







Using learning analytics to improve student transition into, and support throughout the Ist year

Workshop delivered in partnership by the ABLE Project 2015-1-BE-EPPKA3-PI-FORWARD STELA Project - 2015-1-UK01-KA203-013767



Co-funded by the Erasmus+ Programme of the European Union



• Who are we?

• Tinne De Laet

KU Leuven Head of Tutorial Services of Engineering Science promotor ABLE and STELA background = mechanical engineer

• Sven Charleer Doctoral researcher ABLE, KU Leuven

• Tom Broos Doctoral researcher STELA, KU Leuven



- Who are we?
 - Sarah Lawther

Learning and Teaching Officer, Nottingham Trent University

Rebecca Edwards

ABLE Project Officer, Nottingham Trent University

• Maartje van den Bogaard, ABLE project, Universiteit Leiden

• Jan-Paul van Staalduinen STELA project, TU Delft,

- Who is in the audience?
 - stakeholders:
 - tutors/student counsellors,
 - student support staff
 - academic administration staff
 - lecturer
 - policy makers, leader in education
 - researcher
 - learning analytics and institute
 - from institute with/without experience in learning analytics
 - from institute that uses learning analytics to support first-year students
 - what is your current attitude toward learning analytics
 - sceptic
 - doubting
 - enthusiast

- Would you please introduce yourself to the people on your table
 - Who are you?
 - Have you been to any EFYE conferences before?
 - Do you have any advice for first time attenders?

Workshop structure

- Welcome & first discussion
- What is learning analytics?
- First short exploration
- Learning analytics: the partners & the projects
- How can learning analytics support student transition into university?
- What strategies do we need to exploit the insights learning analytics provides?
- Feedback on project ideas
- Conclusion



no universally agreed definition

"the measurement, collection, analysis and reporting of data about learners and their contexts, for purposes of understanding and optimizing learning and the environments in which it occurs" [1]

"the process of developing actionable insights through problem definition and the application of statistical models and analysis against existing and/or simulated future data" [2]

[1] Learning and Academic Analytics, Siemens, G., 5 August 2011, <u>http://www.learninganalytics.net/?p=131</u>
[2] What is Analytics? Definition and Essential Characteristics, Vol. 1, No. 5. CETIS Analytics Series, Cooper, A., <u>http://publications.cetis.ac.uk/2012/521</u>

• no universally agreed definition

"learning analytics is about collecting traces that learners leave behind and using those traces to improve learning" [Erik Duval, 3]

† 12 March 2016



[3] *Learning Analytics and Educational Data Mining*, Erik Duval's Weblog, 30 January 2012, https://erikduval.wordpress.com/2012/01/30/learning-analytics-and-educational-data-mining/

How is learning analytics different from institutional data? [4]

• High-level figures:

provide an overview for internal and external reports; used for organisational planning purposes.

• Academic analytics:

figures on retention and success, used by the institution to assess performance.

• Educational data mining:

searching for patterns in the data.

• Learning analytics:

use of data, which may include 'big data', to provide actionable intelligence for learners and teachers. AND STUDENTS

[4] Learning analytics FAQs, Rebecca Ferguson, Slideshare, <u>http://www.slideshare.net/R3beccaF/learning-analytics-fa-qs</u>

Different level of analytics

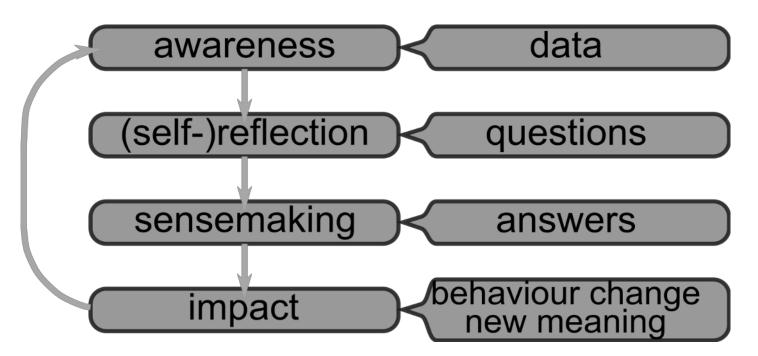
	level	beneficiaries
	course-level	learners, teachers, faculties
	aggregate	learners, teachers, tutors, counsellors, faculties
\langle	institutional	administrators, funders, marketing
	regional	administrators, funders, policy makers
	national and international	national and international governments, policy makers

adapted from http://www.slideshare.net/gsiemens/learning-analytics-educause

data visualization versus predictive analytics

- is showing data enough?
- how to show data to create sense-making and impact?
- is predicting study success/drop out the only thing that matters?
- can both be combined?

learning analytics process model



[Verbert et al. 2013] Verbert K, Duval E, Klerkx J; Govaerts S, Santos JL (2013) Learning analytics dashboard applications. American Behavioural Scientist, 10 pages. Published online February 201, doi: 10.1177/0002764213479363

how to evaluate learning analytics?

- is perceived usefulness enough?
- is increased self-awareness enough? How will you measure this?
- is increased sense-making enough? How will you measure this?
- how could impact be measured?

six critical dimensions of learning analytics

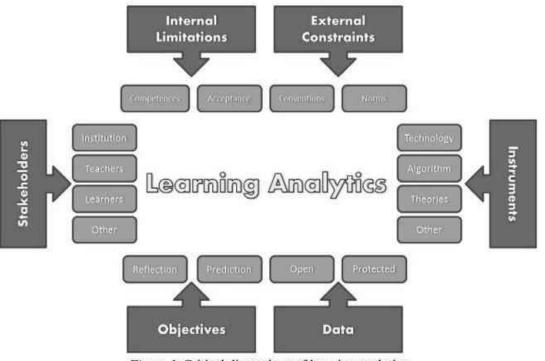


Figure 1. Critical dimensions of learning analytics

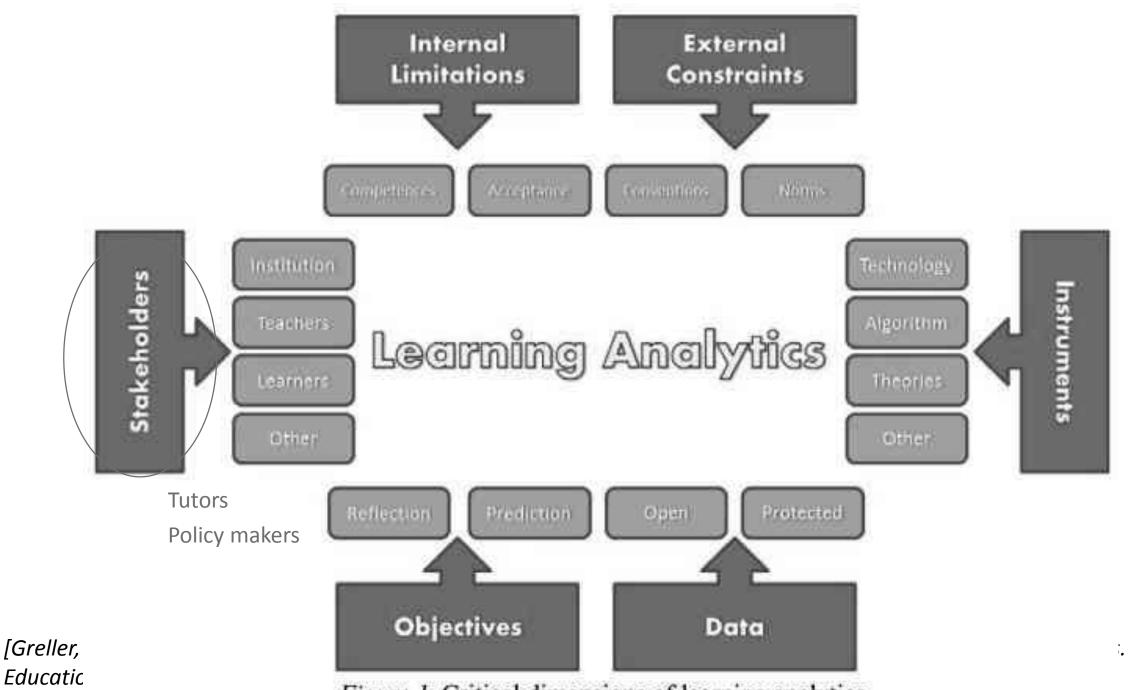
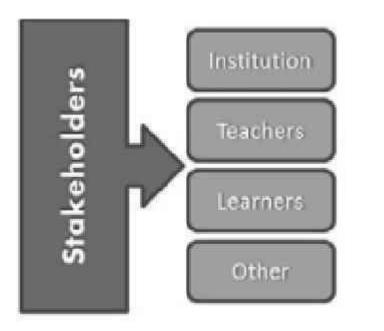


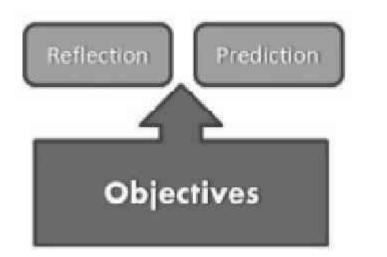
Figure 1. Critical dimensions of learning analytics



data subjects

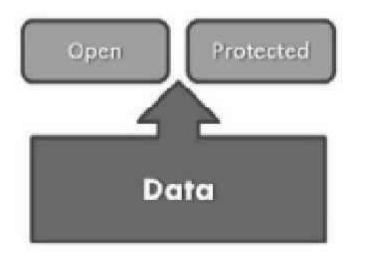
- here: students (could also be teachers)
- data clients
 - students
 - tutors
 - academic administrators
 - policy makers

what are the objectives?

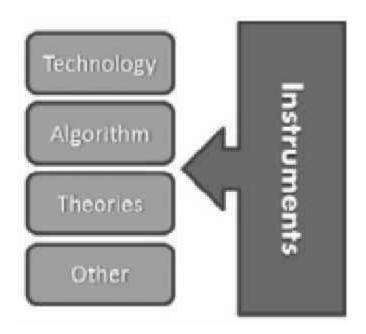


- is it about reflection or prediction?
 - is showing data enough?
 - how to show data to create self-awareness, sense-making and impact?

• which data is available?



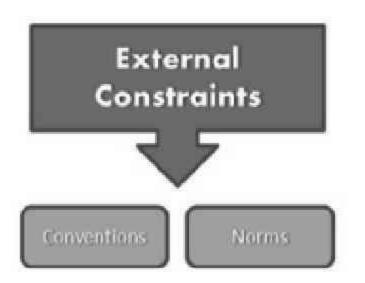
- is it open? protected?
- is it ethical to use the data?



• Technology, algorithms, theories are at the basis of learning analytics

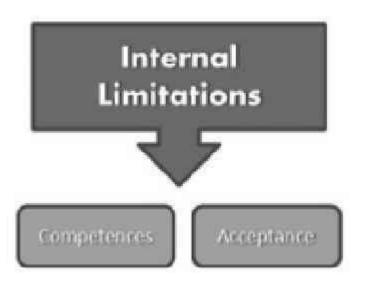
not the focus today

• **except:** pedagogic theories for supporting students



- **conventions:** ethics, personal privacy, and similar socially motivated limitations
- norms: restricted by laws or specific mandated policies or standard

ethics and privacy IS a big issue



- competences: application of learning analytics requires new higher-order competences to enable fruitful exploitation in learning and teaching
- acceptance: acceptance factors can further influence the application or decision making that follows an analytics process



Exploration

Exploration

now that you have some context and before we start spoiling you with our ideas and practices

If you have "carte blanche" for LA in supporting first year experience:

- which goals would you set for your learning analytics intervention?
- what kind of intervention?
- which stakeholders involved?
- how does it fit in different models presented?



Learning analytics at the partners

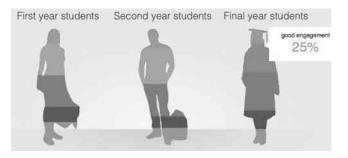
Nottingham Trent University Student Dashboard

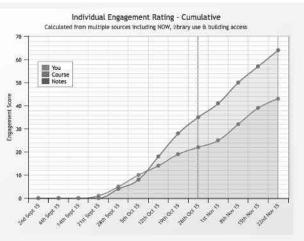


- Whole institution Dashboard
- Focus on engagement, not risk of failure
- Viewed by both students and staff
- Currently 4 key measures
 - Card swipes
 - VLE use & drop box submission
 - Library use
 - We are adding attendance & e-resources
- Strong association between engagement & both progression & attainment
- Challenges remain
 - particularly changing student outcomes

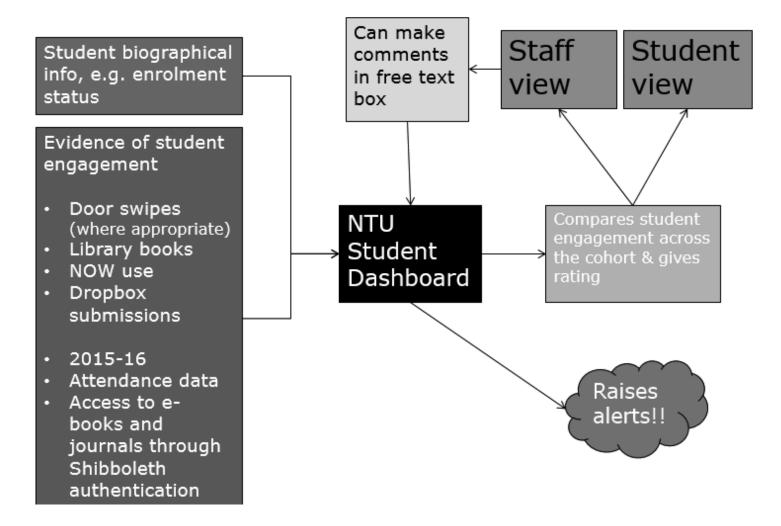


OUTSTANDING SUPPORT FOR STUDENTS

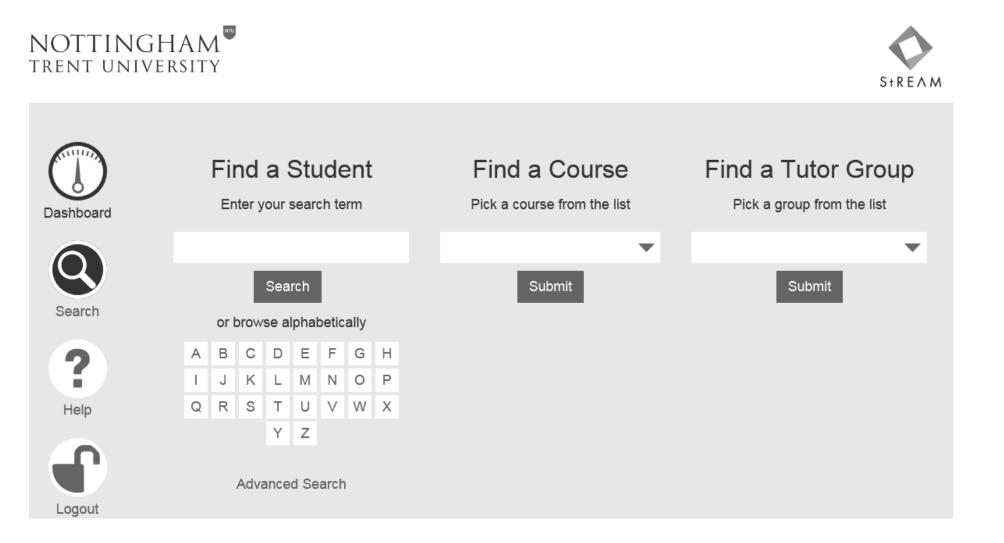




What does the NTU Student Dashboard do?



What does the Student Dashboard look like?



Leiden University

• Leiden University is in the South West of the Netherlands.



Universiteit Leiden The Netherlands

- It has seven faculties in the arts, sciences and social sciences, spread over locations in Leiden and The Hague.
- The University has over 5,500 staff members and 25,800 students.



Leiden University Learning analytics (1)

Matching between student and course

- Online questionnaires on variables pertaining to success in a particular course
- Automated feedback
- Invitations to students at risk to discuss their decisions and needs
- Information is uploaded in central student database and available to student counsellors at all times
- Follow up options for interventions are currently explored

Leiden University Learning Analytics (2)

However:

- The computer systems are not advanced: systems are not linked (yet)
- Privacy regulations are interpreted the conservative way

But:

• Thanks to the experiences with MOOCs and SPOCs awareness of potential benefits beyond the Matching initiative is growing.

Learning Analytics @ **TUDelft**

- No full-scale LA implementation yet in campus digital learning environment
- Web Information Systems research group and Extension School take active interest in LA
- MOOC platform (edX) provides an opportunity to gain valuable experience with LA



Building with Nature

_eadership fo

Next Generation Infrastructures

WROTE STATE STATE



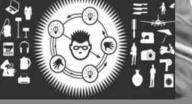
ng Water nent Functional Programming



Data Analysis for your Business

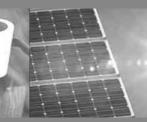


Sustainable Urban Development

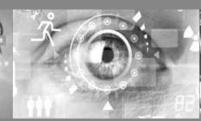


Delft Desigr Approach

Industrial Biotechnolog



Introduction to Solar Energy



Responsible Innovation



⁻reatment of Jrban Sewage



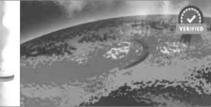
Circular Economy



F



Credit Risk Management



Introduction to Water & Climate Creative Problem

ative Problem /ing



Aeronautical

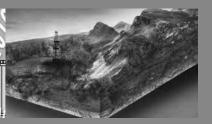
Basics of Transport Phenomena



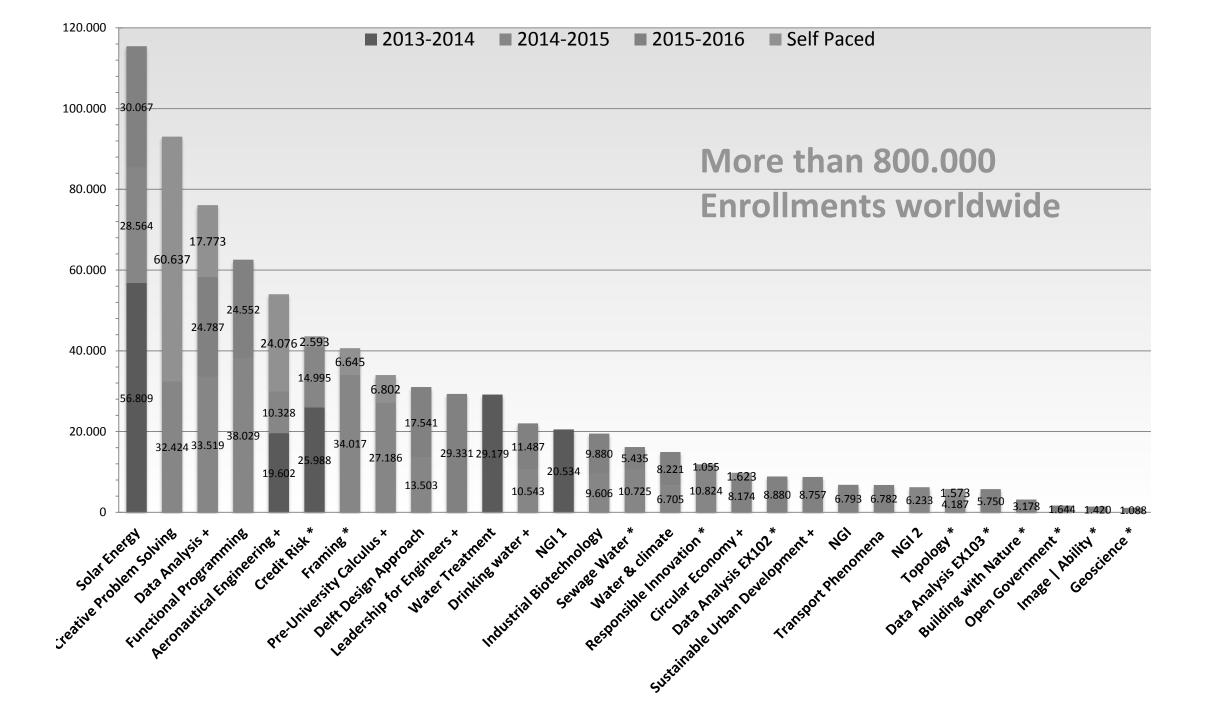
Topology in Condensed Matter



Open Government

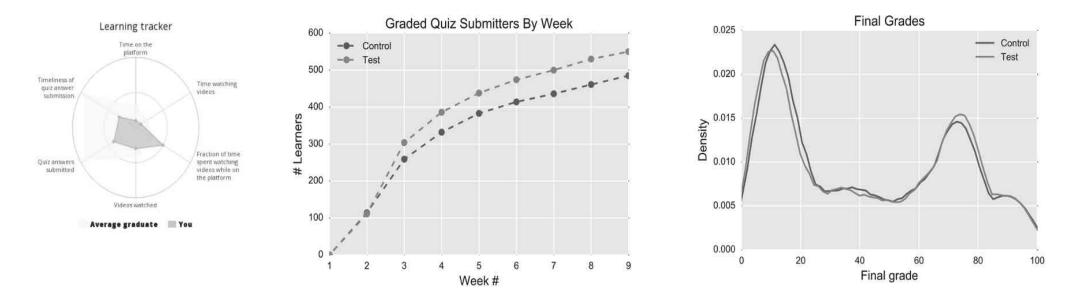


Geology: the earth and its resources



Learning Analytics for Learners

- Dashboard running in current Drinking Water MOOC
- In design stages of second iteration (for April MOOCs)

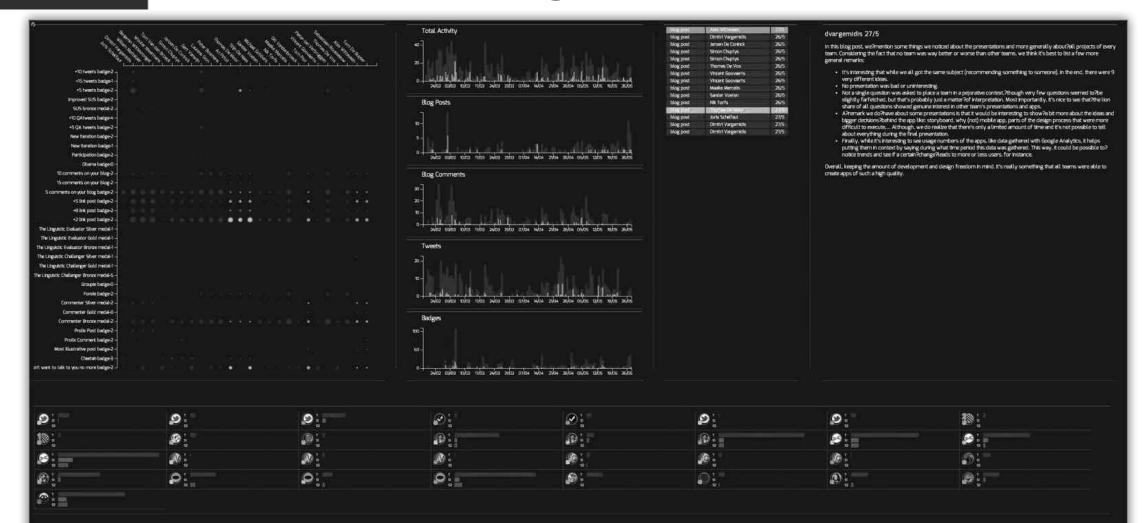


- Demonstration paper accepted at LAK 16 workshop

Daniel Davis, Guanliang Chen, Ioana Jivet, Claudia Hauff, Geert-Jan Houben. "Encouraging Metacognition & Self-Regulation in MOOCs through Increased Learner Feedback" *Learning Analytics for Learners workshop at Learning Analytics & Knowledge*. 2016.

KU LEUVEN

Learning dashboards: time & effort



Set 1		HatO4	#1504	Team LPE	Tem 504	Chi Mavez	Clenterest	Itaam TEV /	Chi Con Carra
- 340	601	Diratra Vargemetts	Tom Van den Brueck	Sert Vanwijn		Sender Voeten	Maalke Marcelis	Gill Vandenbroeck	10m Pint
847	Joris Schelfaut	Willem Mattalaer	Smon Chuptys	Laurens Sipo	Thomas De Molor	Michael Bobbers	Pieter-Jon Verbruggen	Vincent Goovaerts	Alex Witteveen
Sector 1	Wouter Moermans	Benjamin Wittevrorigel	Jeroen De Coninck	Feter Bosmans	Stip Devices	Nik Torfs	Sebastian Rousseave	Thomas De Vos	Tom De Buyser

Abstract to the essential





Charleer, S., Santos, J. L., Klerkx, J., & Duval, E. (2014). Improving teacher awareness through activity, badge and content visualizations. In *New Horizons in Web Based Learning* (pp. 143-152). Springer International Publishing.

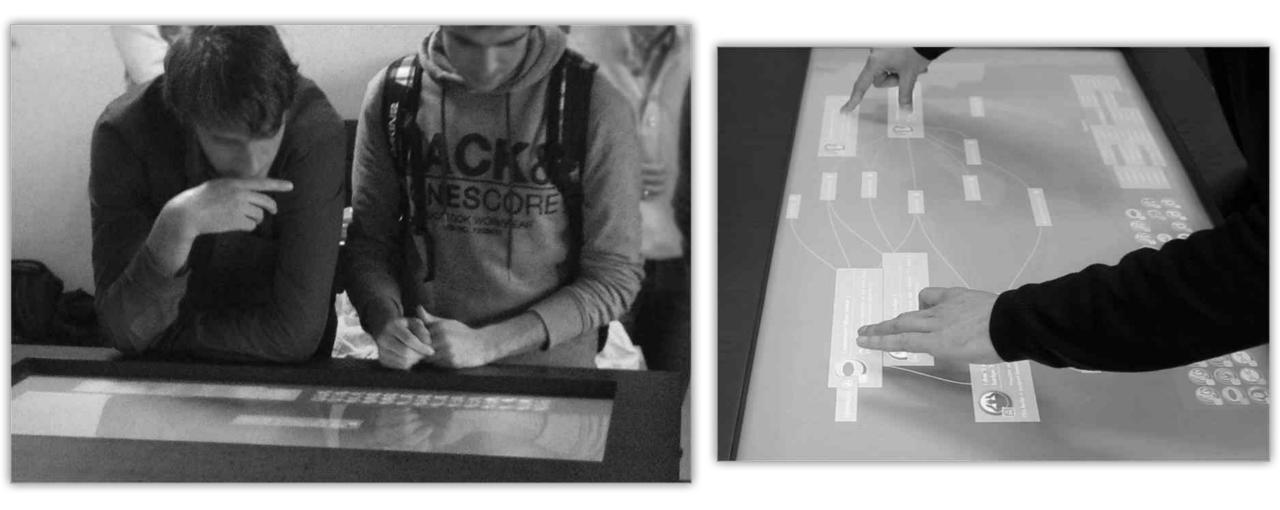
Integrate into the classroom



KU LEUVEN



Collaboration





Collaboration





Introduction to different case studies

Introduction

- there will be three case studies
- each group receives one case study
- each group will get 40 minutes to work on the case
- at the end present one striking result/challenge/question to everyone (2 min.)

Case I

Case II

Case III



Working on cases



Conclusion / Farewell