



Make-A-Vis: Learning Sciences Research

Advancing Public Understanding of Scientific Data

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The Creativity Labs @ UCI

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Overview

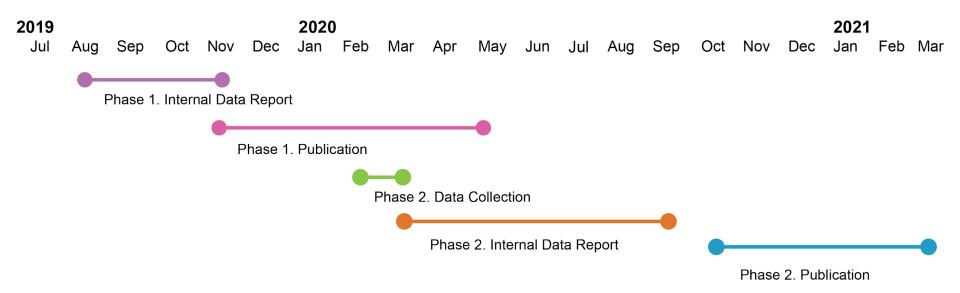
Overview of Research

Phase 1: What type of visualization personalization of the exhibit design leads people to initially engage with and maintain engagement with a data visualization?

Phase 2: How can we increase museum visitors' data visualization learning and literacy? What role does digital and physical construction play in improving learning outcomes?



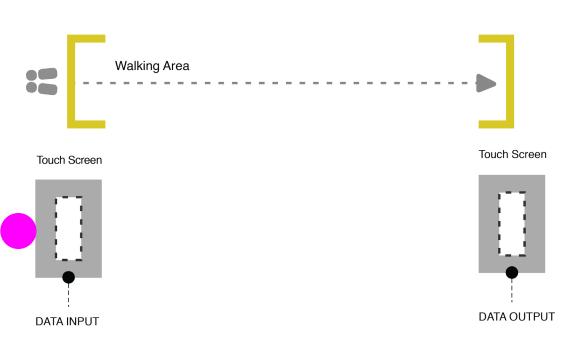
Timeline of the Phases



Overview of Phase 1

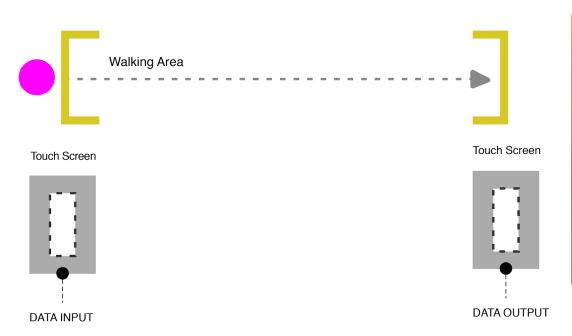
- Importance of data visualization literacy
- Data visualization is frequently taught with curated data sets (e.g., Lyons & Roberts, 2014)
- Within science museums the physical set-up of the exhibit plays a role in scaffolding data visualization engagement (Lyons & Roberts, 2014)
- Constructionist approaches to learning suggest that learning happens best when people construct personally meaningful projects (Papert, 1980; 1992)
- Make-A-Vis: Making meaning of personal, emergent, and real-time data
- Phase 1: Video-data analysis of purposes of engagement and physical engagement typologies at COSI with Make-A-Vis as part of the Walk exhibit

Context





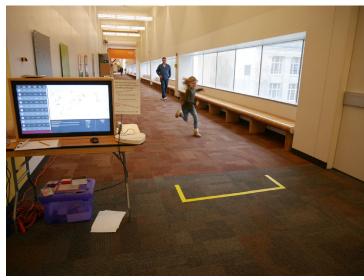




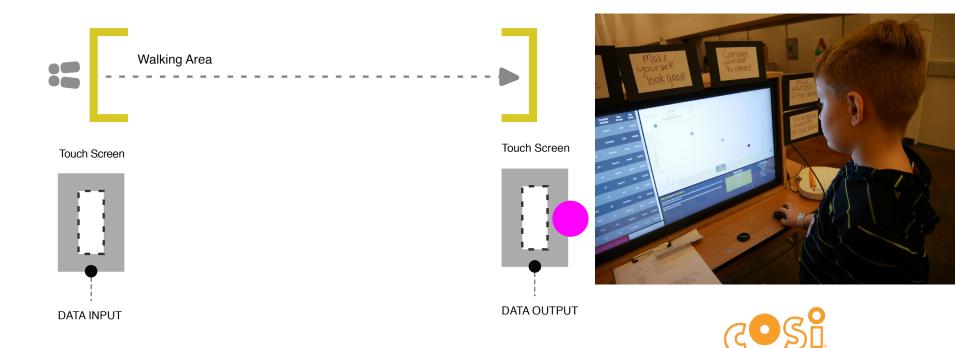










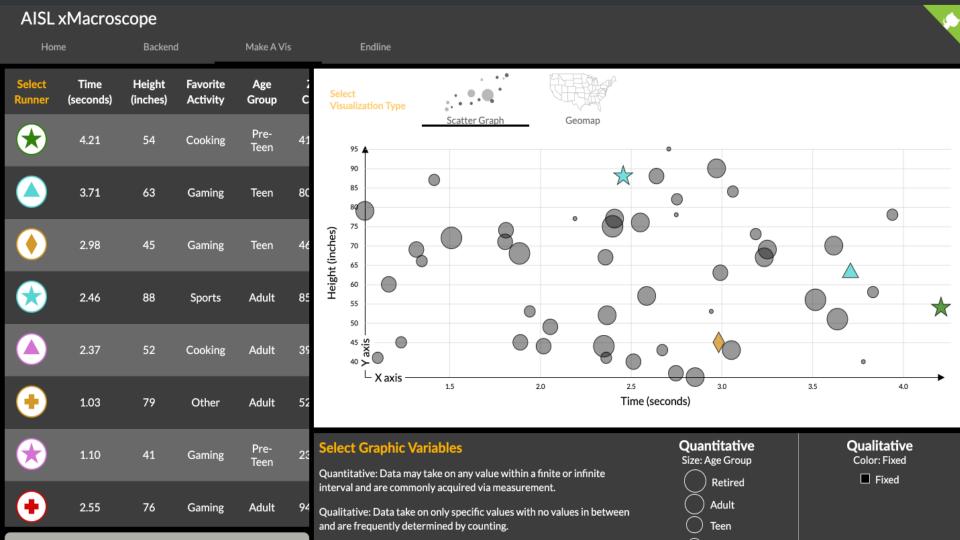


Setting | Task Types

- 1. Find yourself in the data
- 2. Make yourself look good
- 3. Compare yourself to others
- 4. Change your data
- 5. Find a group member in the data







Data Sources

Demographic survey

Paper and pencil self-reports of age, race, ethnicity

Length of engagement

- Minutes of engagement: Starts at entry; ends at exit
- Number of walks

Data queries

Task types visitors engaged with

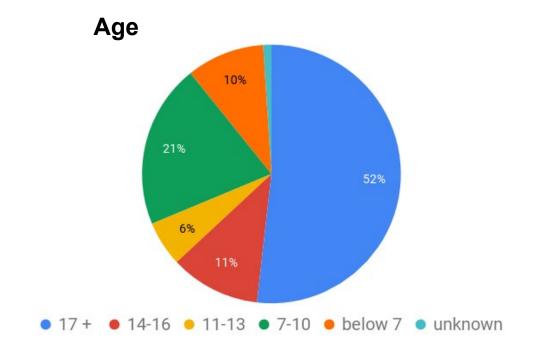
Semi-structured interviews

- What do you enjoy most about the exhibit?
- What do you find surprising about the exhibit?
- How do you recommend to improve the exhibit?



Phase 1 Participants

Number of participants (N)	195
Number of groups (N)	74
Average group size	3
Hours of observation (hours)	20



Phase 1 Participants

Ethnicity

Latinx 8.21%

Race

White 76.92%

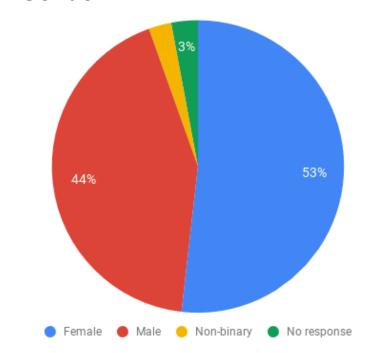
Black 11.79%

More-than- one 9.23%

Asian 3.59%

Other 1.54%

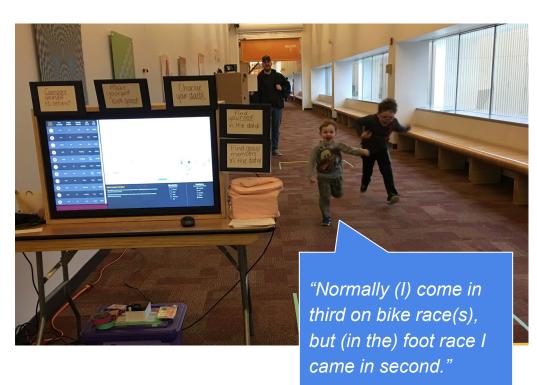
Gender



Emerging Findings

Findings: Length of Engagement

Mean dwell time	5:16 minutes
Minimum dwell time	00:40 minutes
Maximum dwell time	16:13 minutes
Average walks	3 times/group
Minimum walks	1 times/group
Maximum walks	12 times/group
No data entry walks	23 groups

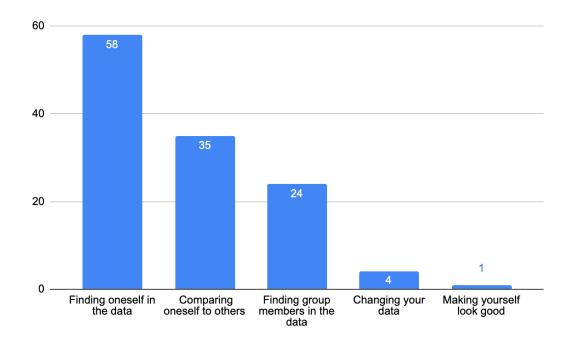


Findings: Engagement with Data Exploration Tasks

Most frequent: Finding oneself in the data

"(I can) see where I live"

Construction of personalized data visualization can support engagement.

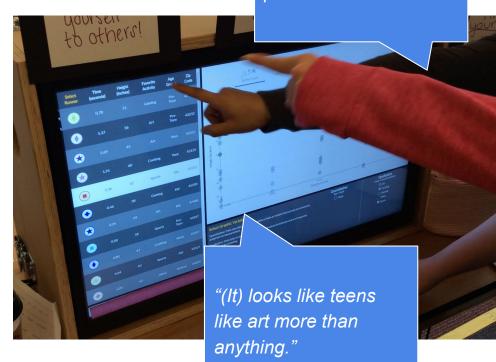


Findings: Comparing Aggregates

Within the emergent, real-time data set, visitors purposefully compared data cases and drew conclusions that may or may not be general truths.

It was the purposeful in vivo sense-making of real time data through comparisons that presented engagement with data visualization literacy.

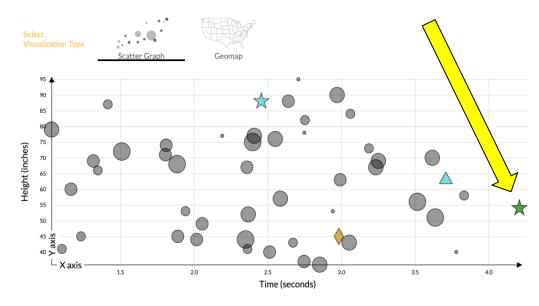
"Retired people and kids walk in (the) same pace."



Findings: Group engagement

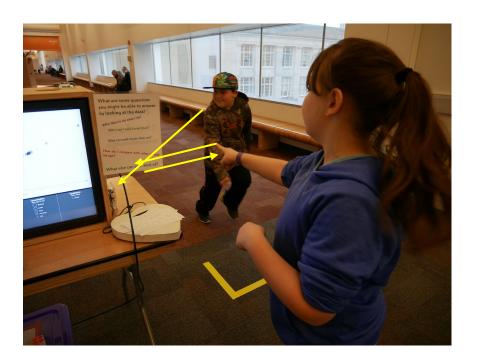
The spread of the exhibit in physical space made it possible for several members of a group to engage at once, through teachable moments, shared data entry, joint creation of visualizations.

Fastest or slowest?



Findings: Physical Exhibit Engagement

Technology transparency: The visible motion sensor lead to inquiries about input/output relationships as visitors observed the accuracy of its reading.



Moving Forward

Phase 1 Analyses under Way

Civic data visualization literacy

Purposes

What purposes of data visualization literacy do visitors engage with a realtime data set?

Physical design

What group
engagement
typologies with the
set-up are linked to
longer-term
engagement?

Phase 2: Learning Objectives

After using the Make-A-Vis,

- 1. participants can find their own data on the screen.
- 2. participants can visualize data variables with graphic variable types.
- 3. participants can compare and cluster data points.
- 4. participants can explore relationships among multiple data variables.
- 5. participant can develop hypotheses based on recognized relationships among multiple data variable types.
- 6. participants can confirm or deny formed hypothesis based on the data visualization.



Thank you!

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