

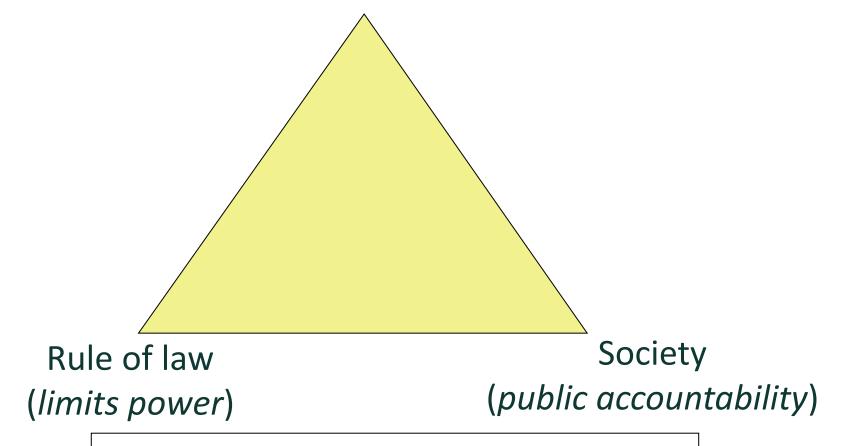
Best Practices on Open Data: the Brazilian experience

Gilberto Câmara (INPE, Brazil) Belmont Forum, co-chair

Foundations of modern democracies

State (power by technical staff)

Fukuyama (2011)



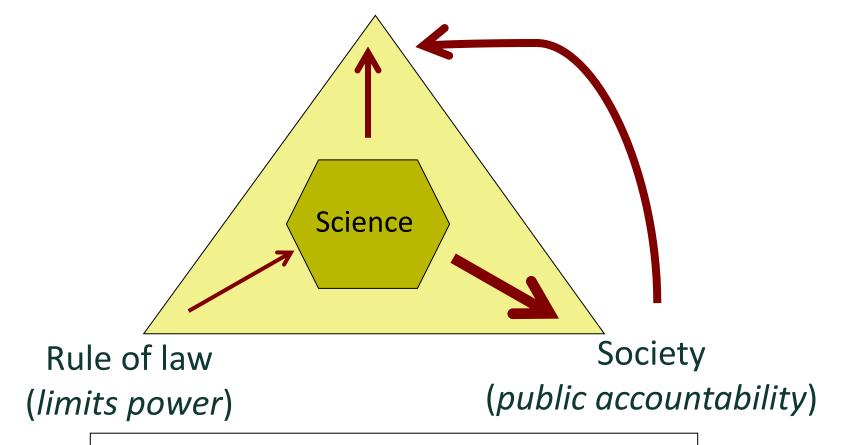


How is the interplay between science, public opinion and government?

Foundations of modern democracies

State (power by technical staff)

Fukuyama (2011)





How is the interplay between science, public opinion and government?



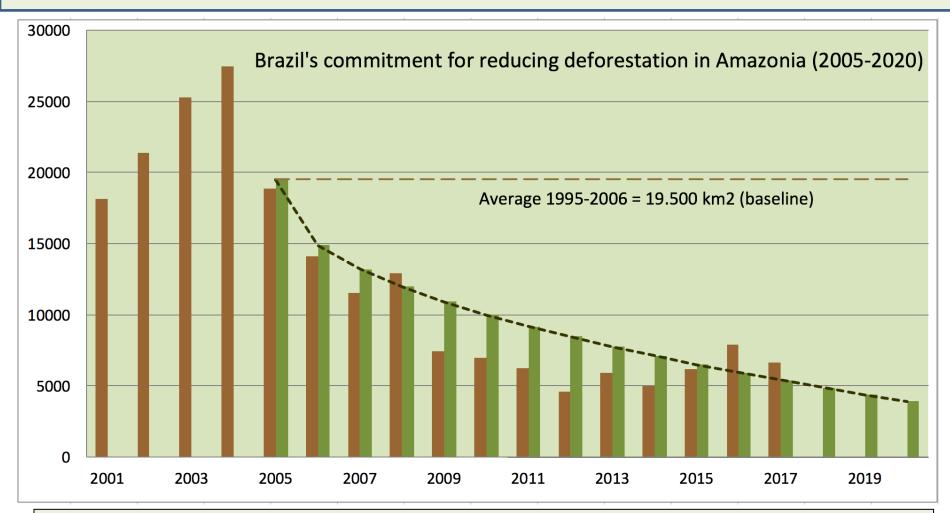
"A few satellites can cover the entire globe, but there needs to be a system in place to ensure their images are readily available to everyone who needs them. Brazil has set an important precedent by making its Earthobservation data available, and the rest of the world should follow suit."

Real-time Deforestation Monitoring



Daily warnings of newly deforested large areas

Transparency builds governance!

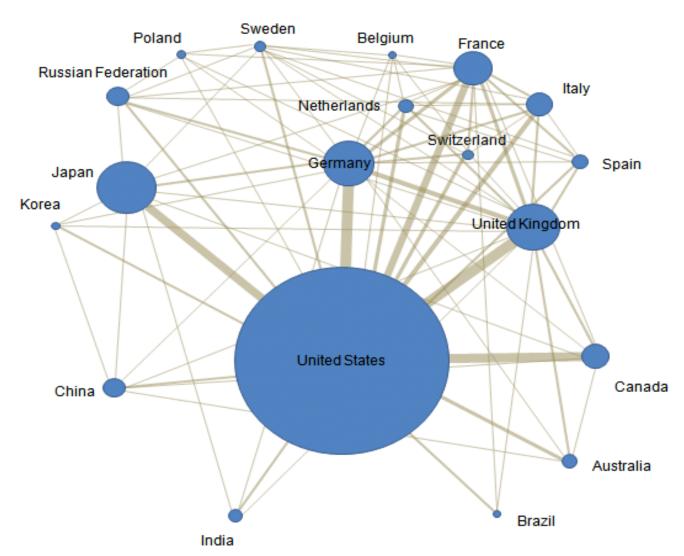


Deforestation in Amazonia (2001-2017): 27,000 km² to 6,200 km²





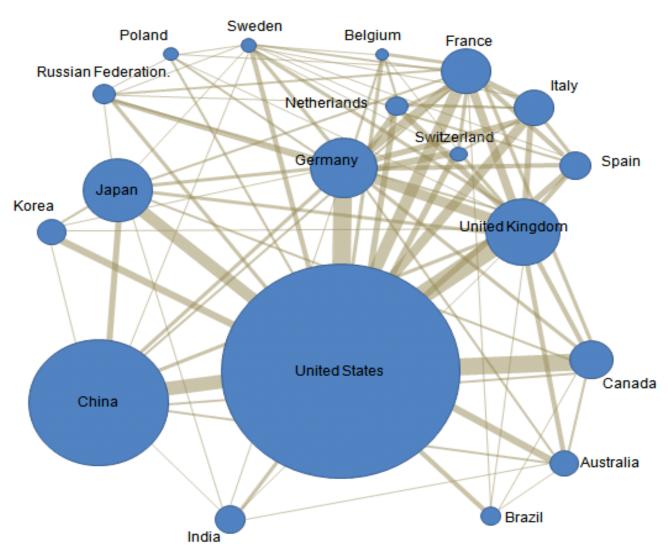
World Science (1998)



OECD



World Science (2008)



OECD







Comment | OPEN

The FAIR Guiding Principles for scientific data management and stewardship

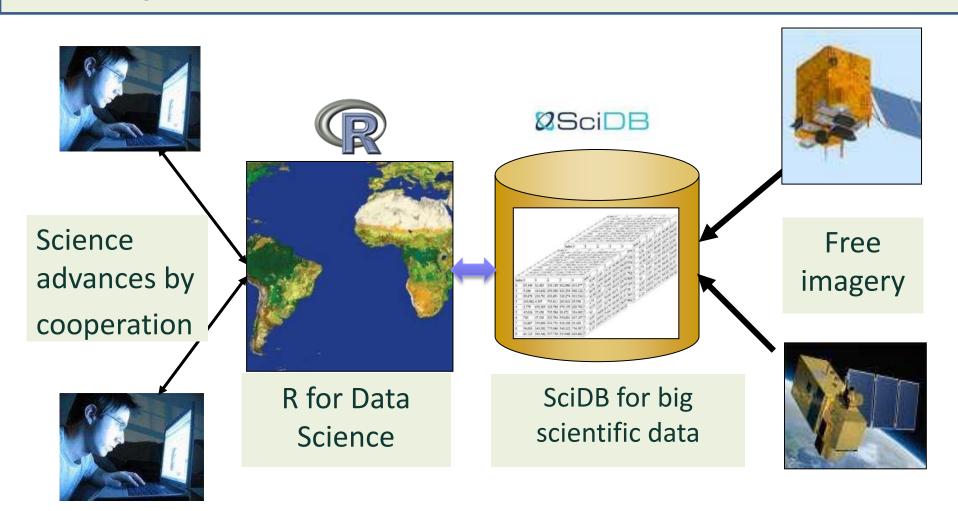
Mark D. Wilkinson, Michel Dumontier [...] Barend Mons [™]

Findable, Acessible, Interoperable, Reproducible





Hard problems need Science!



INPE's Data Cube: advancing eScience



Machine learning with big data

Google Trends, May 2012 - April 2017
Big Data vs Machine Learning search terms



Combine SVM, Random forest, Deep learning, Neural nets with satellite image time series



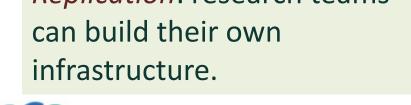
From User Requirements to Data Architectures

Needs of research community

Analytical scaling: algorithms developed at the desktop run on big databases

Collaborative work: share results with the scientific community.

Replication: research teams



System architecture choices



State-of-the-art statistical tools for Data Science

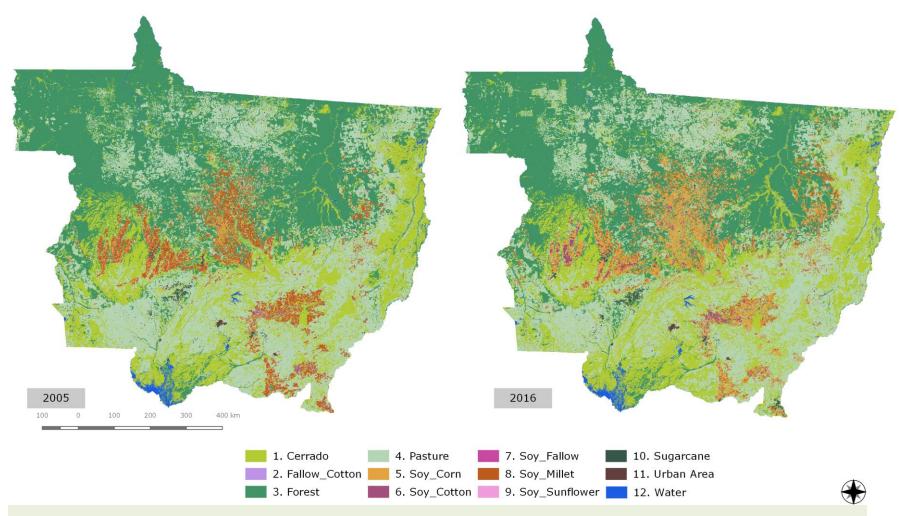


Data management for scientific data





Mato Grosso – Brazil's agricultural frontier



Land change dynamics (2001-2017)



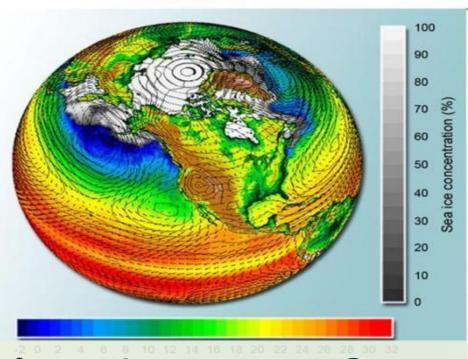
High accuracy for mapping land use change

	1	2	3	4	5	6	7	8	9	UA
1 Cerrado	393	0	0	12	0	0	0	0	0	0.97
2 Fallow-Cotton	0	33	0	0	1	2	0	0	0	0.92
3 Forest	1	0	136	0	0	0	0	0	0	0.99
4 Pasture	6	0	1	357	3	1	0	5	0	0.96
5 Soy-Corn	0	1	1	1	352	18	0	26	4	0.87
6 Soy-Cotton	0	0	0	0	13	376	0	4	0	0.96
7 Soy-Fallow	0	0	0	0	0	0	88	0	0	1.00
8 Soy-Millet	0	0	0	0	25	2	0	199	2	0.87
9 Soy-Sunflower	0	0	0	0	4	0	0	1	47	0.90
PA	0.98	0.97	0.99	0.96	0.88	0.94	1.00	0.85	0.89	

Mato Grosso – Brazil's agricultural frontier Overall accuracy of 94%





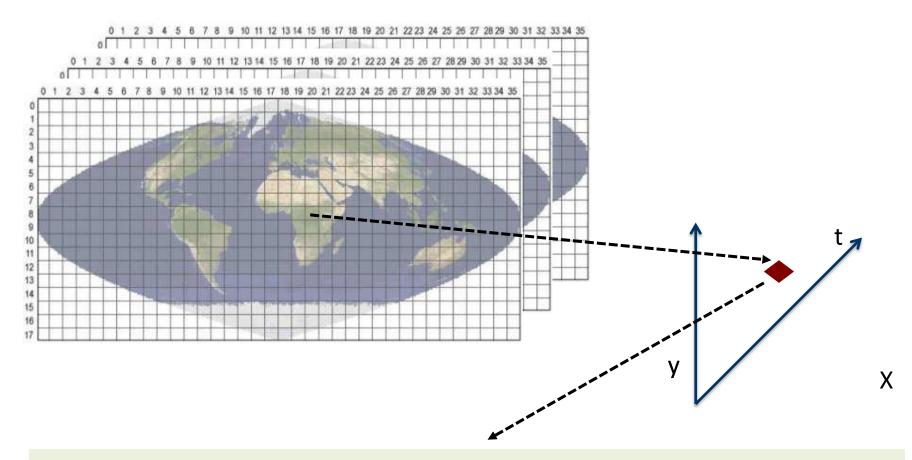


What do these data have in common?





Array databases: all data in a single array

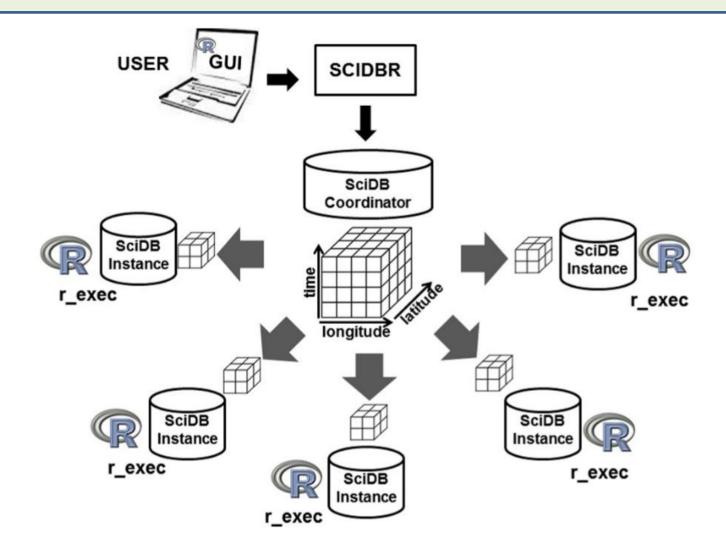


result = analysis_function (points in space-time)



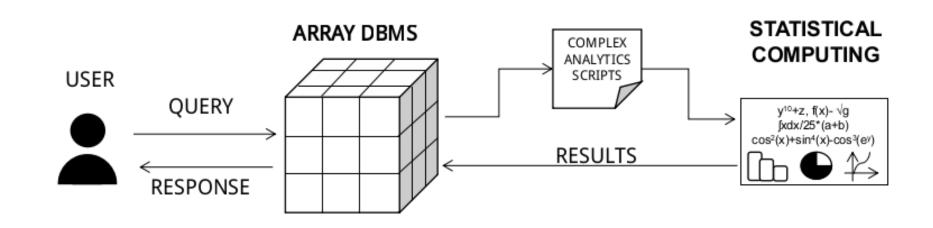


SciDB: efficient processing of R scripts





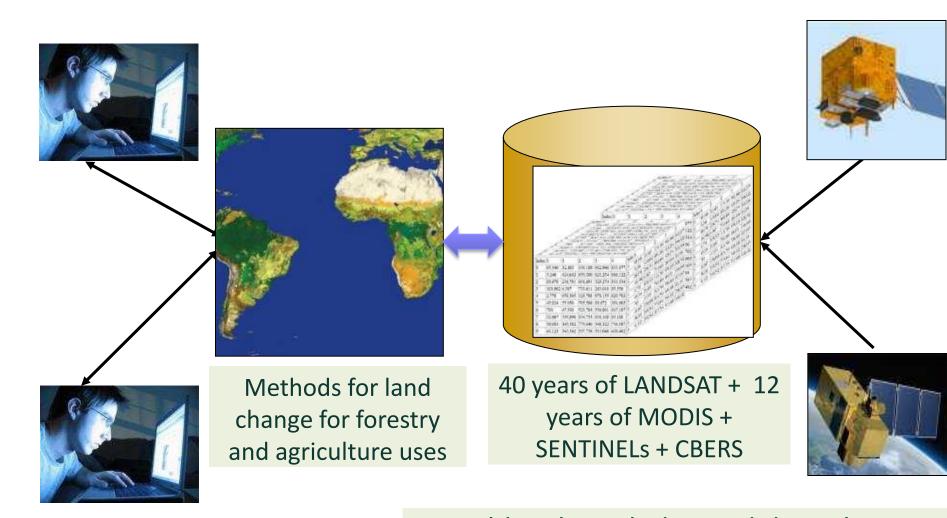
Flexibility with performance



Case Study	Area	Decision	Measures	Proc time	
	(km^2)	dimensions	(GB)	(hours)	
Mato Grosso	900,000	92	135	6	
Cerrado	2,050,000	92	308	13	



Global Land Observatory





Building knowledge and data about global land change