

Learning Analytics

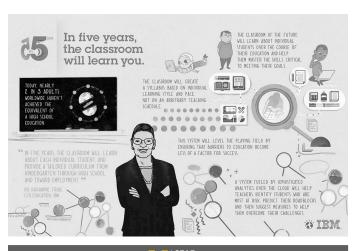
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Learning Analytics

...is the <u>analysis</u> and <u>use</u> of data, generated through normal administrative or academic activities, for the purposes of <u>improving</u> <u>teaching and learning</u>.













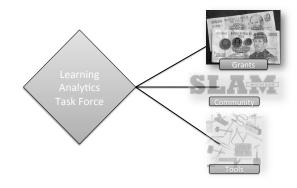




Learning Analytics at U-M

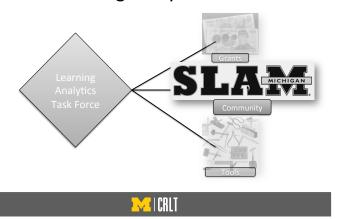


Learning Analytics at U-M

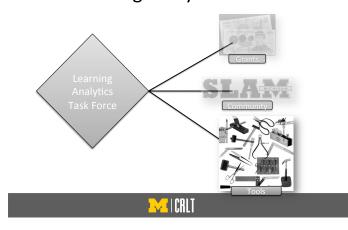




Learning Analytics at U-M



Learning Analytics at U-M



Case Study: Screencasting in Materials Science & Engineering



Joanna Mirecki Millunchick Professor of Materials Science & Engineering

Pedagogical Intervention: Supplemental Videos

Is it worth it?

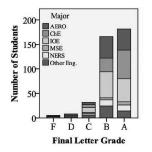


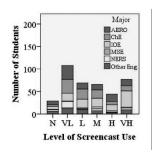


Data Sources

- Student perceptions
- · Learning management system data
 - How many times videos were watched
 - Who watched videos
- Student grades

Screencast usage was positively and significantly correlated to overall performance in the course

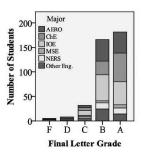


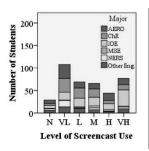




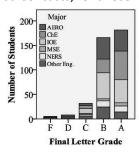


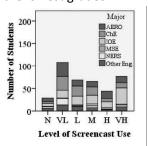
ChE use screencasts at lowest levels, while receiving the highest grades





IOE use the screencasts the most, and receive average grades. Prior to the introduction of screencasts; IOEs received the lowest grades.

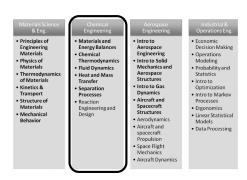




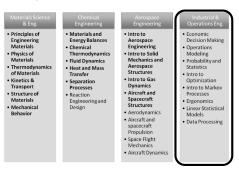




The ChE Curriculum covers many MSE topics

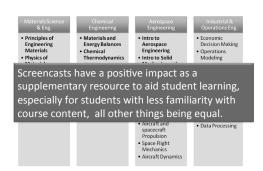


The IOE curriculum does not cover any material science topics









Brainstorm

Get into small groups:

What do you wish you knew about your students' behaviors?





Student Explorer

An Early Warning System for Academic Advisors

Steven Lonn - Project Lead - slonn@umich.edu

Stephen Aguilar - Graduate Student Lead - aguilars@umich.edu

Stephanie Teasley - USE Lab Director - steasley@umich.edu



What is an Early Warning System?

...utilize historical and current educational data in order to determine, in near real time, which students might be at risk of academic failure... EWSs produce "actionable intelligence" that the user can utilize to direct students toward resources or changes in behavior in a timely manner (Arnold, 2010).

Unlike some other EWSs, Student Explorer uses a categorization scheme using live data and does *not* currently use a predictive model for student success.



Data Sources

1. Performance Data

Learning Management System Gradebook Learning Management System Assignments tool Math web homework/testing system

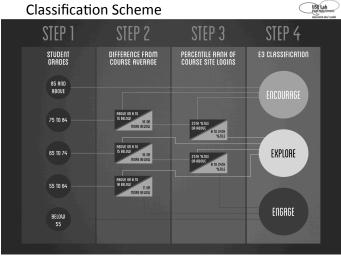
2. Usage Data (Proxy for "Effort")

Learning Management System login data

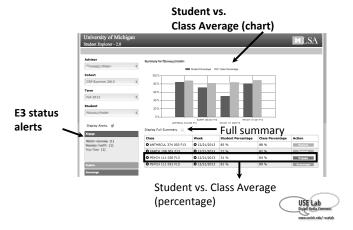
(In development: Other forms of usage data, e.g., file downloads)



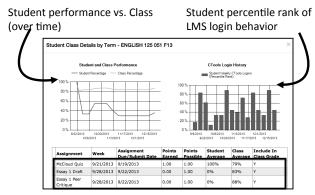
Classification Scheme



Summary Screen

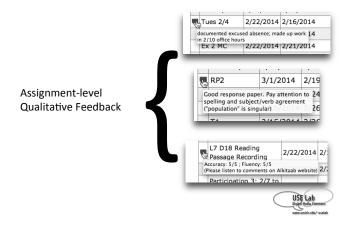


Course Detail



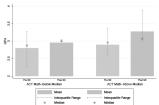
Performance on individual assignments with qualitative instructor feedback

Comments



Does it Work?

 Measured sophomore GPA gains for M-ENGIN population after introduction of Student Explorer.



Currently investigating impact of Student Explorer usage during SY 2013-2014 for M-ENGIN, M-BIO, and CSP Summer Bridge.

Reports / Findings

Lonn, S., Aguilar, S. J., & Teasley, S. D. (In Press). Investigating Student Motivation in the Context of a Learning Analytics Intervention During a Summer Bridge Program. Computers in Human Behavior. doi: 10.1016/j.chb.2014.07.013

Krumm, A. E., Waddington, R. J., Lonn, S., & Teasley, S. D. (2014). A learning management system-based early warning system for academic advising in undergraduate engineering. In (J. A. Larusson & B. White, 605, Learning Analytics: From Research to Practice (pp 103-119). New York: Springer. DOI 10.1007/978-1-4614-3305-7_6

Aguilar, S., Lonn, S., & Teasley, S. D. (2014). Perceptions and Use of an Early Warning System During a Higher Education Transition Program. Proceedings of the Fourth International Conference on Learning Analytics and Knowledge (pp. 113-117). Indianapolis, IN: ACM.

Lonn, S., Aguilar, S., & Teasley, S.D. (2013). Issues, challenges, and lessons learned when scaling up a learning analytics intervention. Paper presented at the 3rd Learning Analytics and Knowledge Conference Leuven, Belgium.

Lonn, S., Krumm, A.E., Waddington, R.J., & Teasley, S.D. (2012). Bridging the gap from knowledge to action: Putting analytics in the hands of academic advisors. Paper presented at the 2nd Learning Analytics and knowledge Conference. Vancouver, BC, Canada. Available in the Proceedings (184-187). doi: 10.1145/2330601.2330647

USE Lab

Krumm, A.E., Waddington, R.J., Lonn, S., & Teasley, S.D. (2012). Increasing Academic Success in Undergraduate Engineering Education using Learning Analytics: A Design-Based Research Project. Paper presented at the Annual Meeting of the American Educational Research Association. Vancouver, BC, Canado

Learning Analytics Readiness



Readiness

Ability and willingness to engage in an action

Readiness

Data

Ability

Governance/ Infrastructure

Culture/

Perception of Readiness

- Existence
- "Easily" matched from different sources

MI CRLT

CRLT

Readiness

Data

Ability

Governance/

Culture/

Perception of Readiness Skill with

- Databases
- Statistics
- Learning Theories

Data

Ability

Governance/ Infrastructure

Culture/ Process

Perception of

Readiness

- Policies that support access to data
- Policies that protect student privacy
- Leadership at multiple levels (administration, faculty, staff)

ICRLT

MI CRLT

Readiness

Data

Ability

Governance/ Infrastructure

Culture/ Process

Perception of

- Clear procedures for accessing data
- Support from those with ability
- Buy-in at all levels (student, faculty, staff, administration)

Readiness

Data

Ability

Governance/ Infrastructure

Culture/

Perception of





Learning Analytics Readiness

