TRUSTED LEARNING ANALYTICS



ITCILO INNOVATIVE LEARNING INTERVENTIONS DIPLOMA PROGRAMME.

Learning Analytics in Portalen für Fachinformationen

Prof. Dr. Hendrik Drachsler

@hdrachsler





WhoAml

- Hendrik Drachsler
 Professor Educational Technologies
 & Learning Analytics
- Research topics:
 Recommender Systems
 Learning Analytics
- Application domains:
 Schools
 HEI
 Medical education



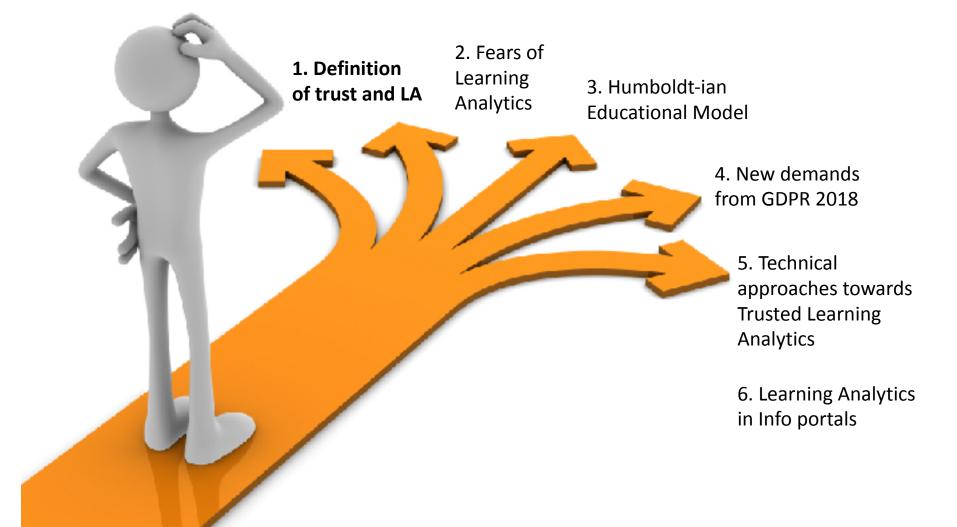








Lecture structure



A definition of Trust



How do you define trust?

Picture by Terry Johnston https://www.flickr.com/photos/powerbooktrance/466709245/

A definition of Trust



Trust is about a firm belief in the reliability, truth, or ability of someone or something.

A trustful relation is mutually based on

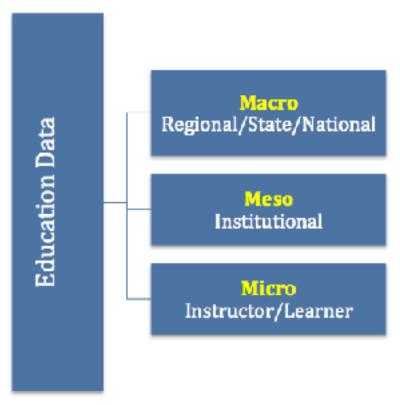
- openness
- truth
- reliability
- integrity
- belief
- faith
- freedom of suspicion

Picture by Terry Johnston

https://www.flickr.com/photos/powerbooktrance/466709245/

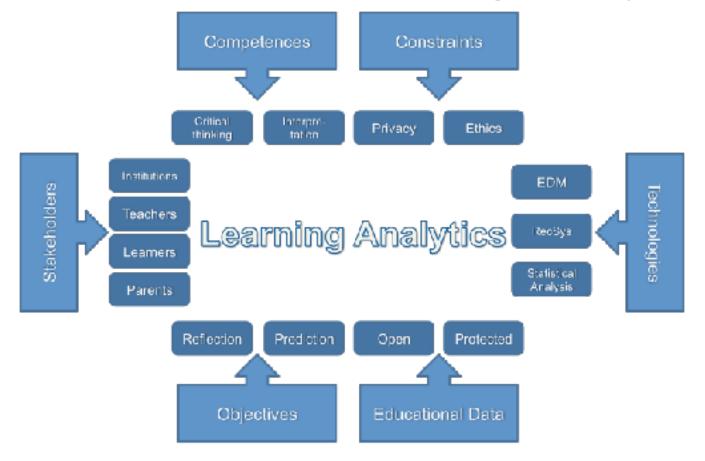
What are Learning Analytics for you?

Learning Analytics

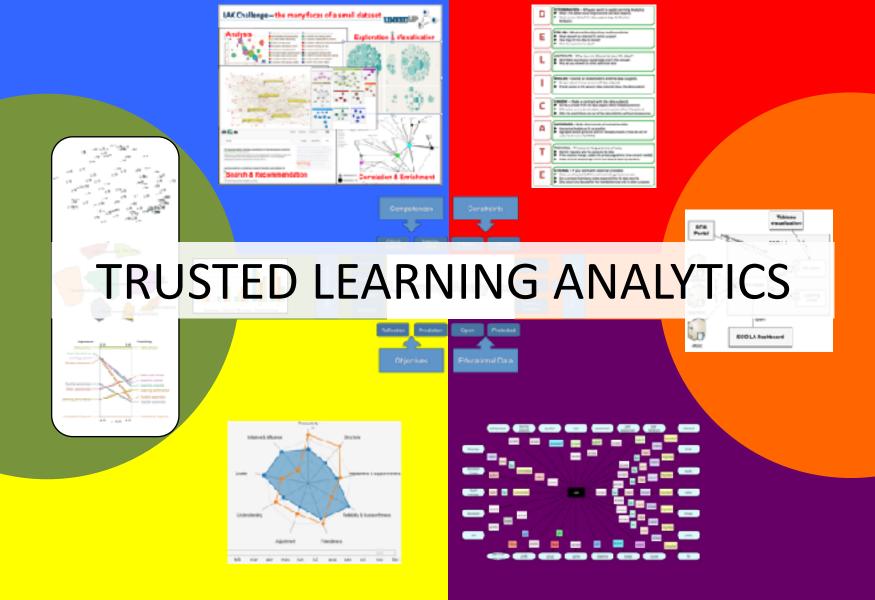


Greller, W. & Drachsler, H. (2012). **Turning Learning into Numbers**. **Toward a Generic Framework for Learning Analytics**. Journal of Educational Technology & Society. http://ifets.info/journals/15_3/4.pdf

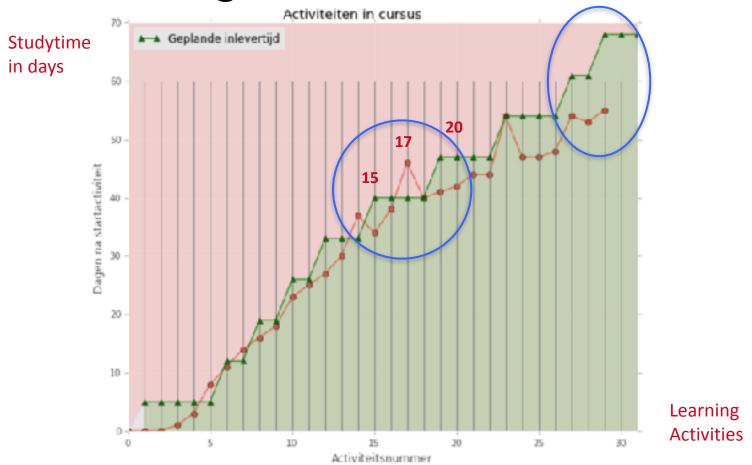
A definition of Learning Analytics



Greller, W. & Drachsler, H. (2012). Turning Learning into Numbers. Toward a Generic Framework for Learning Analytics. Journal of Educational Technology & Society, 15(3), 42–57. http://ifets.info/journals/15/3/4.pdf

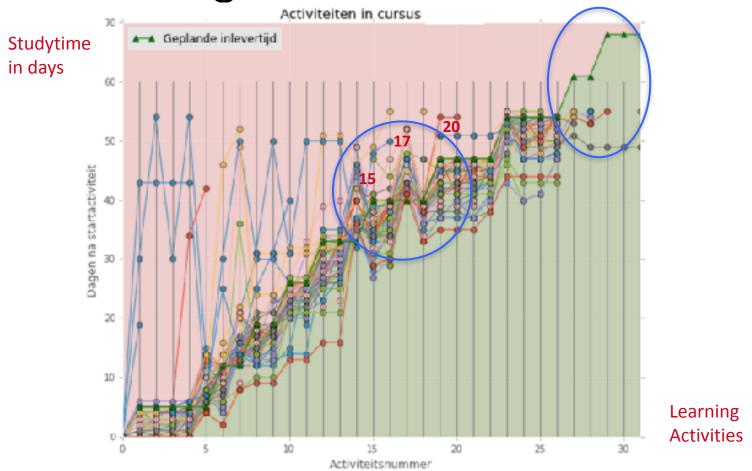


New insights



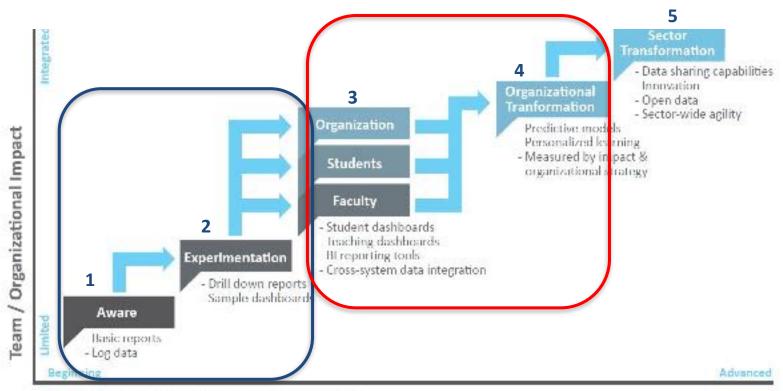
Graph by Rob Koper. **Data science voor de realisatie van online activerend onderwijs**. Presentation given at Dag van het Onderwijs (5 November 2015). Heerlen. The Netherlands

New insights



Graph by Rob Koper. **Data science voor de realisatie van online activerend onderwijs**. Presentation given at Dag van het Onderwijs (5 November 2015). Heerlen. The Netherlands

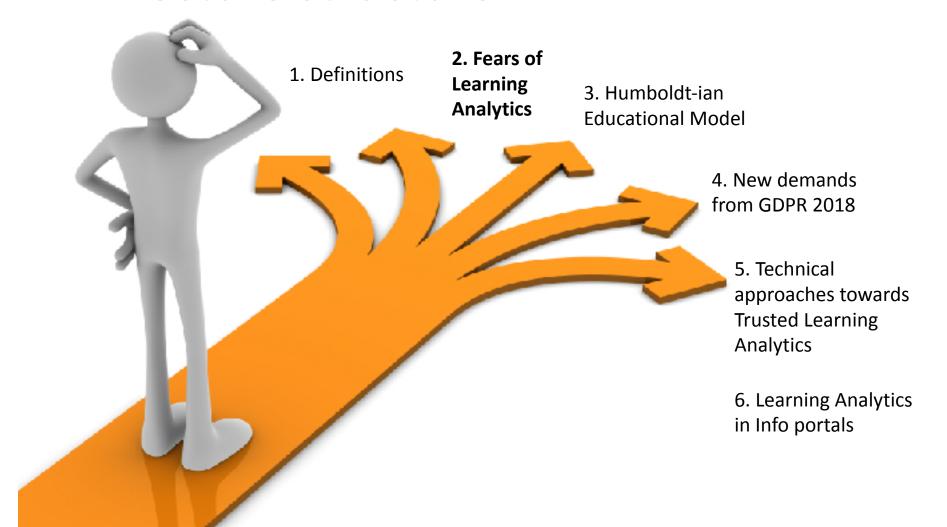
LA Sophistication Model



Maturity of Learning Analytics Deployment

Siemens, G., Dawson, S., & Lynch, G. (2014). Improving the Quality and Productivity of the Higher Education Sector – Policy and Strategy for Systems-Level Deployment of Learning Analytics. Canberra, Australia: Office of Learning and Teaching, Australian Government. Retrieved from http://solaresearch.org/Policy_Strategy_Analytics.pdf

Lecture structure



Do you have any concerns when you think about Learning Analytics in K12 or HEI?

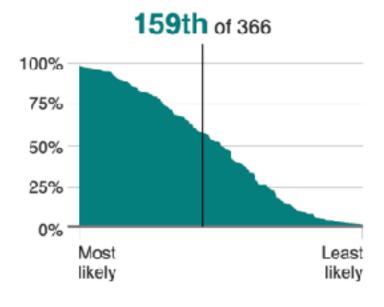
People are afraid of AI and digital technologies

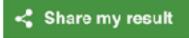


Taxi and cab drivers and chauffeurs

Likelihood of automation?
It's too close to call (57%)

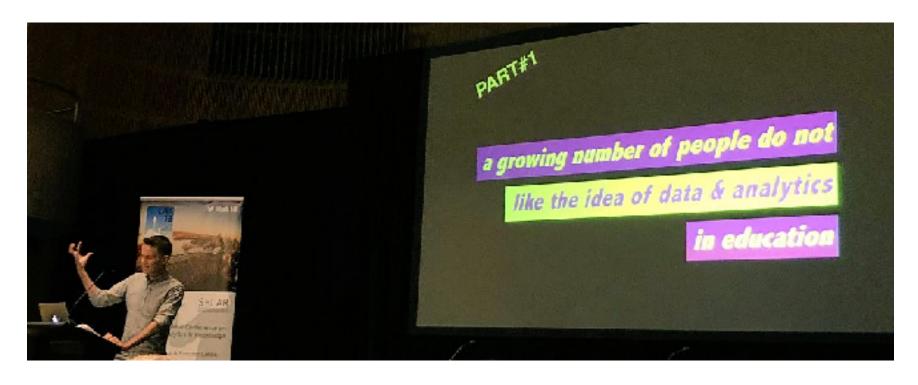
How this compares with other jobs:







Keynote Neil Selwyn @ LAK 2018, Sydney, Australia

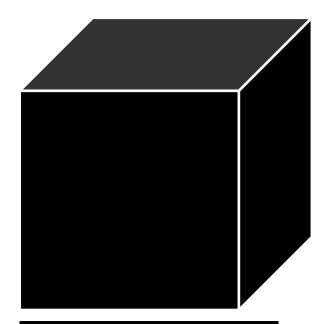


Learning Analytics has a trust problem ...

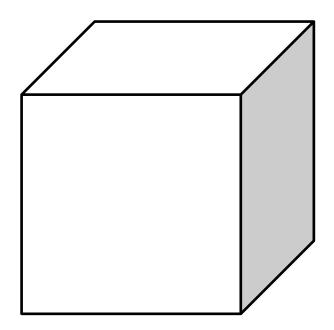
Keynote Neil Selwyn @ LAK 2018, Sydney, Australia



Black box vs. White box

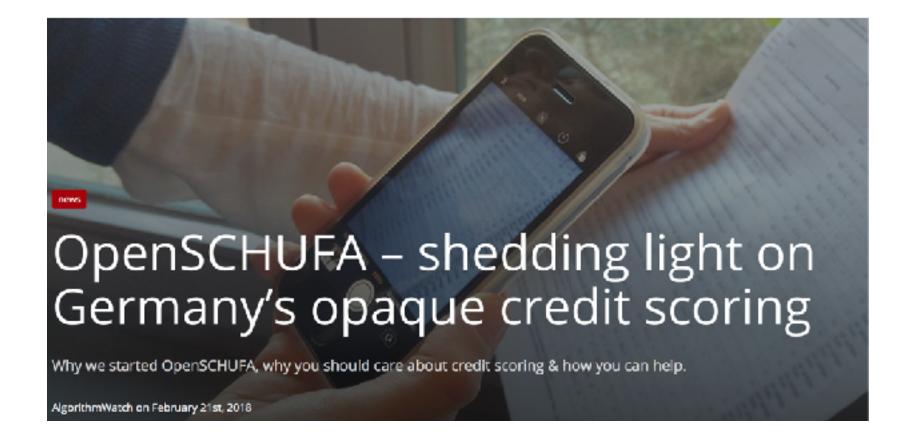


Unknown algorithms
Unknown data collection
Automated decisions
No access to raw data
No control who uses it



Open algorithms
Transparent indicators
No automated decisions
Full access to data
Knowing who accesses your data





Educational Example

BIG DATA

InBloom Student Data Repository to Close

By NATASHA SINGER APRIL 21, 2014 1:21 PM 4 24 Comments



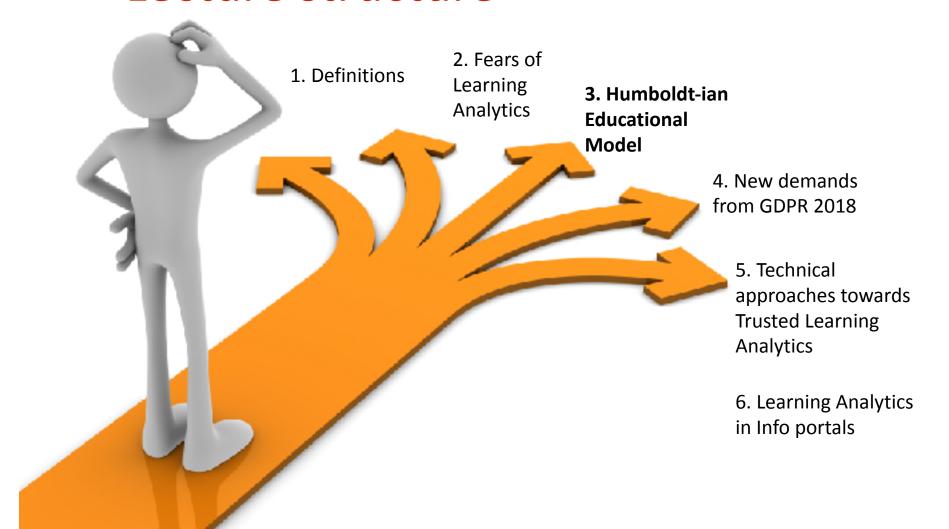
New York State is planning to share your child's confidential information with private corporations

New York State has agreed to share confidential student and teacher data with a Gates-funded corporation called inBloom Inc.

Ignoring the fears and public perception of the application of analytics can lead to a lack of acceptance, protests, and even failure of entire LA implementations.

have been stored

Lecture structure



Is the German Education system prepared for the

challenges of the data-driven society?

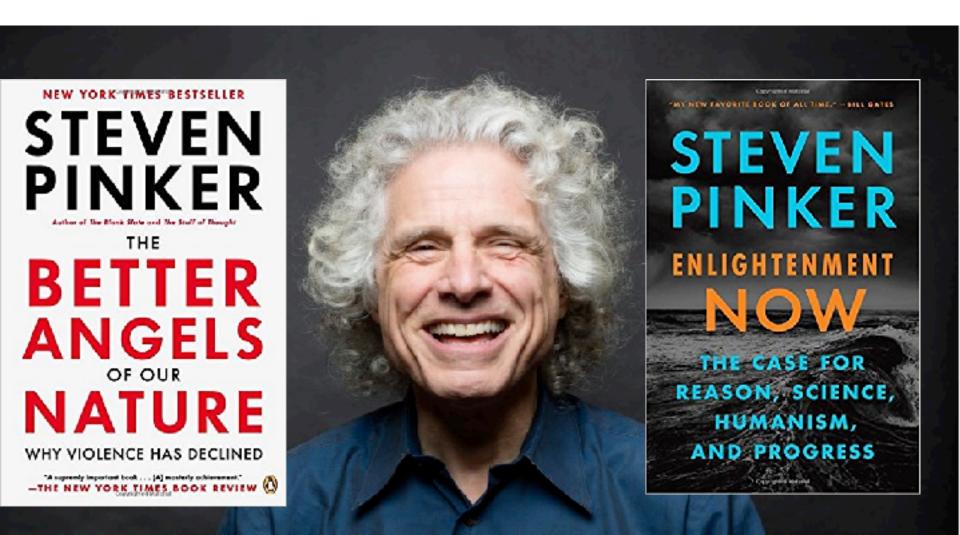
Humboldt-ian Model in the age of Big Data



Designed an education system ...

- based on unbiased knowledge (objective analytics but also critical reflection)
- combining research and teaching
- allowing students to choose their own course of study (personalization)
- develop autonomous reflected individuals (self-regulated learning)
- education beyond vocational training only
- comprehensive general learning (lifelong learning)
- cultural knowledge (formative feedback vs. summative assessment)

Humanismus & Enlightment NOW



Julian Nida-Rümelin Nathalie Weidenfeld

DIGITALER HUMANISMUS

Eine Ethik für das Zeitalter der künstlichen Intelligenz



Digitaler Humanismus

Julian Nida-Rümelin, Nathalie Weidenfeld

Eine Ethik für das Zeitalter der künstlichen Intelligenz



Download Cover



Sachbuch



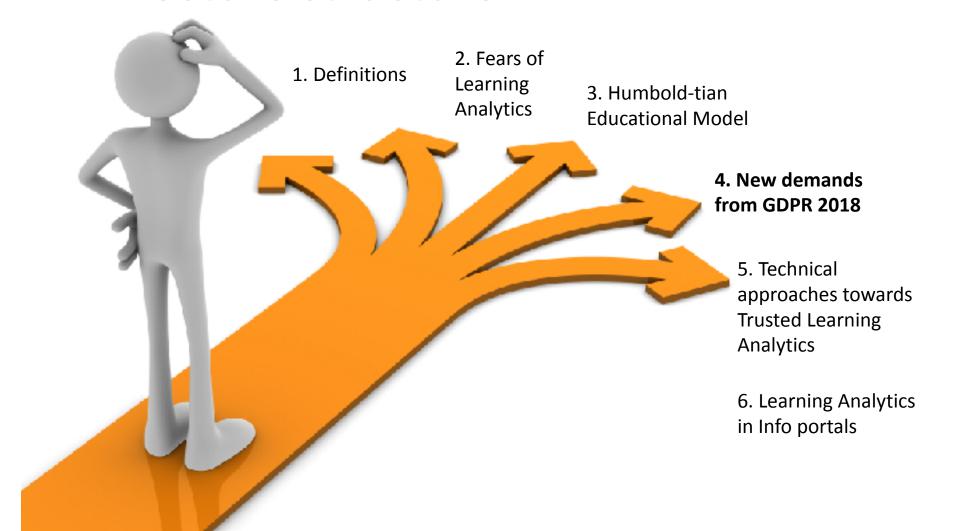








Lecture structure



GDPR 2018

The EU General Data Protection Regulation (GDPR) is the most important change in data privacy regulation in 20 years - we're here to make sure you're prepared.

"the biggest change to data protection law for a generation" Elizabeth Denham, Information Commissioner



GDPR 2018

- Right to be informed
- Right of access
- Right to rectification
- Right to erasure
- Right to restrict processing
- Right to data portability
- Right to object automated decision making

Do your Learning Technology systems support these rights?



Some things are already on its way

DETERMINATION – Why you want to apply Learning Analytics?

What is the addedvalue [Organisational and data subjects]?

what are the rights of the data subjects [e.g., Liu virective 95/46/EC]

EXPLAIN— Be open about your intentions and objectives

What data will be collected for which purpose?

How long will this data be stored?

Who has access to the data?

LEGITIMATE – Wity you are allowed to have the data?

➤ Which data sources you have aready (aren't theyenough)?

➤ Wity are you allowed to rollectadditional data?

INVOLVE – Involve all stakeholders and the data subjects

Be open about privacy concerns (of data subjects)

Provide access to the personal fata collected (about the data subjects)

Training and qualification of staff

COMSENT — Make a contract with the data subjects

Ask for a consent from the data subjects before the data collection

Define clear and understandable concent questions (Yes / No options)

Differ the possibility to opt-out of the data collection without consequences

ANONYMISE - Make the individual not retrievable

Anonymise the data as far as pessible

Aggregate data to generate abstract netadata models (Those do nor fall under EU Directive 95/46/EC)

TECHNICAL — Procedures to guarantee privacy

Monitor regularly who has access to the data

If the analytics change, update the privacy regulations (new consentneeded)

Make sure the data storage fulfills international security standards

EXTERNAL — If you work with external providers

Make sure they also fulfi the national and organizational rules

Sign a contract that clearly states responsibilities for data security

Data should only be used for the intended services and no other purposes

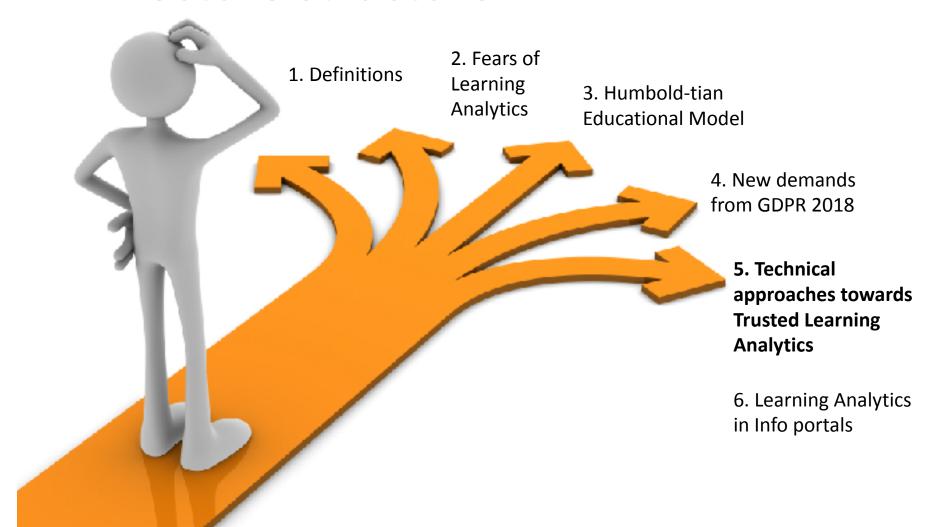
Drachsler, H. & Greller, W. (2016). Privacy and Analytics – it's a DELICATE issue. A Checklist to establish trusted Learning Analytics. 6th Learning Analytics and Knowledge Conference 2016, April 25-29, 2016, Edinburgh, UK.

Online at:

http://www.laceproject.eu/ethics-privacy/

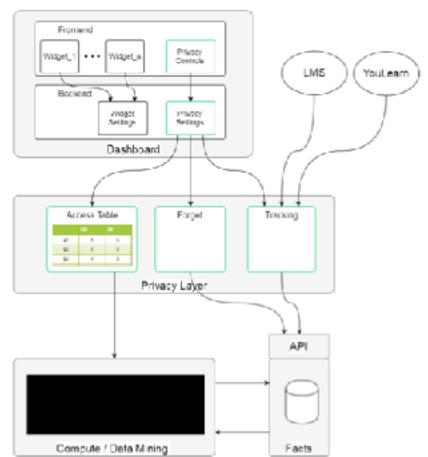


Lecture structure

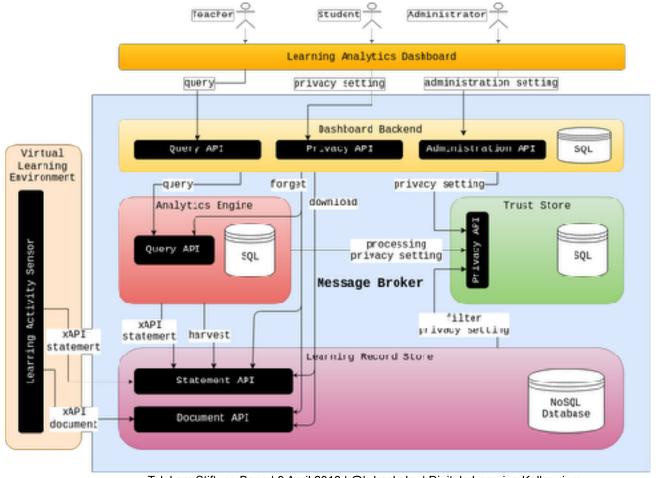


Trusted Learning Analytics Infrastructure

- T-4-LA first GDPR 2018 conform Big Data Infrastructure
- Joined project with GU und OU
- Among 'traditional' learning data we also aim to collect multimodal data.



Trusted Learning Analytics Infrastructure



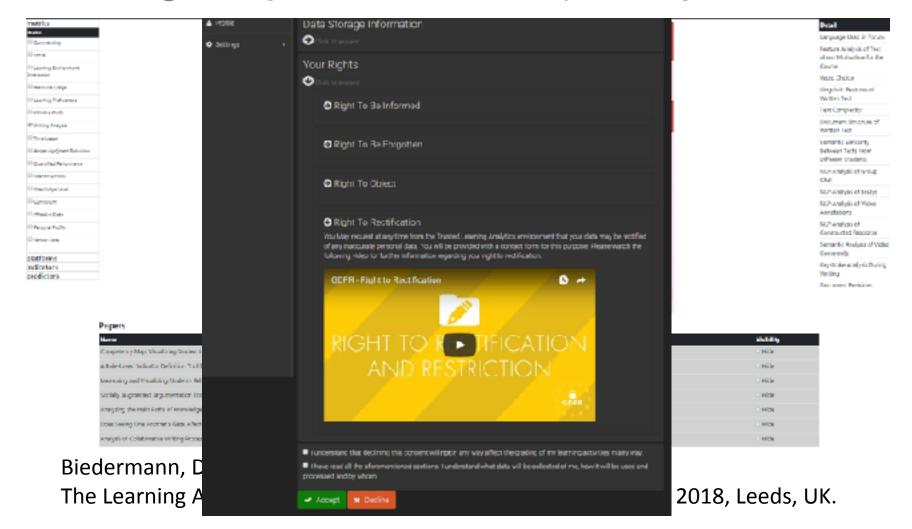
Ciordas-Hertel, G., Schneider, J., & Drachsler, H. (submitted). Towards a Trusted

Towards a Trusted Learning Analytics Big Data Infrastructure.

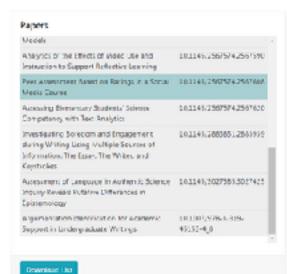
EC-TEL conference 2018, Leeds, UK.

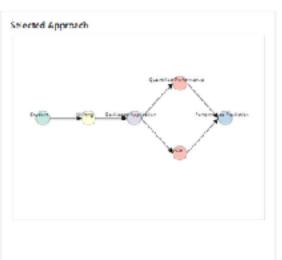
Telekom Stiftung Bonn | 9 April 2018 | @hdrachsler | Digitale Learning Kolloquium

Learning Analytics Indicator Repository

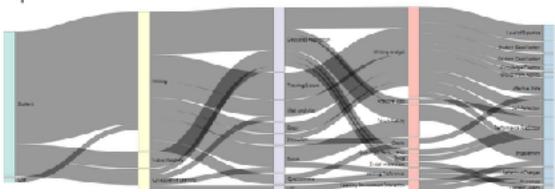


Learning Analytics Indicator Repository



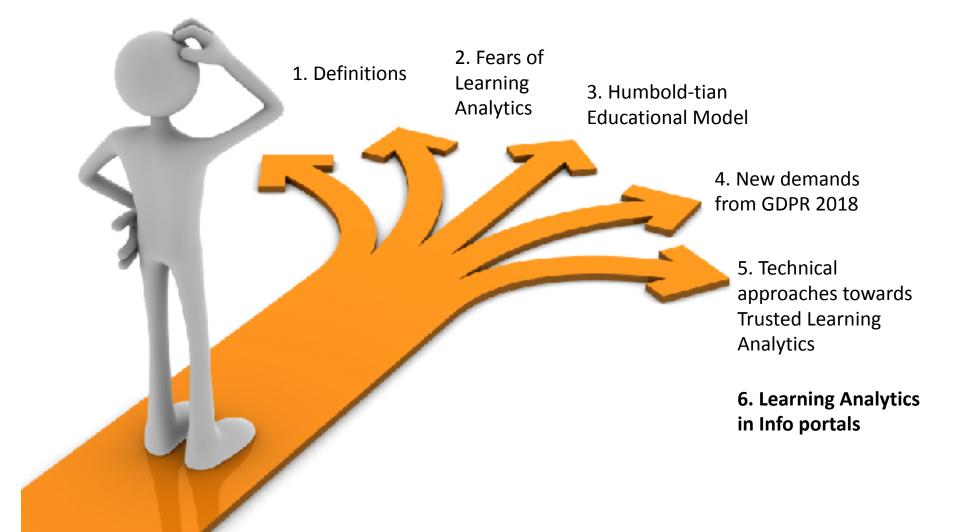


Graph Overview



Biedermann, D.,
Schneider, J., &
Drachsler, H.
(submitted).
The Learning Analytics
Indicator Repository. ECTEL conference 2018,
Leeds, UK.

Lecture structure





Learning Analytics in information portals

Educational Fingerprinting



Information portals and learning

- Information portals offer a broad variety of information
- Users interact with content to gain information / knowledge











Ihr Wegweiser zur Bildung



Erziehungswissenschaft Bildungsforschung Fachdidaktik



How can we faster adapt information portals to individual learners?

• Goal: User centric adaptive (open) educational ressources

• **Problem**: First time users without user account

• **Needed**: Previous Knowledge – without manual user input





User: Colin / Wikimedia Commons / CC BY-SA 4.0



Learner Modeling

Learner Modeling = Data Collection + Profile
 Construction (Brusilovsky)



- Current approaches:
 - Learner accounts
 - Ethernet or browser proxy
 - Browser plugin



Current Data Collection Problems

- Interpretation requires data of several sessions
- Accounts require learner login
- Learning Analysis Proxies or Browser Plugins must be set up manually

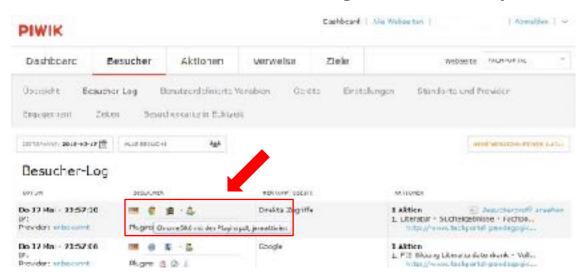
To experience adaptive content, learners have to be ...

- 1. active
- 2. continuous users
- 3. aware, that they can receive adaptive resources in the future



Fingerprinting

- Known web analytics technology for user identification
- Available ubiquitously
- Uses information of user configuration sent by the browser





Scientific Fingerprinting Projects

• Learners are unique due to their web fingerprint

Observed Uniqueness	Panopticlick	Am I Unique?	Cao et al.
N	470.161	118.934	1.903
Unique users	83,6%	89,0%	99,2%
Pub. year	2010	2016	2017



https://amiunique.org/

Your browser fingerprint appears to be unique among the 1,572,321 tested so far.

Currently, we estimate that your browser has a fingerprint that conveys at least 20.58 bits of identifying information.

The measurements we used to obtain this result are listed below. You can read more about our methodology, statistical results, and some defenses against fingerprinting here.

https://panopticlick.eff.org/



Example Fingerprint

Similarity ratio		Similarity ratio	
Attribute	Similarity ratio	Value	
User agent 6	<0.1%	"Mozilla/5.0 (Windows NT 6.1; WOW64; /60.0"	
∧ccept ⑤	54.42%	"text/html,application/xhtml+xml,applica	
Content encoding	46.22%	"gzip, deflate, br"	
Content language 1	3.91%	"de,en-US;q=0.7,en;q=0.3"	
List of plugins 1	<0.1%	"Plugin 0: Shockwave Flash; Shockwave F 40.dll."	



Experimental Design

• Research Question / Hypothesis:

- Can we provide personalised information to visitors of information portals based on their FP?
- Information portal visitors can be modeled according to their FP.

• Approach:

- Identify individual learning models by a questionnaire
- Assign Fingerprinting data to learner models
- Validate extracted correlations by a second user group

Possible Outcome:

 Different learning behaviour of users with advanced PCs vs. Users with outdated PCs.





Outlook / Vision

- Create a data driven "Bildungsindex" fed by the search terms and FPs of different users that visit an information portal for education.
- Continuous expert survey to generate an national "Ifo Geschäftsklima Index" for education



Many thanks for your attention!

Questions now or later:



@hdrachsler





Slides: http://bit.ly/TrustedLA









