Developing assessments of geological thinking and reasoning and studying their affordances for studying spatial cognition



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NSF DRL# 0733286, NSF-DRL# 1008649; NSF-DGE# 0742503;NSF-GeoEd# 1034643; NSF-TUES#1022755; US Dept of Education # R305A090170

Some concepts addressed this week

- Knowing your user
- · Change blindness
- Cognitive load
- · Peer learning
- Spatial skills
- · Measuring transfer



Questions for you to ponder....

- How do you know students are learning what you think they are learning?
- Who is falling through the cracks and how do you know?
- How do you know students are transferring skills acquired in one context to another?
- How do you guide learners to learn (really learn) what you want them to learn?



My program of research

 How people learn with rich visual representations to do reasoning, make inferences, solve problems, etc.

Key here is to:

- Find exemplary domains
- Design ecologically valid tasks
- Develop appropriate methods and items for collecting and analyzing data



Program of research, cont'd

How to best support learning with rich visual representations.

Key here is to:

- Develop assessments that can *differentiate* amongst students in terms of their learning
- Develop assessments that can capture student's learning processes and link these to conceptual learning outcomes-
 - called performance assessment.



Today I will...

1) Present some important concepts in characterizing learning with visual representations in

- highly 3-D domains, e.g. Geology, Architecture, and/or
- semantically-rich domains, e.g., Physics, Biology, Chemistry.

2) Describe some work to illustrate techniques used to study learning. Broad range of these with *differing methodologies* presented here

- bring to the fore some important Cognitive Science constructs and some key assessment ideas.



Framework for Cognitive Research in Visual Domains

- Regarding learning processes for visualizations, learning is viewed as an active and constructive process.
- View largely influenced by "Levels of processing" (Craik & Lockhart, 1972):

- the nature of a learner's processing of the target material largely determines the learner's memory representations, reasoning, etc. for that material. *

 Characterizing learning processes and their resulting conceptual understanding is central in the Learning Sciences.



Cont'd

- The levels of processing framework, originally developed for text, has been shown to be applicable with complex visual stimuli.
- Complex or semantically rich visualizations
 - i.e., those with domain-specific symbol systems
 - require a conceptual knowledge base
 - (different from simple, iconic representations like stop signs).



What's special about visualizations from an informationprocessing perspective?

- Comprehending or interpreting complex visualizations is difficult because
 all the information is presented to the learner simultaneously.
- This is in contrast to textual information sources in which the information follows the structure of the text (Larkin & Simon, 1987).

- Thus in the case of text, knowledge acquisition follows the structure of the text.

- Clearly not possible for visualizations; then what guides knowledge acquisition?



Information processing (cont'd)

Differences in representational format have direct ramifications on students' information-processing...

- when all information is presented simultaneously, prior knowledge is needed to guide knowledge acquisition.
- what if students lack prerequisite knowledge and skills? (addressed later).

In terms of information processing, visualizations & simulations can provide...

- Perceptual cues which supports rich inference-making (and these inference-making processes may differ from those used on textual information).



Gobert, J. (2005a) in In *Visualization in Science Education*, J. Gilbert (Ed.), Springer-Verlag Publishers, Dordrecht, The Netherlands.

Information-processing, cont'd

Thus, there is a continuum to the degree of *degree of visual isomorphism* to the objects/processes represented by various types of informational formats.

- Textual representations describe in words various aspects of science phenomena--non-isomorphic to the things they represent.
- Diagrams/illustrations represent spatial structure, but are static.
 inferences are required to reason about dynamic processes.
- Models and simulations represent objects' spatial structure as well how they function dynamically or over time.



Student Difficulties in Learning from Visualizations

Simply providing a diagram or model as an adjunct to text does not facilitate or promote deep understanding because:

- increased *cognitive load* on learners (Sweller, et al, 1990) particularly when the students do not recognize that the "things" being described are the same (e.g. a text and a graphic; Gobert, 1994).

- students *lack the necessary domain knowledge* in order to guide their search processes through diagrams in order to understand the relevant information you want them to acquire (Lowe, 1989, 1999; Head, 1984; Gobert, 1994; Gobert & Clement, 1999).



How does expertise fit in?

- B/c visualizations present all information simultaneously, *prior knowledge* is critical in guiding knowledge acquisition.
- Not only do experts have more knowledge but experts represent knowledge in ways that are important to tasks in the domain.



Expertise in visual domains

- Expert-novice studies have been conducted in:
- · electronics (Egan & Schwartz, 1979),
- engineering (Vicente, 1991, 1992; Bedard, 1993)
- architecture (Akin, 1979; Chase & Chi, 1981; Gobert, 1989; 1994; 1999),
- geographical map reading (Ormrod et al., 1986; Gilhooly et al., 1988; Thorndyke & Stasz, 1980), and
- topographical map reading (Chang et al., 1985).



I will post this powerpoint with full reference list

Expertise

Experts in visual domains have domain-specific schemata

 perceptual and cognitive structures evolved through experience

- especially adapted for performance in their respective domain (Brewer & Nakamura, 1984; Schank & Abelson, 1977; Rumelhart & Norman, 1975).

- Schemata influence the <u>amount</u> and <u>manner</u> in which information is processed and encoded in memory
 - account for the superior recall and inference-making by experts (Chang, Lenzen, & Antes, 1985; Gilhooly et al., 1988; Head, 1984).



 How to promote deep learning in students who (of course) lack significant amount of prior knowledge....?



Model-Based Teaching & Learning

- Model-based learning, a synthesis of cognitive psychology and science education...
 - Involves formation, testing, and reinforcement, revision, or rejection of mental models.
 - Mental modeling building has been used as a pedagogical framework in many domains including geology to promote students' deep understanding of complex causal processes, e.g., convection (Gobert & Clement, 1999; Gobert, 2000).
 - For Geoscience, deep understanding requires the *integration* of spatial, causal, temporal knowledge into a rich mental model.



Gobert, J. & Buckley, B. (2000). Int. J. Science Education, 22(9), 891-894.

Model based learning involves a dialectic between internal and external models





Affordances of a representation depends on student characteristics

Affordance

First used by J.J. Gibson (1977), a perceptual psychologist

- what is made possible for the learner, etc. INDEPENDENT of the learner's skills, knowledge, etc.
- D. Norman, (1988, The Design of Everyday Things)
- From HCI community extended this
- success of affordance is DEPENDENT on learners' skills, prior knowledge, etc.



To date, 2 types of studies

- Studying the *affordances* of learning with representations of different kinds, in different domains, etc.
- Studying how to support students' in learning with representations.



Overview of methodologies & technologies

METHODOLOGY / TECHNOLOGY	GOAL
Think aloud protocols	Characterize knowledge acquisition, reasoning processes, structure of knowledge
www	Support peer learning across the US/World & scale learning materials worldwide
Google Earth	Develop/implement rich visualizations beyond typical size and time scale
Logging (of students' interactions with simulations)	Characterize learning & adaptively tutor students in real time
Educational data mining	Analyze huge amounts of student log files



Domain of Architecture

- · Using think aloud and drawing protocols, I characterized e-n differences in
 - knowledge-acquisition processes while learning form plans,
 - inference making about 3-dimensions from plans (2-D) with edited plans.
- Task is analogous to reading topographical maps: hierarchy of line weights represents relative heights.





Experts were more systematic than novices in knowledge acquisition patterns.





Experts outperformed novices on both 2-D and 3-D comprehension.





Figure 2: 2-dimensional & 3-dimensional comprehension, experts vs. novices.



Regarding Spatial Visualization skills



- Spatial visualization skills ("ability to manipulate or transform spatial patterns") were statistically related:
 - 3D search (in think aloud protocols)
 - 2-D & 3-D comprehension



Gobert, J. (1999). Expertise in the comprehension of architectural plans. In Visual And Spatial Reasoning In Design '99, John S. Gero and B. Tversky (Eds.)

Representations in Geology

- Like architecture, in geology 3-dimensional information is inferred from in 2-dimensional visualizations by differing line types and hierarchies of pen weights.
- Processes that occur over time (glaciation, mountain formation, etc.) must also be inferred from more subtle aspects of representations (as well as from prior knowledge).



Research on learning with visualizations in Plate Tectonics

Excellent domain to do this because:

- large use of visualizations
- important role of mental model building in understanding the hidden causal mechanisms, e.g., convection.

Task:

-- asked students to draw models of the inside of the earth & of the causal mechanisms responsible for plate tectonics.

(NSF-REC# 9806141 for next three studies).



See also, Gobert, J. (2005b). In Journal of Geoscience Education.

Typical models of structure of earth of middle school students Type 0= 10.6%, Type 1=89.4%



Type 0: Spatially Incorrect Model



Type 1: Spatially Correct Model



Affordances (not!) of spatially incorrect model

Task: To depict processes inside the earth





Affordances of spatially correct model



E: "Why is the magma hot?"S: "... the mantle is right near the core."

30



From Gobert, J. (2000). International Journal of Science Education, 22(9), 937-977.

Curriculum design study for Plate Tectonics to support students' model-based learning (NSF-REC# 9980600).

- 2,000 middle and high school students from CA and MA collaborated on-line in 2 X 2 peer groups about the differences in PT using WISE (Web-based Inquiry Science Environment; Linn & Hsi, 2000).
- The curriculum engaged students in many inquiry-oriented, modelbased activities.



Gobert, J., & Pallant, A., (2004). Fostering students' epistemologies of models via authentic model-based tasks. *Journal of Science Education and Technology. Vol 13(1), 7-22.*

What's on your plate curriculum (mtv.concord.org)

- Draw, in WISE, their own models of plate tectonics phenomena (mtn formation, volcanoes, or earthquakes).
- Participate in an on-line "field trip" to explore differences between the East and West coast in terms of earthquakes, volcanoes, mountains
 - beginning with the most *salient* differences (earthquakes).
 - Using visual representations, learn about location of earth's plates (to develop relationship between plate boundaries and plate tectonic phenomena).
 - Using simulations, learn about causal mechanisms involved in plate tectonics, i.e., convection & subduction (scaffolded by reflection activities to integrate knowledge). **
- Learn to critically evaluate their peers' models which in turn serves to help them think critically about their own models.



Cont'd

In doing these activities, students develop...

- Content knowledge (spatial, causal, dynamic, temporal)
- Epistemological understanding of the nature of scientific models.
- Inquiry skills for model-building and critiquing their peers' models. *

Findings: Content and epistemological gains in all classes

Significant model revisions as well...



EastWest @ East OW PairID 13

Complete ID SP13G13805_5Ea

Nanes

Model Type Mountain Building

Model Picture



Critique:

W13: We have evaluated your model. The parts that helped us to understand the process you were modeling were the labels that you had("anticline and syncline"). We will also make the following suggestions that will help us better to understand your model. Label the colored part of you model. Put all the important labels on your model(enough for a person who hasn't learned this before to get the idea).

Critique Score

Comments on example 1....

- in model t1, "local" model, focused on crustal layer, no causal mechanisms
- in model t2, revised model is more detailed and shows a cut away section to reveal convection currents (causal mechanism).
- model is improved in both content and communication.





Comments on Example 2....

- in model t1, no causal mechanisms
- peer's critique asks for arrows to show direction of movement of plates, also they ask for a cross-section view.
- model t2, provides cross section showing plate movement & direction, magma layer is also added.



Technology to trace students' learning with models (mac.concord.org)

Principal & Co-Principal Investigators

Paul Horwitz, Concord Consortium, Pl Janice Gobert, formerly of Concord Consortium, Co-Pl Robert Tinker, Concord Consortium, Co-Pl Uri Wilensky, Northwestern University, Co-Pl

The Modeling Across the Curriculum project addresses:

- how technology can support student learning & assessment with visualizations and simulations
- how technology allows for scalability (over 400 schools in 31 countries).



mac.concord.org; IERI #0115699 www.concord.org



In response to IERI rpf, we set the following goals...

We developed content in the form of microworlds:

- Newtonian Mechanics (Dynamica)
- Genetics (BioLogica)
- Gas Laws (Connected Chemistry, developed by Wilensky et al)

We developed technology, i.e., Pedagogica, a powerful infrastructure:

- Delivers curricular materials
- Logs learners' interactions with models

-- provides performance assessment data for both researchers and teachers.

We measured:

• HOW students are learning with models across trials, tasks, and domains using log files.



Affordances of Students' Experimentation with Microworlds:

A Haphazard Strategy





Student's Systematic Strategy, e.g., vary one ball at a time (good strategy in the absence of prior knowledge).





Steps for analysis of log data*

An iterative process of

- task analysis of inquiry tasks (Ericsson & Simon, 1980)
- development of rubrics
- hand scoring logs
- writing algorithms for auto-scoring logs
- validating auto-scoring algorithms
- validation of summary/concise reports for export to 3rd party statistics programs



Buckley, B., Gobert, J. & Horwitz, P. (2006). Using Log files to Track Students' Model-based Inquiry. In the *Proceedings of the Seventh International Conference of the Learning Sciences (ICLS), Mawah: NJ: Erlbaum, pp.57-63.*

What is auto-logged on task 3

Auto-scoring of % trials in which students:

- Varied one ball (systematic, CVS used)
- Repeated trials (haphazard)
- Equal & extreme pairs (systematic)
- Closer to & further from the goal
- Goals flips.



Why do we care?

Because we can use logs to predict learning:

- i.e., Significant correlations found between gain score and
 - total time on task (r= 0.23)
 - reading time (r= 0.20)
 - # trials (r= 0.20)
 - % equal pairs (r= 0.23)



Performance Assessments in Dynamica

- Model-based learning, e.g., systematicity is auto-scorable in students' actions.
- Well-suited to assessing students' systematicity because of the well-defined domain, and numerical form of data.
- There is a relationship between inquiry strategies and post-test gain scores; e.g., # trials, and using the equal pairs strategy.



•

Current work: Science Assistments

Janice Gobert, Pl

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Project Overview

- Science Assistments is a technology-based environment for assessing and assisting middle school students on inquiry *.
- Our activities are based on guided inquiry & experimentation with microworlds.
- We are developing autotutoring of students' inquiry.



Hello! You are going to be a scientist today and conduct experiments in a virtual laboratory!



Students *learn* and are *assessed* while they do inquiry with microworlds

- With microworlds, students:
 - develop a hypothesis,
 - design & conduct an experiment
 - analyze data & warrant their claims, and
 - communicate findings (NSES, 1996).
- Because we log all students' actions, we can tutor students in real time.





Example of our State change microworld...





Student makes a hypothesis

Scientific Process: Explore Hypothesize Experiment Analyze data It's time to build a hypothesis. Use the boxes below, choosing parts of the sentence, to produce your hypothesis.

Hypothesis Builder:		
If I change the Choose One 🛊 so that it Choose 💲		
the Choose One Choose Choose		
Add Statement Statement number 1 is stored at the end of the table		
Hypotheses	Tested	Analyzed
If I change the amount of ice so that it increases the time the ice takes to melt increases		
Would you like to test this hypothesis now, or a more hypotheses now and test them all later? Let's go experiment Let me add more hypo	dd theses	Ţ
Note: the current hypothesis is the one that is highlighted.		
I need to explore more I'm ready to run my experiment		



Student tests hypothesis: Trial 1

Scientific Process:ExploreHypothesizeExperimentAnalyze dataRun different trials of experiment to test your hypothesis. The table will capture your data. Click on 'Show table'

My Current Hypothesis: 1. If I change the **amount of ice** so that it **increases** the **time the ice takes to melt_increases**

Show hypotheses list





Student tests hypothesis: Trial 2

Scientific Process: Explore Hypothesize Experiment Analyze data

Run different trials of experiment to test your hypothesis. The table will capture your data. Click on 'Show table' to see your data so far.

My Current Hypothesis: 1. If I change the **amount of ice** so that it **increases** the **time the ice takes to melt_increases**

Show hypotheses list





If student is not testing their hypothesis ...

Scientific Process: Explore Hypothesize Experiment Analyze data

Show hypotheses list

Run different trials of experiment to test your hypothesis. The table will capture your data. Click on 'Show table' to see your data so far.

My Current Hypothesis: 1. If I change the **amount of ice** so that it **increases** the **time the ice takes to melt increases**



3 levels of hints



Student interprets data from table

		Inde	pendent V	ariables	5	Depende	ent Variab	les	
Trial Number	Hypothesis Number	Has Cover	Container Size	Heat Level	Liquid Amount	Melting Temp(°C)	Boiling Temp(°C)	Time(sec) Melting	Time(sec) Boiling
1	1	true	Large	Low	100 grams	0	100	5	35
2	1	true	Large	Low	300 grams	0	100	16.25	102.5



Student warrants claim with their data

Scientific Process: Explore Hypothesize Experiment Analyze data Now it's time to look at the table of data and analyze.

My Current Hypothesis: 1. If I change the **amount of ice** so that it **increases** the **time the ice takes to melt_increases**

Show hypotheses list

	Independent Variables				Dependent Variables				
Trial Hypothesis Number Number	Has Conta Cover Size	ner Heat Level	Liquid Amount	Melting Temp(°C)	Boiling Temp(°C)	Time(sec) Melting	Time(sec) Boiling		
1 1	true Larg	e Low	100 grams	0	100	5	35		
2 1	true Larg	e Low	300 grams	o	100	16.25	102.5		
Data Interpretation When I changed the	n: Choose One	🔹	so that it	Choose	\$				
the Choose One		Choose	÷.1	am basın	g this on:	Data from	trial: choo	ose a trial 🌻	
compared to data f	rom trial:	noose a trial	🗘 . This sta	atement	does suppo	irt 🌲 🗖	y hypothe	sis.	
Add Statement									
		Interpreta	ations						
1: When I changed the amount of ice so that it increased the time the ice took to mel 1 increased. I am basing this on data from trial 1 compared to data from trial 2. This statement does support my hypothesis							+		
							~		2
								E C	\mathbb{C}
Go back. I need mo	re data. I	m done. Let	's do another	r hypothesis	. I'm a	all done			333
								A M	<u>n 6</u>



If student interprets data incorrectly ...

Scientific Process: Explore Hypothesize Experiment Analyze data Now it's time to look at the table of data and analyze.

My Current Hypothesis: 1. If I change the **amount of ice** so that it **increases** the **time the ice takes to melt_increases**

Show hypotheses list

		Independent Variables					Depender	nt Variables		
Trial Number	Hypothesis Number	Has Cover	Container Size	Heat Level	Liquid Amount	Melting Temp(°C)	Boiling Temp(°C)	Time(sec) Melting	Time(sec) Boiling	
1	1	true	Large	Low	100 grams	o	100	5	35	
2	1	true	Large	Low	300 grams	o	100	16.25	102.5	

Data Interpretation:

When I changed the amount	of ice 🔹 so that i	it decreased 🗘
the melting point	decreased \$. Lem basing this on Data from trial: 1
compared to data from trial:	2 ‡ . This	Look what happened to the
Add Statement		dependent variable when you change the amount of ice . You base some good trials to come up
	Interpretations	with a correct interpretation.
1: When I changed the amount increased. I am basing this of statement does support my h Go back. I need more data.	nt of ice so that it increase n data from trial 1 compar hypothesis I'm done. Let's do anot	her hypothesis. I'm all done

3 levels of hints



Communicate findings

Pretend you are explaining your conclusions about the effects of the <u>amount of</u> <u>substance</u> on each of the dependent variables to a friend who did not do the experiment. Discuss how you conducted this experiment and how you came to your conclusions. Be as specific as possible (example: mention the independent variable and its effects on each of the dependent variables).



...Students' skill level for content understanding and inquiry skills

are generated

For Inquiry skills: Hypothesizing

- IV, DV

Conduct trials

- Testing the IV/ their hypothesis
- Using CVS

Interpret Data

- Correct claims
- Warranted claims

Communicate Findings



To code communication skills (under development)

- To code open responses to react in real time:
 - "Nonsense" detector
 - "Too short response" detector
 - "Not correct terms"



Educational Data Mining to auto-analyze log files

- Once logs are labeled, use EDM to determine what fine-grained logged features correspond to specific inquiry skills.
 - Build detectors over feature sets, i.e., aggregates of logged actions.
- Validate detectors (see Sao Pedro et al, Montalvo et al, 2010 on our website).



Using Detectors to Predict Performance

Using our detectors as a basis for assessing skills, we can:

- (1)Predict skill proficiency before a student starts a new activity
- (2)Test the relationship between a skill honed in our learning environment and other transfer measures of inquiry
- See papers on our website (Sao Pedro et al, Montalvo et al) for techniques and results.



Sao Pedro, Baker, Gobert, Montalvo, & Nakama (in review)

Research like this has important implications for...

1) Learning Sciences

- provides a bird's eye view into the "black box" regarding students' learning processes with greater validity than previous measures

2) Intelligent Tutoring Systems

- is an essential component of ITS which can fade scaffolding as a student's skill level increases.

3) Science Education

- is critical for formative assessments for teachers
- richer assessments of science learning

4) Scalability

- our materials can reach any learner on the web.



Implications for some future research for you

- GE to support students' understanding of spatial, dynamic, and temporal processes.
 - Fine tune items and assessment to capture different facets of knowledge.
- GE to develop in students a sense of how we know what we know about geologic processes
 - Important to deep learning (& scientific literacy)
- In terms of assessment, you can relate students' knowledge acquisition processes to their learning outcomes by *logging* students' actions.



Question to ponder

- What are the prerequisite skills needed to leverage from what Google Earth has to offer for learners?
 - variance of these in students
 - can get an empirical handle on these to better measure learning.



Thank you!

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