

# Learning Analytics or how to improve the teaching and learning processes from data

Dr. José V. Benlloch-Dualde Dr. Lenin-Guillermo Lemus-Zúñiga

Computer Eng. Dept. (DISCA) Universitat Politècnica de València







### Index

- ✓ Educational data explosion✓ Defining Learning Analytics
- ✓ Data Sources
- ✓ Applications of Learning Analytics
- ✓ Current topics of interest
- ✓ Conclusions





### Data Explosion

 $\checkmark$  More and more data is available.

✓ Data-driven decision making is common in many fields.

✓ Higher education has traditionally been inefficient in its data use.







### **Special Considerations**

Addressing organizational resistance

The complex nature of educational data

Ethical and legal concerns



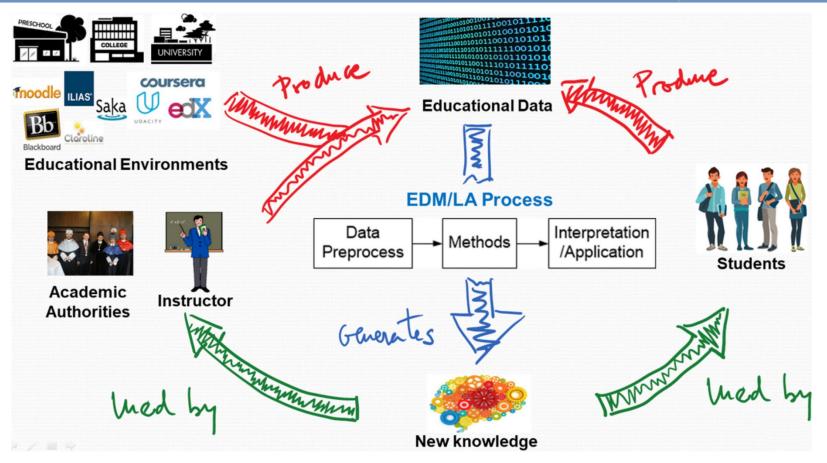


### **Defining Learning Analytics**

- Learning analytics is the measurement, collection, analysis and reporting of data about learners and their contexts, for purposes of understanding and optimising learning and the environments in which it occurs.
  - Long, P., & Siemens, G. (2011, September 12). Penetrating the Fog: Analytics in Learning and Education | EDUCAUSE. EDUCAUSE. https://er.educause.edu/articles/2011/9/penetrating-the-fog-analytics-in-learning-and-education
- Learning Analytics is the development and application of data science methods to the distinct characteristics, needs, and concerns of educational contexts and the data streams they generate for the purpose of better understanding and supporting learning processes and outcomes.
  - Wise, A. F. (2019). Learning Analytics: Using Data-Informed Decision-Making to Improve Teaching and Learning. In Contemporary Technologies in Education (pp. 119–143). Springer International Publishing. https://doi.org/10.1007/978-3-319-89680-9\_7







Romero, C., & Ventura, S. (2020). Educational data mining and learning analytics: An updated survey. Wiley Interdisciplinary Reviews: Data Mining and Knowledge Discovery, 10(3). https://doi.org/10.1002/WIDM.1355





## Could Learning Analytics help us to improve student learning in our particular context?





### Data Sources



#### Prior educational background

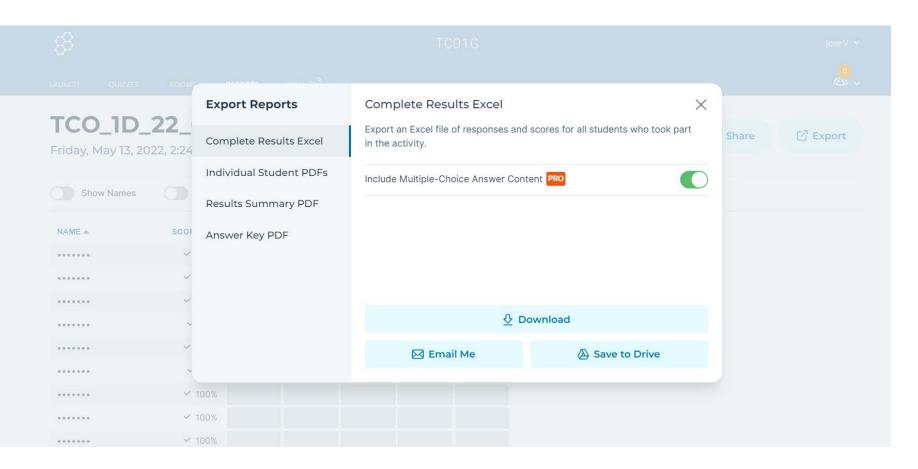
#### Data from physical space

Log data (educational environments) Artifact data created by a student

Academic performance







Obtaining reports from Socrative quizzes





ų	isername	10_09_0	-	1_11_	-	-	-	_	-		-		-	-		13_00	_	15.2		14,4.		_	the second s		Agregado		Nota (1)
			5		2		5		3		4		5		5		5		5		5	5		6	55	0	0,9
			6		4		3		2		3		5		5		5		4		5	5		6	53	0	0,9
			6		4		4		2		3		5		3		5		4		4	5		5	50	0	0,8
			3		1		5		3		3		5		5		4		4		5	5		б	49	0	0,8
			4		2		5		3		-4		4		4		4		3		5	5		6	49	0	0,8
			6		4		4		2		3		5		4		5		3		5 NP1	1		5	46	1	0,7
			5		4		3		2		з		5		4		5		3		4	5		3	46	0	0,7
			4		4		4		3		3		4		3		3		3		5	5		4	45	0	0,7
			4		4		3		3		3		4		3		5		3		2	- 4		5	43	0	0,7
			5		2		5		3		3		5		2		4		4		4	5	NP12		42	1	0,7
			5		1		5		3		3		4		3		3		4		4	5	NP12		40	1	0,6
			5		4		- 4		3		3		4		5	NP8			3		5 NP1	1		4	40	2	0,6
			5		2		4		2		2		5		3		4		3		3 NP1;	1		5	38	1	0,6
			5		4		4		3		3		4		3		4		3		5 NP1	1	NP12		38	2	0,6
			5		4		4		2		3		4		2		3		2		0	4		3	36	0	0,6
		NP1			4		- 4		2	NP5			3		3		4		3		3	- 4		5	35	2	0,5
			4		2		2		2		3		4	N	P7		3		2		5	4		4	35	1	0,5
			4		2		4		2		3		3		2		4		2		2	5	NP12		33	1	0,5
			5		4		2		3		3		5		2		2		2		4 NP1	1	NP12		32	2	0,5
			5		2		4		3		3		4		4		4		3	NP10	NP1	1	NP12		32	3	0,5
			5		2		2		1		2		4		1		2		2		2	5		4	32	0	0,5
			5		2		3		2		2		3	N	P7		3		2		0	5		4	31	1	0,5
			4		2		3		2		3		3		0		3		1		4	5	NP12		30	1	0,5
			41	NP2			2		2		3	NP6		N	P7		3	NP9			5	5		4	28	4	0,4
			3		2		3		0		3		3		2		0		2		1	3		5	27	0	0,4
			4		2		4		2	NP5			3		3		4		2		O NP1	1	NP12		24	3	0,4
			4		2		4		2		3		3		2	NP8			2		2 NP11	1	NP12		24	3	0,4
			51	NP2			4		2		2		3	N	P7	NP8			2	NP10	NP1	1		4	22	5	0,3
			11	NP2			1		1		2		1		2		3		2		0	3		2	18	1	0,3
			31	NP2			3		1		2	NP6		N	P7	NP8		NP9		NP10	NP1	1	NP12		9	8	0,1
# items			31	NP2			1		0		2	NP6		N	P7	NPB		NP9		NP10	NP1	1	NP12		6	8	0,1
			3 1	NP2			1	NP4			2	NP6		N	P7	NP8		NP9		NP10	NP1	1	NP12		6	9	0,1
		NP1		NP2			2		0			NP6			P7	NP8		NP9		NP10	NP1		NP12		5	9	0,0
		NP1	-	NP2		NP3	-	NP4	-	NP5	-	NP6			P7	NP8		NP9		NP10	NP1		NP12		0	12	0,0
	NP1		NP2		NPS		NP4		NP5		NP6			P7	NP8		NP9		NP10	NP1		NP12		0	12	0,0	
		6		4		5	1100.20	3		5		5		5		5		5		5	5		6	59		5/4	
	ANP		4		9		2		3		4		7		10		9		7		8	15		-	Total items		
	#P		11		26		33		32		31		28		25		26		28		27	20		20	reser mento		

Aggregated scores for all students who took part in the activities (Socrative Quizzes)





📰 What? Select activity t	o report.
Activity:	Visits
🛗 When? Select time per	iod to report.
Period:	Last 7 days (PoliformaT server time zone)
Select users to re	port.
Users:	All
Necify how results	Its should be presented.
Totals by:	User <ul> <li>Tool</li> <li>Event</li> <li>Resource</li> <li>Resource action</li> <li>Date</li> <li> </li></ul>
Number of results:	Limit to: 0
Presentation:	Table 🗸
Generate report S	ave report Back

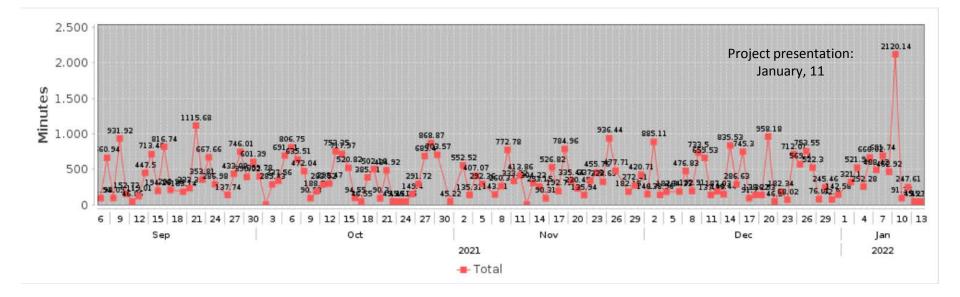
How to get a new report from *PoliformaT* (UPV Learning Management System?





Oct 6, 2021

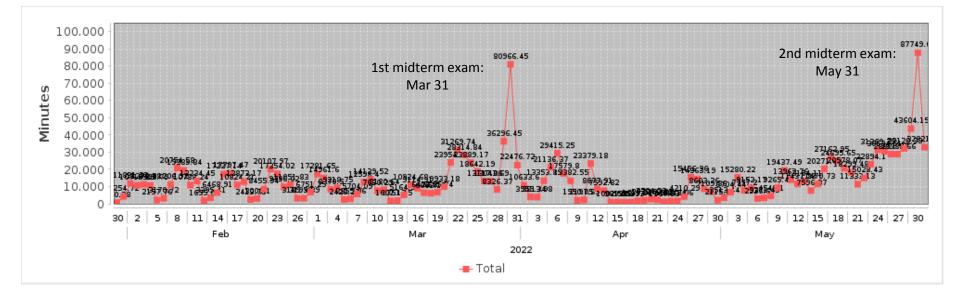
806,8



Site: "Elr" (DOC 33421 2021)	Date	Duration (min.)	
	Jan 9, 2022	2120,1	
Activity type: Presence Time	Sep 21, 2021	1115,7	
Date range: Sep 6, 2021 - Jan 13, 2022	Dec 20, 2021	958,2	
	Nov 25, 2021	936,4	
User selection type: Role	Sep 9, 2021	931,9	
Role selected: Student	Dec 2, 2021	885,1	
	Oct 28, 2021	868,9	
Report generated: Jun 2, 2022 4:07 PM CEST	Dec 14, 2021	835,5	
	Sep 16, 2021	816,7	



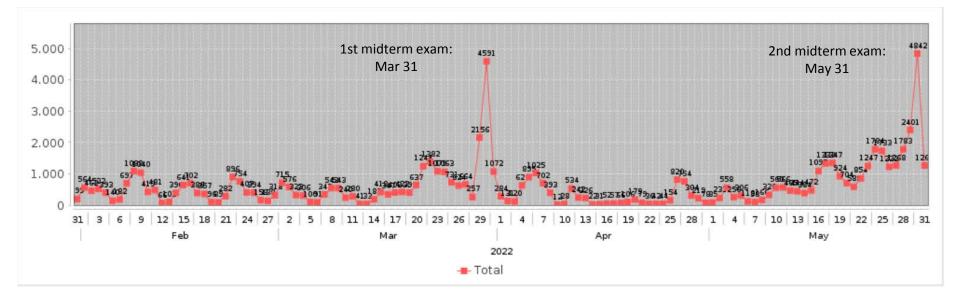




	Date	Duration (min.)
	May 30, 2022	87749,1
Site: "Tecnología de computadores GII" (GRA_11544_2021)	Mar 30, 2022	80966,4
Activity type: Presence Time	May 29, 2022	43604,1
	Mar 29, 2022	36296,5
Date range: Jan 30, 2022 - May 31, 2022	May 31, 2022	32821,8
User selection type: Role	May 28, 2022	32120,3
	May 24, 2022	31369,8
Role selected: Student	Mar 22, 2022	31269,7
Report generated: Jun 02, 2022 4:25 PM CEST	May 25, 2022	30534,5
Report generated. Juli 02, 2022 4.25 TWI CEST	Apr 6, 2022	29415,2







	Date	Total
Site: "Tecnología de computadores GII" (GRA_11544_2021)	May 30, 2022	4842
Activity type: Events (Select by tool)	Mar 30, 2022	4591
	May 29, 2022	2401
Tools selected: Resources	Mar 29, 2022	2156
Date range: Jan 30, 2022 - May 31, 2022	May 24, 2022	1784
User selection type: Role	May 28, 2022	1783
	May 25, 2022	1733
Role selected: Student	Mar 22, 2022	1382
Report generated: Jun 2, 2022 4:45 PM CEST	May 18, 2022	1347
$\frac{1}{10000000000000000000000000000000000$	May 17, 2022	1333





### Example 1. Course Signals (Purdue Univ.)

Aims to help students understand their progress at a course level

Data sources:

- ✓ Student performance
- ✓ Engagement (interaction with the LMS)
- ✓ Past performance and students characteristics

These components are weighted and fed into a predictive algorithm to produce a 'traffic light' showing how at risk each student is considered to be.

A range of interventions can be taken by the instructor. Arnold, K. E., & Pistilli, M. D. (2012). Course signals at Purdue. *Proceedings of the 2nd Int. Conf. on Learning Analytics and Knowledge*, 267–270. https://doi.org/10.1145/2330601.2330666







### Example 2. Multimodal LA

Aims to predict the final academic performance using multiplesource and multimodal data from blended learning environments.

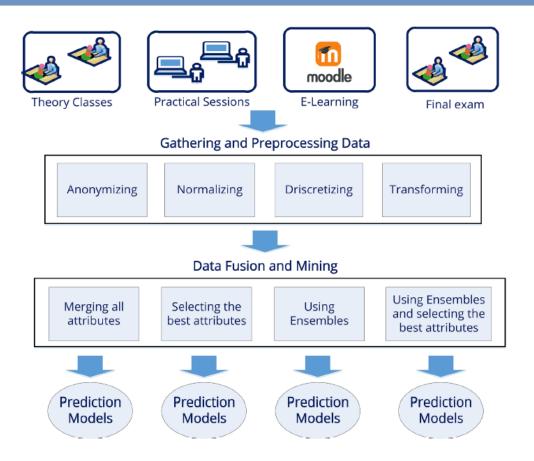
Data sources:

- ✓ Theory classes (data extracted from videos)
  - Attendance / Location / Attention / TakeNotes
- ✓ Practical sessions
  - Attendance / Score
- ✓ Online Moodle sessions
  - ✓ Quiz scores / Forum actions / Tasks uploaded / Time spent
- ✓ Final exam mark

Chango, W., Cerezo, R., & Romero, C. (2021). Multi-source and multimodal data fusion for predicting academic performance in blended learning university courses. *Computers & Electrical Engineering*, *89*, 106908. https://doi.org/10.1016/J.COMPELECENG.2020.106908







"The best prediction models show us that the level of attention in theory classes, scores in Moodle quizzes, and the level of activity in Moodle forums are the best set of attributes for predicting students' final performance in our courses."





### Current topics of interest of LA research

- ✓ Collaborative learning and teamwork group
- ✓ Dashboards and visual learning analytics
- ✓ Early warning systems
- ✓ Measuring self-regulated learning
- ✓ Game learning analytics
- ✓ Multimodal learning analytics
- ✓ Providing personalized feedback
- ✓ Writing analytics

Romero, C., & Ventura, S. (2020). Educational data mining and learning analytics: An updated survey. Wiley Interdisciplinary Reviews: Data Mining and Knowledge Discovery, 10(3). https://doi.org/10.1002/WIDM.1355





### Conclusions

According to the conclusions of a recent study about LA in European HE:

Teaching and support staff are found to be the main users of LA...

In contrast, there is little evidence of active engagement with students or using LA to develop self-regulated learning skills. We highlight the importance of grounding LA in learning sciences and including students as a key stakeholder in the design and implementation of LA.

Tsai, Y. S., Rates, D., Moreno-Marcos, P. M., Muñoz-Merino, P. J., Jivet, I., Scheffel, M., Drachsler, H., Delgado Kloos, C., & Gašević, D. (2020). Learning analytics in European higher education—Trends and barriers. *Computers & Education*, *155*, 103933. https://doi.org/10.1016/J.COMPEDU.2020.103933

An invitation to work together (instructors, administrators, data scientists, learning scientists and why not, students).



# THANK YOU

José V. Benlloch-Dualde jbenlloc@disca.upv.es

