



MONASH
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Centre for Learning Analytics Monash

Unlocking the full potential

Next critical steps for learning analytics

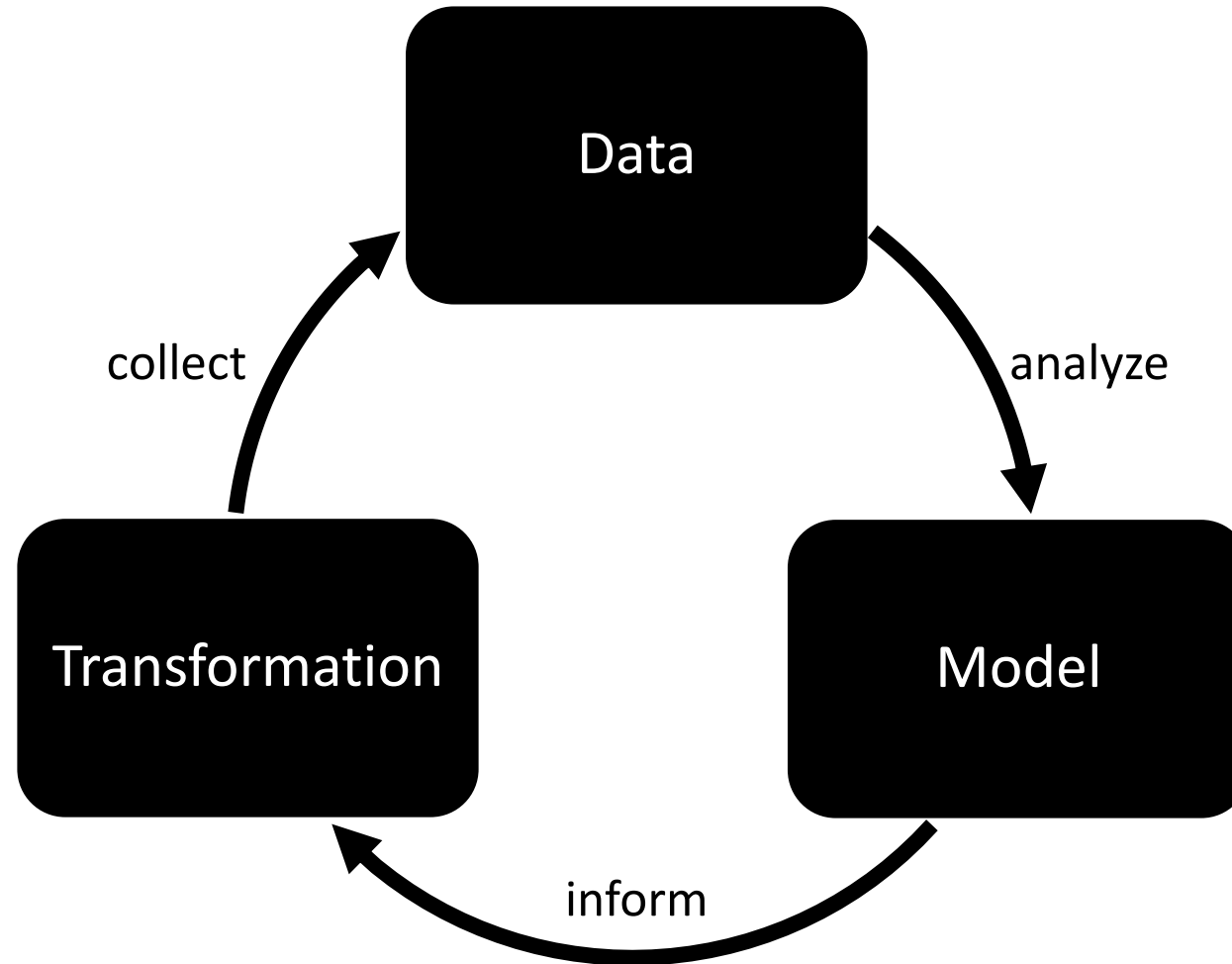
Dragan Gašević
@dgasevic

Oct 20th, 2022
Hong Kong Metropolitan University
[Online]

Learning Analytics

Much promise and high interest

Closing the loop



Challenge

Are measurements and results in
learning analytics reliable?

Challenge

How we grow uptake of
learning analytics?

Key takeaway

We need to get serious about
the quality of data and models

Key takeaway

Humans are central for
adoption of analytics

DATA – MODEL – TRANSFORMATION –
FINAL REMARKS

Challenge

Can we trust measurements in learning analytics?

Challenges

Are data we use good proxies for
what we want to measure?

Data quality can't be fixed with AI (garbage-in-garbage-out)

Three strategies for improving data

Improving validity



Introducing meaning to clicks

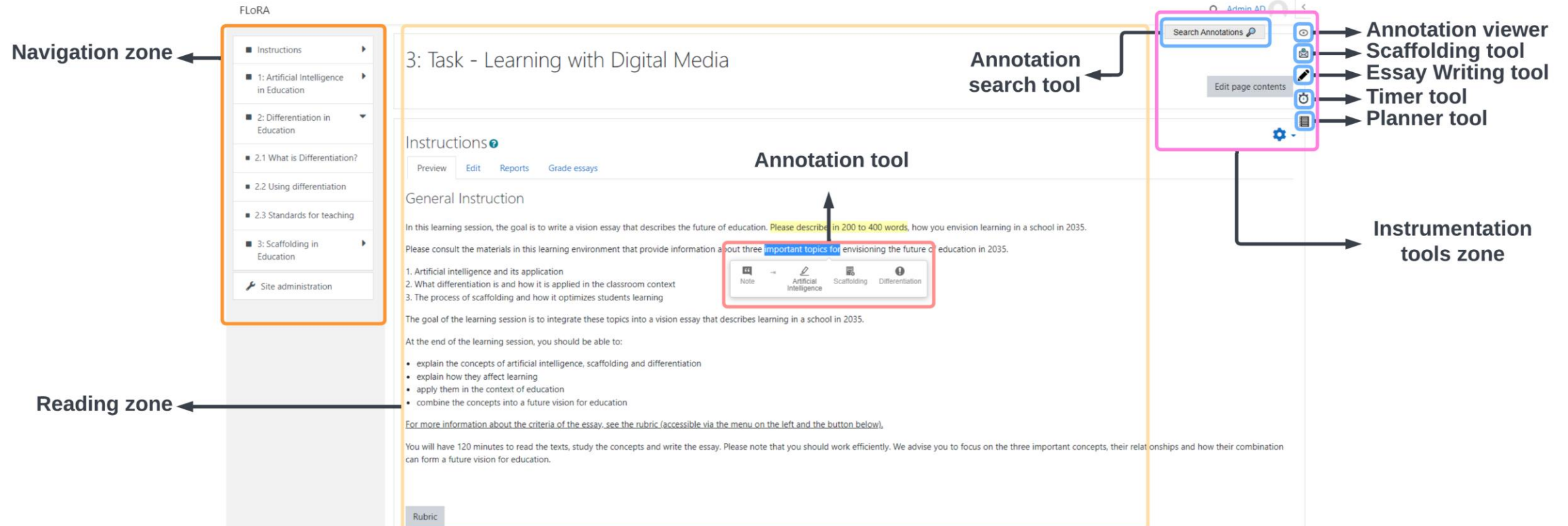
Trace-based self-reports

Improving validity



Create special instrumentation tools

FLoRA



+ + extensions

<https://floraproject.org>

FLoRA

The screenshot displays the FLoRA interface for a task titled "3: Task - Learning with Digital Media". The main content area shows instructions for writing an essay. A sidebar on the left contains a navigation menu with categories like "Instructions", "1: Artificial Intelligence in Education", and "2: Differentiation in Education".

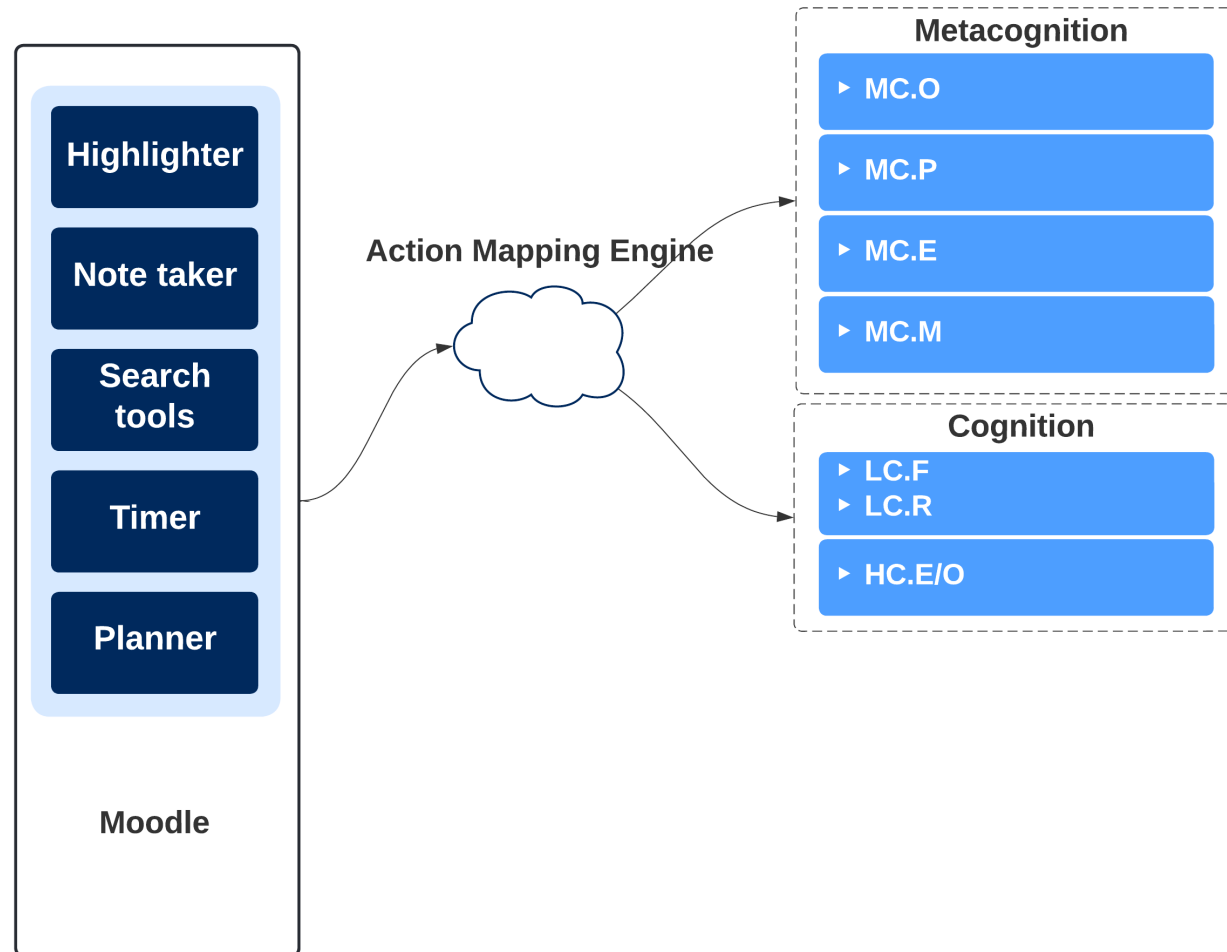
Annotations are shown in a side panel on the right, titled "Annotations 2". The annotations are color-coded and linked to specific parts of the text:

- Highlighted text:** A yellow box highlights the text "write a vision" in the instructions.
- Additional Notes:** A red box highlights a rich text editor for adding notes to the annotation.
- Annotation labels:** A purple box highlights the "#Artificial-Intelligence" tag and the "Add new tags" input field.
- Post options:** An orange box highlights the "Post to Only Me" dropdown menu and the "Cancel" button.
- Annotation post result:** A green box highlights the final published annotation, showing the text "about three important" and the "#Scaffolding" tag.
- Annotation tools:** A blue box highlights a toolbar with icons for "Note", "Artificial Intelligence", "Scaffolding", and "Differentiation".

Arrows point from these callouts to the corresponding elements in the interface. The text "write a vision" is highlighted in yellow in the main content. The text "classroom context" is highlighted in blue. The text "about three important" is highlighted in yellow. The text "accessible via the menu on the left and the button below" is highlighted in blue.

Mapping trace data to processes

Instrumentation tools



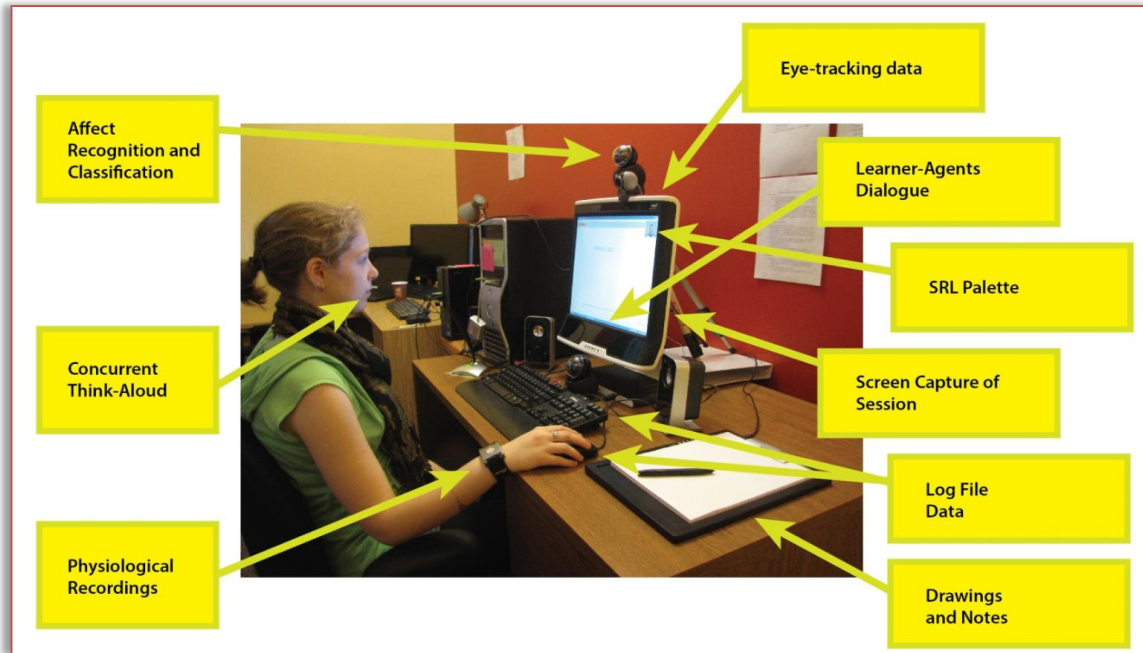
Improving validity



Combining multiple data channels

Multimodal and multichannel data

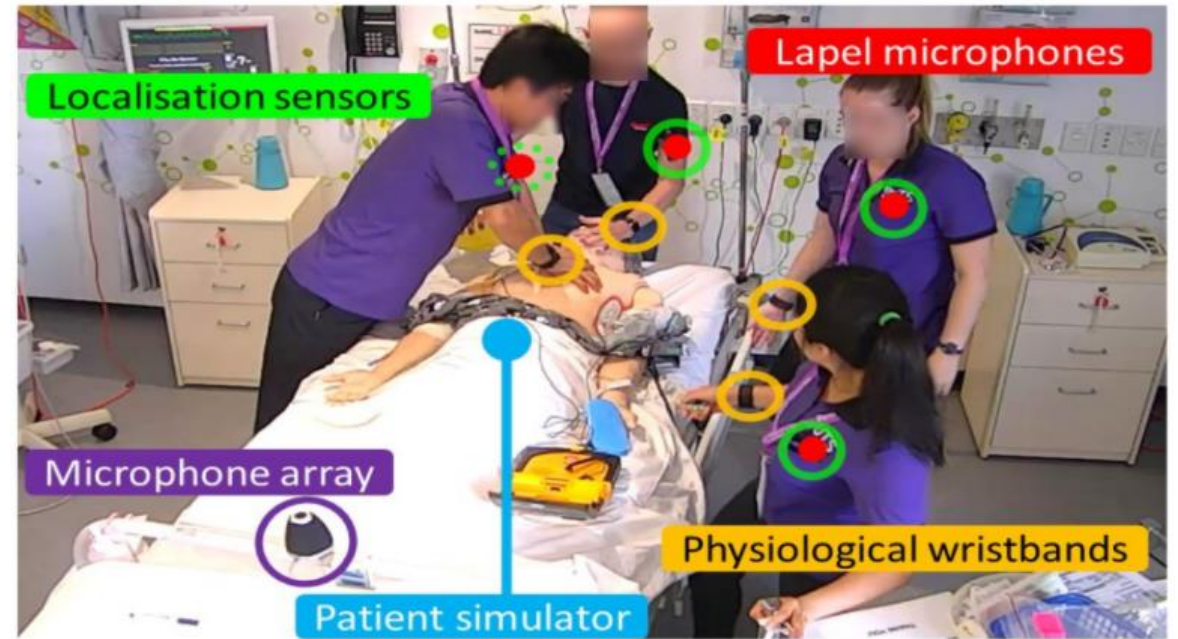
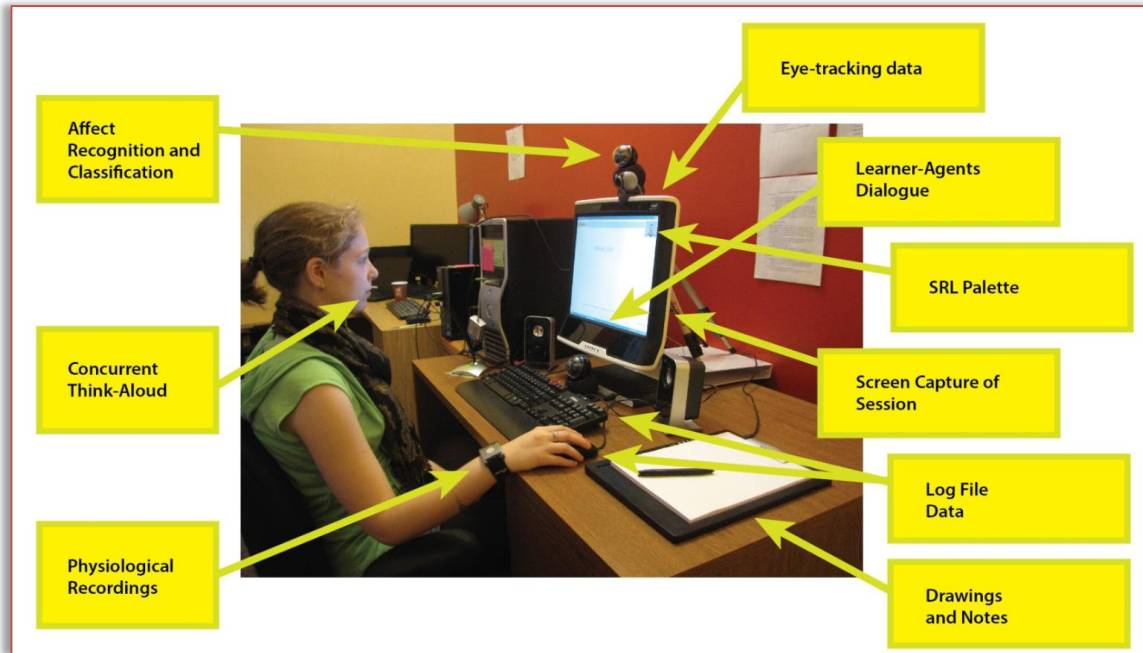
Self-regulated learning



Multimodal and multichannel data

Self-regulated learning

Teamwork



From multichannel data to constructs

Raw trace data

15:06:06	/learn/announce
15:07:34	/learn/content
15:10:22	/learn/announce
15:11:01	/learn/content
15:12:27	/learn/content?type=detail&id=1002579286
17:49:58	/info
17:51:44	/learn/announce
17:51:46	/learn/content
17:52:02	/learn/content?type=detail&id=1002579307
17:52:38	/learn/content?type=detail&id=1002579307&cid=1002813724
17:56:32	/learn/content?type=detail&id=1002579307&cid=1002813725
20:44:19	/info

~ ~

20:44:30	/learn/announce
20:44:32	/learn/content
20:44:34	/learn/content?type=detail&id=1002579275
20:44:41	/learn/score
20:44:41	/learn/custom?id=1002062038
20:44:42	/learn/announce
20:44:44	/learn/content?type=detail&id=1002579275&cid=1002813499
20:44:45	/learn/content?type=detail&id=1002579275&cid=1002813500
11:53:47	/info
11:53:50	/learn/announce
11:53:52	/learn/content?type=detail&id=1002579275&cid=1002813500
10:05:40	/learn/content
10:05:45	/learn/content?type=detail&id=10



Multi data channels:
 Navigational log data
 Peripheral data
 Eye-tracking data

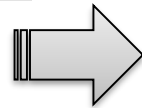
From multichannel data to constructs

Raw trace data

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```

~



Learning actions

Action label 1
Action label 2
Action label 2
Action label 3

~

```

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10:05:40 /learn/content
10:05:45 /learn/content?type=detail&id=10
  
```

Action label 4
Action label 2
Action label 3
Action label N



Multi data channels:
Navigational log data
Peripheral data
Eye-tracking data



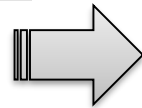
Actions such as:
Relevant_reading
Write_essay
Note_editing

From multichannel data to constructs

Raw trace data

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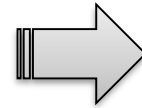
~



Learning actions

Action label 1
Action label 2
Action label 2
Action label 3

~



SRL processes

Process 1
Action 1 -> Action 2
Process 2
Action 2 -> Action 3

~

```
20:44:30 /learn/announce
20:44:32 /learn/content
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10:05:40 /learn/content
10:05:45 /learn/content?type=detail&id=10
```

Action label 4
Action label 2
Action label 3
Action label N

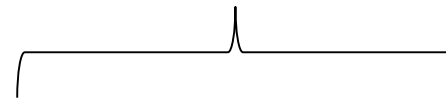
Process 3
Action 4 -> Action 2 -> Action 3
Process N
Action N -> Action N



Multi data channels:
Navigational log data
Peripheral data
Eye-tracking data



Actions such as:
Relevant_reading
Write_essay
Note_editing



Mapping with SRL theory:
Process 1 -> Orientation
Process 2 -> Monitoring
Process 3 -> Evaluation

From multichannel data to constructs

Raw trace data

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10:05:40 /learn/content
10:05:45 /learn/content?type=detail&id=10
  
```

Multi data channels:
 Navigational log data
 Peripheral data
 Eye-tracking data

Learning actions

Action label 1
 Action label 2
 Action label 2
 Action label 3

Action label 4
 Action label 2
 Action label 3
 Action label N

Actions such as:
 Relevant_reading
 Write_essay
 Note_editing

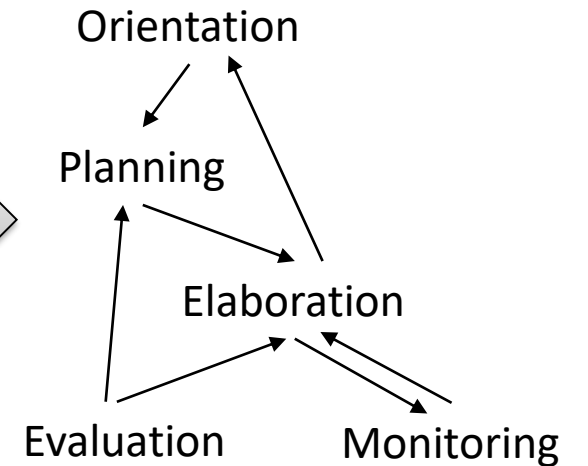
SRL processes

Process 1
 Action 1 -> Action 2
 Process 2
 Action 2 -> Action 3

Process 3
 Action 4 -> Action 2 -> Action 3
 Process N
 Action N -> Action N

Mapping with SRL theory:
 Process 1 -> Orientation
 Process 2 -> Monitoring
 Process 3 -> Evaluation

Self-regulated learning



Modelling (e.g.,
 process mining, ENA)

Validation

Using other data sources to improve
and validate measurement

Measurement sensitivity

Data channels differ in
what they can capture

DATA – **MODEL** – TRANSFORMATION –
FINAL REMARKS

Learning context

To what models can be generalized in learning analytics?

What shapes generalizability?

Instructional conditions shape learning analytics results

What shapes generalizability?

Students matter the most in learning analytics

Opportunity

Analytics of learning strategies

Analysis methods

Analytics of learning strategies

Unsupervised machine learning

+

Sequence mining

Process mining

Network analysis

Key findings (1/3)

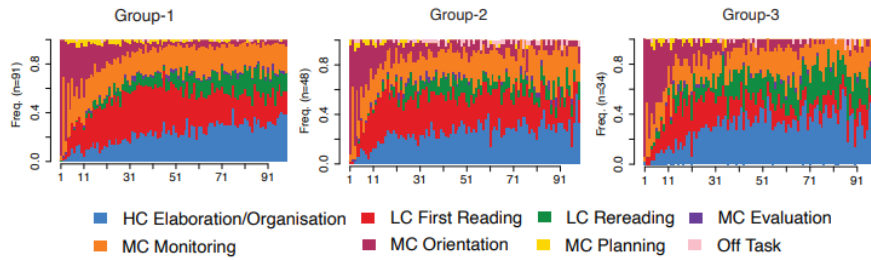
Analytics of learning strategies

Regulation of strategies is consistent with relevant theory

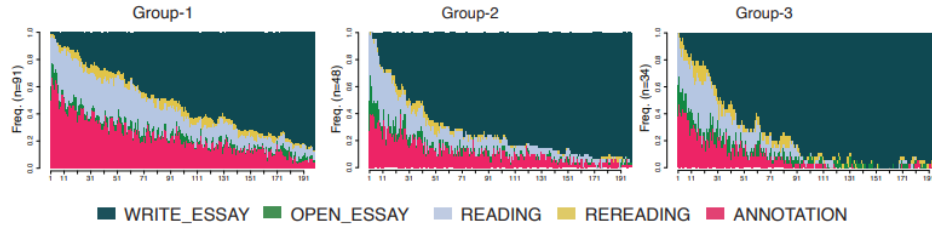
Characterizing learning strategies

- Group 1 - Read First, Write Next
- Group 2 - Read and Write Simultaneously
- Group 3 - Write Intensively, Read Selectively

A. State distribution plot of SRL processes



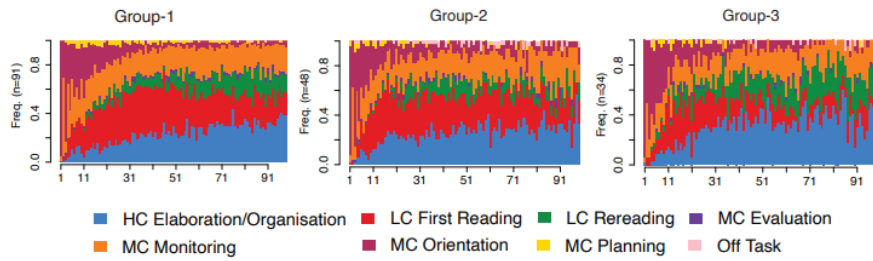
B. State distribution plot of learning actions



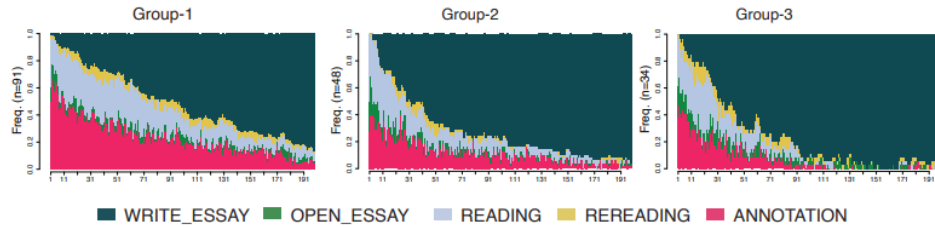
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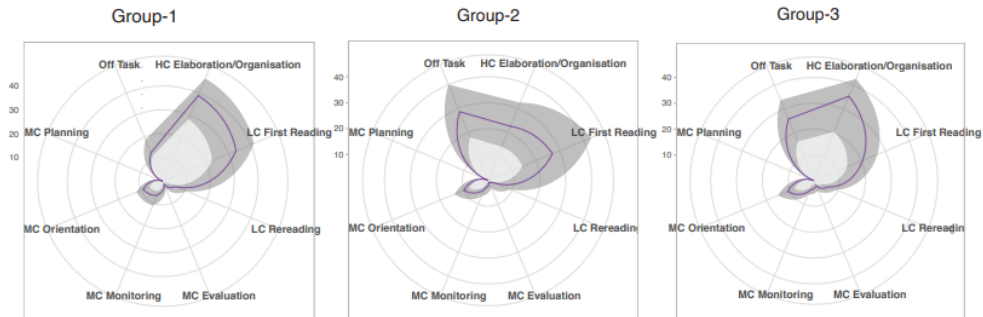
A. State distribution plot of SRL processes



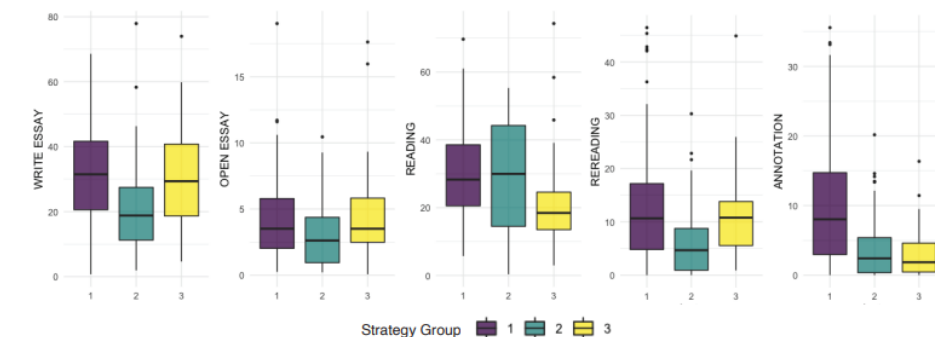
B. State distribution plot of learning actions



C. Distribution of time duration of SRL processes



D. Distribution of time duration of learning actions

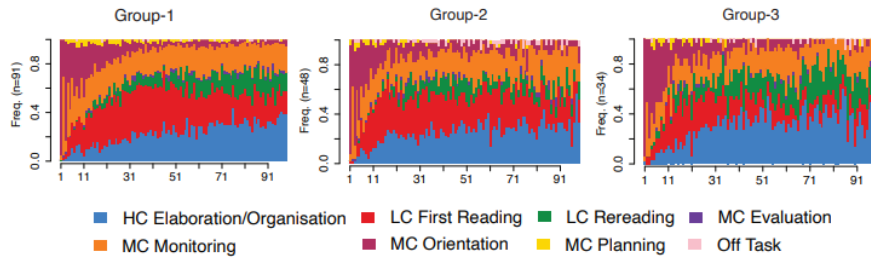


er Graaf, J., ... & Gasevic, D. (2022). Effects of Internal and External Conditions on Strategies of Self-
of the 12th International Learning Analytics and Knowledge Conference (pp. 392-403).

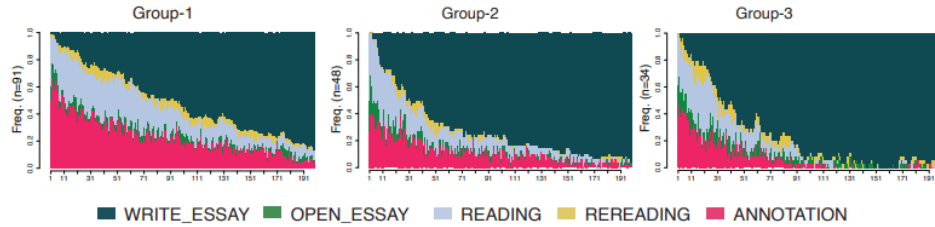
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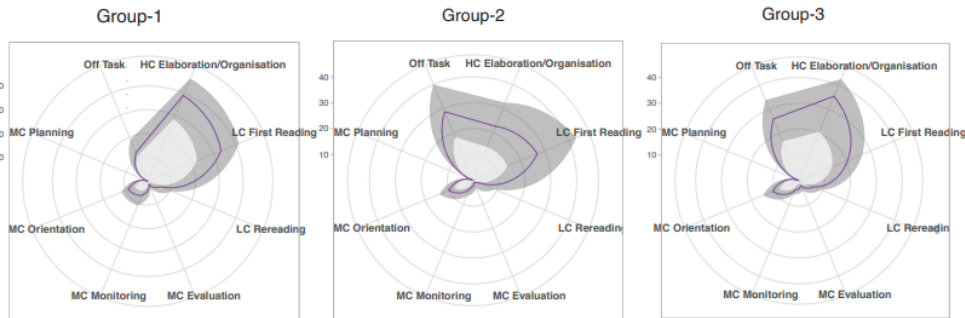
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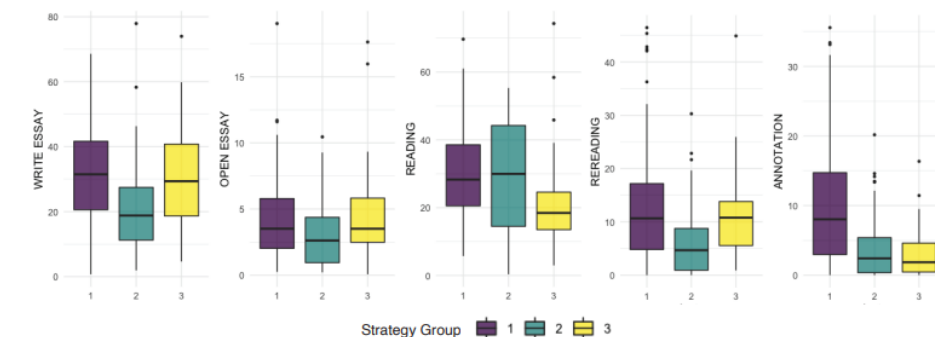
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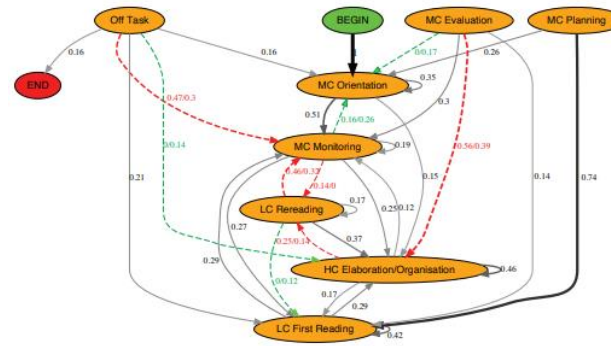
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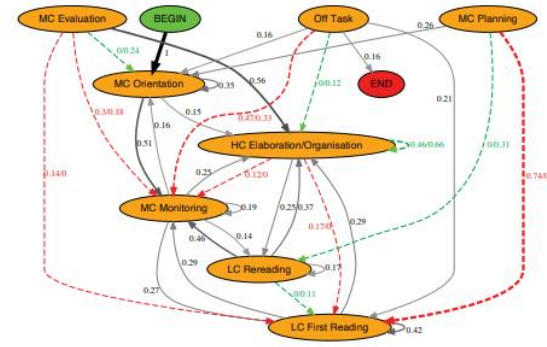
D. Distribution of time duration of learning actions



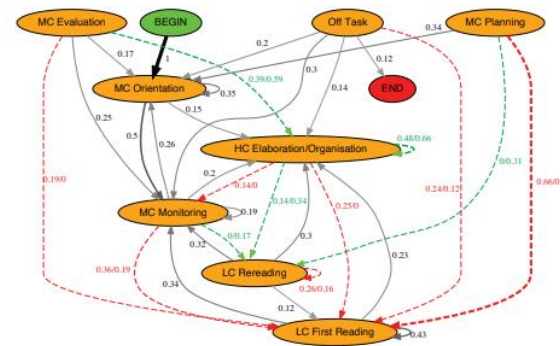
E. First Order Markov Models of SRL processes



FOMM Comparison between Group-1 and Group-2



FOMM Comparison between Group-1 and Group-3



FOMM Comparison between Group-2 and Group-3

of Internal and External Conditions on Strategies of Self- and Knowledge Conference (pp. 392-403).

Key findings (2/3)

Analytics of learning strategies

Strategies are predictive of academic performance

Key findings (3/3)

Analytics of learning strategies

Explain underlying learning processes and mechanisms

Ultimate goal

Models of *individual* learners

An idiographic approach

Identify learning signatures of individual learners

Directions

Beyond accuracy – model *fairness*

Directions

Explainable analytics to support *learning about learning*

DATA – MODEL – **TRANSFORMATION** –
FINAL REMARKS

Interaction

Dashboards

Interaction

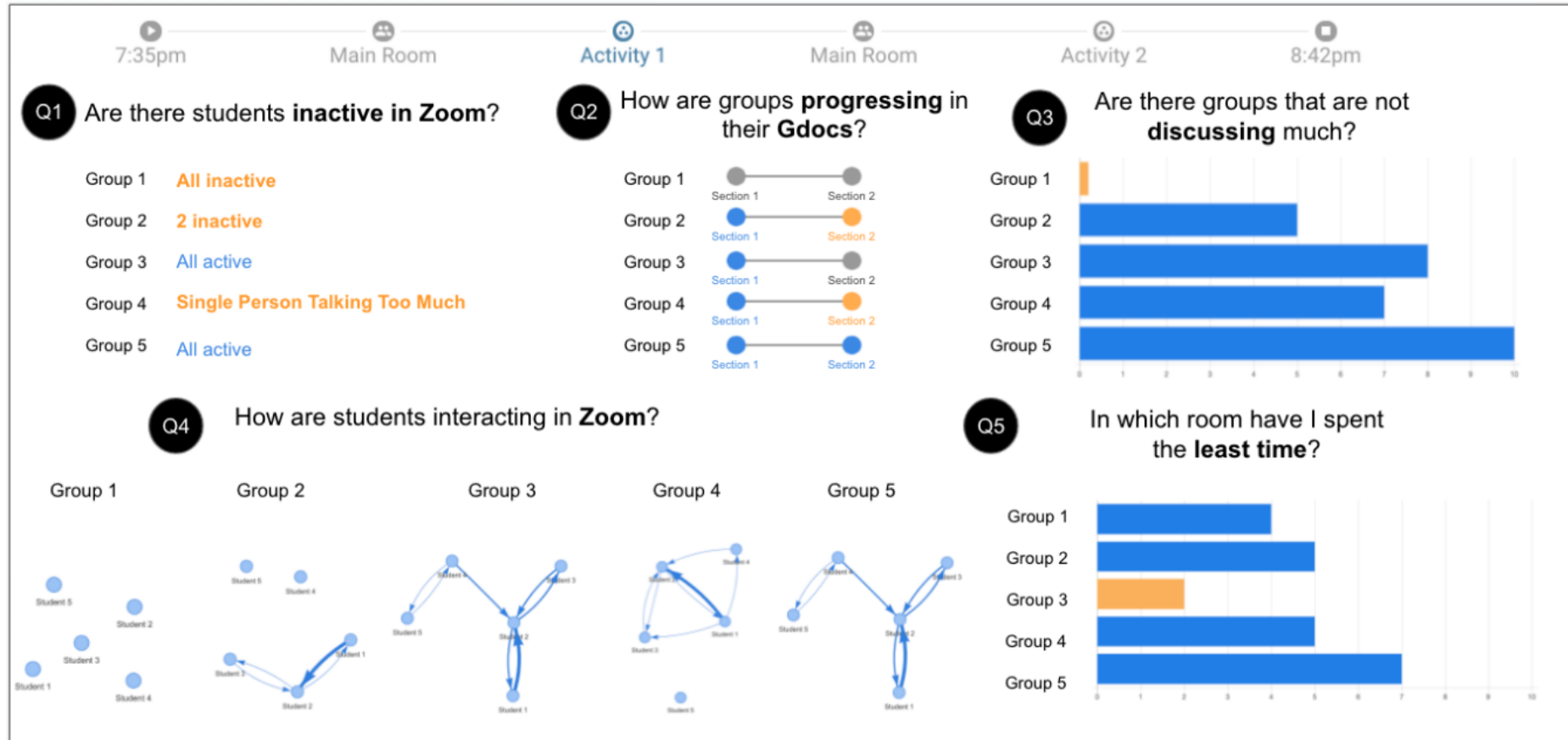
Dashboards can be harmful



Direction

Analytics in the loop
(human is already in the loop!)

Data storytelling



Recommendations for teachers

Individual Group
Close ✕

Justine Made at least one critical connection ✓ ?

Intervention

Justine is:

- balancing issues stakeholders care about

You might suggest that Justin thinks about:

- how land use changes affect indicators in the model, and how that can help balance issues stakeholders care about

Chat

KNOW HOW TO DISABLE THIS OUT.

Bel: ● 6/29/16 02:45 PM
as we change things the indicator graphs would change things, like more commercial zones increased sales and more industrial zones increased the Carbon monoxide and job

Bel: 6/29/16 02:45 PM
good

Justine: 6/29/16 02:45 PM
That sounds right

Justine: 6/29/16 02:45 PM
I feel I need to know more about zoning and its implications.

Ryan: 6/29/16 02:45 PM
I think it was very hard to change anything without having repercussions in a different category. It was definitely challenging to try and satisfy all of the different demands of the stakeholders.

Ryan: 6/29/16 02:45 PM
Several

Justine: 6/29/16 02:45 PM
I tried to change carbon monoxide but clearly it didn't work.

Nic: ● 6/29/16 02:45 PM
I feel like I was just randomly changing industrial plots to open space/wetlands for more nesting sites and less carbon emissions but I feel like if I really knew how to successfully zone, the outcome would be better.

Nic: 6/29/16 02:45 PM
Also several indicators

Justine: ●●● 6/29/16 02:45 PM
It seems almost impossible to please every stakeholder because you have to sacrifice jobs and sales to reduce carbon emissions and increase nesting sites, so we will have to compromise.

Recommendations for teachers

Individual Group
Close X

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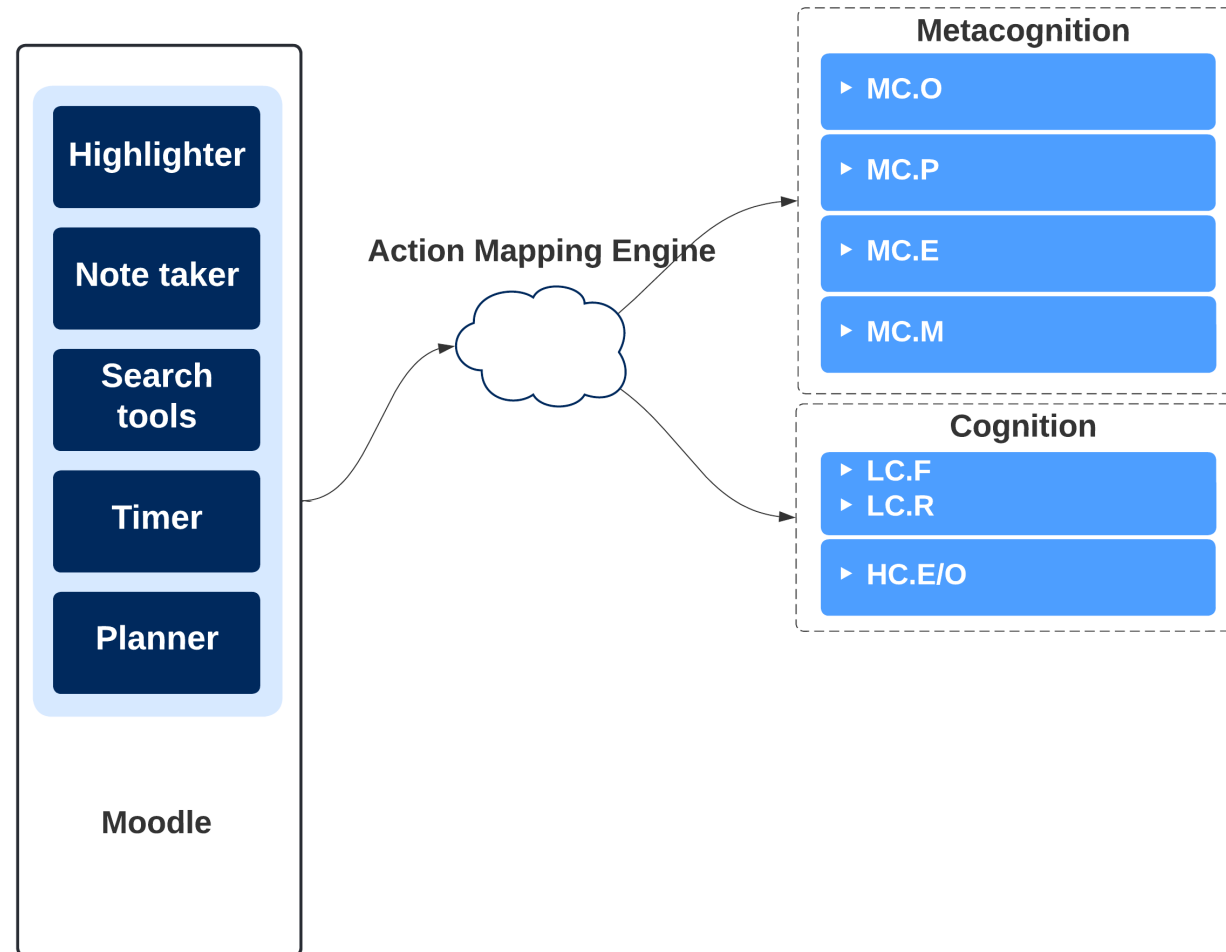
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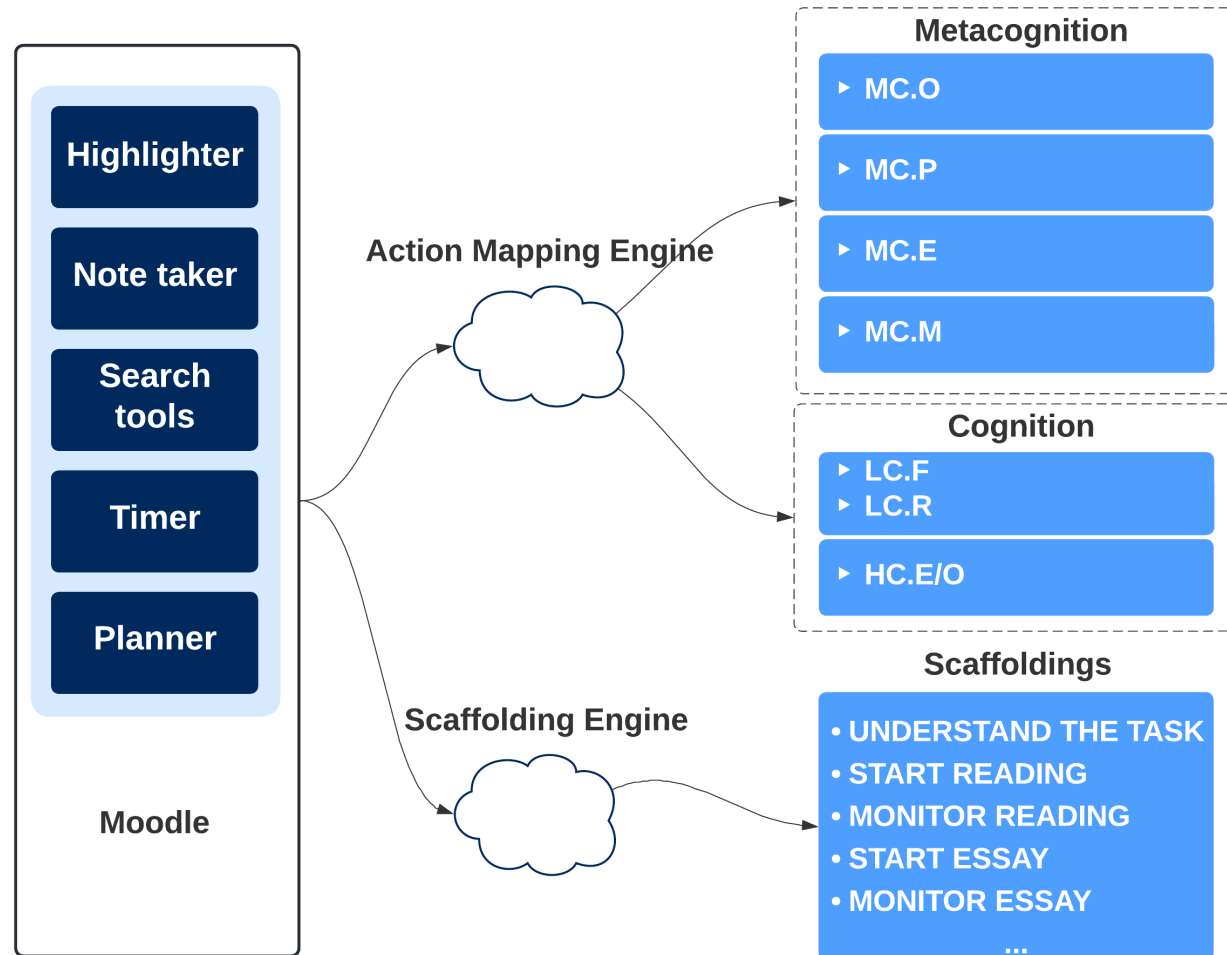
Mapping trace data to processes

Instrumentation tools



Analytics-based personalized scaffolding

Instrumentation tools



Analytics-based personalized scaffolding

The screenshot displays the FLoRA (Learning Analytics) interface. On the left, a navigation menu lists sections: Instructions, 1: Artificial Intelligence in Education, 2: Differentiation in Education, 2.1 What is Differentiation?, 2.2 Using differentiation, 2.3 Standards for teaching, 3: Scaffolding in Education, and Exit to Home Page. The main content area is titled '3: Task - Learning with Digital Media' and contains 'Instructions' and 'General Instruction'. A 'Monitor essay' dialog box is overlaid on the page, containing the text: 'It is important to write relevant information and check your writing. Which are the most helpful steps for you to understand the text so as to do the task? (Please select from the recommended options below)'. Three options are presented as buttons: 'Check the essay rubric' (highlighted in yellow), 'Edit your essay', and 'Check the learning goals and instructions'. A 'Create Checklist' button is at the bottom right of the dialog. A white arrow points from the 'Check the essay rubric' button to a white box labeled 'Scaffolding tool' at the bottom right. Another white arrow points from the top right corner of the interface to the 'Scaffolding tool' box.

Open challenge

Integrating personalized scaffolds into task design

Opportunity

Analytics to enhance feedback quality

Automatic detection of properties of feedback

Opportunity

Towards automatic feedback

Opportunity

Towards automatic feedback

Automatic feedback increases student performance

Opportunity

Towards automatic feedback

No evidence that
human feedback is more effective than automatic feedback

Challenges

Towards automatic feedback

No evidence that automatic feedback eases instructors' workload

Challenges

Towards automatic feedback

Main method used for automatic feedback provision is the comparison with a desired answer in some subject

DATA – MODEL – TRANSFORMATION –
FINAL REMARKS

Stop the game of low hanging fruit
and start measuring what matters

Moving away from
the idea of *homogenous* learner models

Human-centred learning analytics



MONASH
University



Centre for Learning Analytics Monash

Unlocking the full potential

Next critical steps for learning analytics

Dragan Gašević
@dgasevic

Oct 20th, 2022
Hong Kong Metropolitan University
[Online]