

# Open learner models (OLMs) and learning analytics dashboards: A systematic review

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# Overview

Why OLM research is relevant to LAK (Judy)

A little history and background about OLMs (Judy)

Results & Highlights (Ioana)

Comparing OLMs and LAK dashboard (Dan)

Implications for future work (Dan)

# Why is OLM research relevant for LAK?

What was the motivation for this work?

# Motivation

To bring together two strands of research:  
Learner-focused learning analytics  
Open Learner Models (OLMs)

# A little history

OLMs .....

1970s

Artificial Intelligence in Education (AIED)

Intelligent Tutoring Systems (ITS)

Bloom's famous 2-sigma (1984)

Student models

Learner models

~ 1999: **Open learner models**

Learning analytics

Learner-focused learning analytics

1970s

Artificial Intelligence in Education (AIED)

Intelligent Tutoring Systems (ITS)

Bloom's famous 2-sigma (1984)

Student models

Learner models

~ 1999: **Open learner models**

Learning analytics

Learner-focused learning analytics

OLMs are “designed” to model the learner, initially for personalisation of teaching but more recently as independent interfaces .... Just like learning analytics interfaces

What is an Open Learner  
Model (OLM)?



How is it *similar* to learner oriented learning analytics?

# Both OLMS and learner facing learning analytics

Provide an interface for the learner

The interface should enable a learner to do core metacognitive tasks eg monitor their progress, support planning, support self-reflection

The interface should help learners have more agency - more control and awareness

How is it *different* from  
learning analytics?

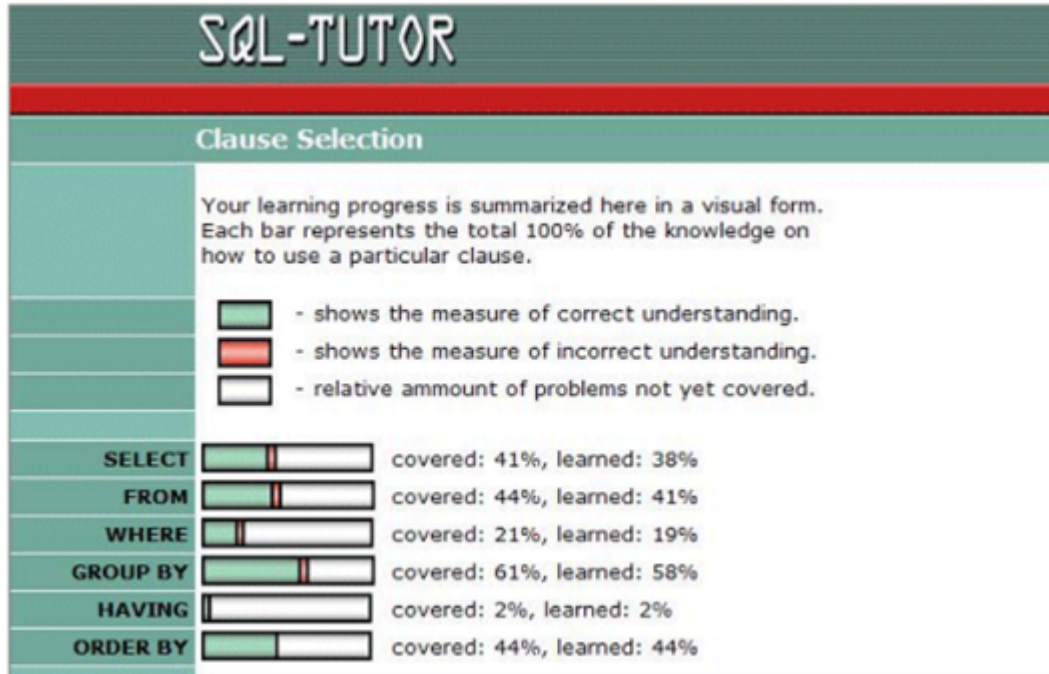
OLM - Developed in parallel to (research) platform  
LAD - Developed independently

# OLMS versus learner facing learning analytics

A simplified view of the design of an OLM in 4 steps:




1. Design the “ontology” of the model - what components will be modelled
  - eg The learner can write simple loops in C
2. Design the evidence sources to reason about each component
  - eg create an environment where students do programming tasks
3. Reason from the evidence to conclude the level of knowledge
  - eg. analyse the detailed correctly completed tasks, the ones that were done incorrectly, the ones still not attempted, taking account of potential for slips and guesses
4. Create an interface that enables the learner to answer questions
  - eg Have I mastered writing of C loops to the required level

# One elegant example of an OLM



# Wayang Outpost My Progress

« Go back to Learning Hut

Topic	Progress	Performance	Remarks	Actions
Volumes		<p>Mastery Level</p> <div><div>30</div></div> <p>Problems Done : <b>3</b> Total Problems : 4 <a href="#">Learn More &gt;</a></p>	<p>Got last problem right! Do you want to try more problems like this?</p> <p><a href="#">Comment &gt;</a></p>	<p><a href="#">Continue »</a></p> <p><a href="#">« Review</a></p> <p><a href="#">Challenge</a></p>
Number Sense		<p>Mastery Level</p> <div><div>93</div></div> <p>Problems Done : <b>8</b> Total Problems : 18 <a href="#">Learn More &gt;</a></p>	<p>Skill mastered! Do you want to try more challenging problems, or a new topic?</p> <p><a href="#">Comment &gt;</a></p>	<p><a href="#">Continue »</a></p> <p><a href="#">« Review</a></p> <p><a href="#">Challenge</a></p>
XY Linear Functions and Relationships		<p>Mastery Level</p> <div><div>45</div></div> <p>Problems Done : <b>12</b> Total Problems : 25 <a href="#">Learn More &gt;</a></p>	<p>Don't like reading? Have the computer read aloud - click the read aloud button.</p> <p><a href="#">Comment &gt;</a></p>	<p><a href="#">Continue »</a></p> <p><a href="#">« Review</a></p> <p><a href="#">Challenge</a></p>
Circles and Arcs		<p>Mastery Level</p> <div><div>0</div></div> <p>Problems Done : <b>0</b> Total Problems : 16 <a href="#">Learn More &gt;</a></p>	<p>Untried topic- Would you like to try this topic now?</p> <p><a href="#">Comment &gt;</a></p>	<p><a href="#">Try this »</a></p>

(Muldner et al., 2015)

## Skill Meter

<input type="checkbox"/> [ 提取 信息 大小 值 ]	Extract an embedded list
<input type="checkbox"/> [ 且 且 且 且 且 且 ]	Extract info from an embedded list
<input type="checkbox"/> [ 提取 且 ]	Extract info from a list
<input type="checkbox"/> [ 并 并 并 并 并 并 并 ]	Deleting an extra node from the parameter list
<input type="checkbox"/> [ 且 并 并 且 子 ]	Coding a variable
✓ <input type="checkbox"/> [ 子 且 并 且 子 子 且 并 并 ]	Declaring a function parameter
✓ <input type="checkbox"/> [ 并 并 并 并 并 并 并 并 ]	Coding a function name
✓ <input type="checkbox"/> [ 并 并 并 并 子 并 并 并 ]	Coding DEFUN
✓ <input type="checkbox"/> [ 并 并 并 并 子 并 并 并 ]	Remove N Items
✓ <input type="checkbox"/> [ 并 并 并 并 子 并 并 并 ]	Skip over Items
✓ <input type="checkbox"/> [ 并 并 并 并 子 并 并 并 ]	Work From the Back of the List
✓ <input type="checkbox"/> [ 并 并 并 并 子 并 并 并 ]	Extract the Last Item
✓ <input type="checkbox"/> [ 且 且 且 且 子 并 并 并 ]	Extract the Nth Item
✓ <input type="checkbox"/> [ 并 并 并 并 子 并 并 并 ]	Coding LIST - embedded lists involved

# Methodology



# Methodology

- Search Method
- Inclusion Criteria
- Code book & coding
- Compare to LAD review (Bodily, 2017)

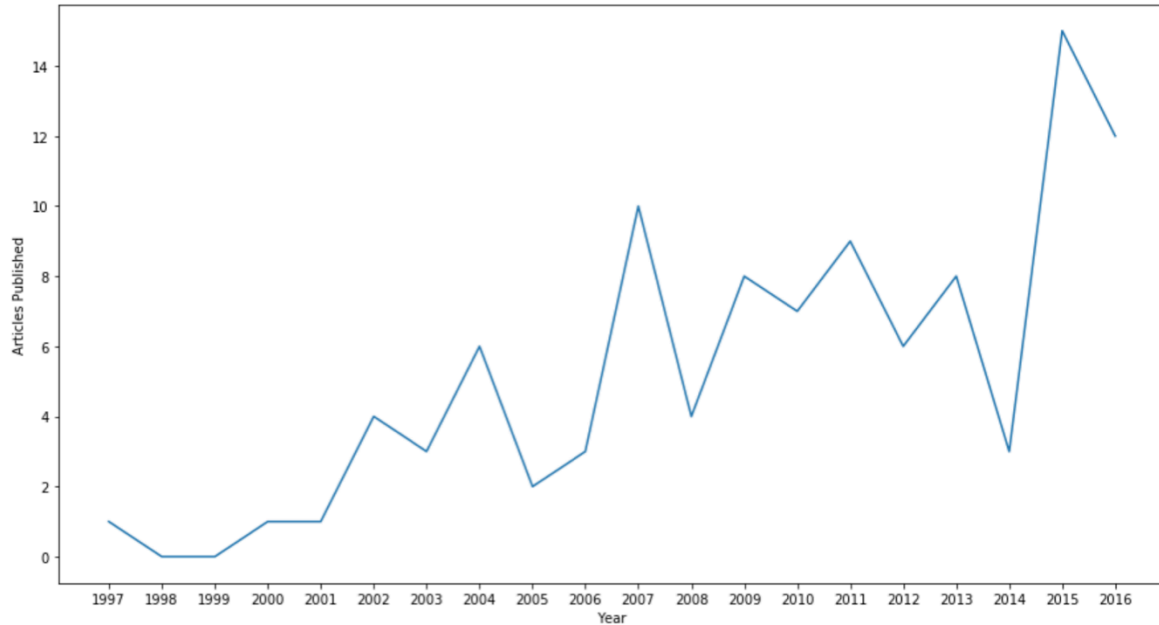
# Key Findings & Highlights

# Trends in OLM Research

- Publication venue
- Authors
- Top cited articles
- Publications over time

# Trends in OLM Research

- Publication venues
  - small overlap between LAD and OLM
- LAD review papers to date have not purposely included OLM research in their inclusion criteria




# Central Themes in OLM

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## Keywords


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intelligent tutoring systems

learning analytics 

self-regulated learning


self-assessment

learner model 


reflection


visualization/visualisation

intelligent tutoring system

user trust 

learner independence

 OLM community is more aware of LA community than vice versa

 SRL and reflection are a focus -> suggesting the purpose of opening the model to the learner.

 Inspectable or negotiated uncommon in LAD

# Central Themes in OLM

Abstract

paper

system

approach

study

results


social 

support

based

knowledge

learners

adaptive 

 Rise of OSSM

 intent to personalise or adapt instruction to learners

# OLM Data & Modeling



# OLM Data & Modeling

Category	# of OLMs	% of OLMs
Single type of data	62	57.9%
Behavioral Metrics	35	32.7%
Multiple applications	6	5.6%
Input provided by the user	42	39.3%
Complex Modelling	40	37.2%

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# OLM Evaluations

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Category	# of OLMs	% of OLMs
Authentic evaluation	42	39.3%
Evaluation	80	74.8%
Multiple evaluations	11	10.3%
Formal domain	53	49.5%
Tertiary education	58	54.2%
Secondary education	12	11.2%

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👉 Authentic evaluation	42	39.3%
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# Comparing OLM & LAD

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Category	LAD	OLM
Evaluation percentage	59%	75% 
Behavioral metrics	 75%	33%
Assessment data	37%	100% 
Comparison	38%	52% 
Interactive	31%	81% 

# Comparing OLM & LAD

- **Evaluation:**
  - OLM is a more mature field (1997 vs 2011)
- **Data:**
  - Tracking activity traces of learners is indeed at the core of learning analytics;
  - assessment data is not a prerequisite for useful dashboards
  - visualizing assessment can provide a solid foundation for learning analytics dashboards to support student retention
- **Comparison:**
  - OLM used for interpretation and reflection by students - interesting point to adopt in LAD
- **Interaction:**
  - LAD belief that a dashboard is a single screen of important information, understood at a glance.
  - Two shortcomings:
    - Trust & negotiation
    - Lack of user input



# Comparing OLM & LAD

Category	LAD	OLM
Behavioral metrics	✓75%	33%
Assessment data	37%	100%✓

# OLMS versus learner facing learning analytics

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# Limitations

- Only include articles that introduce a new OLM
- Search sources and searching in titles, abstracts and keywords: missing relevant literature
  - Expert checks and adding articles based on the most prominent authors in the field
- Comparison with LAD: using one review that covered Jan 2005 - June 2016.

Implications for the future



# Recommendations

- Unify terminology for LAD and OLM
- Learn from each other
  - Eg, develop LAD in parallel with platform