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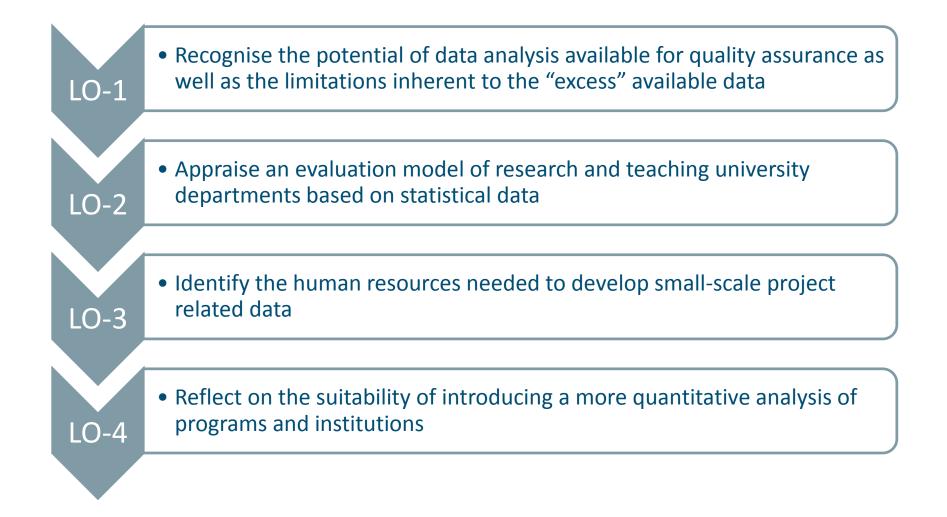
per a la Qualitat del Sistema Universitari de **Catalunya**

Big (and Small) Data meets Quality Assurance

Anna Prades, Maribel Quirós, Maria Giné, Lorena Bernáldez 10th European Quality Assurance Forum 21th November, London



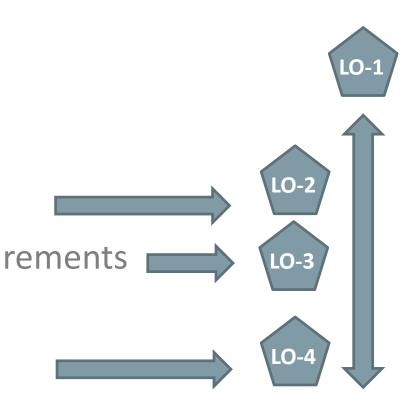






Context

- Why do we need data?
- Case study
- Human resources & skills requirements
- The future
- Discussion









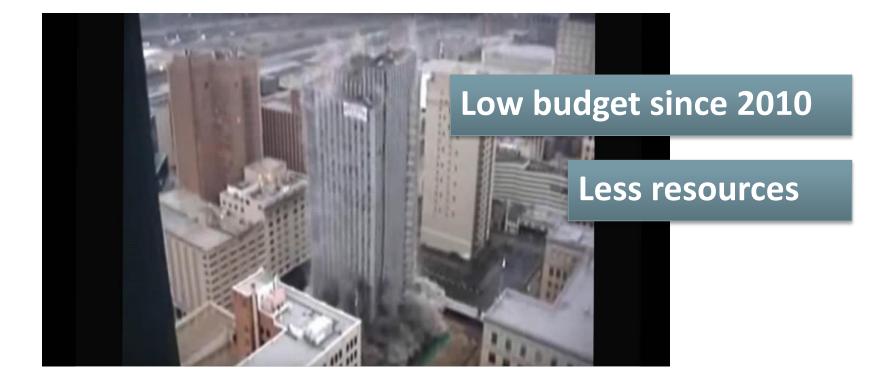
Agència per a la Qualitat del Sistema Universitari de **Catalunya**

The purpose of AQU Catalunya is the assessment, accreditation and certification of quality in the sphere of the universities and higher education centers of Catalunya.











UNEIX

Sources: universitiy pre-enrollment ddbb, research centers ddbb (CERCA), universities microdata (academic results) + teaching data + AQU surveys



unieix.

Graduates employment survey (Ba, Ma and PhD)

- Editions: 2001, 2005, 2008, 2011 and 2014 Population: 3 years after graduations (public + private universities)
- 153 variables (n= +/- 50,000)



Graduates satisfaction survey

- 1st edition: 2014
- Set of 22 common questions



Employers survey

- 1st edition: 2014
- 1000 employers (from Health and Education to SMEs)



Accreditation reports

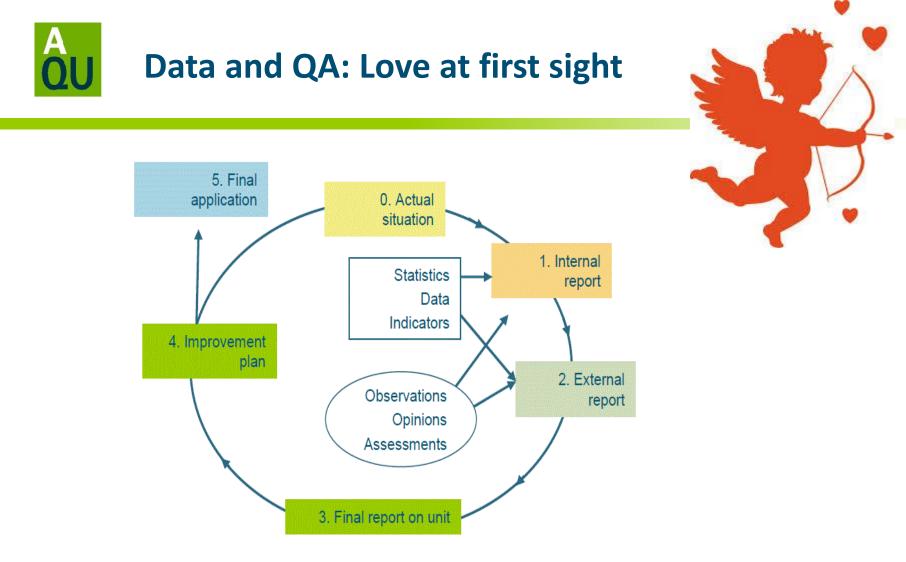
Up to date 150 assessment reports (SQL database)

OU Context: Uneix + Surveys -> winddat.aqu.cat





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"The quality assessment "spiral" is a circular and cyclical process which begins with collecting and systematizing information".

Westerheejden, 1996: 274

9



Why do we need to use data?

Standard 1.7

Standards and Guidalians for Ownity Assumates in the European Higher Education Area (Stat) Approximation for the State of State Approximation for the State of State of State Manual Association State Association (Spin States) Institutions should ensure that they collect, analyse and use relevant information for the effective management of their programmes and other activities

Quality culture



Data driven decision making



Turning data into **actionable** insights





It's cool!!!







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Subject area	Sub-area	N⁰
		Departments
	Fine Arts	6
Humanities	Philology	22
numanties	Philosophy	5
	Geography and History	21
	Political Sciences and Sociology	5
	Law	17
Social Sciences	Economics and Business	20
	Education	19
	Journalism and Communication and Library Science	7
	Psychology	11
Experimental Sciences	Biology	6
Experimental Sciences	Geology	8
	Mathematics and Physics	19
	Chemistry	10
	Biomedicine	21
Health Sciences	Pharmacy	4
	Nursing	7
	Medicine and Surgery	16
	Veterinary Science	4
	Architecture and Civil Engineering	12
Engineering & Architecture	Mechanical and Production Engineering	28
	Agricultural Engineering	5
	ICT Engineering	14
Total		287



Case Study

Туре	Code	Description
Research	RIND01	Sum of TOTAL income generated in 4 years by PhD- holders / number of PhD-holders
	RIND02	Sum of income from EUROPEAN SOURCES generated in 4 years by PhD-holders / number of PhD-holders
	RIND03	Sum of income from STATE SOURCES generated in 4 years by PhD-holders / number of PhD-holders
	RIND04	Sum of income from CATALAN SOURCES generated in 4 years by PhD-holders / number of PhD-holders
	RIND05	Sum of income from CONTRACTS AND AGREEMENTS generated in 4 years by PhD-holders / number of PhD-holders
	RIND06	Sum of income from COMPETITIVE FUNDING SOURCES generated in 4 years by PhD-holders / number of PhD-holders
	RIND07	Sum of income from NON-COMPETITIVE FUNDING SOURCES generated in 4 years by PhD-holders / number of PhD-holders
	RIND08	Current research premiums from regional government / no. people with current research premiums in the category Principal Investigator
	RIND09	Sum of TOTAL income generated in 4 years by: 2-FULL PROFESSOR, 3-SENIOR LECTURER, 16-UNIVERSITY SCHOOL FULL PROFESSOR
	RIND10	Sum of TOTAL income generated in 4 years by ALL OTHER TEACHING STAFF (not categories 2,3 or,16, above)

Fundraising

- Income generated
- Competitive vs not competitive
- National vs International

Research awards / distinctions

By number of PhDs



Case Study

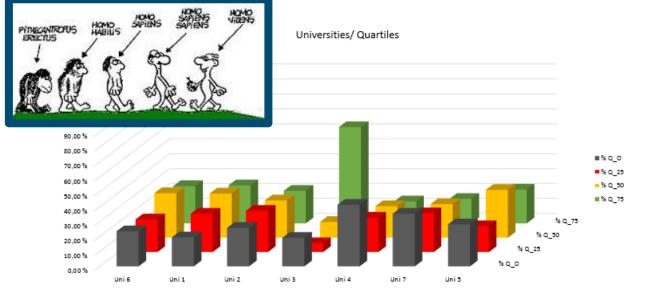
Туре	Code	Description
Teaching		Full-time teaching staff (A+B) with net teaching capacity > 0 / teaching staff (permanent + adjunct) * 100
		Part-time teaching staff (A+B) with net teaching capacity > 0 / teaching staff (permanent + adjunct) * 100
	PIND05	Permanent teaching staff >= 60 years / total permanent teaching staff * 100
	PIND06	Permanent teaching staff >= 45 years and < 60 years / total permanent teaching staff * 100
	PIND07	Permanent teaching staff >= 35 years and 45 < years / total permanent teaching staff * 100
	PIND08	Permanent teaching staff < 35 years / total permanent teaching staff * 100
	PIND10	Six-yearly research premiums from regional government * 6 / five-yearly teaching premiums from regional government * 5
		Total teaching and research staff + International researchers / Total teaching staff * 100
	PIND12	Total teaching and research staff + EU researchers - 15 + USA & CANADA / Total international * 100
		% permanent teaching staff in (23 Special Services and 22 Services) / teaching staff (permanent + adjunct)
	PIND15	% net teaching capacity of permanent teaching staff / potential teaching capacity of permanent teaching staff
	PIND17	% PhD-holders / total teaching staff
	PIND18	Age indicator (sum of all groups - over-60) / over 60
	PIND19	Distribution of classroom teaching hours; % according to teaching staff category

Teachers

- Academic categories
- Permanent vs non permanent
- Age
- National vs International







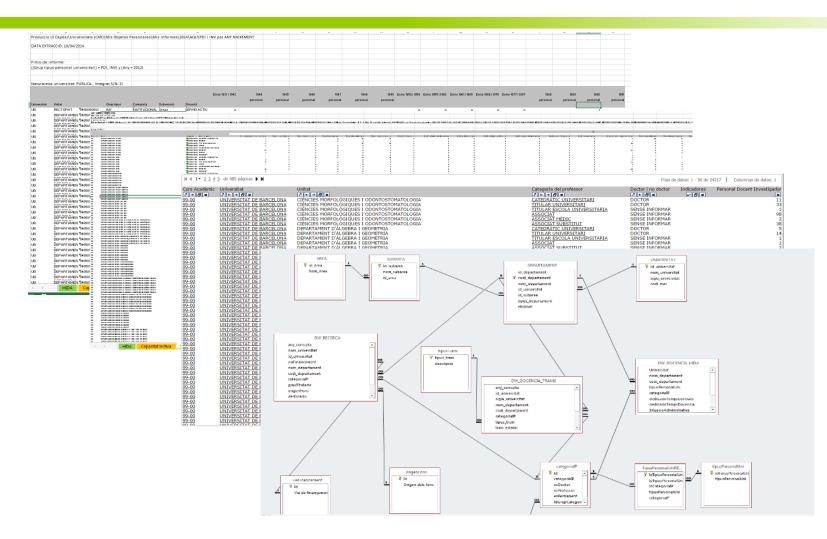
Universities

	N [⊉] Dept.	Nº Ind.	% Q4	% Q3	% Q2	% Q 1
Uni 6	106	2014	23,44 %	21,80%	29,79 %	24,98 %
Uni 1	57	1073	19,38 %	25,63%	29,54 %	25,44 %
Uni 2	42	798	25,69 %	27,57%	24,94 %	21,80 %
Uni 3	8	152	19,08 %	5,92%	10,53 %	64,47 %
Uni 4	24	456	41,23 %	23,03%	21,05 %	14,69 %
Uni 7	26	494	35,22 %	25,91%	22,47 %	16,40 %
Uni 5	24	456	28,07 %	17,32 %	32,02 %	22,59 %
Total	287	5443	25,79%	23,06 %	27,28 %	23,87%

Quartiles classification of the previous indicators

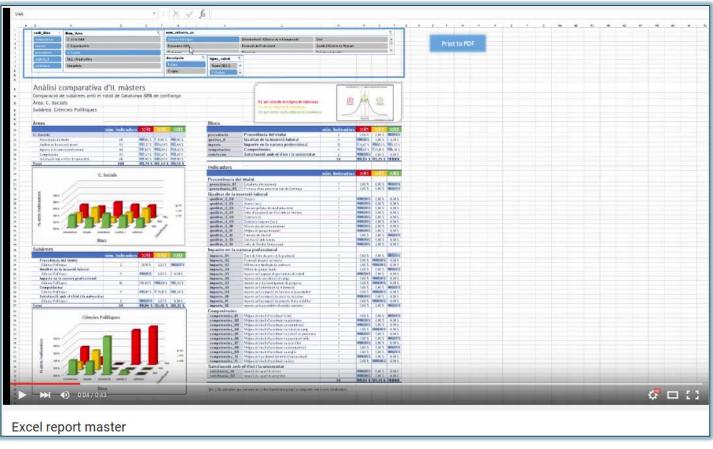


Case Study: the backoffice



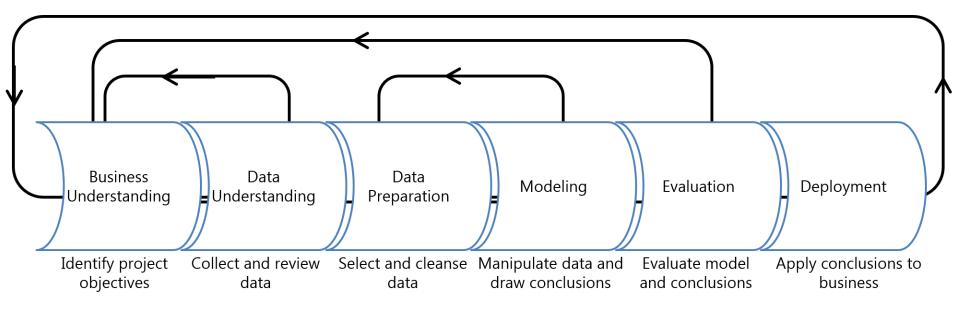


An extrapolation of the Case Study: Masters employment indicators



https://youtu.be/1fadu1fvf28





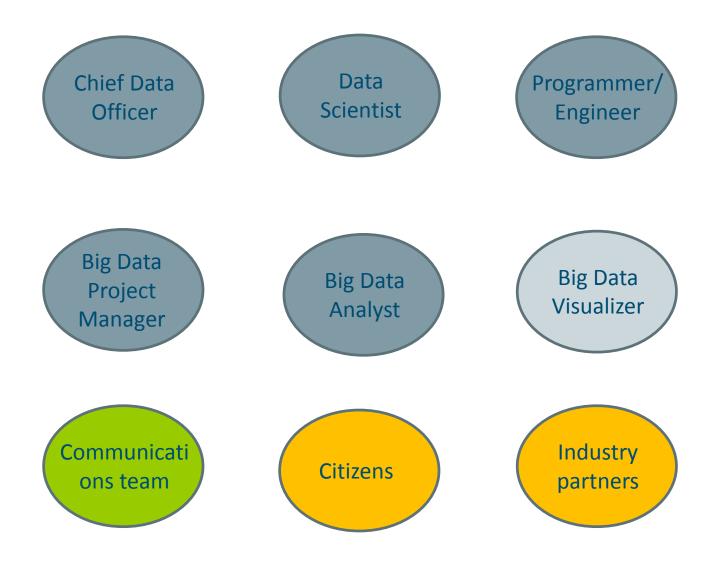
CRISP DM(2000)

Image from Data Science and Machine Learning Essentials (edX), by Cyntia Rudin and Stephen F Elston



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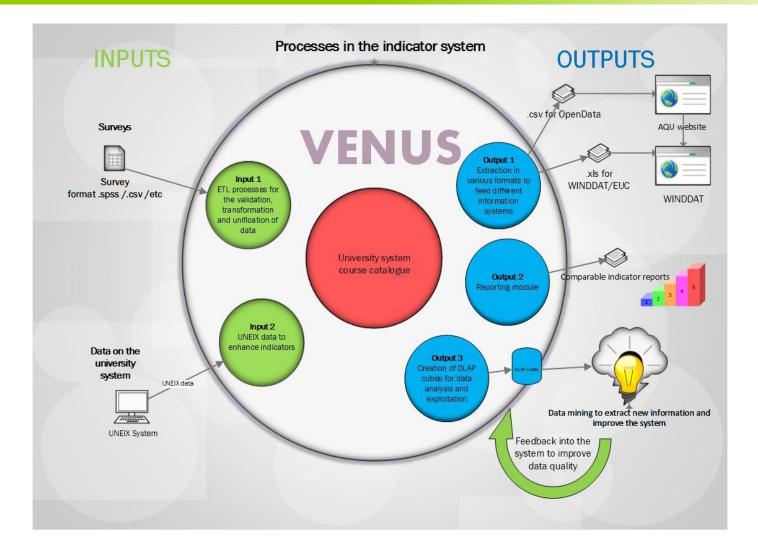
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The future: a preliminary list

- Predictive models of professional success
- Detect good practices regarding skills gap
- Real-time teaching staff renewal (upcoming retirements)
- Institutional KPI for teaching and research
- Predictive model of accreditation results: a selfassessment tool
- Possible use of existing data to simplify the assessment procedures







Thank you for your attention

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