Next Steps for Big Data in Education

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My Current Role in Data-Intensive Research in Education

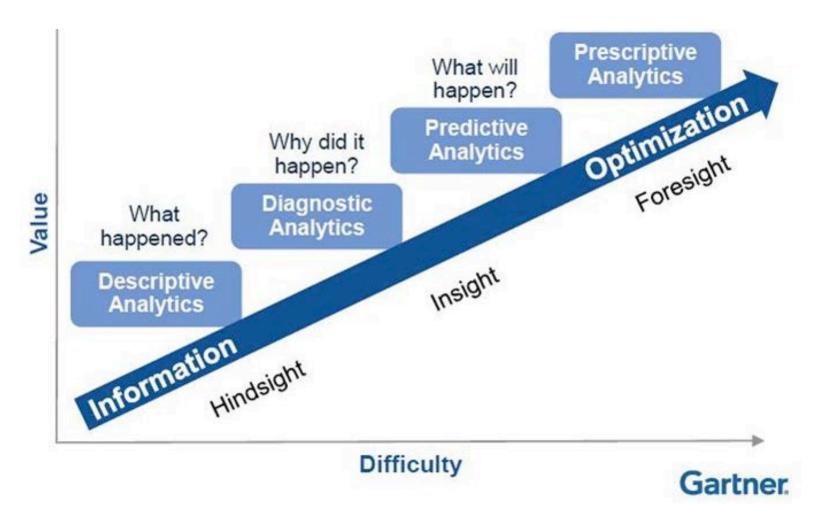
- Organized a two workshop sequence on dataintensive research for NSF and the field: insights from relatively mature data-intensive research initiatives in the sciences and engineering were applied to nascent dataintensive research efforts in education
- Confront "big data" issues in my design-based research in ecosystems science education

Definitions

- <u>Big Data</u> is characterized by the ways in which it allows researchers to do things not possible before (i.e., Big data enables the discovery of new information, facts, relationships, indicators, and pointers that could not have been realized previously).
- <u>Data-intensive research</u> involves data resources that are beyond the storage requirements, computational intensiveness, or complexity that is currently typical of the research field.
- <u>Data science</u> is the large-scale capture of data and the transformation of those data into insights and recommendations in support of decisions.

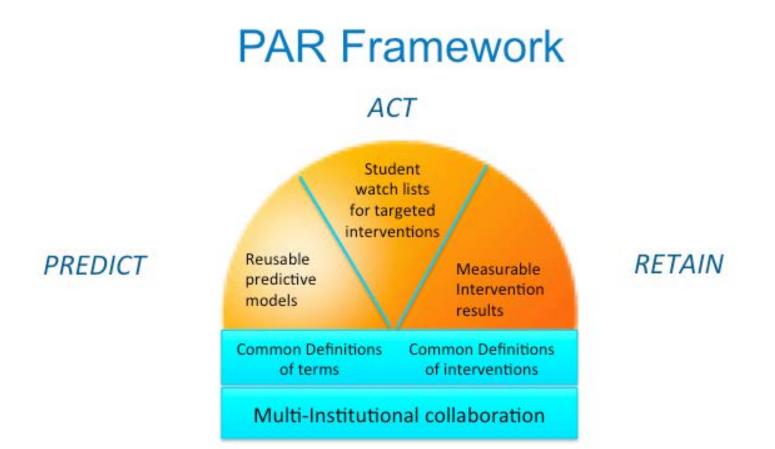


From Hindsight to Foresight



From Description to Prescription

- Determine students' probabilities of failure (*predictions*)
- Determine which students respond to which interventions (*uplift modeling*)
- Determine which interventions are most effective (*explanatory modeling*)
- Allocate resources accordingly (cost benefit analysis)



Scalable, cross-institutional improvements that support each individual student's success

PAR Data Inputs

Student Demographics & Descriptive

Gender Race Prior Credits Perm Res Zip Code HS Information Transfer GPA Student Type

Student Course Information

Course Location Subject Course Number Section Start/End Dates Initial/Final Grade Delivery Mode Instructor Status Course Credit

Student Financial Information

FAFSA on File – Date Pell Received/Awarded – Date

Student Academic Progress Curent Major/CIP Earned Credential/CIP

Course Catalog

Subject Course Number Subject Long Course Title Course Description Credit Range

Lookup Tables

Credential Types Offered Course Enrollment Periods Student Types Instructor Status Delivery Modes Grade Codes Institution Characteristics

Possible Additional **

Placement Tests NSC Information SES Information Satisfaction Surveys College Readiness Surveys Intervention Measures

** Future

Descriptive and Predictive Insight

PAR Benchmarks Descriptive Analytics

Cross Institutional

Student/degree/major level insight into:

- What *did* the retention look like for students entering in the same cohort
- How does your institution compare to peer institutions / institutions in other sectors
- 3. How *did* performance vary by student attributes

PAR Models Predictive Analytics

Institutional Specific insight into:

- 1. What students are being retained over time?
- 2. Which students are currently at risk for completing and why?
- 3. Which factors are directly correlated to student success?
- 4. What is the predicted course completion rate for a particular program?



Outcome Measures

- Credits Earned
- Credits Earned- No Dev Ed
- Credit Ratio
- Credit Ratio- No Dev Ed
- Retention
- Credentials Earned (Any, Bachelors, Associates, Certificate)

Predicting retention aimed at taking action - finding the most important factors

Factor Map ATTRIBUTE_VALUE												
HIGH_SCHOOL_GPA_(1.618, 2.864] GPA		A_(2.509, 3.126]				GENDER_Male		STATUS_Full		VETERAN_Yes		
ifies	ded	i.	ciai s	ness			_		GE	Time		,
MAJOR_Humanifies	MAJOR_Undecided	MA IOP Social	Sciences	Busi	Busi							
MAJOP	MAJOF		Sc	MAJOR Business								
GPA_(1.180, 2.509) MAJO		R_He	alth						Γ			
GPA_<=1.180						_				MAJOR_Physical and Mathematics Sciences		al and ciences
								MAJOR_Biological Sciences	STATUS_Part Time		VETERAN_No	
6.58 1 0.2 0.4 0.6 FAIL_RATE RISK_RATIO												

Risk Factors							
ATTRIBUTE_VALUE	FAIL_RATE 🔻	RISK_RATIO					
GPA_<=1.180	0.80	6.58					
HIGH_SCHOOL_GPA_<=1.618	0.78	3.71					
HIGH_SCHOOL_GPA_(1.618,	0.52	2.48					
GPA_(1.180, 2.509]	0.50	4.12					
STUDENT_TYPE_Community	0.50	2.28					
MAJOR_Health Sciences	0.50	3					
STATUS_Part Time	0.46	1.53					
MAJOR_Humanities	0.45	2.75					
DELIVERY_MODE_Online	0.43	1.67					
MAJOR_Undecided	0.42	2.56					
RACE_ETHNICITY_Asian	0.42	1.22					
MAJOR_Social Sciences	0.41	2.48					
MAJOR_Business	0.40	2.42					
COURSE_CAMPUS_Central	0.39	1.24					
VETERAN_No	0.39	1.46					
GENDER_Female	0.37	1.08					
RACE_ETHNICITY_Black	0.37	1.08					
RACE_ETHNICITY_White	0.36	1.04					
GENDER_Male	0.35	1					
RACE_ETHNICITY_Hispanic	0.34	1					
COURSE_CAMPUS_West	0.33	1.05					

Key Research Questions

- Can we <u>detect</u> problems that students are having as they are happening, through automated analysis?
- Can we provide real-time feedback to students and educators in response to the problem detection?
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<u>3 E's of Immersive Learning</u>

Engagement

Students are motivated to do well, see the relevance of their learning, and increase in self-efficacy

Evocation

Immersive interfaces can evoke a wide spectrum of authentic performances with embedded support

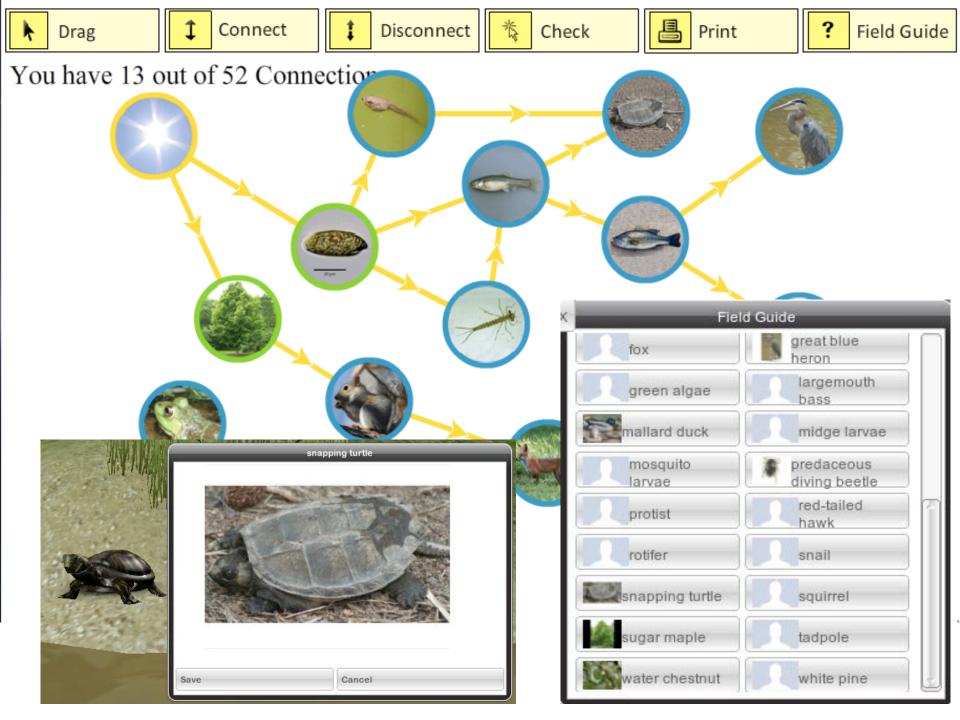
> Evidence

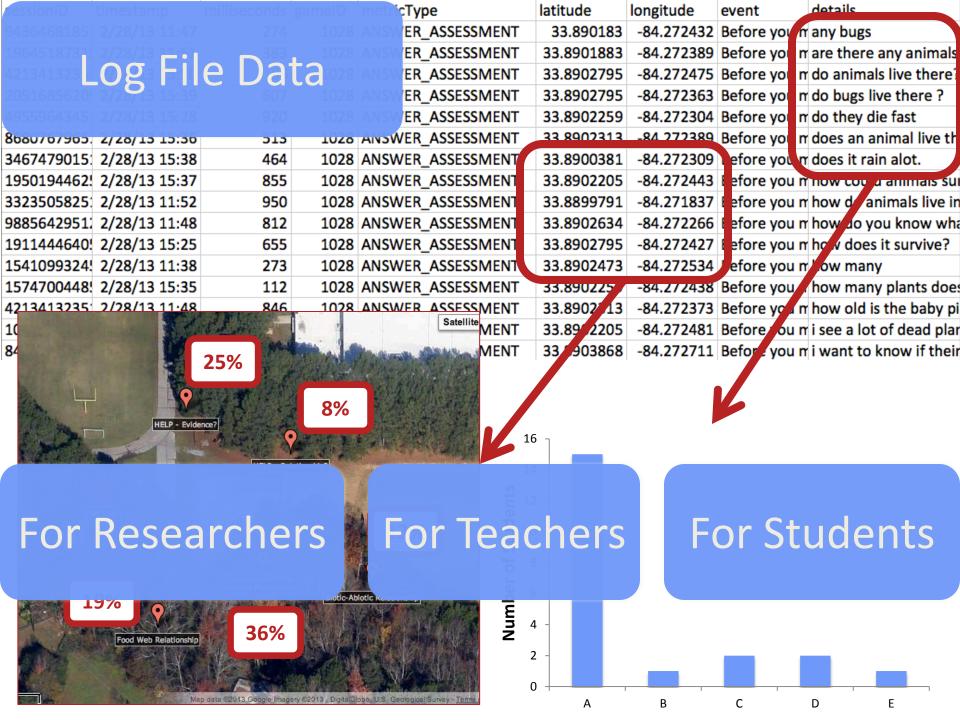
Log files, chat logs, shared notebooks, and similar artifacts provide a rich evidentiary trail



EcoMUVE – Multi-User Virtual Environment







Det alface and 5

Collaborative construction of concept maps





(Conner Flynn)

Augmenting Real World Ecosystems http://ecomobile.gse.harvard.edu

GoPro Cameras Capture EcoMOBILE Experience



P2: Now we need to write two things we see that could affect dissolved oxygen.

P1: <mark>Plants</mark>.

S: Plants.

P2: Plants. And...you guys would rather say...rain, or the dead matter?

S: Mm, dead matter, maybe. 'Cause-

P1: (Why the) dead matter is bacteria?

S: —Yeah. The bacteria. And we don't know, you know, how long...this has been...

P2: You got plants already? Plants, 'cause they release dissolved oxygen into the water.

P1: This could

P2: ... Provide food for bacteria, increasing their population and increasing their need for dissolved oxygen.

P1: The bacteria and -

P2: And um -

P1: And causing an increase in population.

P2: Yeah, increasing their population and their need for dissolved oxygen.

S: [Student talking to other student] Quinn.

P1: Um, provide food for bacteria, increasing population?

P2: Mhm. [Partner 1 continues typing in Evernote]

Evernote: **Plants** could release dissolved oxygen into the water and dead matter could provide food for bacteria, increasing the bacteria population and their need for dissolved oxygen.



EcoMUVE

- MUVEs promote self-efficacy in science
- Simulate experiences otherwise impossible in school settings.
- Explore time and scale
- Opportunities to take on roles, work in teams
- Shared immersive experience that contextualizes learning and supports inquiry

(Ketelhut et al. 2010, Metcalf et al. 2011)

EcoMOBILE

 Greater fidelity and sensory richness, physical interactions with organisms and environments.



- Self-directed collection of real-world data and artifacts.
- Facilitated use of cameras, recording devices, probes, GPS, mapping, graphing, augmented reality.



Inquiry skills?

Collaboration?

Leadership?

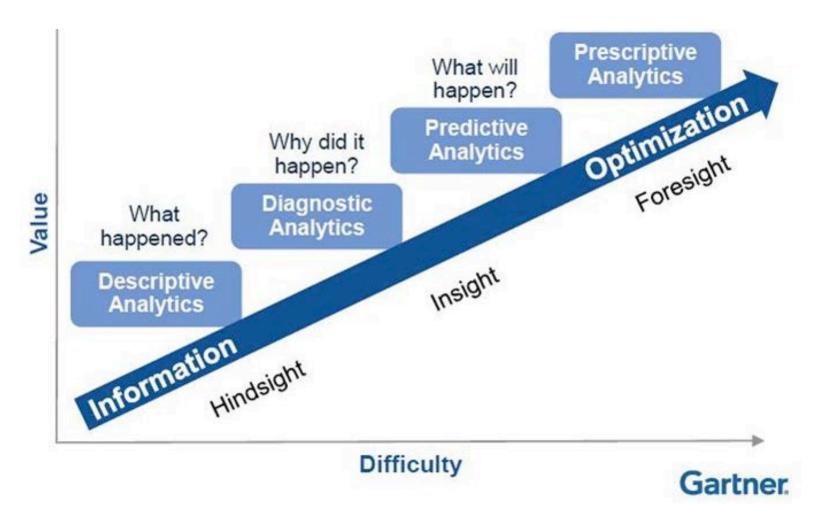
Self-efficacy?

Metacognition?

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From Hindsight to Foresight



Imagining the Possible

Edith S. Gummer Ewing Marion Kauffman Foundation







NSF Advancing Data Intensive Research in Education June 1-2 2015



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District and School Information

Type	Name	Description
-	District Azrestation	Haw Data: Eurrent accreditation datus for Districts.
180	District Calendar Days and Hears	Rev Dita: Comparison of district's rates par information.
-	School Detrict by Representative District	List of achoel districts by Weasuri Representative Datrict,
*	Actual District by Renatorial District	List of school extricts by Hissouri Benatorial District.
-	School District by Congressional District	Lait of school districts by Divited States Congressionel District
1	School Didnitte with Congruttional District	List of School Districts with their United States Congressional Detrict
-	School Districts Map - County Houndaries	Hap of Malalan school patricts with county boundaries.
-	School Districts Mag - Heuse District Boundaries	Hap of Masouri school districts with House district boundaries.
-	Activate Districts Hap - Senate District Reandance	Hap of Hazzan school detrotic with Senate district boundaries.
1	Schull Districts Map - Congressional District Representation	May of Missian school districts with compressional portrol bioinducies.
8	District Enrollment by Groom	Raw Deta: District enrollment by grade level for 1991-2014 (sizte as of 10/25/2014).
3	District En/olimient 2007-2013	Raw Data: Hulb-year enrolment by district.
	Endrict Braudiment 2014	have Gata. District divisionent by illight - ranked by percent of total Mate escalment.
6	Detroit Dropout and Graduation Rates 2007-2001	Rev Deta: 3007-3011 Dropout and Graduation Rates by Oetrict,
-	Census of Technology 2011	Report: Summary Report of annual Centus of Technology measures.

Macro-Level Data

District and Building Graduation and Dropout Indicators

Tipe	Name	Description
181	District Annual Dropault Rate	Raw Ceta: District dropout information, Tecluites total dropouts and next intrakouts.
8	Detrict Graduets Analysis	Kaw Data: Datrict graduate analysis, includes placement status or previous year's productes.
151	District Distantion Kate	Navi Cata: District graduation rates, Includes total graduates and race breaksure.
181	Four-Year District Graduation Rate	Raw Deta: Feur-Vear State and District Graduation Rates
- 23	Hair-Hear Clatrict Graduation Rate	PGP: Huge-Year State and Electrics Graduation Hates
181	Building Annual Dropout Rate	Raie Cete: Building (impout information, Includes total dropouts and race breakauts,
8	Building Graduate Analysis	Rain Cetz: Bullding graduate analysis, lockulas planement atatus en previews year's graduates.
81	Building Isratuation Kata	Raw tista: Building graduation rates, includes total graduates and race breakouts.
183	Oraduptes 18 Hanth Postsecondery Follow-up (3P3P- C11)	Report: Number of graduates who have enrolled in Polatecondary education within 16 membro of graduation. This report is being published as on

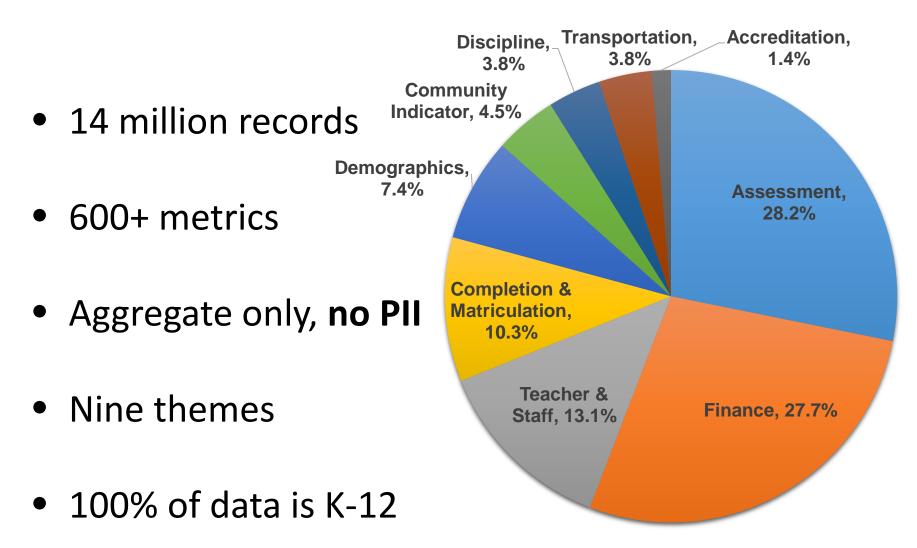
District and Building Student Indicators

100		
191	Building Propertions Attendance Rate	Raw Date: Building propertional attentions rate (student), attending at least 90% of the time), teams, 2009-2013, includes 4-8 and 9-12 rates.
8	Building Destgline Incidents	Haw Data: Building classifies information, includes annalment, surpline of incidents, and rates by offenia and removal types.
8	Building Demographic Cata	Raw Opta: Building demographic data. Includes annuliment information by cace, Data as at 15/05/2014 .
1981	Building Attendance Rate	Naw Cela: Bulleting attrivulance rates.
8	District Proportional Attendance Rate	Raw Data: District propertienal attivistance rate (students attending 80% of the time). Fears 2009-2013. Includes K-6 and 9-12 rates.
193	District Discipline Incidents	Ray Data: Datest classifier information, trickates enrollment, number of incidents, and refers by offerin, and removal types.
131	District Demographic Data	Raw Date: District demographic data. Includes excellenced information by rece. Only as of DEDM/2014
(8)	District Attendance Rate	Raw Data: District attandance rate. Includes K-EJ and K-B rates.
Type	Name	Cestorphies

District and Building Education Staff Indicators

Type	Name.	Description
3	Elebrich Faculty Information	Raw Data: Detrict faculty interhabors. Inclusion assary, degree, and years of experiences. Data as of 10/09/2014
3	District Student Staff Ration	New Date: District staff ratios, incluse student to inschers and inschers to administrator ratios, Date as of 10/05/2014
18	Building Faculty Information	Raw Data: Building faculty information, includes asjary, plagmer, and years of experience. Data as of 10/09/2014
81	Bullong Raters Staff Katon	Ree Data: Building staff rates. Unclude student to teachers and student to advenserator rates, Cata as of 10/09/2014
161	Building Certification	Raw Ceta: Building bracker certification information. Includes number of trachers with a valid certificate and highly qualified trachers.
(8)	District Certification	Raw Data: Building teacher certification information. Includes number of teachers with a valid certificate and highly qualified teachers.

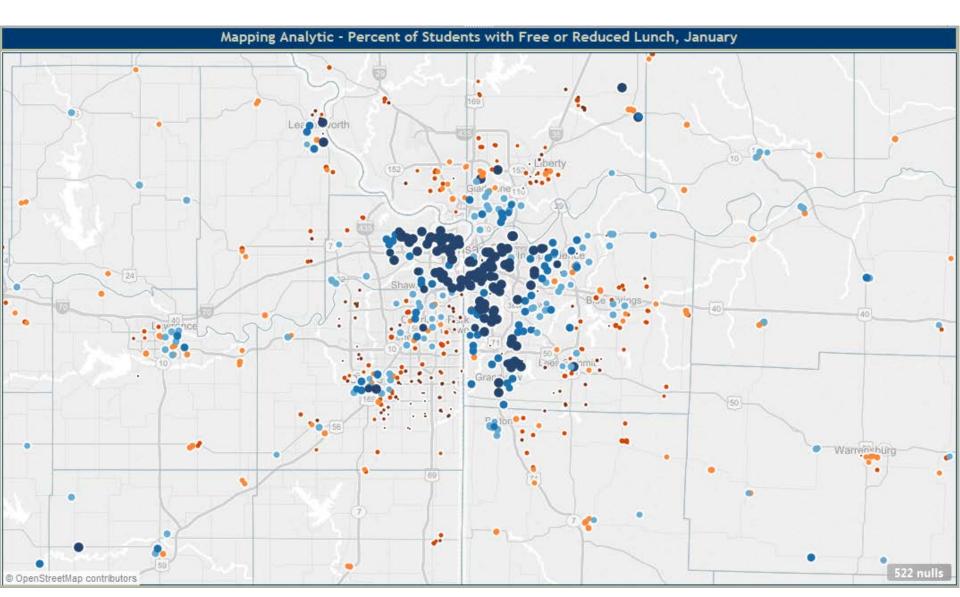
The Data Behind EdWise



Beginning of School Year

2013

Tool 1: Mapping



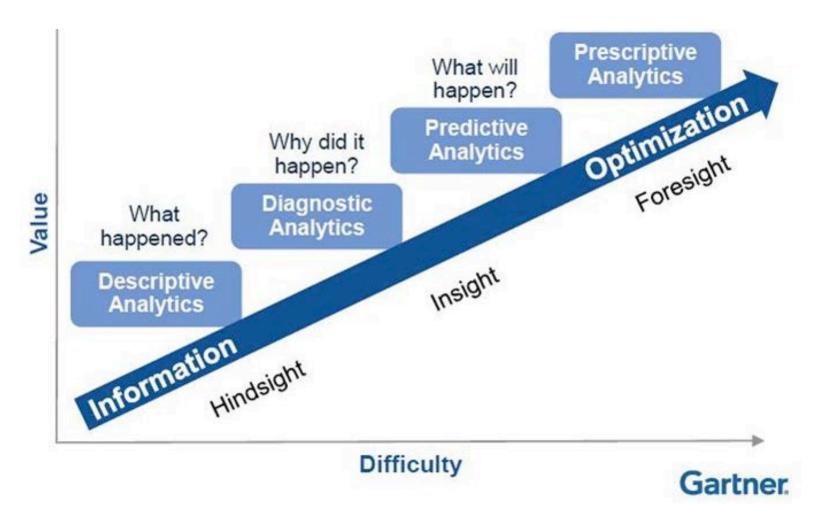
Tool 2: Find a District/School

Metric Measurement Level Beginning of School Year State Grade/Assessment Group Distri	cts/Schools /	Matching Selected Metrics			
Building (All) (All)					
Kansas	Abilene/Ab	vilene High School			
Metric Measurement Level Beginning of School Year State Grade/Assessment Group	Districts/S	chools Matching Selected			
District	State	District/School			
Metric Category [A]	Missouri	Ferguson-Florissant R-II			
Fall Enrollment		Hazelwood			
		Kansas City 33			
Metric Value [A]		St. Louis City			
10000 50749					
Metric Category [B]					
Students Per Teacher					
Metric Value [B]					
0 15					
Metric Category [C]					
Percent of Students with Free or Reduced Lunch, January					
Matria Malua (C)					
Metric Value [C]					
0.5 31.167					
Metric Category [D]					
Percent of Non-White Students within the Fall Enrollment					
Metric Value [D]					
0.5					

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Questions to Ponder

- To what types of behavioral data could we now apply these methods?
 - Micro-level data (e.g., each student's second-by-second behaviors as they learn)
 - Neso-level data (e.g., teachers' patterns in instruction; students' patterns in retention)
 - Macro-level data (e.g., aggregated student outcomes for accountability purposes) Gummer's work with EdWise
- What are the barriers to collecting, storing, sharing and analyzing these data?
- How can we build human and organizational capacity to use evidence-based findings effectively?