





Southern African Association for Institutional Research

## Closing the loop!

(of Knowledge, Action and Intervention)





### Welcome

- 4the Meeting: Closing the Loop
- Aim: SAHELA
- Role of Stakeholders
- Disclaimer
- The road ahead





Southern African Association for Institutional Research

# A Framework for Learning Analytics





### What is Learning Analytics?

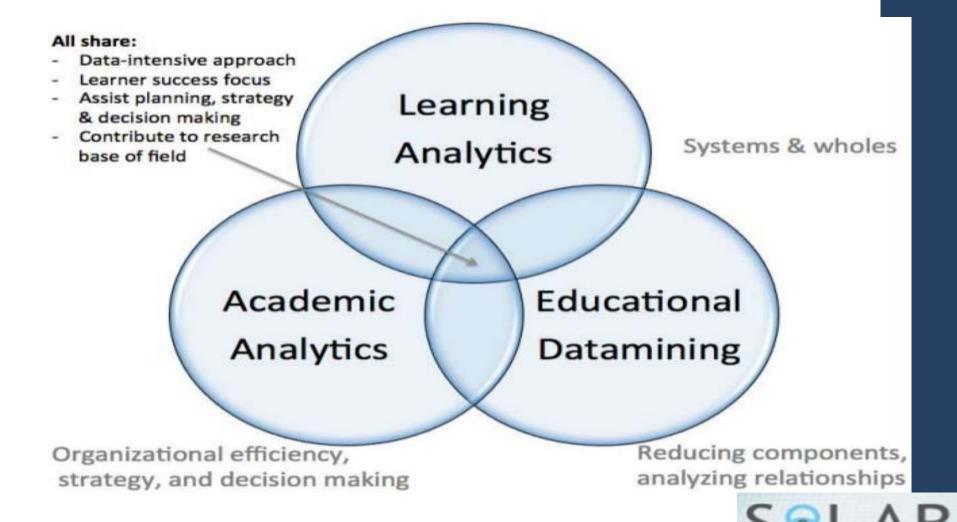
Learning Analytics and Knowledge Conference, 2011

Learning analytics is the measurement, collection, analysis and reporting of data about learners and their contexts, for the purpose of understanding and optimizing learning and the environments in which it occurs.

## Academic or Learning Analytics

Type of Analytics	Level or Object of Analysis	Who Benefits?
Learning Analytics	Course-level: social networks, conceptual development, discourse analysis, "intelligent curriculum"	Learners, faculty
	<b>Departmental</b> : predictive modeling, patterns of success/ failure	Learners, faculty
Academic Analytics	Institutional: learner profiles, performance of academics, knowledge flow	Administrators, funders, marketing
	Regional (state/provincial): comparisons between systems	Funders, administrators
	National and International	National governments education authorities

## Nexus of analytics and data mining



## Analytics in Higher Education

#### **Learning Analytics**

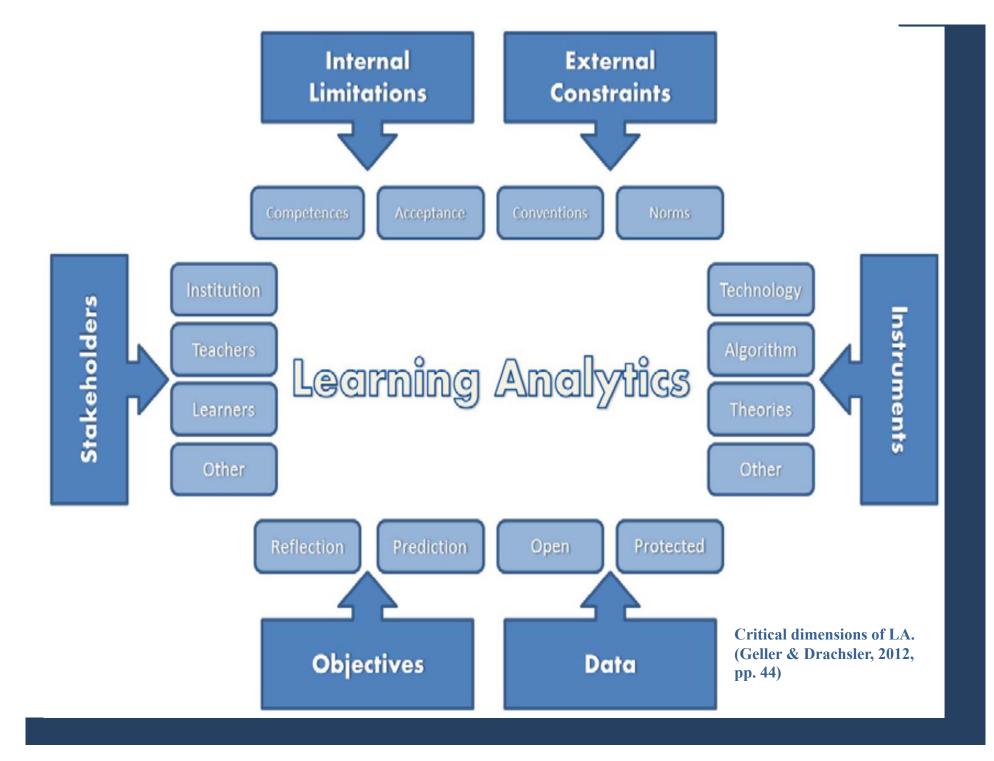
Best way to teach and learn

#### **Learner Analytics**

Best way to support students

#### **Action Analytics across an institution**

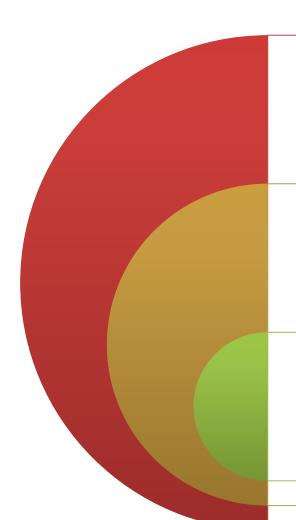
Best way to operate a college inclusive of functions including enrollment management, HR management, facilities management



#### Objectives of LA

- Reflection/ description of:
  - students based on normative data
  - lecturers reflecting on the learning process of their students
  - institutions evaluating and reflecting on groups of students
- Prediction suggests a statistical modelling of the data about the student learning process in order to provide support to students with similar characteristics to those of the target group modelled in the analysis.

## Levels of Analysis



## Descriptive Statistics

- Used for benchmarking and reporting
- Subjective interpretation
- Most common starting point

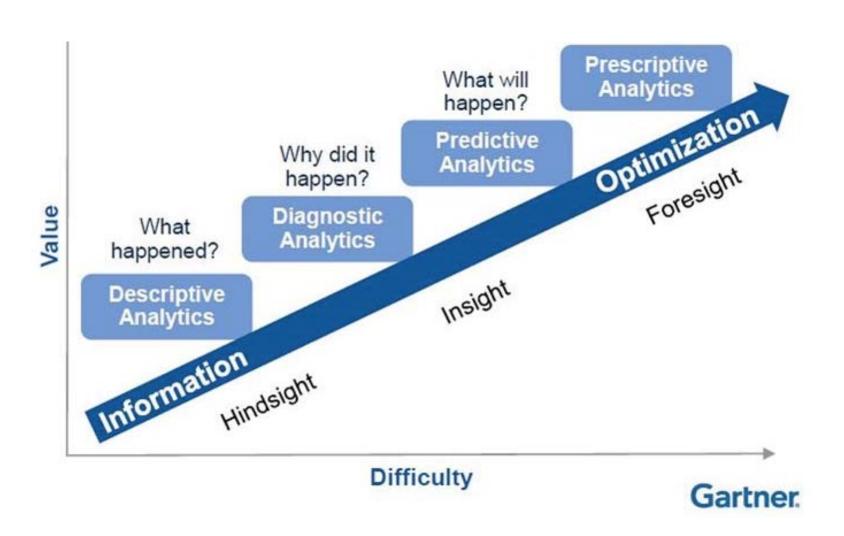
## Inferential Statistics

- Used to make judgments
- Lower confidence levels for prediction
- 9-20% of solutions

## Predictive Analytics

- Extract information to predict
- Used to detect patterns and predict future outcomes
- level for prediction
- 1-5% of solutions

#### From Hindsight to Foresight



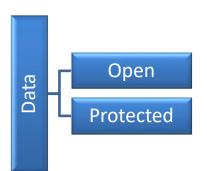
## Different Levels of Insight

#### Descriptive Analytics

- 1. How many logins, page views, and other metrics have occurred over time?
- 2. What were the course completion rates for a particular program over time? What were the attributes of the students who didn't successfully complete?
- 3. Which tools are being used in courses the most?

#### **Predictive Analytics**

- 1. Which students are exhibiting behaviors early in the semester which put them at risk for dropping or failing a course?
- 2. What is the predicted course completion rate for a particular program? Which students are currently at risk for completing and why?
- 3. Which tools and content in the course are directly correlated to student success?



#### Data for LA

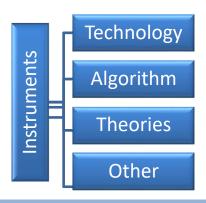
- Student record data from LMS and other sources.
   This is more than we can sense and goes beyond numbers and categorical labels to include things like text, images and video
- Mash-ups of data from different sources to facilitate learner-oriented services and personalisation
- LA data is protected or proxy access, which limits the evaluation of LA models - Anonymisation is one means of creating access

## Data dimension questions

- Who collects data?
- Are all the variables you need to answer your question(s) available? If not, do proxies exist?
- In what format?
- At what level is the data?
  - Per course
  - Per student
  - Per term
  - Per login to the LMS
- How reliable are the data?
- What other data are available unstructured?

## Data dimension questions

- What impact does ethical aspects have on the way data is collected and shared?
- Who receives this information?
  - Advisors
  - Other support service providers
  - Lecturers
- What do they do with it?
  - Contact the student with what messages?
  - Offer assistance what kind?



#### Instruments for LA

- The technology element refers to the analytical application of:
  - educational data mining
  - classical statistical techniques
  - data visualisation
- Statistical algorithms that are used by the technology applications to transform data into information
- The theoretical constructs, such as self-regulated learning, should be operationalised appropriately and underpinned by theory

## Statistical analysis

- Descriptive analysis
- Bayesian models
- Percentiles
- Cluster analysis
- Neural networks
- Regression analysis
- Decision Trees (e.g. CART)
- Other 'qualitative' data (survey, speech, text)

#### Multiple technology solutions in each application category

	Education Planning	Counseling and coaching	Risk Targeting and Intervention	Transfer and Articulation
Legacy ERP/SIS/LMS		pleSoft  Jenzaba	Blackboard  Desire	earning Technology
Vendor point solutions	connectedu° campuscrui campuscrui		Starfish retention solutions.  hobsons  empower  C CIVITAS	parchment- credentials unleashed Nuventive
Homegrown point solutions	Sinclair's Valencia's LifeMap  Austin Peay's Degree Compass	STUDENT SUCCESS PLAN  Central Piedmont's Online Student Profile	Signal Si	
Direct-to-student	connectedu education & career management my Edu		PERSISTENCE +PLUS	College Fransfer.Net*

## Higher Education Analytics Landscape

**Institutional Analytics (HR, Finance, Enrollment)** 

(Blackboard, COGNOS)

#### **Student Success Analytics**

"Retention CRM"
Analytics

(Starfish EARLY ALERT and Starfish CONNECT, Hobsons)

At-Risk Student
Predictive
Analytics

(Noel-Levitz, Mapworks, EAB, J. Gardner, Campus Labs, CIVITAS Learning) Instructional Analytics

(Knewton, MyLabs, ALEKS)

Program
Effectiveness
Analytics

(Starfish INSIGHT v.1)

Starfish INSIGHT v.2-v.3

## Instruments dimension questions

- Who analyses the data?
- What data are analysed?
- What assumptions are being made about the analysis/findings?
- What are the input variables and relation to outcome measures?
- What subgroups are used?
- What are the levels of analyses and why?
- What type of analytics is appropriate?

## Instruments dimension questions

- Which technologies are available for LA?
- Which type of analyses (algorithms) are appropriate to answer the strategic goals?
- Invest in "open" or "black box" analytics?
- What theoretical/conceptual frameworks are used?
- Does the analytics lead to actionable activities?

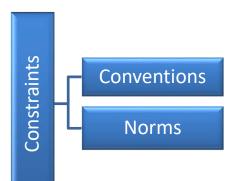


#### Internal limitations of LA

- The instruments and data sources are always used within the constraints of internal limitations such as:
  - the knowledge and analytical skills of the staff working with LA models and data
  - institutional culture and processes of change management operative in an institution
  - A reluctant or slow buy-in by stakeholders can also be an internal limitation.

## Limitations dimension questions

- What skills are needed to analyse/interpret and report on the LA data?
- To whom are reported?
- To whom should there be reported?
- How can you ensure that LA initiatives are accepted widely?
- Do users have access fixed or dynamic reports?
- What interventions are associated with LA?
- How are results fed forward in the action enquiry framework – monitoring and evaluation?

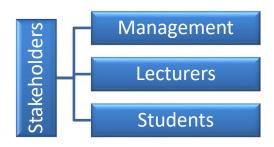


#### External constraints of LA

- The external constraints of LA focus on:
  - Conventions refer to ethics, personal privacy,
     and similar socially motivated limitations
  - Norms are restrictions imposed by laws or specific mandated policies or standards. These could refer to the institutions policies, practices, programs and processes

## External constraints questions

- How can inappropriate or incorrect interpretations of data be managed?
- How is personal and institutional privacy managed?
- What are the ethical implications of not acting on student data?
- How is POPI integrated into the LA initiative?
- How do we ensure that what we measure is what matters?
- What are the policies, processes or practices that guide LA initiatives?
- What impact does restricted reporting have for LA?



#### Stakeholders of LA

The Stakeholders are the focus of the workshop:

- Students
- Lecturers
- Institutional management

#### Handout: Strategic Plan

Activity (20 min)
University of Pretoria

List three key performance indicators that your institutions is focused on.

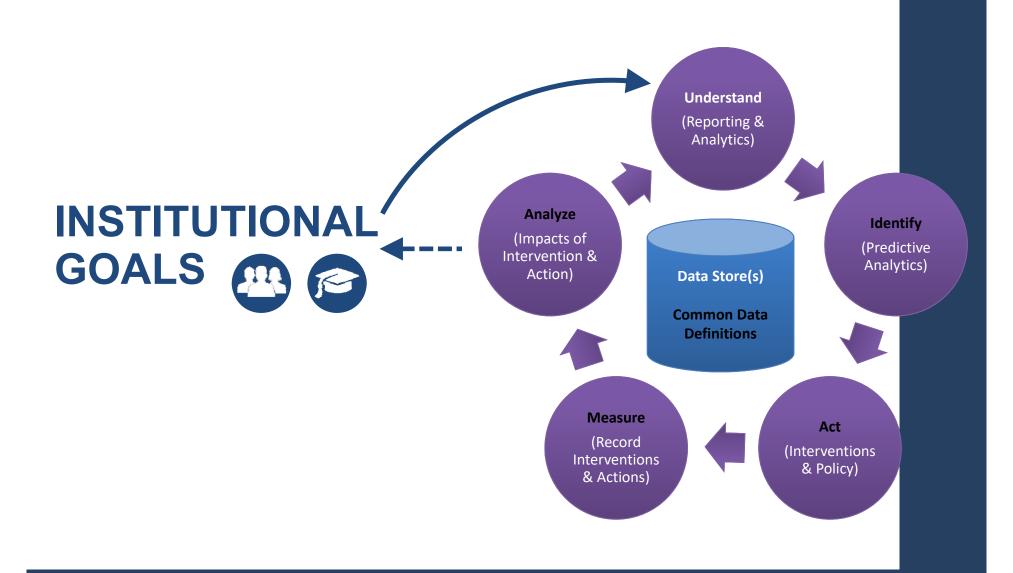
What are the appropriate data to measure these goals?

How are the findings reported so that it is both understandable and actionable?

How is insight fed back into the process of addressing institutional goals and creating new ones?

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## **Analytics in Higher Education**



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#### Handout: LA Framework

Internal

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External

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## Session 1 handout

1.	In your own words, please explain what Learning Analytics is.
2.	To your knowledge, are there any projects at your institution that make use of Learning Analytics?
No	
Yes	
If the	ere are any projects, please explain briefly:





Southern African Association for Institutional Research

# State of Learning Analytics in South Africa





## "a South African Study"

- The Learning Analytical context is still in its infancy in South Africa but multiple institutions are displaying promising practices pertaining to the further development of the research field.
- This lead to the conceptualisation of this research project, which directly stems from the inaugural SAHELA conference in 2013.
- The SAHELA 2013 participating institutional representatives were approached to provide detail as to the further development of an Learnin Analytical culture at their institutions.



## **Participants**



















#### Critical dimensions of Learning Analytics

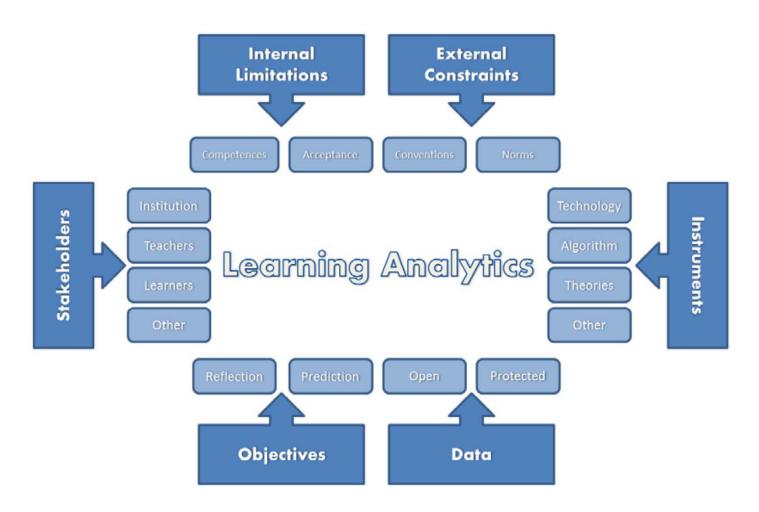


Figure 1. Critical dimensions of learning analytics. (Geller & Drachsler, 2012, pp. 44).



#### **Survey Dimensions**

#### **Stakeholders**



Where are analytics located in the institution?

Who are the various stakeholders that benefits from the analytics? (e.g. data used to assist students, faculties/schools, departments, institutions...)

What are the objectives of analytics at your institution? (E.g. descriptive and/or predictive)?



**Objectives** 

#### Data



What data is being collected? (e.g. learner characteristics, engagement, interventions, evaluations)

Are there specific times and events used to collect student and institutional level data?

What analytics systems are being used What analytical tools and dashboards are available?



Instruments



## **Survey Dimensions**

### Limitations



Are staffs responsible for the analytics trained and knowledgeable?

### **Constraints**



What processes does the institution have in place to deal with any legal or ethical issues? Who has access to the data?

### **Evidence of outcomes**



Any outcomes or achievements with regards to incorporating analytics? What interventions are taken as a result of the analytics?

Future analytical driven innovations that aim to use student data to optimise their success?

Further developments

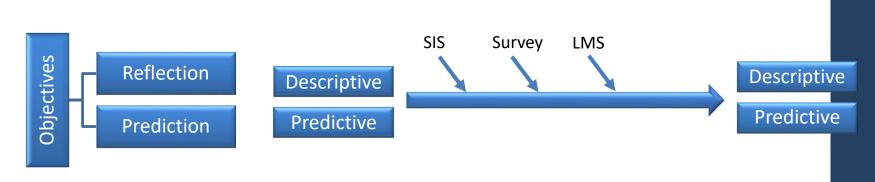




## Rubric

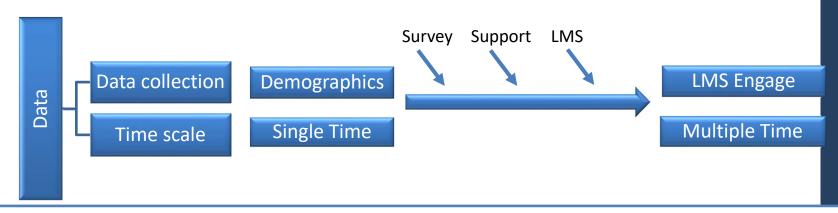


Stakeholder: As the system matures, the stakeholders move from a meso level to a micro level and as the practices move from a highly decentralised data environment to a controlled centralised data environment as the system starts combatting data silos

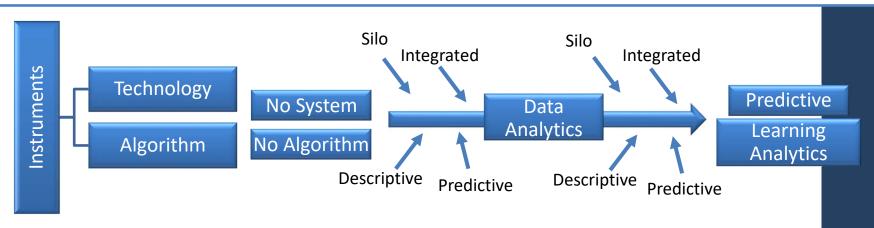


Objectives: The concept of Learning Analytics as part of the "big data" movement is to consolidate multiple data sources to provide a broader understanding of stakeholders. As the objectives strengthen the system includes both descriptive and prediction models

## Rubric

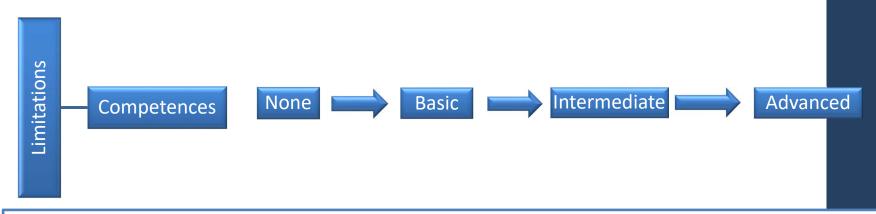


Data: To explore the abilities of Learning Analytics is to engage with more data and more frequently. This dimension as it matures move from a single to more complex data source and from a single point of data collection to a multiple time stamp

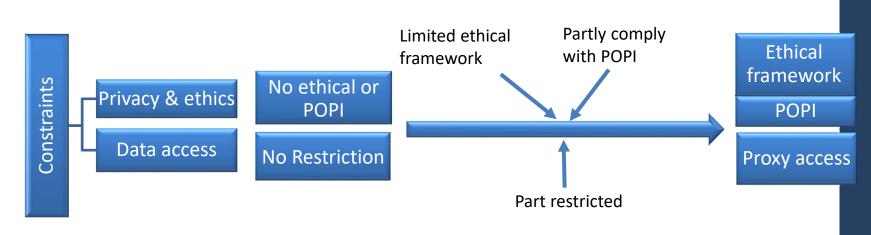


Instruments plays a pivotal role and its critical to reflect on the systems and methodology we use in an effort to better understand our students. This dimension matures from a Data Analytical focus to a predictive Learning Analytical focus

## Rubric



Limitations: Capacity development internally to an institution can be a tremendous task and creating a culture of evidence and adopting analytical tools and techniques can be a limiting factor



Constraints: Two major constraints to the evolution of Learning Analytics is the sensitive matter of ethics and the governance of student data

# Results

Code description	NMMU	UFS	UWC	Wits	TUT	UJ	UNISA	UP	Dimension
									max score
Stakeholder: Analytics	3	4	2	3	3	3	3	3	5
Stakeholder: Data users	4	4	3	3	4	3	5	5	5
Stakeholder: Practice	3	2	2	2	3	2	2	3	3
Objective: Reflection	4	4	4	4	4	4	4	4	5
Objective: Prediction	3	3	2	3	3	3	4	3	5
Data: Time scale	4	4	4	4	4	4	4	4	5
Data: Data collection	4	4	4	4	4	4	4	4	5
Constraints: Privacy and Ethics	3	3	3	2	3	2	3	3	3
Constraints: Data access	3	3	3	2	2	3	2	3	3
Instruments: Technology	3	3	3	3	3	3	3	3	5
Instruments: Algorithm	3	4	3	3	3	4	5	4	5
Limitations: Competence	3	3	3	4	3	3	4	3	4
TOTALS:	40	41	36	37	39	38	43	42	53

## Summary of findings

- Magnitude of data available from MIS, SIS data, survey results, in some instances interventions are evaluated and prediction models are based on the data, however, LMS data is not used widely to inform teaching and learning.
- SMART data
- Interventions were inferred from data > outcomes NOT assessed – Planning for outcomes assessment prior to embarking on the intervention
- 'Sophistication of the analytics system is not in centralisation but rather decentralisation' (Jan Lyddon)





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# Learning Analytics for students





# Contextual questions

- What interactions do your students have with the systems, policies and process that contribute to their learning experience?
- How much of your students' educational outcomes are due to internal factors or external factors?
- Think of a specific type of student at your institution that you want to understand better?
- What do students do with their data?
- What data is most useful for students?
- How do students use their data to better their learning?

# Learning Analytics for students

Students leave behind a digital footprint, this footprint can be seen on LMS, VLE, assessments, library and entering/exiting the campus.

According to Niall Sclater, Learning analytics is the process of using this data to improve learning and teaching:

- Students are not always aware of their performance in relation to their peers.
- Students are not always granted access to the data that they generate and their thus able to apply action.
- Students need continues formative feedback so that institutions can assist learners in transforming their learning and gain a better understanding of how they learn in relation to others.

How do some institutions combat this?





220 terabytes of web data6 petabytes other data



3 times the amount of ALL the U.S. academic research libraries contents combined



IBM stipulate that users create 2.5 quintillion bytes per day... To put this into perspective, 90% of all the data ever has been created in the last two years

# Learning Analytics for students

### There is a lot of data out there...













What data is most useful for students?

### What information should be shared with students?

- Monitoring data (Includes academic progress, level of engagement)
- Comparative data (Compare a students' progress with peers)
- Useful data (Information about exam times, class time tables)

### How should this information be acted upon?

- Students should receive prompts and recommendations
  - Wellness centres, tutorials, advising, counselling
- Communication facilities with staff and students
- Providing consent to what data is used for learning analytics

## How students use their data

VLE at the University of Maryland, Baltimore County

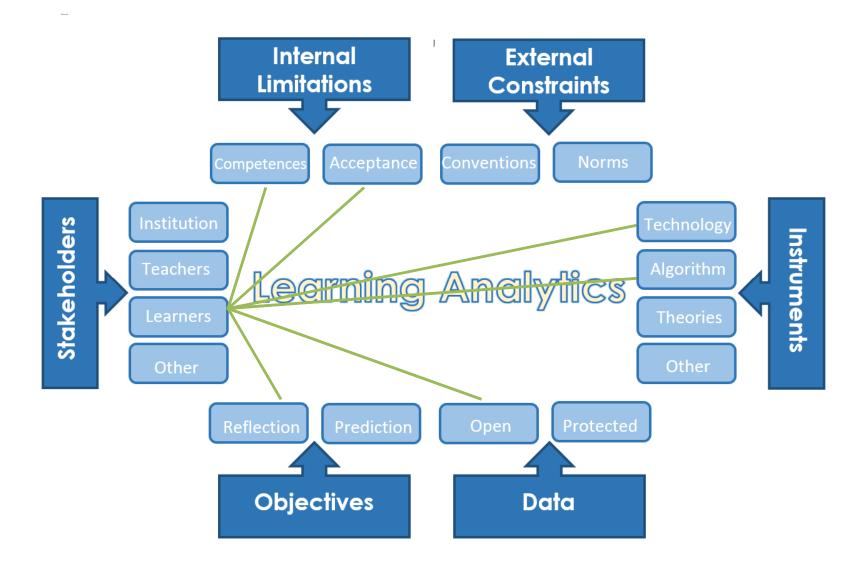


Students submit and view their data using the "Check My Activity tool" The CMA was developed for students to compare their own learning activity in a course against their peers

"obsessive status-checking tendencies" of students enabled:

- Frequent feedback on what their peers are experiencing
- Feedback of their own activities

Further research determined that 92% of the students used the VLE and of those 91.5% used the CMA tool.



## How students use their data

### Early Alert at the University of New England



Complex system with a simplex interface, Leece and Hale in 2009 stipulated that the aim was to develop a dynamic automated process that capture the learning wellbeing of students:

E-motion: students use emoticons and textboxes to express how they Feel each day linked to their subjects.









The Vibe: Words entered in the textboxes are aggregated into a word cloud and updated every 10min so the rest of the cohort can see what the other students are thinking and sharing.



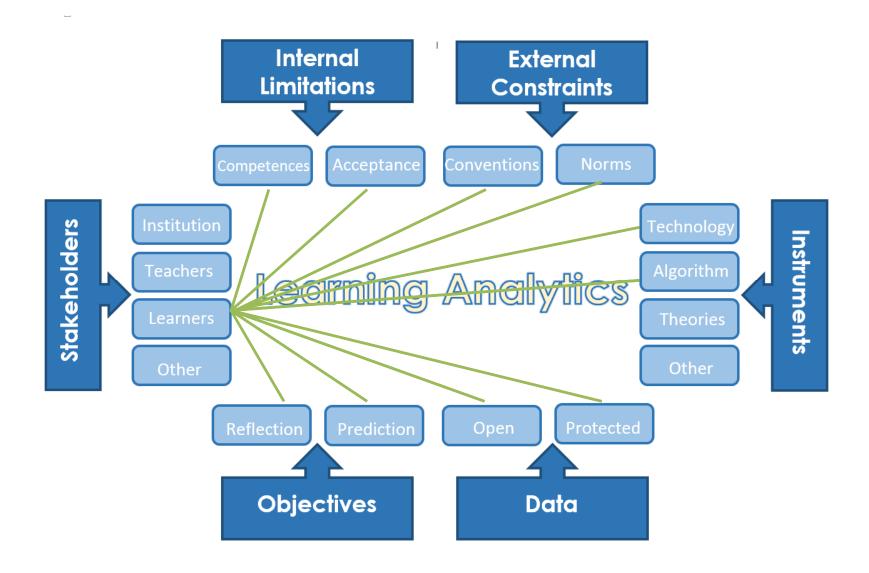
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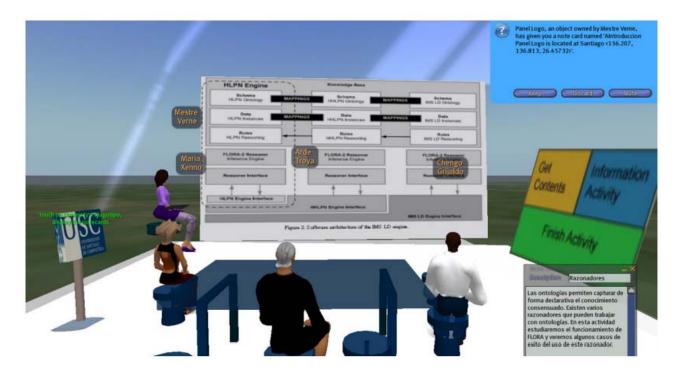
Automated Wellness Engine (AWE): Complex system that analysis data from 34 Indicators which contributes to the classification of risk.

The system is powerful in the sense that it analyses data in real time using seven corporate data systems every night. The following day support staff is informed based on the 34 triggers which students need further support.



## What's next?

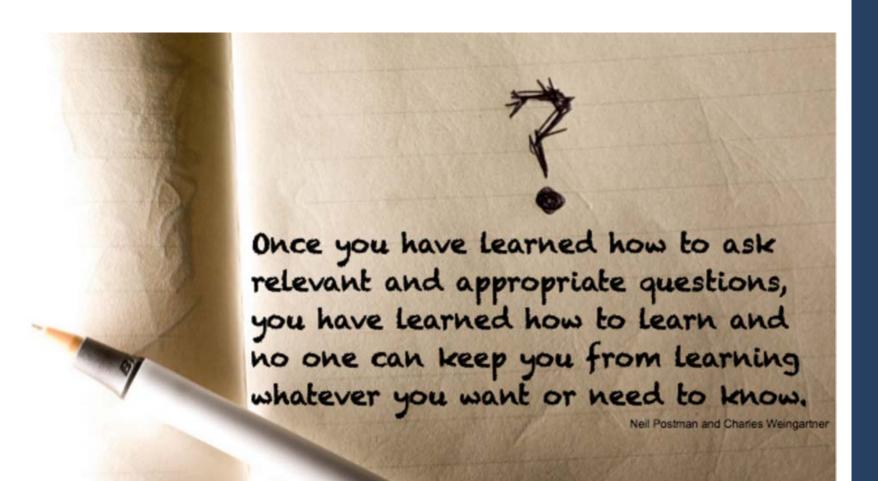
OPENET4VE platform, is a virtual world learning environment Augmenting the learning experience



International Conference on Virtual and Augmented Reality in Education

### Session 2 and 3 handout

3.	In your option what data should <b>NOT</b> be shared with students?
4.	Do you think a learning analytical approach where students are key decision makers would enhance teaching and learning?
No	
Yes	
Pleas	se explain briefly:







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# Learning Analytics for lecturers





# Overview

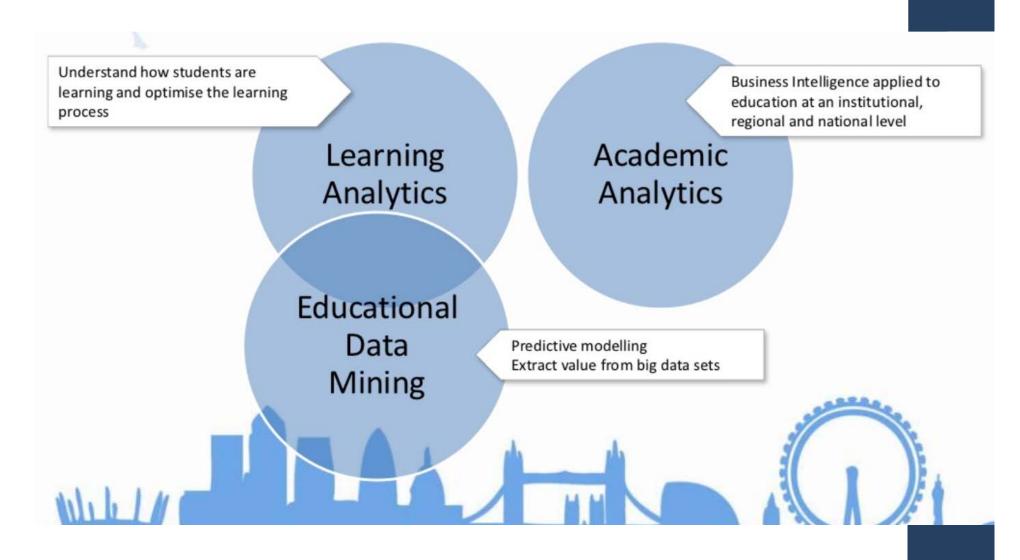
- What do we have?
- What do we do?
- What do we need?

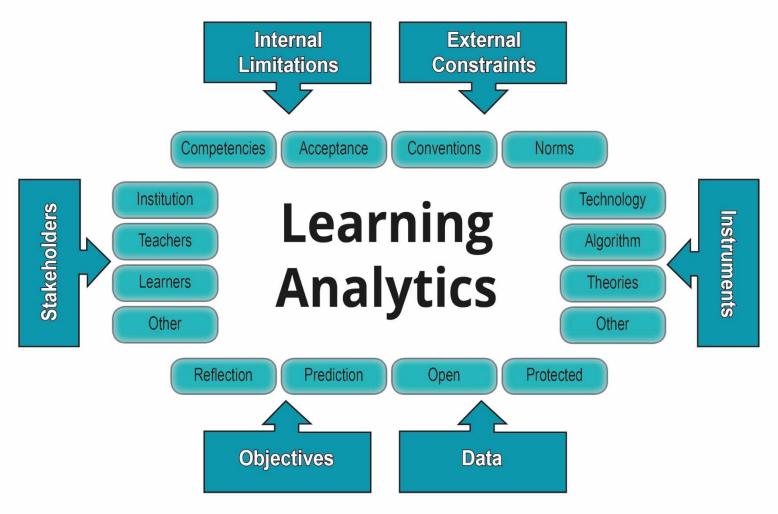
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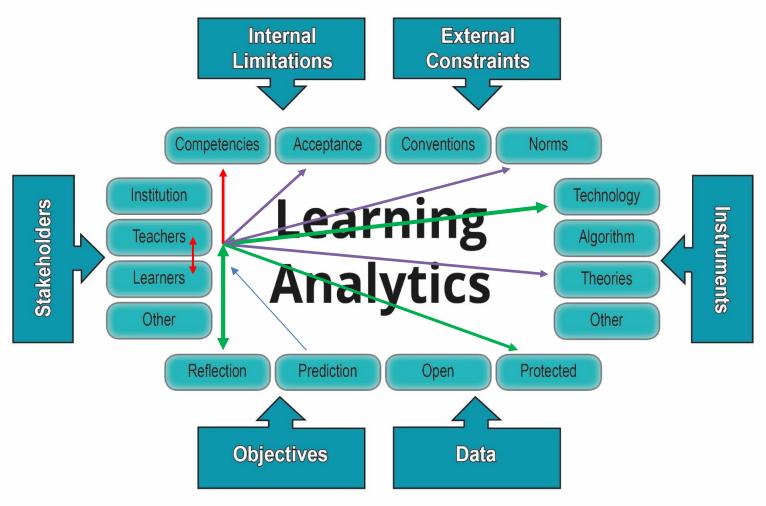
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## Academic or Learning Analytics



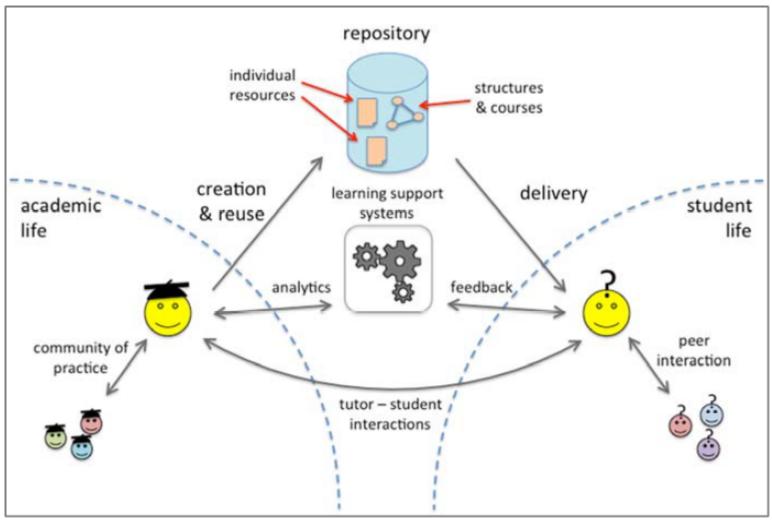


Critical dimensions of LA. (Geller & Drachsler, 2012, pp. 44)



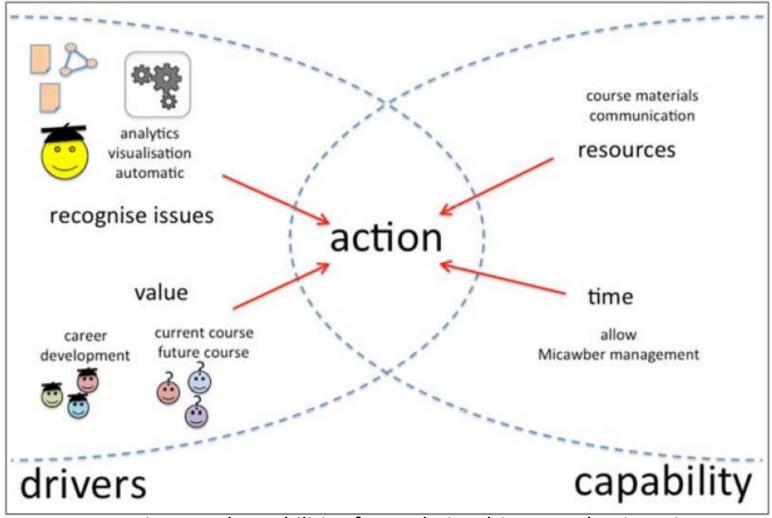
Critical dimensions of LA. (Geller & Drachsler, 2012, pp. 44)

# Learning Analytics for Lecturers



Learning resource lifecycle: actors, agents and events

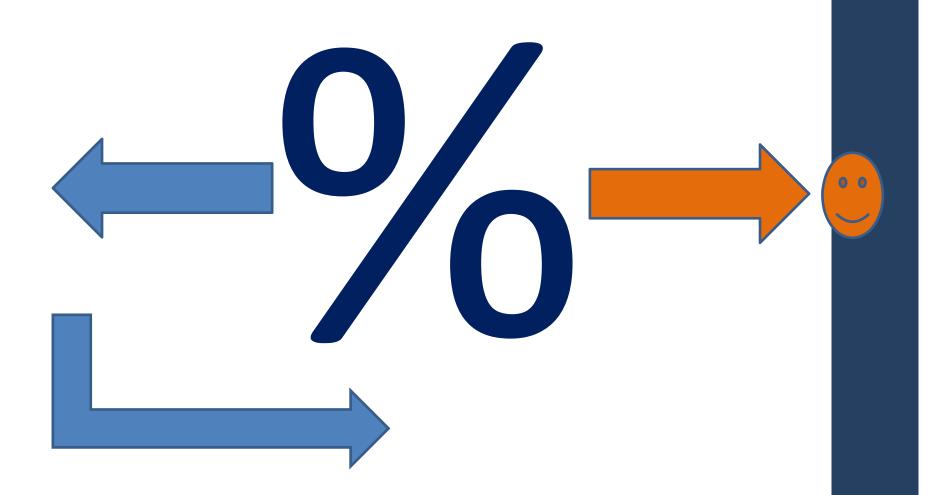
# Learning Analytics for Lecturers



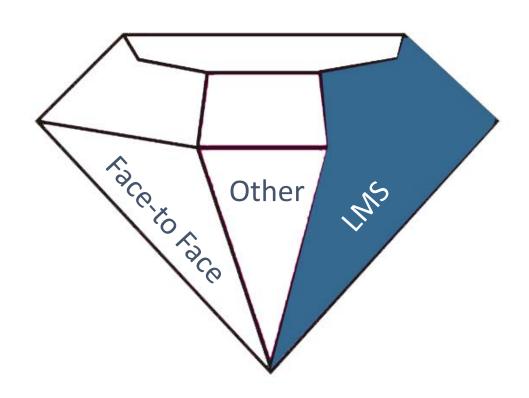
Drivers and capabilities for analytics-driven academic action

## WHAT DO WE HAVE?

Learning Management System Data



# Blended model: Impact on data value



# Lecturers capture grades in LMS

CRADE DISTRIBUTION

Comments	)	Progress mark
Admission	87	81.00
Admission	66	60.00
Admission	39	31.00
Admission	73	81.00
Admission	60	57.00
Admission	70	72.00
Admission	73	72.00
Admission	93	93.00
Admission	61	60.00
Admission	54	62.00

GRADE DISTRIBUTI	ON
Greater than 100	0
90 - 100	68
80 - 89	298
70 - 79	349
60 - 69	316
50 - 59	233
40 - 49	141
30 - 39	81
20 - 29	34
10 - 19	12
0 - 9	17
Less than 0	0

### ▼ Grade Center

Needs Grading

Full Grade Center

Assignments

Group: Group: Progress Mark 1 - 39%

Group: Group: Progress

Mark 40 - 44%

Group: Group: Progress

Mark 45 - 49%

Group: Group: Progress

Mark 50 - 54%

Group: Group: Progress Mark 55 - 64%

Group: Group: Progress

Mark 65 - 69%

Group: Group: Progress Mark 70 - 74%

Group: Group: Progress

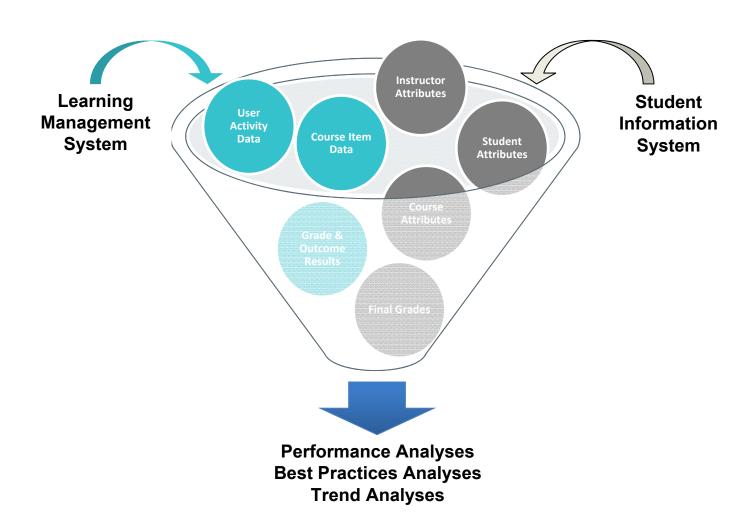
Mark 75 - 100%

Tests

## WHAT DO WE DO?

Student progress data

# **Analytics for Learn**



# Categories of Analytics for Learn

Grades

Student Course Summary

**Grade Center** 

**Course Design** 

**Course Summary** 

**Course Items** 

User Activity

**Session Activity** 

**Course Activity** 

Course

**Item Activity** 

**Submissions** 

**Forum Submissions** 

# Lecturers capture grades in LMS

### Learn Course At A Glance

### **Learn Course Information**

Instructor:

Course ID:

Academic Semester: 2016 Academic Term

Status: Unavailable

Instruction Method: Full-time Contact

Students Enrolled: 1623

Faculty: Department:

### ITEM COUNT (SAME INSTRUCTION METHOD)

ITEM	TEM COURSE DEPARTME		% DIFFERENCE
Assessment	182	47.0	
Content	223	252.0	I
Tool	1,060	160.0	

#### % OF ITEMS ACCESSED (SAME INSTRUCTION METHOD)

ITEM	COURSE	DEPARTMENT AVG	% DIFFERENCE
Assessment	15.4%	22.2%	
Content	30.6%	30.0%	
Tool	0.4%	1.4%	

#### **ACTIVITY (SAME INSTRUCTION METHOD)**

ITEM	COURSE AVG	DEPARTMENT AVG	% DIFFERENCE
Accesses	82	54.9	
Minutes	807	685.3	
Interactions	600	360.6	
Submissions	27	8.1	
		-2	00% -100% 0% 100% 2009

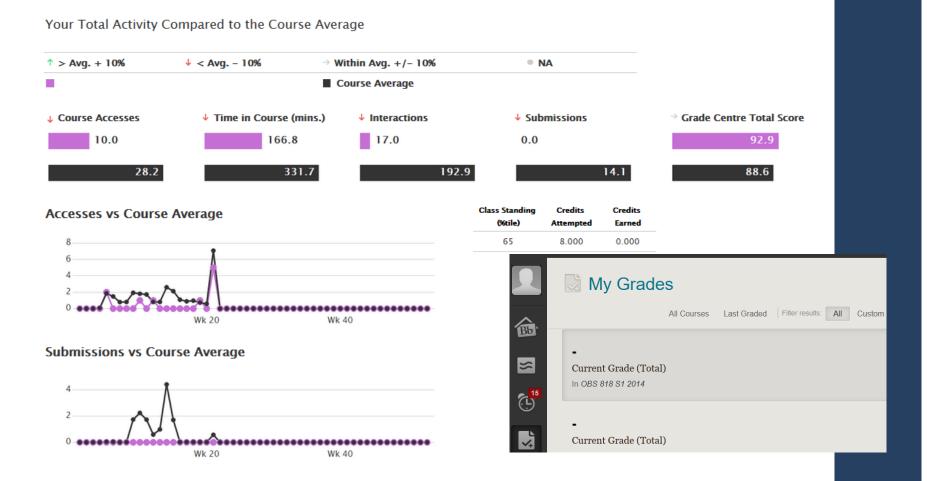
## Lecturers capture grades in LMS

Student Activity Summary (1623 St	tudents)										1
							COURSE AC	CESSES	MINU	TES	INTE
STUDENT NAME *	Academic ‡ Plan	Gender ‡	Race ‡	Age⊙	DATE OF LAST ÷ ACCESS	DATE OF LAST \$ SUBMISSION	STUDENT \$	AVG	STUDENT \$	AVG	STUDENT \$
Куансун, в (волише) 5053)	No SIS Match	F	African	43	08/07/2016	18/05/2016	49 🗸	82.3	251	806.9	28
(0000003)	MEd	F	White	39		11/02/2016	0 1	82.3	0	806.9	
	BCom Hons	F	African	37	21/06/2016	04/05/2016	111 1	82.3	1472 1	806.9	70
	No SIS Match	F	African	34	13/06/2016	14/04/2016	26 🗸	82.3	322	806.9	22
	Baccalaureus Educationis	F	African	34	27/06/2016	18/05/2016	178 1	82.3	2170	806.9	187
	Baccalaureus Commercii	F	Coloured	32	21/07/2016	19/05/2016	172 1	82.3	2359 1	806.9	167
	BCom Hons	F	African	32	28/06/2016	24/05/2016	76	82.3	2210 1	806.9	100
	Baccalaureus Commercii	М	African	31	25/06/2016		15 🗸	82.3	65	806.9	15
	Baccalaureus Commercii	M	African	30	25/06/2016	18/05/2016	44 🗸	82.3	392	806.9	30
	Baccalaureus Educationis	F	African	30	18/06/2016	23/05/2016	58 ↓	82.3	1407	806.9	77
	Baccalaureus Commercii	М	African	29	18/07/2016	03/05/2016	85	82.3	1696	806.9	•

## Lecturers capture grades in LMS

	COURSE	ACCESSES		MINU	ITES		INTERACT	TIONS		SUBMIS	SIONS		GRADE CEN	TRE	SCORE
DATE OF LAST \$ SUBMISSION	STUDENT ‡	AVG		STUDENT ‡	AVG		STUDENT \$	AVG		STUDENT ‡	AVG		STUDENT ‡	-	NVG
18/05/2016	49	<b>↓</b> 8	32.3	251	<b>↓</b> 80	6.9	288 ↓		599.9	6	Ψ.	26.6	35.0%	4	65.5%
11/02/2016	0	↓ 8	32.3	0	↓ 80	6.9	0 4	ŗ	599.9	1	1	26.6		<b>4</b>	65.5%
04/05/2016	111	1 8	32.3	1472	<b>↑</b> 80	6.9	707 1	į	599.9	22	<b>4</b>	26.6	77.0%	1	65.5%
14/04/2016	26	↓ 8	32.3	322	<b>↓</b> 80	6.9	221 ↓		599.9	4	1	26.6	9.0%	4	65.5%
18/05/2016	178	1 8	2.3	2170	<b>↑</b> 80	6.9	1872 1	į.	599.9	50	1	26.6	93.0%	1	65.5%
19/05/2016	172	1 8	32.3	2359	1 80	6.9	1678 1	į.	599.9	67	1	26.6	70.0%	->	65.5%
24/05/2016	76	→ 8	32.3	2210	<b>↑</b> 80	6.9	1007 1	į	599.9	46	<b>T</b>	26.6	42.0%	4	65.5%
	15	↓ 8	32.3	65	<b>↓</b> 80	6.9	152 ↓	ŗ	599.9	0	1	26.6	6.0%	4	65.5%
18/05/2016	44	↓ 8	32.3	392	↓ 80	6.9	309 ↓	ţ	599.9	13	<b>4</b>	26.6	61.0%	$\rightarrow$	65.5%
23/05/2016	58	↓ 8	32.3	1407	<b>1</b> 80	6.9	775 🕇	ţ	599.9	35	1	26.6	87.0%	1	65.5%
03/05/2016	85	→ 8	2.3	1696	<b>↑</b> 80	6.9	1086 1		599.9	27	$\rightarrow$	26.6		<b>4</b>	65.59

## Student report in LMS



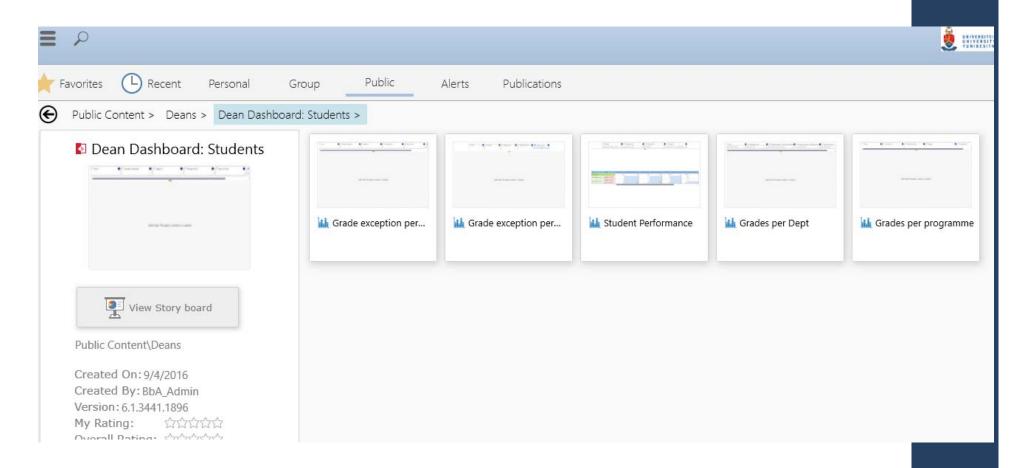
#### **MONITOR PROGRESS**

Dashboard: Student grades

## Real-time access to grades

- Access Analytics for learn dashboards
  - Student progress
  - Module design
  - Overview of LMS use per faculty and department

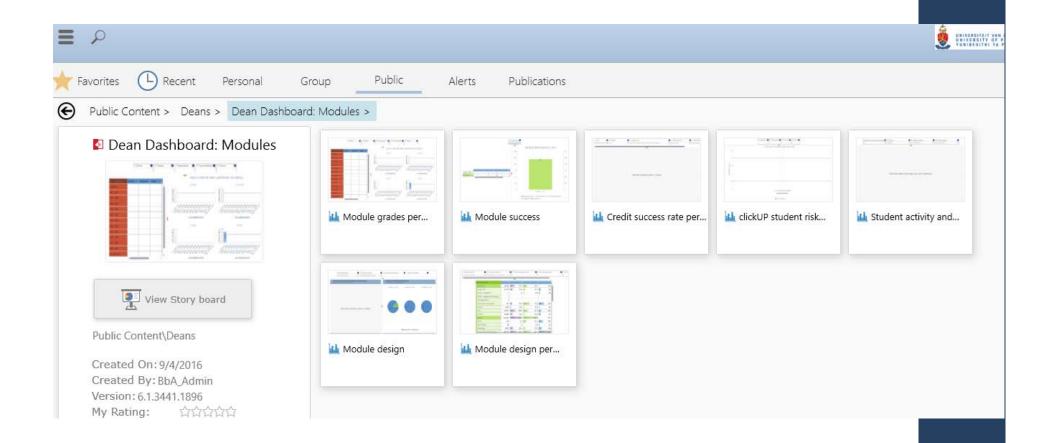
## Dashboard: Student progress



#### **MONITOR PROGRESS**

Dashboard: Module success

#### Dashboard: Module success



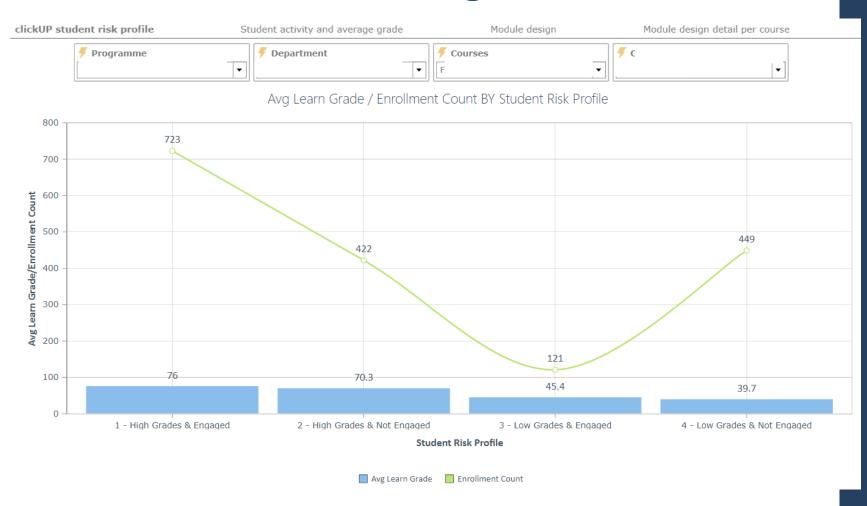
#### **MONITOR PROGRESS**

Dashboard: Course design

## Course design



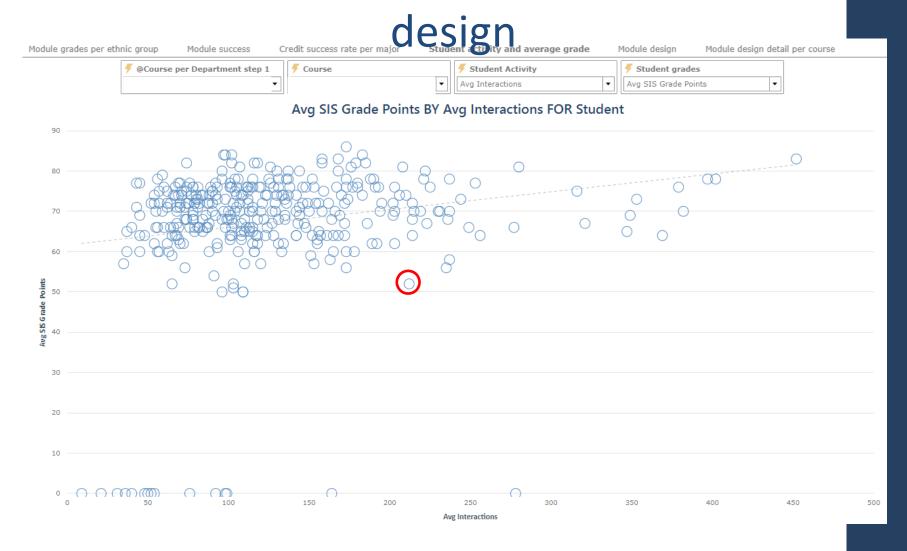
# Grades, student engagement and course design



# Grades, student engagement and course design



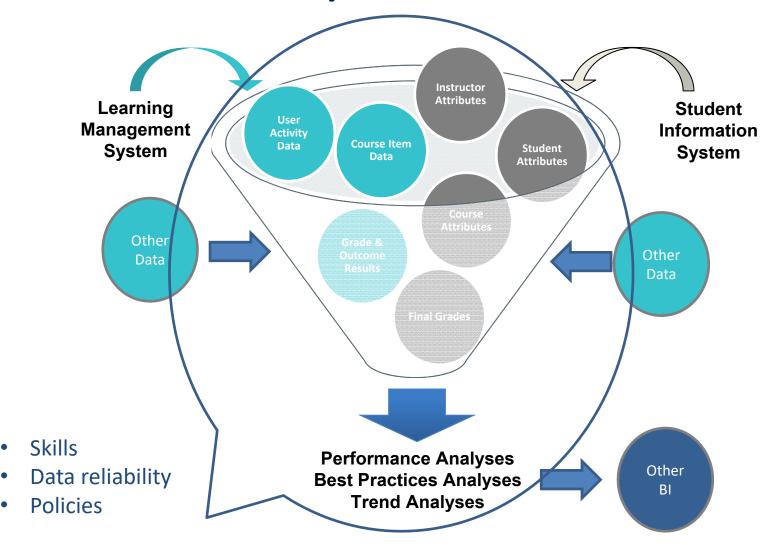
#### Grades, student engagement and course



#### WHAT DO WE NEED?

Data Integration

## **Analytics for Learn**



## **Analytics for Learn**

			Avg SIS Grade	Points	Avg Learn	Grade	
	PUM 120	1 - High Grades & Engaged			•	30.0	^
	PAD 122	1 - High Grades & Engaged			0	64.0	
@tuks.co.za	PUT 120	1 - High Grades & Engaged			0	66.0	
	PUF 110	1 - High Grades & Engaged		66.0	0	58.0	
	AIM 111	2 - High Grades & Not Engaged		78.0		90.6	
	PUM 120	1 - High Grades & Engaged				30.0	
TARCHAIL COM	PAD 122	3 - Low Grades & Engaged				52.0	
@GMAIL.COM							
	PUF 110	1 - High Grades & Engaged	0	61.0		60.0	
	PUM 120	1 - High Grades & Engaged				21.7	
Otube so as	PAD 122 2 - High Grades & Not Engaged						
@tuks.co.za	PUT 120	1 - High Grades & Engaged				56.0	
	PUF 110	2 - High Grades & Not Engaged		61.0		62.0	
	PUM 120	2 - High Grades & Not Engaged				21.7	
	PAD 122	3 - Low Grades & Engaged			0	46.0	*





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# Learning Analytics for institutional management

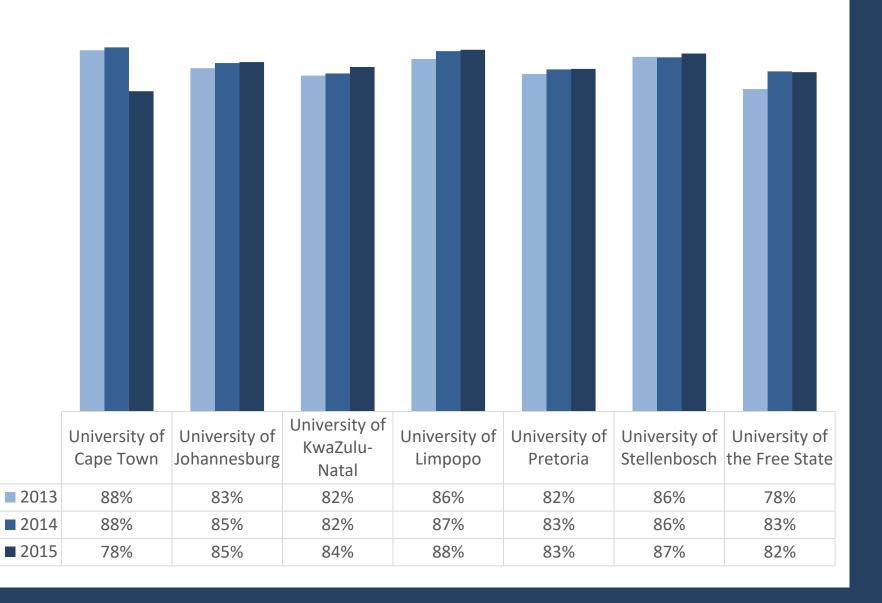




## Analytics for institutional management

- Examples of indicators prepared by Departments for Institutional Planning:
  - Number of entering students per faculty and their demographic data
  - Examination statistics
  - Success rates
  - Graduation rates
  - Retention rates

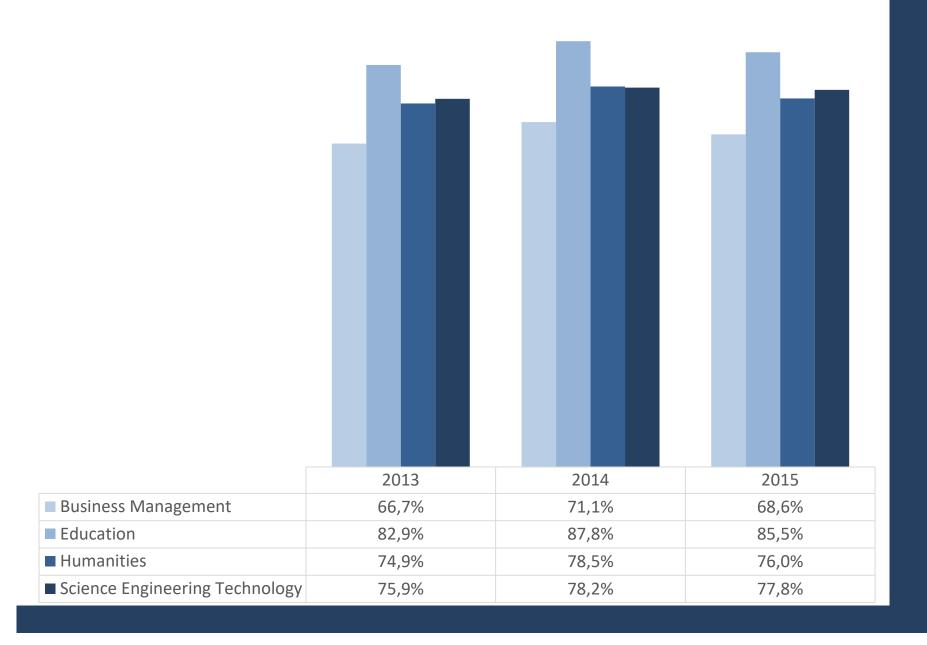
## UG Degree Credit Success Rate for contact mode students



## LA for institutional management

 How can institutional management evaluate their faculties performance?

#### Undergraduate degree credit success rate by CESM



## Graduation rate of faculties - first time entering, 3 year degree (2010 cohort)

FACULTIES	Baseline	n (2012 Grad)	n+1 (2013 Grad)	n+2 (2014 Grad)	Total Grad	Dropout	Enrolled (still busy)
	1035	35%	14%	4%	53%	43%	5%
	1007	14%	19%	8%	42%	49%	9%
	1884	38%	24%	9%	71%	20%	9%
	164	52%	13%	4%	69%	30%	1%
	622	24%	22%	10%	56%	33%	11%

## LA for institutional management

- Who is it that we should admit?
- And once we admit students to our campus, are we treating those students equitably?

#### Who should be admitted?

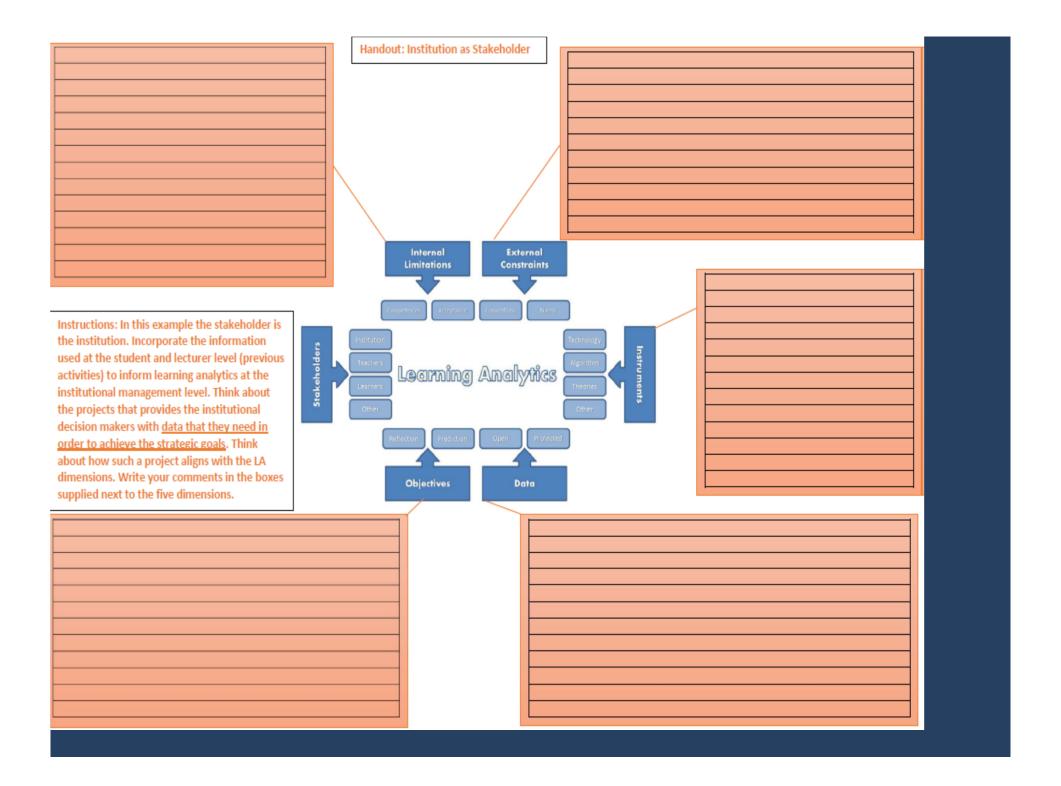
Final Mark	Mathematics	English	Sciences		Percentile
70	91	85	91		90
65	87	82	87	<del>~</del>	80
61	85	79	84	$\mathbf{S}$	70
58	82	77	81	inal:	60
56	79	75	78	of F	50
54	76	72	75	Percentile of Final Mark	40
50	73	70	72	rcer	30
47	69	67	68	Pe	20
41	63	63	60		10

## LA for institutional management

- What do the programs which we fund actually do?
- We run a lot of different programs on our campuses that are aimed at improving the student experience.
- We have many different student support programs like:
  - Living learning communities
  - Mentoring
  - Tutoring
  - Orientation
- Understand the impact that each program has on the students who participate - all different kinds of students

# Cross-tabulation of academic risk cluster and participation in FSA session/s

Academic cl	uster	Numb	Total		
		Zero session	One session	Two > sessions	
At risk	Count	28	27	12	67
	% within FSA sessions	46.7%	32.1%	21.4%	33.5%
Borderline	Count	24	29	22	75
	% within FSA sessions	40.0%	34.5%	39.3%	37.5%
Not at risk	Count	8	28	22	58
	% within FSA sessions	13.3%	33.3%	39.3%	29.0%
TOTAL	Count	60	84	56	200
	% within FSA sessions	100%	100%	100%	100%



## **Activity**

- Incorporate the information used at the student and lecturer level (previous activities) to inform learning analytics at the institutional management level.
- Provide actionable information to shift the needle on the strategic goals created in session 1.
- Use the G&D Framework as guide, you could also develop a SA version as adaptation.

## Additional questions

- What are the variables that drive LA outcomes across the three levels? (transformation, student success rates and graduation out-put – in minimum time);
- How do the dimensions of the G&D framework impact LA at the various levels? (Stakeholders, Data, Ethics, Training, Analysis, etc.);
- How should the sector position itself strategically to facilitate institutions to practically implement LA at student, lecturer and management levels and do so in a coordinated fashion?

## Additional questions

- What question(s) are you trying to answer?
   What problems are you trying to solve?
- What variables are suggested in the literature? Institutional knowledge? Other institutions?
- How do you measure the impact of the institutional programs?
- What process and procedures will need to change?

## Session 5 handout

1.	Are you aware of any Learning Analytical driven projects where the Institution is the main stakeholder?
No	
Yes	
2.	Does your institution measure the impact of student support programs on a regular basis (e.g. annually)?
No	
Yes	
Pleas	se explain briefly:





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## Closing the loop





# **Activity**University of Pretoria

## No to secret sauce analytics

Do you have "magic number" mentality

Are you using a black box of formulas they can neither be shared nor explained?



Have you recognised and acknowledge you are turning governance over partially to a formula (but this formulas is not shielded from view)

Are these formulas part of a library of open formulae that others can both test, review and comment on?





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## Closure



