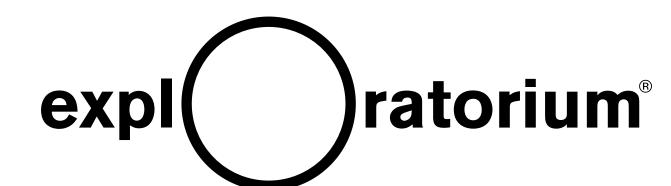
# **Educational Data Mining Approaches for Digital Libraries**

Mimi Recker, Beijie Xu, Bart Palmer *Utah State University*Sherry Hsi, Rob Rothfarb *Exploratorium*Project Site: http://edm.usu.edu









#### **RESEARCH GOALS**

- Collect Web usage data in the Instructional Architect and the Exploratorium's Learning Resources Collection and apply Web metrics and data mining techniques to infer teacher behaviors and learning online.
- Contribute new knowledge regarding Knowledge Discovery from Data (KDD) and the application of Educational Data Mining (EDM)to digital libraries.

### **DATA SOURCES**

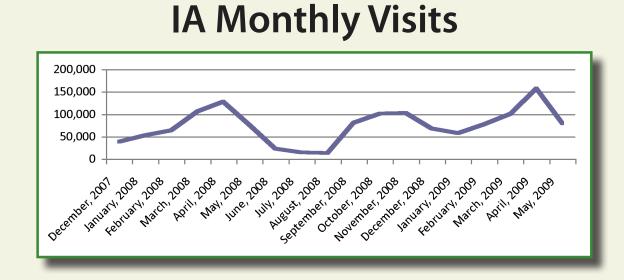
Instructional Architect (IA)	Exploratorium LRC (ELRC)	Other Data Sources
Web usage: Web server log Google Analytics IA relational database  Other: Registration profile Online survey	Web usage: Web server log Google Analytics  Other: Online survey Search term capture Interviews (tbd)	U.S. Census Data  NCES Demographic Data

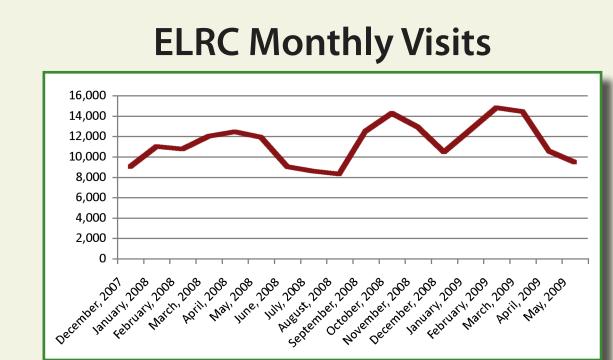
## WEB METRICS

Metric	Definition
Visits	Sequence of user transactions from a single IP within a certain time period (considered a more robust measure than "hits")
Unique visitors	A measure of a site's true audience size
Page views	Number of times Web pages are accessed during a single visit, including repeated viewing of the same page
Visitor origin	Country or region of visitor origin
Operating Sys/Browsers	Operating system and browser used by visitors
Length of visit	Time spent by visitor on site
Entry/Exit Pages	First and last pages accessed by visitors
Referrer	Last page the user visits before landing on the site
Bounce rate	The percentage of visitors who "bounce" away to a different site

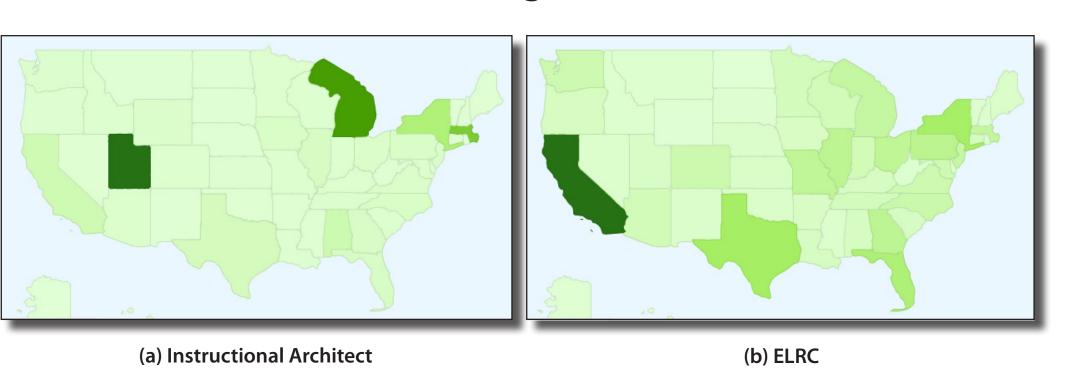
This material is based in part upon work supported by the National Science Foundation under Grant Number 0840745. Any opinions, findings, and conclusions or recommendations expressed in this material are those of the authors and do not necessarily reflect the views of the National Science Foundation.

#### **EXAMPLE 1**



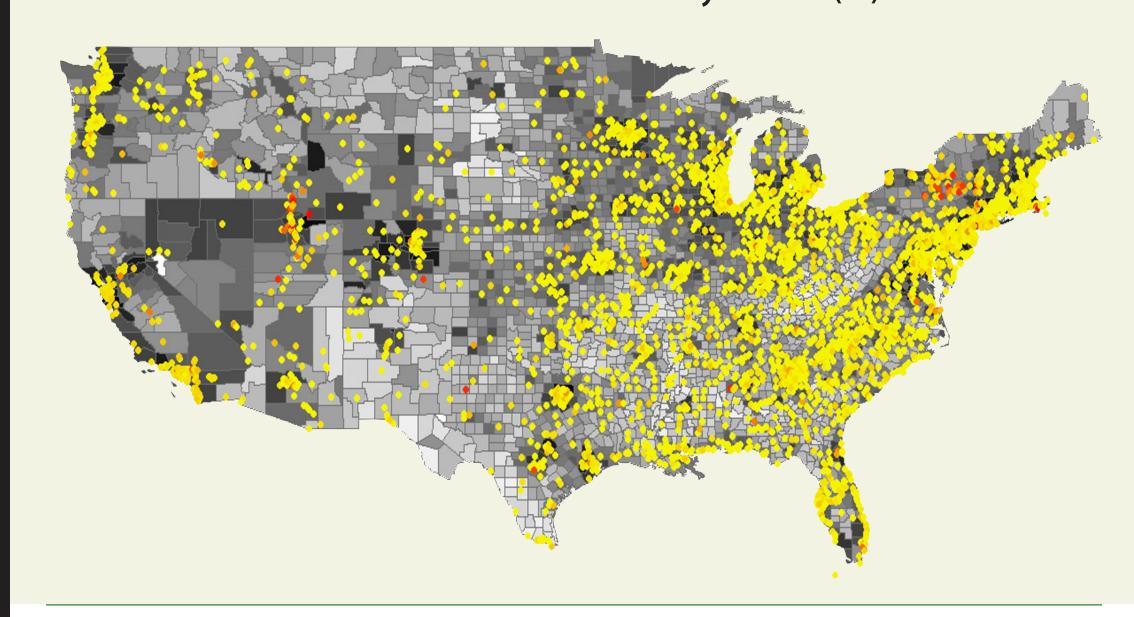


## **EXAMPLE 2**Visits from the Contiguous United States



#### EXAMPLE 3

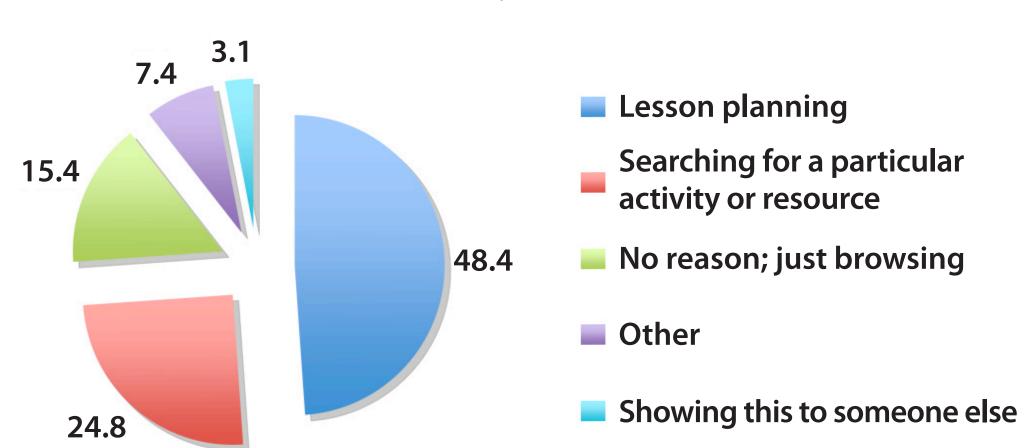
GeoAnalysis
Visits Overlaid on Median Family Income (IA)



#### **EXAMPLE 4**

Combining with Pop-up Survey in Application (ELRC)

Visitor Purpose (July-Sept 2009, N = 714)



#### **EDUCATIONAL DATA MINING**

#### **Knowledge Discovery from Data /Ed Data Mining Process**

#### **Phase I** Data Preprocessing

- Data cleaning and path completion
- Data integration
- Data selection
- Data transformation

**Phase II** Applying Data Mining Algorithms

**Phase III** Interpretation and Post-processing; Evaluation and presentation

#### **DATA MINING APPROACHES**

Approach		Application for the IA		
Algorithm	Usage	IA data	Intention	
Clustering	Cluster data into groups	Use of resources Use of projects	Identify teacher groups	
Association	Examine the relations among variables	Online behaviors		
rule		Use of resources Use of projects	Identify the relations among	
Sequential pattern	Examine the pattern in time-series transactions	Online behaviors	teacher-related features	
		Click-stream data	Identify the IA pages often accessed together	
	l l			

## LATENT CLASS ANALYSIS— CLUSTERING RESULTS

		Cluster <b>1</b>	Cluster <b>2</b>	Cluster <b>3</b>	Cluster <b>4</b>	Cluster <b>5</b>
Cluster Size		0.29	0.24	0.21	0.16	0.10
Indicators	Range					
percent_student_projects	0–1	0.96	0	0.34	1	0.44
percent_private_projects	0–1	0	1	0.29	0	0.32
percent_copy_projects	0–1	0	0.15	0.30	0	0.32
percent_public_projects	0–1	0.96	0	0.40	0	0.43
Visit stickiness	0.22-8.93	1.16	0.62	0.99	0.71	2.98
<b>Project stickiness</b>	0.15-8.06	1.99	0.20	0.66	0.32	2.12
Resource stickiness	0-29.91	1.06	0.39	0.77	0.57	4.22
Browse others	0–1	0.44	0.44	0.20	0.59	0.21

Cluster 1: Goal-oriented, Willing-to-share

Cluster 2: Inactive, Less motivated

Cluster 3: Lukewarm performers

Cluster 4: Goal-oriented, Less willing-to-share, Less devoted

**Cluster 5:** Productive, Active, Adapters