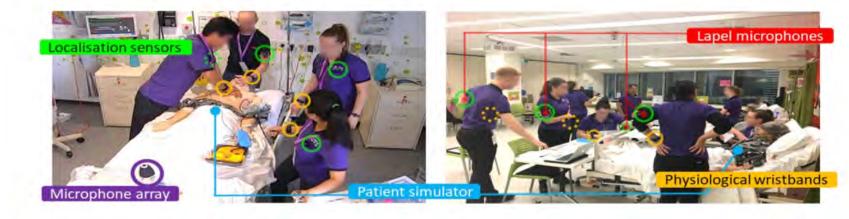


Figure 1: Multimodal analytics in healthcare scenarios: in a simulation-controlled room (left) and in the classroom (right).

Vanessa Echeverria, Roberto Martinez-Maldonado, and Simon Buckingham Shum. 2019. Towards Collaboration Translucence: Giving Meaning to Multimodal Group Data. In Proceedings of ACM CHI conference (CHI'19). ACM, New York, NY, USA, Paper 39, 16 pages. <u>https://doi.org/10.1145/3290605.3300269</u>

# OUR CONTEXT: HIGH PERFORMANCE TEAMWORK in NURSING

Simulation wards are used in universities and hospitals

Up to 6 teams in action at once at UTS

1 instructor

"Analytically cloaked": no data amenable to computational analysis, to inform debriefs immediately after exercises



Co-design methods were used to gain insights from students and educators about running and performing simulation exercises

What were educators' fantasy "superpowers"?

At what points would students value data-driven feedback?



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. *Australian Conference on Human-Computer Interaction, OZCHI 2018*, Melbourne, pp.205-214. <u>https://doi.org/10.1145/3292147.3292168</u>

Student: "Tutors have a big class so they can't supervise everyone at the same time. It's hard for them to give feedback to every student"

Student: "Time is against us and we don't have enough time for debriefing. Limited time needs a clear direction. Otherwise you're just going to get many different feelings coming back at you"

Student: "I think that a better feedback would give me perspective, because when you are in the simulation you can't see where you're positioned, you can't see how you're talking"

Instructor: "omniscience" to see what's going in every team, in parallel, in detail

> Instructor: "Capturing students' body positioning, or movement, would help them visualize the whole activity, to complete the picture."

### **INSPIRATION: "SOCIAL TRANSLUCENCE"**

Tom Erickson et al. (CHI'99) on the challenge of providing missing social cues in online platforms

#### Translucence *≠* Transparency

In f-f social spaces/places, we use translucence to disclose specific information at an appropriate fidelity, as with frosted glass doors and windows.

Visibility of socially significant information Awareness of others' presence or actions Accountability of people's own visible actions



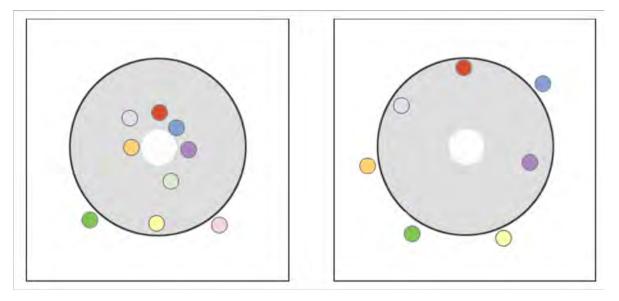
H. Schnädelbach and D. Kirk, *People, Personal Data and the Built Environment* (Springer Series in Adaptive Environments). Springer International Publishing, 2018.

Erickson, T., Smith, D. N., Kellogg, W. A., Laff, M., Richards, J. T., & Bradner, E. (1999). Socially translucent systems: social proxies, persistent conversation, and the design of "babble". In *Proceedings of the SIGCHI conference on Human Factors in Computing Systems* (pp. 72-79). ACM.

### **GROUP CHAT PROXY**

A visualization indicating at a glance an important aspect of the history, or current state, of an online social space

Figure 2. A social proxy for a group chat in the Babble system: (a) an active chat (b) after chat has ceased.

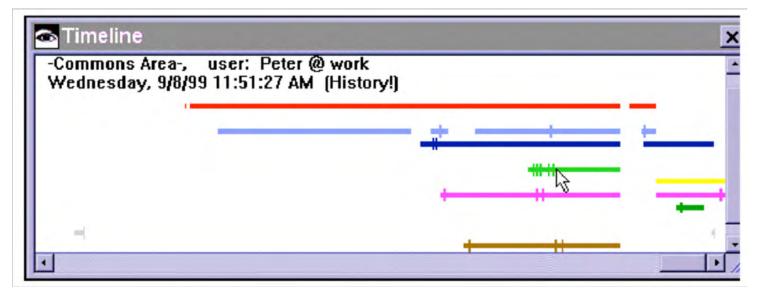


Erickson, Thomas. (2009). 'Social' systems: Designing digital systems that support social intelligence. AI & Society, 23. 147-166. 10.1007/s00146-007-0140-3.



A visualization indicating at a glance an important aspect of the history, or current state, of an online social space

Figure 3. The timeline proxy shows users' presence in the chat room as flat lines and their posts as blips, thus showing activity over time.



Erickson, Thomas. (2009). 'Social' systems: Designing digital systems that support social intelligence. Al & Society, 23. 147-166. 10.1007/s00146-007-0140-3.

### **ONLINE LECTURE PROXY**

A visualization indicating at a glance an important aspect of the history, or current state, of an online social space

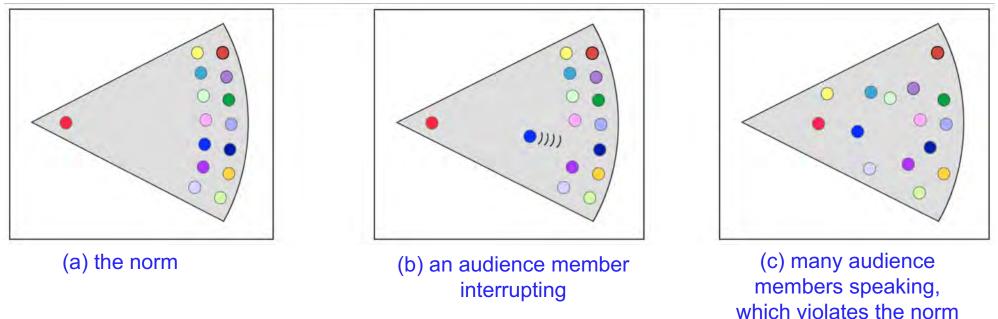


Figure 4. Three instances of the lecture proxy

Erickson, Thomas. (2009). 'Social' systems: Designing digital systems that support social intelligence. AI & Society, 23. 147-166. 10.1007/s00146-007-0140-3.

# TRANSLUCENCE PROXIES FOR COLLOCATED COLLABORATION?

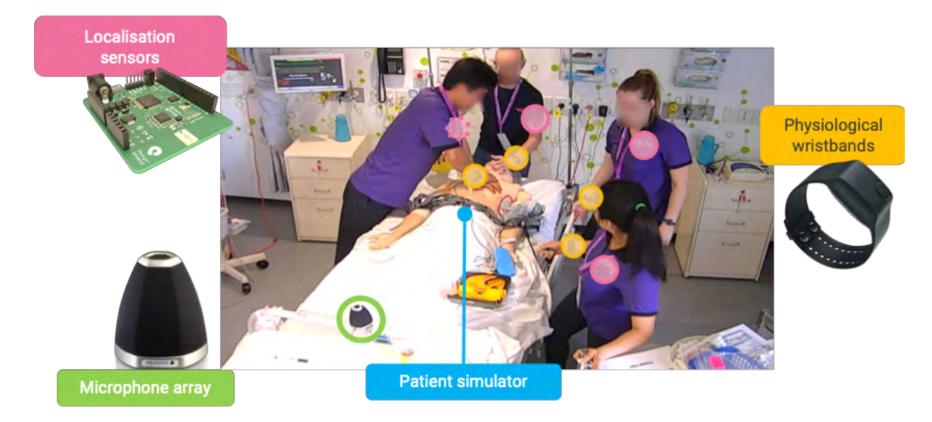
Team simulations can be cognitively and emotionally intense

There's too much going on to see it all, or remember it all

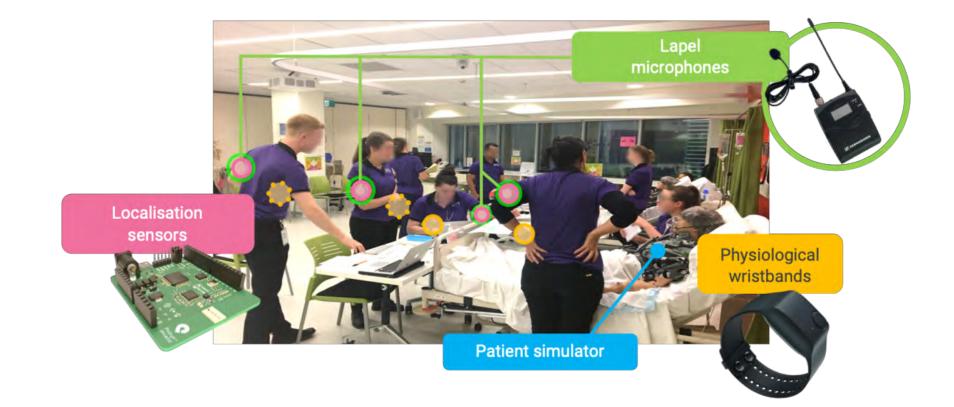
Inspired by Social Translucence proxies, could we devise visual proxies for collocated collaboration?



### THE ANALYTICS CHALLENGE: SENSE AND CAPTURE ACTIVITY

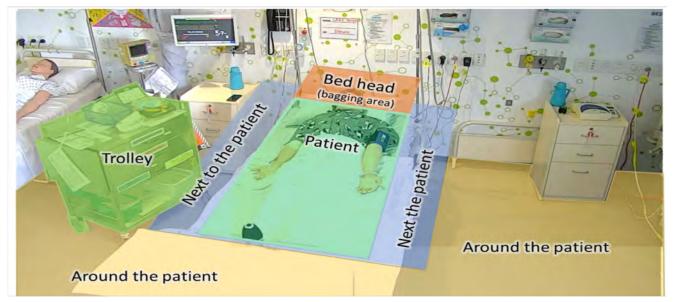


### THE ANALYTICS CHALLENGE: SENSE AND CAPTURE ACTIVITY



## THE ANALYTICS CHALLENGE: MODELLING THE SEMANTICS OF LOCATION

#### Clinical expertise informed the modelling of 5 meaningful zones for positional data



- the patient's bed for cases where nurses are located on top of or very close to the patient
- ii) next to patient for cases where nurses are either side of bed
- iii) around the patient for cases where nurses are 1.5 to 3 metres away
- iv) bed head where nurses commonly stand to clear the airway during CPR
- v) trolley area where nurses access medication or equipment

### **KEY FEATURES OF COLLOCATED COLLABORATION:**

#### ACAD: Activity-Centred Analysis & Design framework

**The SET** — physical and digital space and objects; input devices, screens, software, material tools, furniture

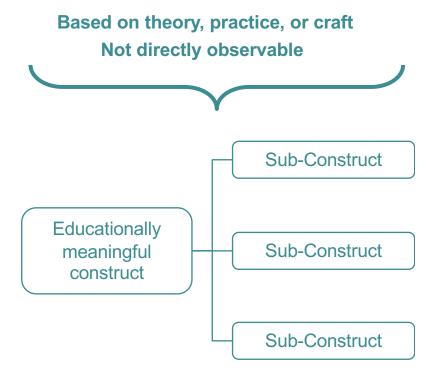
**The EPISTEMIC TASKS** — implicit and explicit knowledge oriented elements that shape the participants' tasks and working methods

The SOCIAL SITUATION — the variety of ways in which people might be grouped together (e.g. dyads, trios); scripted or emerging roles; and divisions of labour

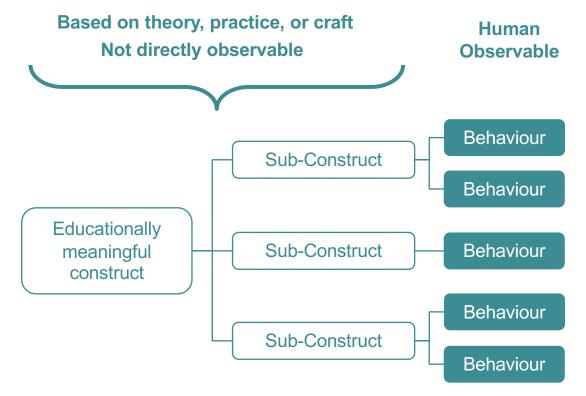
**AFFECTIVE RESPONSES** — and extension to ACAD, building on evidence from healthcare simulation research

Martinez-Maldonado, R., Goodyear, P., Kay, J., Thompson, K., and Carvalho, L. (2016) An Actionable Approach to Understand Group Experience in Complex, Multi-surface Spaces. *SIGCHI Conference: Human Factors in Computing Systems, CHI 2016, 2062-2074.* <u>https://doi.org/10.1145/2858036.2858213</u>

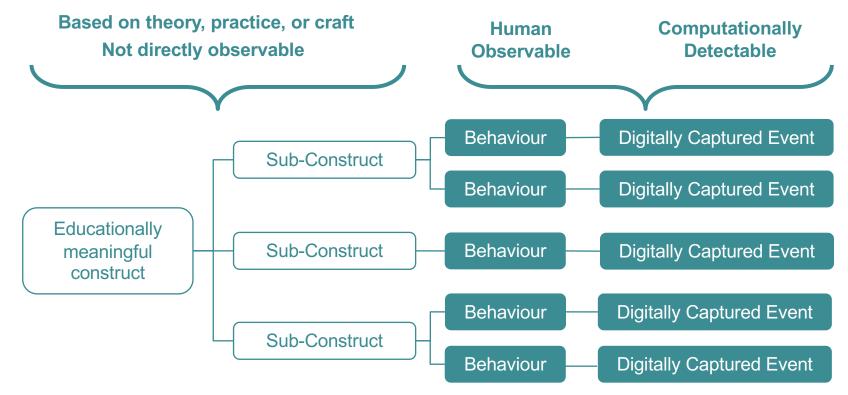
In the fields of Assessment Science and Learning Analytics, the challenge is to forge principled mappings between the qualities that we want to assess, and behavioural evidence

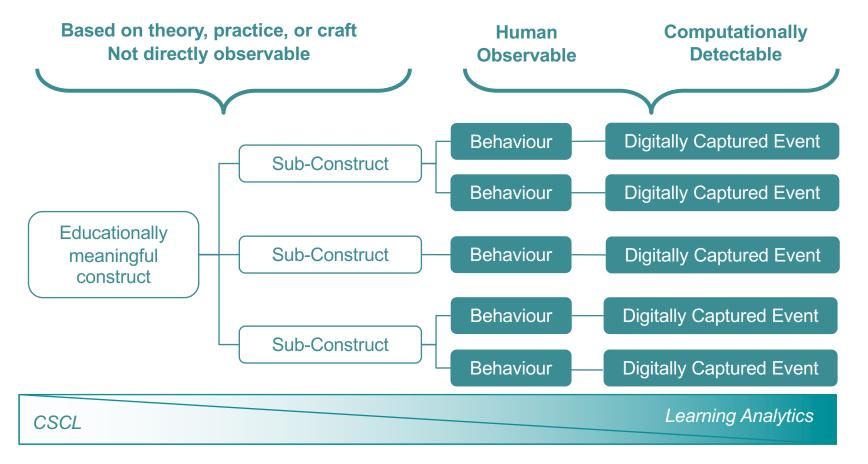


In the fields of Assessment Science and Learning Analytics, the challenge is to forge principled mappings between the qualities that we want to assess, and behavioural evidence



In the fields of Assessment Science and Learning Analytics, the challenge is to forge principled mappings between the qualities that we want to assess, and behavioural evidence, as detected by sensors

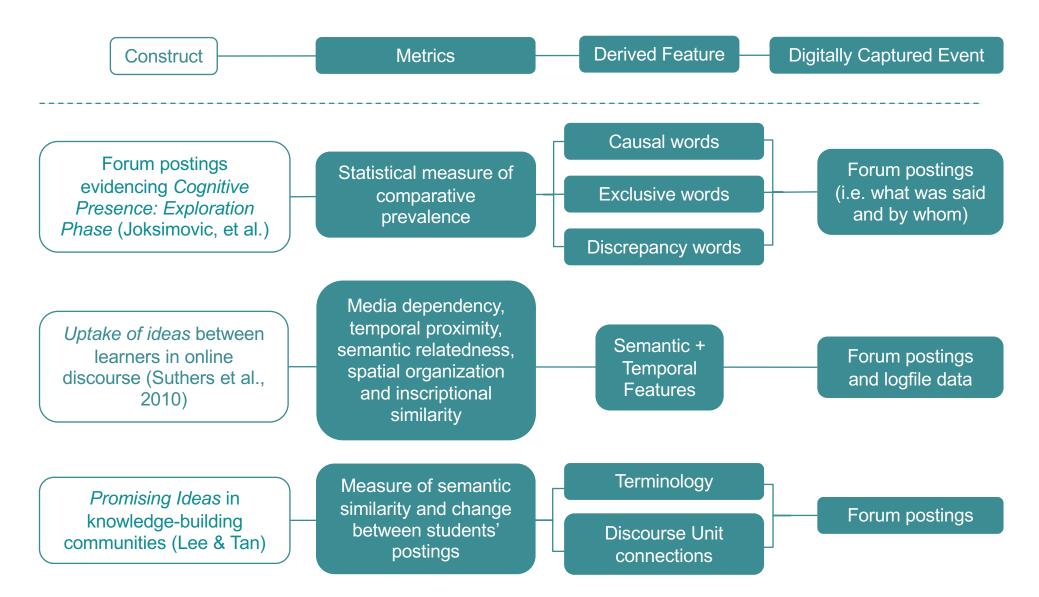




Wise, A., Knight, S. & Buckingham Shum, S. (In Press). Collaborative Learning Analytics. International Handbook of Computer-Supported Collaborative Learning. Springer.

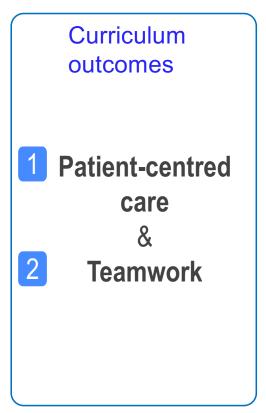
Construct	Metrics	Derived Feature Digitally Captured Event
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Wise, A., Knight, S. & Buckingham Shum, S. (In Press). Collaborative Learning Analytics. International Handbook of Computer-Supported Collaborative Learning. Springer.



Wise, A., Knight, S. & Buckingham Shum, S. (In Press). Collaborative Learning Analytics. International Handbook of Computer-Supported Collaborative Learning. Springer.

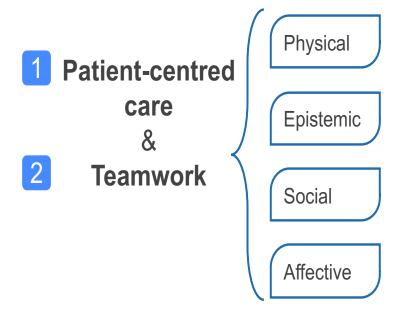
From multimodal logs to higher-order constructs:



From multimodal logs to higher-order constructs:

Curriculum outcomes

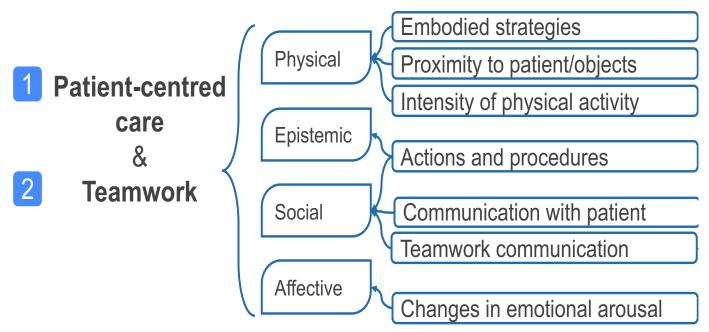
Constructs for collaborative activity (from ACAD Framework)



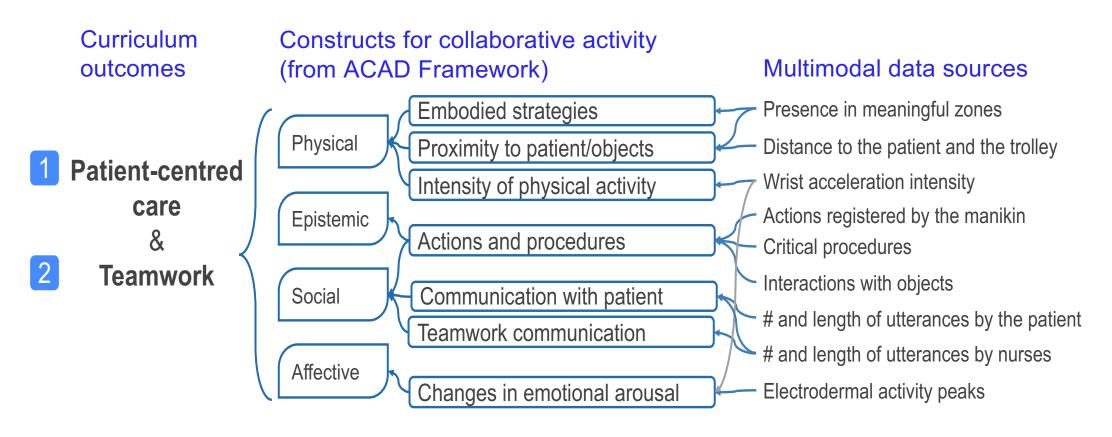
From multimodal logs to higher-order constructs:

Curriculum outcomes

Constructs for collaborative activity (from ACAD Framework)

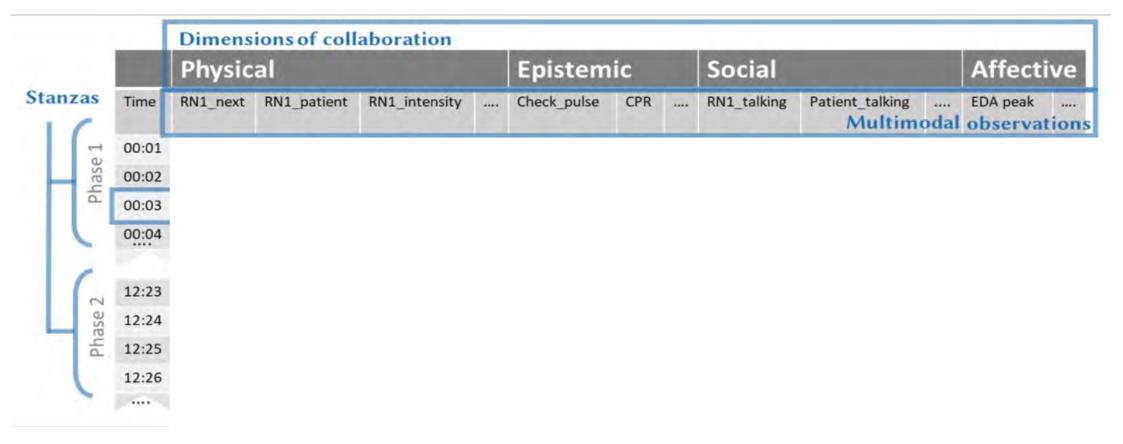


From multimodal logs to higher-order constructs:



	Dimensions of collabora	ation		
- 1	Physical	Epistemic	Social	Affective

1.	Dimens	ions of coll	aboration					
	Physic	al		Epistem	ic	Social	Affective	
Time	RN1_next	RN1_patient	RN1_intensity	 Check_pulse	CPR	 RN1_talking	EDA peak	



		Dimens	ions of coll	aboration								
		Physical				Epistemic			Social	Affective		
anzas	Time	RN1_next	RN1_patient	RN1_intensity		Check_pulse	CPR		RN1_talking	Patient_talking Multimodal	EDA peak	 tions
-	00:01	1	0	low	1	0	0	1	0	1	0	1
Phase	00:02	1	0	low	Þ.	1	0	Þ	0	1	0	
P	00:03	1	0	low		1	0		1	0	0 Seg	gment
C	00:04	1	0	low	Þ	1	0	Þ	1	0	0	
6	12:23	0	1	high		0	1		1	0	0	
Phase.	12:24	0	1	high		0	1		0	0	1	
Phi	12:25	0	1	high		0	1		1	0	1	
	12:26	0	1	moderate		0	0		0	0	0	
-												

#### **Combining data sources to operationalise constructs**

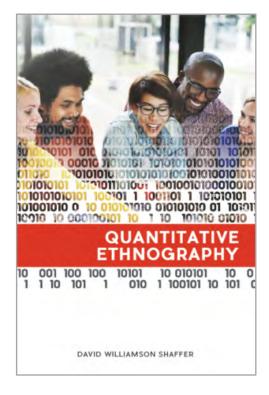
		Dimens	ions of coll	aboration									
			Physic	al			Epistem	ic		Social		Affect	ive
as	Time	RN1_next	RN1_patient	RN1_intensity		Check_pulse	CPR		RN1_talking	Patient_talking Multimoda	EDA peak	 tions	
1	00:01	1	0	low	1	0	0	1	0	1	0	1	
DCDI I	00:02	1	0	low	Þ.	1	0	Þ	0	1	0		
	00:03	1	0	low		1	0		1	0	0 Seg	ment	
	00:04	1	0	low	►	1	0	Þ	1	0	0		
1	12:23	0	1	high		0	1		1	0	0		
	12:24	0	1	high		0	1		0	0	1		
-	12:25	0	1	high		0	1		1	0	1		
	12:26	0	1	moderate		0	0		0	0	0		

Modelling decisions: how to map data type(s) to constructs

Segments can be added by machines or humans

# N.B. "Quantitative Ethnography" (Shaffer, 2017)

An emerging community committed to harmonising quantitative and qualitative methodologies to analyse (large scale) human activity data



0

Society for Learning Analytics Research



LAK18 Keynote Address

https://www.youtube.com/watch?v=LjcfGSdIBAk

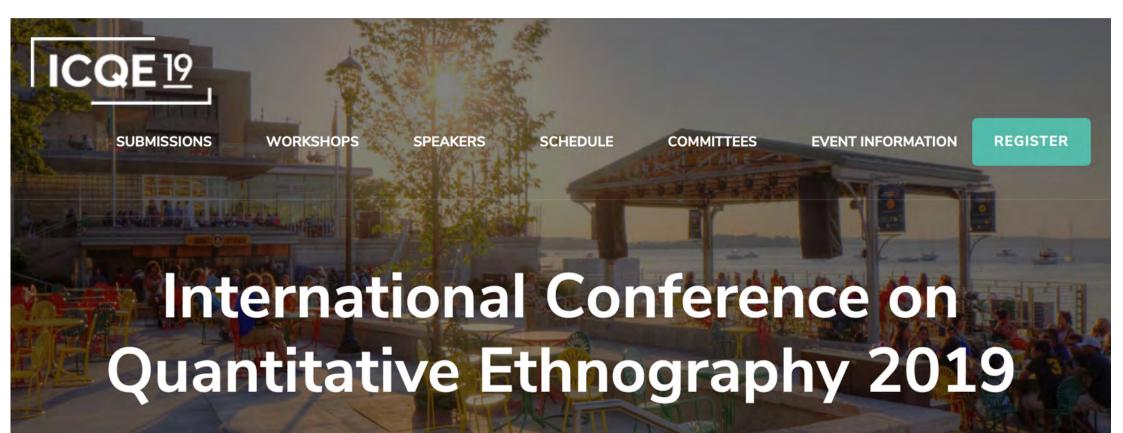
http://www.quantitativeethnography.org

### The quant/qual distinction has dissolved. Each has methods to enrich the other

"In the age of Big Data, we have an opportunity to expand the tools of ethnography — and history, and literary analysis, and philosophy, and any discipline that analyzes meaning — by USING statistical techniques not to supplant grounded understanding, but to expand it. To use additional warrants to support the stories that we tell about the things people do, and reasons they do them."

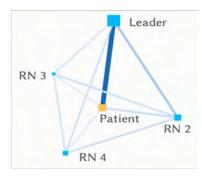
David Williamson Shaffer, *Quantitative Ethnography*, 2017, p.398

#### icqe19.org

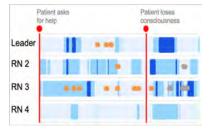


October 20-22, 2019 Madison, Wisconsin

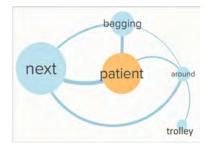
## THE HCI/FEEDBACK CHALLENGE: Making activity visible through proxies



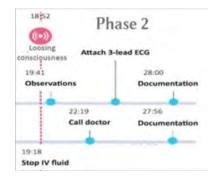
Patient-centred verbal communication, and within nursing team



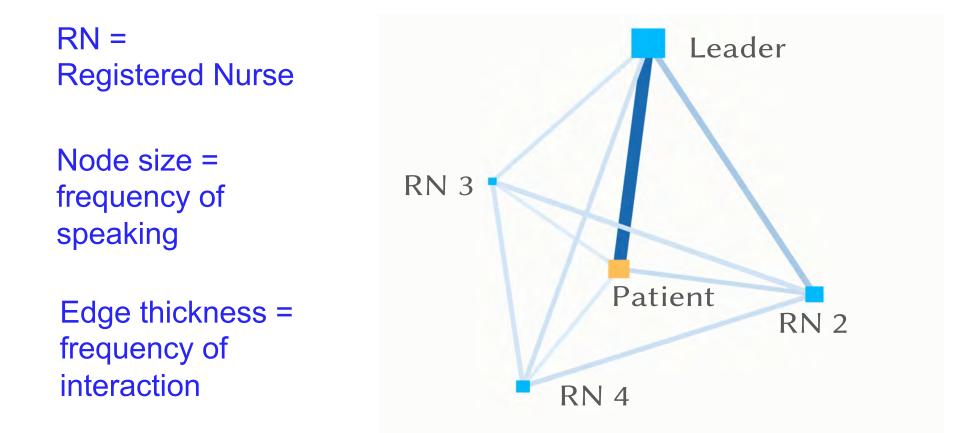
Affective/cognitive arousal via EDA peaks

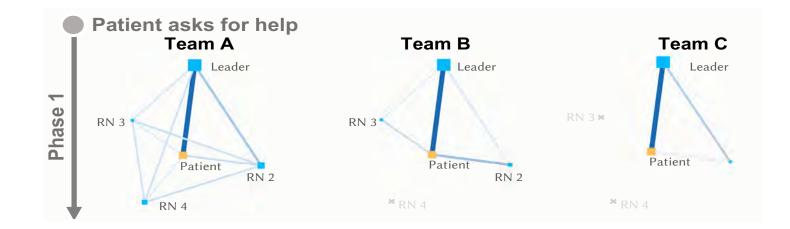


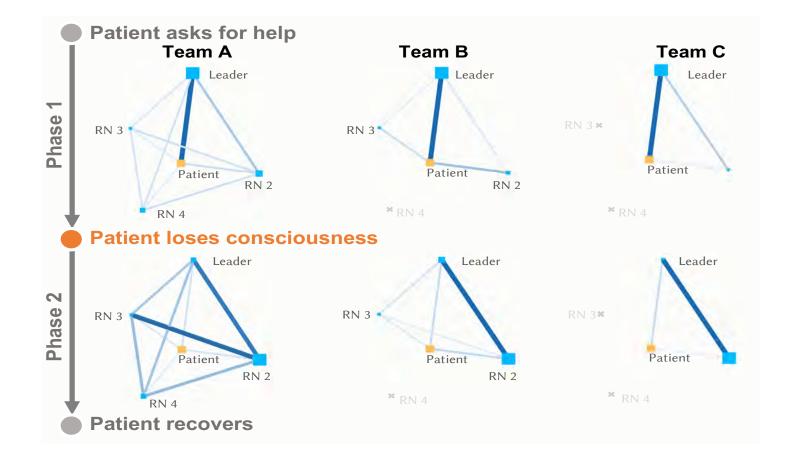
Patient-centred movement around the simulation zones

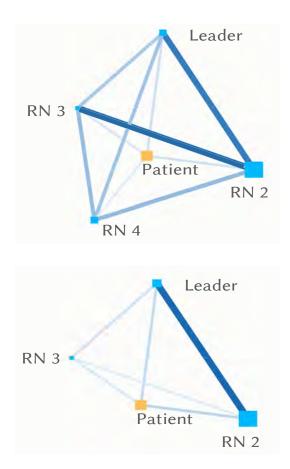


Critical actions performed by nurses









- Excerpt 1: Nurses in Team A communicating effectively
- 1 RN2  $\Rightarrow$  Leader: Put the head up.
- 2 Leader  $\Rightarrow$  RN2: one, two (giving oxygen to the patient)
- 3 RN2 ⇒ Everyone: I am going to do one more... twenty-one, twenty-two, twenty-three ... (doing CPR and counting aloud)
- 4 RN2  $\Rightarrow$  RN3: You take the next round please.
  - RN3 ⇔ RN2: Ok!

5

4

- 6 Leader  $\Rightarrow$  Everyone: one, two (giving oxygen to the patient)
- 7 RN4 ⇒ Everyone: Guys, I am going to start, I am going to do the defib now.

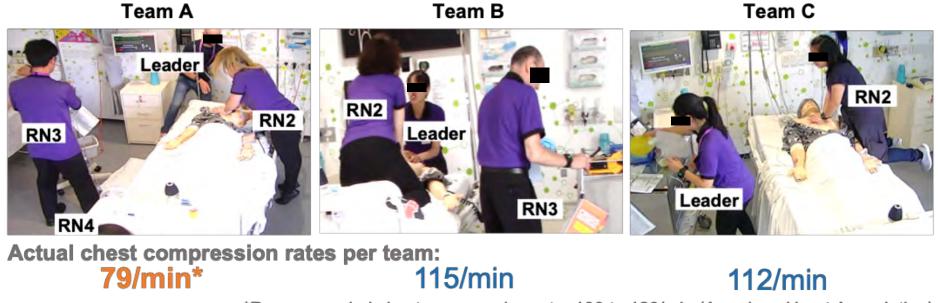
Excerpt 2: Nurses in Team B communicating less effectively

- 1 Leader  $\Rightarrow$  RN2: I am going to check the airway.
- 2 RN2 ⇒ Leader: ...and I will need this one (*pointing to the aging mask*) ...so, should I start?
- 3 Leader ⇒ RN2: Yes!
  - RN2 ⇒ RN1: one, two(*doing CPR and counting aloud*)...twenty-nine, thirty
- 5 Leader ⇒ Everyone: one, two (giving oxygen to the patient)

# **COLLABORATION TRANSLUCENCE:** Patient-centred movement

Three different positions from which nurses performed chest compressions:By the bed (A)Over the patient (B)Over the bed (C)

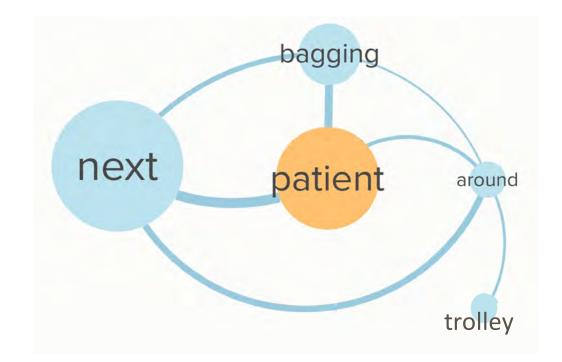
**Chest compressions in Phase 2** 



\*Recommended chest compression rate: 100 to 120/min (American Heart Association)

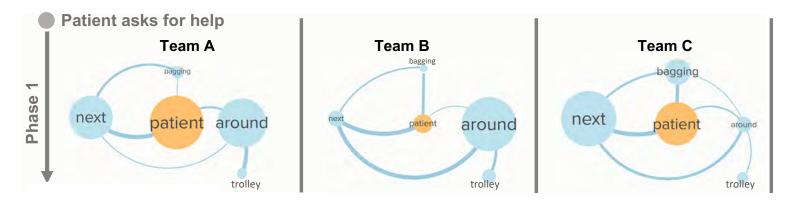
# **COLLABORATION TRANSLUCENCE: Proxy for Patient-Centred Movement**

Each circle represents one zone of interest around the patient's bed, size reflecting relative location, links showing the transitions between zones



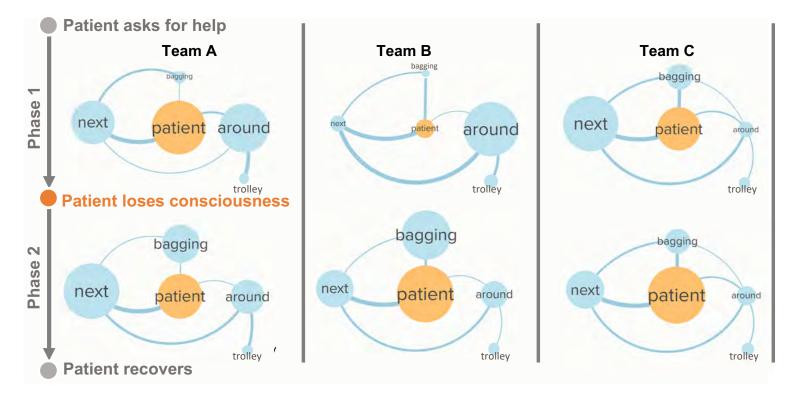
# **COLLABORATION TRANSLUCENCE:** Position and Movement

Each circle represents one zone of interest around the patient's bed, with links showing the transitions among zones.



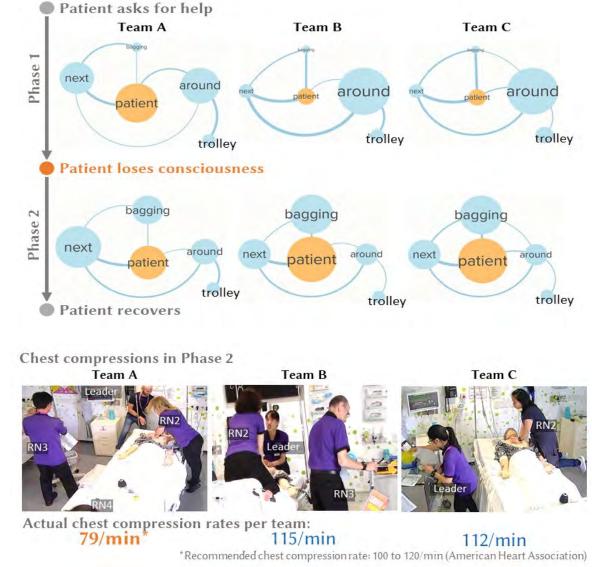
# COLLABORATION TRANSLUCENCE: Position and Movement

Each circle represents one zone of interest around the patient's bed, with links showing the transitions among zones.



# COLLABORATION TRANSLUCENCE: Position and Movement

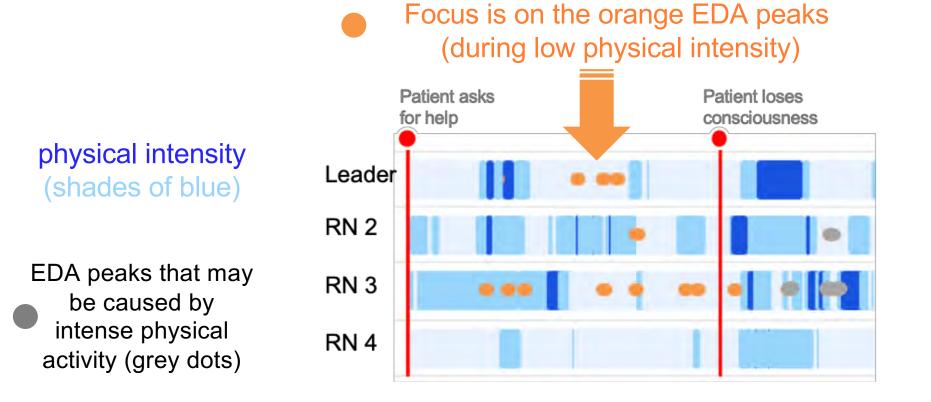
To assist reflection, the proxies can be compared and contrasted between phases, or within phases across teams, optionally coupled with video



# **COLLABORATION TRANSLUCENCE:**

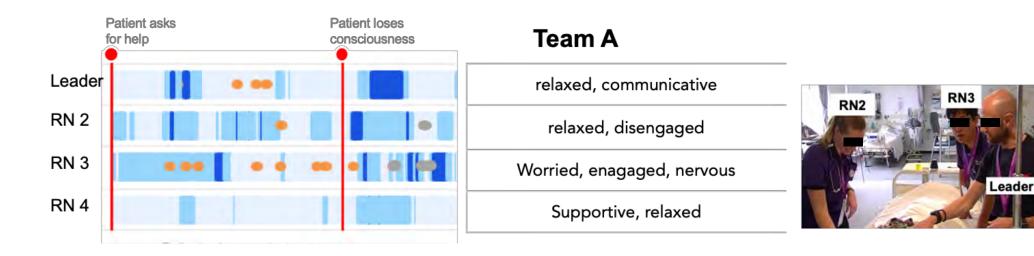
### **Affective response (= stress? engagement?)**

Physiological arousal as measured by Electrodermal Activity (EDA) peaks



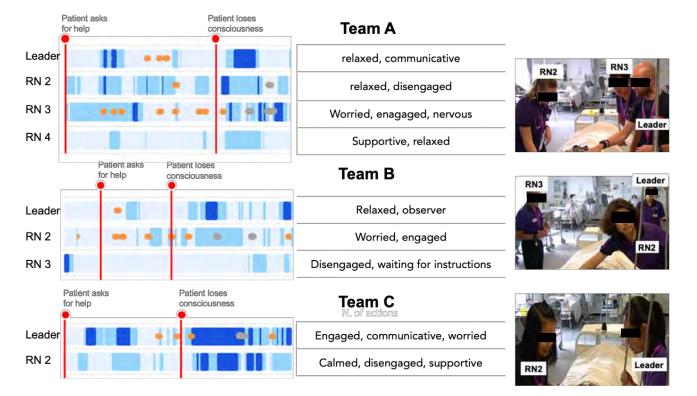
# **COLLABORATION TRANSLUCENCE:** Affective response (= stress? engagement?)

Human analysis of the videos confirmed that both *EDA peaks*, *and their absence*, may signify a range of responses in nurses



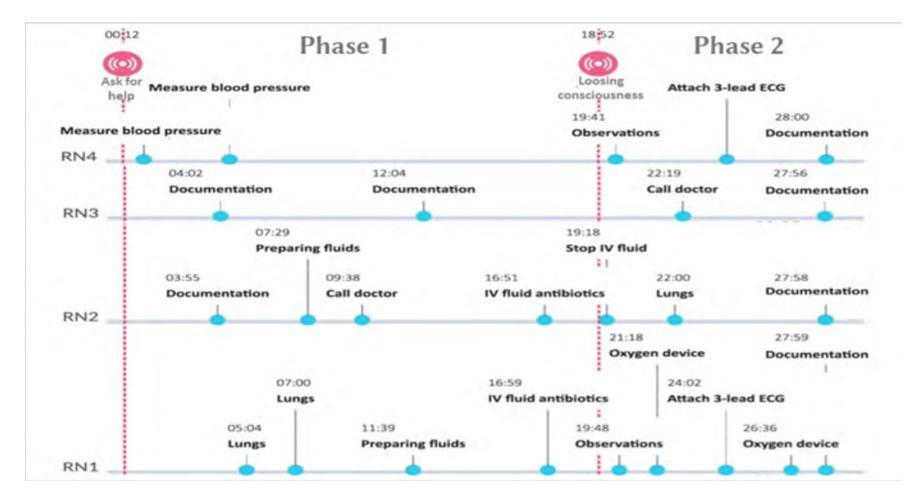
# **COLLABORATION TRANSLUCENCE:** Affective response (= stress? engagement?)

Human analysis of the videos confirmed that both *EDA peaks*, *and their absence*, may signify a range of responses in nurses



## **COLLABORATION TRANSLUCENCE:**

### **Patient-centred team coordination of tasks**



## COLLABORATION TRANSLUCENCE: Patient-centred team coordination enhanced version using "data storytelling" principles (Echeverria et al. 2018)

#### Team 2

After the patient lost consciousness, the team reacted slowly

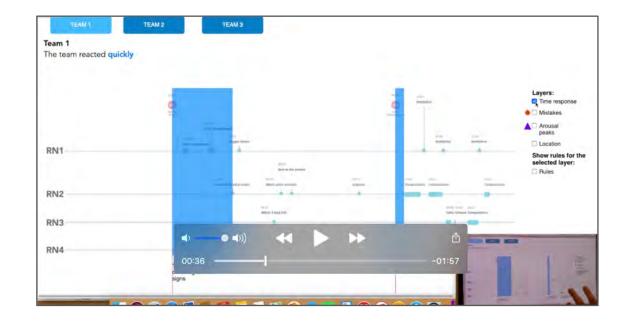


Echeverria, V., R. Martinez-Maldonado, R. Granda, K. Chiluiza, C. Conati and S. Buckingham Shum (2018). Driving Data Storytelling from Learning Design. In *Proceedings of the International Conference on Learning Analytics & Knowledge*, LAK'18, 131-140. ACM. <u>https://doi.org/10.1145/3170358.3170380</u>

### **COLLABORATION TRANSLUCENCE:**

### **Patient-centred team coordination (enhanced version)**

# Video demo



(detailed evaluation being written up)



"...while RN4 and RN2 were doing the fluids I was staying with the patient. It is good to step back and look at what each person was doing, one thing at the same time, I think it shows you how you worked as a team"

"it seems like a lot was done in clumps, you [RN3] were talking to the patient, looking for information while others were doing the observations, that seems practical to me"



(detailed evaluation being written up)

"... particularly if [students] have no arousal peaks. That for me is a concern. What are you actually...? What's happening? Why are you...? Why weren't you engaging with that? Or what could be...? Yes, I think it's interesting. [...] And it's also important because we need to let students understand that that's okay. We actually want them to have a little bit of a stress response in all situations because it does sort of stimulate their thinking and makes [students] aware. It improves [students] awareness." [E5]

(detailed evaluation being written up)

"So, I am assuming RN1 has called the doctor and then told them to get the resus trolley. But there's no other delegation in here because there are no activities [referring to the actions shown in the timeline]. Nothing being done, it's all reactive, it's not proactive." [E8].

(detailed evaluation being written up)

"I think it gives [students] something to look at. And show them location, like, for example, if you [team 3] are supposed to be interacting with the patient but you're standing at the end of the foot of the bed, at least you can show [students] that this is where you were and ask them why." [E5]

(detailed evaluation being written up)

"Unless we show [students] recordings, they won't remember what they actually did. [Using Team-IN] some really would remember exactly how they would do it differently next time" [E5]

"It [Team-IN] would be really helpful for students in terms of a reflection, if you gave them really structured reflection questions and this information, and asked them to reflect on what they were doing, whether it was accurate or not, how they're engaging with the patient and other team members, what they were thinking and feeling at the time, it would be a really valuable tool for deep reflection." [E3]

# Academics' proposed uses of the Team Timeline to prompt student reflection

performance e.g. "Am I doing a good job? Am I getting things done?" [E7]

**arousal** e.g. "Can you talk me through what you were feeling in this moment. Can you tell me what you were thinking at this point?" [E2]

**actions or mistakes performed** e.g. "So, let's have a look at these compressions, and as I said, you know, they're too shallow. So, tell me a bit about what depth the compressions need to be at, and have you achieved that? How do you know you're achieving that and the rate of compression?" [E2]

# Academics' proposed uses of the Team Timeline to prompt student reflection

**response time** e.g. "Do you think that that's a definitive time frame to do that in? Or looking at the ABCDEFG algorithm, when would you hope to provide your first shock?" [E8]

**wrong positioning during the simulation** e.g. *"What were you doing over here? Because you didn't go and get the resus trolley but what took you over here? And then this person stood the most time off to the side and did nothing. And do you think that's a fair way to behave in a team?"* [E8]

**unsafe practice** e.g. *"Is this clinically safe for your patient? Are you performing in a way that's safe for your patient?"* [E5]

### **HOW CLOSE TO FULLY AUTOMATED FEEDBACK?**

RN 3

next

RN 2 RN 3

18357

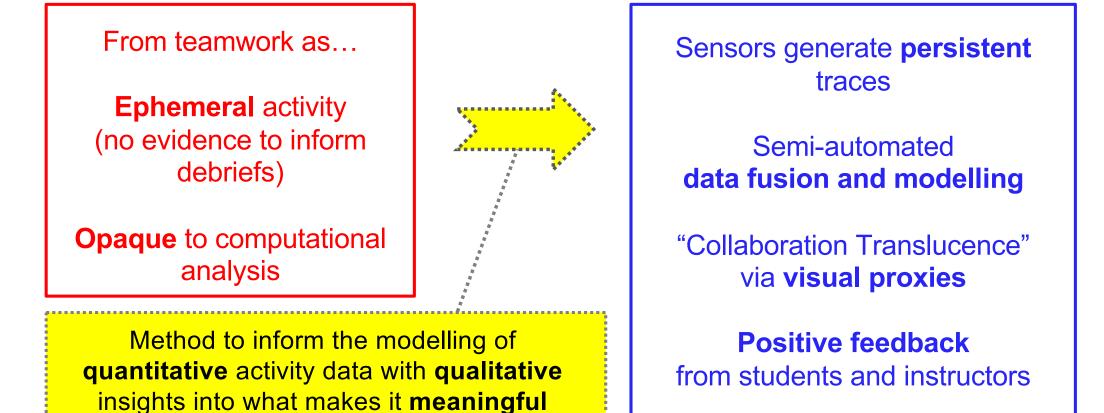
6-0

19-38 Stop IV fluid

22:39 Call doctor

	Collaboration Proxy	Data source	Manual interventions		Automated
Patient RN 4	Patient- centred speech interaction	Audio from video recording	Speech interaction manually annotated	$\rightarrow$	Sociograms generated from speech onset/offset logs
patient energy	Patient- centred movement	X,Y positions and pre- defined zones using indoor localization		$\rightarrow$	Zone transition networks generated from localisation data
	Physical Intensity and Affective reaction	EDA and accelerometer from Empatica wristband	Wristband data download	→	EDA timelines generated from wristband data
Phase 2 starts 3-lead ECG 28 CC Documentation 27.5% Documentation	Teamwork Timeline	Timestamped actions from observation tool	Nursing actions logged by an observer	$\rightarrow$	Timelines generated from action logs

# SUMMARY: A METHODOLOGY TO GENERATE COLLOCATED COLLABORATION ANALYTICS



### **POTENTIAL FUTURE TRAJECTORIES...**

