

11 | Teaching data contexts: An instructional lens

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In textbooks, data often appears next to text that provides context and guidance about how to interpret that data. In the real world, however, disaggregated data often appears disconnected from any text that students might look to for explanation. Bits of data are snagged, aggregated and displayed, often with little context (Weinberger 2007). We encounter them as infographics, unattributed news, data factoids, Google “answers” and even T-shirts (Figure 1). Educators can just accept these “answers” or they can apply an instructional framework of information literacy reasoning (ACRL 2016) to data and treat them as opportunities for inquiry.

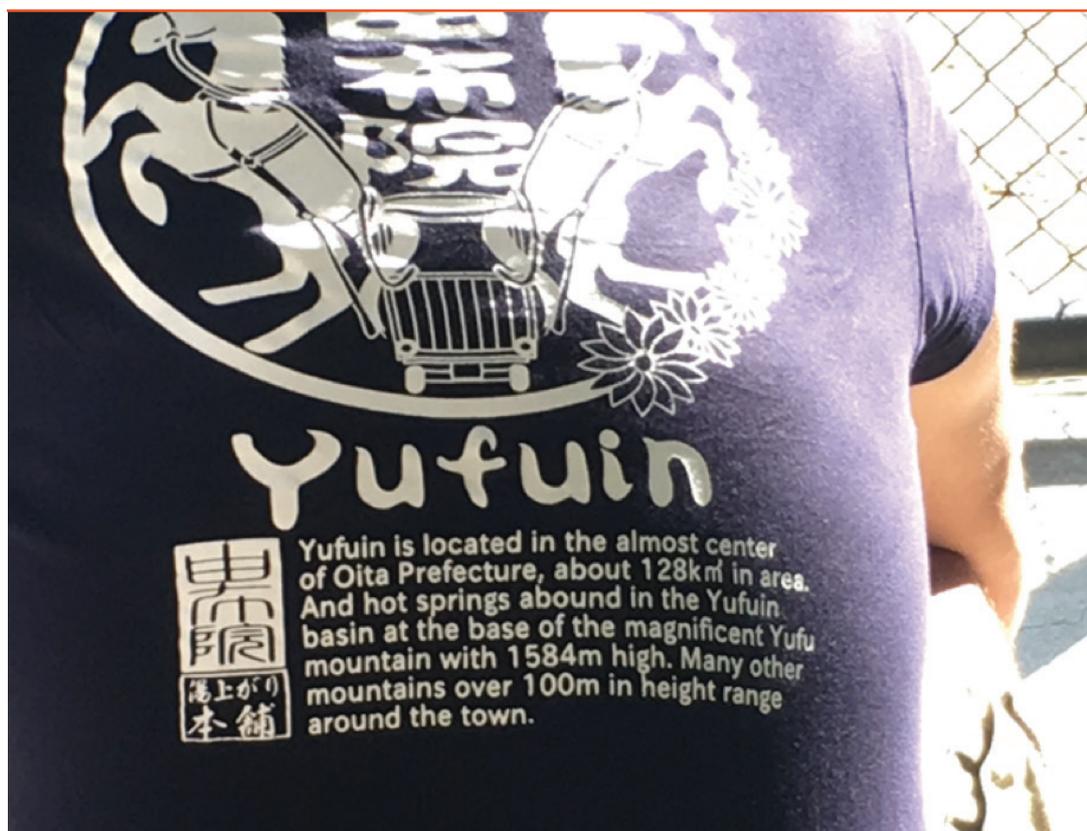


Figure 1: Data presented on a t-shirt at a baseball game. Copyright Debbie Abilock, used with permission.

A statistic doesn't speak for itself, by itself. As readers, we contextualize statistics to make meaning or evaluate an argument. As authors, we extract and manipulate statistics from data sources to use as evidence within an argument. As the Association of College and Research Librarians' *ACRL Framework for Information Literacy* (2016) points out, information is constructed and contextual. Sources reflect the "creators' expertise and credibility, and are evaluated based on the information need and the context in which the information will be used ... [V]arious communities may recognize different types of authority. It is contextual in that the information need may help to determine the level of authority required." The same is true of data. As you have seen elsewhere in this book, we must train ourselves to think of numerical data not as truth but as a reflection of the world in which the data was defined, collected, and discussed.

The following scenarios follow a school librarian who is focused on including data literacy in her instruction, specifically framing it within an understanding of data *context*. By no means are her responses the only possible ones; they depend on the data that's both available and relevant to the expertise of the students, the goals of the teacher, and our own inclinations and understandings.

Scenario 1: Area: Why context matters

A student needs the area of Alaska for a project he's doing about the impact of glaciers on global warming. Because he is in the early phases of research, his librarian recommends that he use the open Web to gain easy-to-read, easy-to-access basic information. He uses the search terms [size of Alaska] in an early Google search (note: the brackets are a convention representing the search box but are not characters entered into the search box). Google's algorithms can now predict the likely kind of information the searcher desires (likely because the word "size" is part of the query) and offer an immediate answer in lieu of a link to a potential answer (see Figure 2).

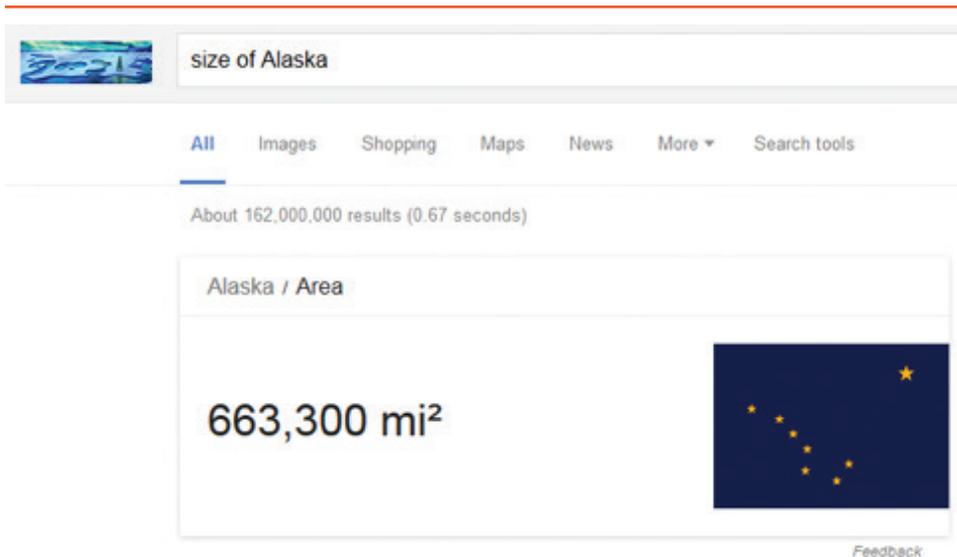


Figure 2: Google search result for [size of Alaska]

Alaska must be a pretty big state, he thinks. Another quick search confirms that it's the largest ... but why are three area numbers listed in the Wikipedia chart: total area (his number), land area, and water area (Figure 3)? He assumed that he'd get a single number answer — a fixed and immutable "area" statistic. He asks the librarian if he should use the total area, which includes water bodies like inland lakes, rivers and even the territorial waters around the land, or the land area which, in the case of Alaska, includes glaciers — that are melting.

This question is important to the student because glaciers are water ... but also solid, like land. The librarian knows that everyday words may imply quantities, counting and measuring but that students may not grasp their significance.

The librarian explains to the student that certain everyday words like area imply a measurement or quantity but, because it is used in various situations, its definition may be ambiguous or imprecise. In this case geologists, geophysicists and other scientists are precise about the area measure they are using (e.g., land or water), so that they can compare the appropriate numbers with

others who are measuring glacial recession. Since the whole class will encounter the same ambiguity in area measurements, the librarian decides that this nuance merits greater attention and begins to curate specific information that will help students understand which number to use and why.

List of U.S. states and territories by area

From Wikipedia, the free encyclopedia

This is a complete **list of the states of the United States and its major territories** ordered by *total area*, *land area*, and *water area*. The water area numbers include inland waters, coastal waters, the Great Lakes, and territorial waters. Glaciers and intermittent bodies of water are counted as land area.^[1]

Contents [hide]

- Area by state/territory
- Area by division
- Area by region
- See also
- Notes
- References
- External links



Area by state/territory [edit]

State/territory	Total area ^[2]			Land area ^[2]			Water ^[2]				
	Rank	sq mi	km²	Rank	sq mi	km²	% land	Rank	sq mi	km²	% water
 Alaska	1	665,384.04	1,723,337	1	570,640.95	1,477,953	85.76%	1	94,743.10	245,384	14.24%
 Texas	2	268,596.46	695,662	2	261,231.71	676,587	97.26%	8	7,364.75	19,075	2.74%
 California	3	163,694.74	423,967	3	155,779.22	403,466	95.16%	6	7,915.52	20,501	4.84%
 Montana	4	147,039.71	380,831	4	145,545.80	376,962	98.98%	26	1,493.91	3,869	1.02%
 Alaska Territory	5	434,600.20	1,114,047	5	434,600.45	1,114,044	99.78%	40	200.45	523	0.26%

Figure 3: "List of U.S. states and territories by area" from Wikipedia

An instructor curates to help students develop context or background specific to their needs and the learning goal. Curation is not just linking to a bunch of sources about climate change or glaciers. Google does that. For a librarian to curate for a project means having a conversation with the teacher about the learning goal and the students' needs, then evaluating and selecting from the "glut" of sources just those which provide the context needed for the students' investigation. Curation is a targeted instructional strategy librarians can use to build just-in-time background (Abilock n.d.).



Examples of quantifiers found in everyday contexts

- all
- area
- distance
- every
- fewer
- heavy
- light
- many
- more
- much
- none
- progress
- some
- trend
- volume

Knowing that other schools might also study the effects of climate change on Alaskan glaciers, the librarian taps into social curation tools and searches curated curricular resources built by librarians for use in their academic, special library and school communities (Valenza et al. 2014). The results are abundant, to say the least, so she runs her thinking by the teacher. She proposes limiting this initial curation to specific pages from the U.S. Geological Survey, the major source of U.S. earth science data. Her final list (below) is highly selective and annotated to clarify why each site is relevant and important:

- » **Definition from the U.S. Geological Survey's Glaciology Project (http://www.usgs.gov/climate_landuse/clu_rd/glacierstudies/massBalance.asp)** This site explains how the area and thickness of glaciers are calculated. The USGS' Glaciology Project measures how glaciers respond to climate change in order to both predict and prepare for the impact of glacial changes. Read their description of the significance of this glacial mass research to climate change study (http://www2.usgs.gov/climate_landuse/clu_rd/glacierstudies/benchmarkGlaciers.asp).

- » **Visual explanation from Researchers at the Centre for Quaternary Research at Royal Holloway, University of London (<http://www.antarcticglaciers.org/modern-glaciers/introduction-glacier-mass-balance/>)** This site charts explaining glacier mass balance and graphs showing trends of mass balance over time written by a glaciologist as part of her commitment to education.
- » **Arctic Sea Ice Thickness Maps Data from the Center for Polar Observation and Modeling (CPOM) (<http://www.cpom.ucl.ac.uk/csopr/seaice.html>)** CPOM compares satellite radar signals that bounce off ice vs. water and, together with ice concentration and types data, can produce accurate thickness measurements in near-real time. For additional insight on how satellites can be used to measure glaciers, see <http://www.indiaenvironmentportal.org.in/files/file/Arctic%20sea%20ice%20warm%20winter.pdf>.
- » **Glacier Mass Balance Data from the National Snow and Ice Data Center (<https://nsidc.org/data/g10002>)** Historical data from 1945-2003 is available for download. The National Snow and Ice Data Center (NSIDC) manages data and supports research about the cryosphere.

She plans to do a mini-lesson to introduce these sources, then help students find data in CPOM and NSIDC related to their selected glaciers. The teacher, who is much more interested in getting into the action-research project, will use the U.S. Climate Resilience Toolkit framework (<https://toolkit.climate.gov>), a source that the librarian found, to help the class define the problems, develop solutions based on what they've learned, and decide on actions they might take.

In this first scenario, the librarian recognizes that the student believes he is searching for an unambiguous term that stood for a single number. Rather than treat this as one student's confusion, the librarian reframes the misconception in a way that can help the entire class recognize that scientists collect specifically-defined measurements over time in order to see trends, make predictions and take action. Her just-in-time curated resources focus on the area measurements that students need for their class' science project. Her annotations explain why these sources are relevant and authoritative, modeling the critical thinking she wants students to use when evaluating other sources for the project. Collaborative planning between the teacher and librarian take into account their respective strengths and goals.



Strategies

- **Listen for verbal quantifiers.** These are contextual clues to opportunities for data literacy instruction.
- **Identify relevant curators.** Reach out for help from social media curators and collaborative curation venues with demonstrated expertise in this particular topic in order to develop your own background and recognize the most useful resources for this specific research.
- **Understand student learning goals.** Understand the purpose of the assignment so that your sources are highly relevant rather than “just-in-case” general pointers.
- **Annotate curated sources.** Write descriptions that evaluate the data context and assess their authority for this topic, as these become models for the students' own evaluative thinking.

Try this

For state or country reports, ask students to find the area in square miles (or square kilometers) using Google search results, Wolfram Alpha, a government source or other appropriate sources. What reasons might account for differences? (For more guidance, please see <https://www2.census.gov/geo/pdfs/reference/GARM/Ch15GARM.pdf>.)

Here are examples of results for Alaska:

Area in square miles	Source
570,640.95	United States Census (http://www.census.gov/quickfacts/table/LND110210/00,02)
663,300	Google (https://www.google.com/search?q=area+of+alaska&ie=utf-8&oe=utf-8)
661,957	Alaska Government publication (p.8) (http://labor.alaska.gov/research/pop/estimates/pub/popover.pdf)
665,400	Wolfram Alpha (https://www.wolframalpha.com/input/?i=area+of+alaska)

Here are examples of results for Great Britain:

Area in square miles	Source
84,440	Wolfram Alpha (https://www.wolframalpha.com/input/?i=area+of+Great+Britain)
88,745	Google (goo.gl/PFYucQ)
80,823 (given as 209331.1 sq. km)	United Nations (http://islands.unep.ch/ICJ.htm#943)
94,525	Nations Encyclopedia (http://www.nationsencyclopedia.com/economies/Europe/United-Kingdom.html)

Scenario 2: Population: How data is constructed and contextualized

American history classes are studying shifts in people's views about immigration. The teachers plan to contrast the colonial immigrants' attitudes toward the indigenous Native American population with U.S. citizens' responses to recent immigrant groups.

The librarian knows that the teachers want students to infer historical attitudes from primary sources like letters, journals, diaries, and drawings that they've used before. She realizes that she has an opportunity to add data as another type of primary source: a census and a survey. She plans to ask students to draw inferences about societal attitudes by comparing them over time.



Examples of primary source data by discipline

- **Humanities** – text mining from unstructured data in archives and manuscripts
- **Social sciences** – tabulated data collected from surveys, polls, census
- **Natural sciences** – tabulated data collected from clinical trials, controlled studies

She reasons that students should learn about the U.S. Census data because it's the main source of U.S. population data. Besides providing the population counts that are used to determine the number of seats each state has in the House of Representatives, the results are used "for many important but overlooked political, economic, and social decisions that end up affecting our daily lives" (U.S. Census Bureau, "Why it's important" n.d.).

Although *statistics* from public opinion polls are reported ubiquitously today, she is less certain about how to find the *primary sources* of public opinion survey *data* from which particular statistics are extracted. Many LibGuides from colleges and universities point to the SDA Archive, which uses the General Social Survey (GSS) and the American National Election Study (ANES), two major sources of public opinion research in the United States. However, after trying to search them herself, she decides that they are too complex for high school students. Instead she discovers the beta GSS Explorer (<https://gssdataexplorer.norc.org/>), which students can easily use to search, extract, analyze, and even visualize statistics from their longitudinal public opinion surveys run since 1972.

At the social studies departmental meeting she proposes that she teach students to use a public opinion survey and the census to trace public attitudes. She'll use Alaska as a model, reserving the thirteen original colonies and their states for the students' own research. She quickly shows teachers how to search the U.S. Census to find population figures, starting with American Indians and Alaskan Natives in Alaska's 2010 census figures (U.S. Census Bureau, "Quick Facts: Alaska," n.d.). Then, using five slides, she explains how the Census's evolving definitions of "Indian" mirrors the societal shifts in attitudes toward race and immigrants:

- » **Slide #1** Definition #1: The 1787 Constitution established the census to allocate the number of representatives from each state and determine how taxes would be divided among the states. Initially Indians weren't included for either purpose — *they were not considered federal or state citizens because they didn't pay taxes* — so they weren't counted.
- » **Slide #2** Definition #2: The American Indian wars and large-scale removals exacerbated distinctions between tribal membership vs. citizenship for the U.S. Marshals, the 1860 census counters, because they were told to count Indians — but only those *who had "renounced tribal rule."* (See Figure 4.)

A photograph of a document with torn edges, containing instructions for counting Indians for the 1860 census. The text is printed in a serif font and reads: "5. Indians.—Indians *not taxed* are not to be enumerated. The families of Indians who have renounced tribal rule, and who under State or Territorial laws exercise the rights of citizens, are to be enumerated. In all such cases write "Ind." opposite their names, in column 6, under heading "Color.""

Figure 4: Instructions on counting Indians for the 1860 census (U.S. Census Bureau 1860).

- » **Slide #3** The census takers were instructed to *exclude Indians on reservations, those that were roaming unsettled areas and any Indians living in Alaska* (Collins 2006). Although people thought that Indians had clear physical characteristics, the census had no instruction on how to identify them, especially those that were of mixed race living within the general population. No surprise that the marshals found it difficult to count everyone.
- » **Slide #4** Definition #3: The 1890 census instructed that “all Indians” were to be counted – except *neither Aleuts nor Eskimos were included as Indians* until Alaska became a state (1959).
- » **Slide #5** Definition #4: When the government began mailing the census to homes (1960), people were asked to *self-identify as Indian*. However, *people of Hispanic origin self-identified as Indian* in large numbers in the 2010 census, muddying the definition still further (U.S. Census, “Instructions” n.d.; Decker 2011).

She demonstrates how to use Pew’s language timeline to pinpoint when racial terms were added or changed in census questions (Pew Research Center for Social and Demographic Trends 2015b). Then she suggests that, when historical periods and immigrant groups are discussed in class, teachers refer to that timeline and prompt students to consider why the government would want to count that group, in that way, at that time.

When classes are ready to look at contemporary attitudes, she will devote another library period to having students read a historical analysis of attitudes toward immigrants that researchers derived from opinion-polling archives (DeSilver 2015). During a third library period, she’ll help students create accounts at the GSS Explorer in order to select and visualize statistical data, evaluate survey methods, and discuss their reflection of attitudes. In a fourth library period, she will ask students to contrast a census

with a survey as a formative assessment. In particular, they'll use a Pew Research survey, looking for potential bias or distortions in the types of questions, then compare it with the proposed data collection plan for the next decennial census (Pew Research Center for U.S. Politics and Policy 2016; Cohn 2015).

After both the science and social studies projects are finished, she'll want to give students a short online quiz in which they compare the kinds of primary sources used in science vs. social science. For now, she'll use the C3 social studies standards during library class (see Figure 5) to build their specific awareness about the forms of primary source data that various social scientists use to answer questions in their particular field.

Wrap-up

In this second scenario, the librarian recognizes that the history teachers have overlooked data as a type of primary source. She uses the U.S. Census, a comprehensive longitudinal public data aggregation of U.S. demographic and economic information, because of its importance in government policy decisions as well as in the electoral process. Since students often encounter surveys and polls in popular culture, her goal is to have students distinguish between a survey (which samples a representative subset of the population to make estimates about the entire population) and a census (which aims to gather data from every person in a country). In addition to her explicit focus on how changing definitions and question wording impact data collection and interpretation, she embeds this sequence of lessons within her longer-term scope and sequence plans to integrate vertically across grades and to make curricular connections across disciplines.

WAYS OF KNOWING	CIVICS/ GOVERNMENT	ECONOMICS	GEOGRAPHY	HISTORY
	POLITICAL SCIENTISTS SAY...	ECONOMISTS SAY...	GEOGRAPHERS SAY...	HISTORIANS SAY...
DIMENSION 2				
DATA SOURCES NEEDED TO ADDRESS QUESTIONS	Government policies, policy pronouncements, political poll results, statistics, leadership efforts, political behavior; observations of local conditions, interviews; news reports	Statistics and lots of them in as real time as possible (labor, capital, credit, monetary flow, supply, demand)	Spatial and environmental data; statistics, map representations, GIS data to measure observable changes to the planet; indicators of territorial impact	Accounts from the recent recession and from hard economic times in the past, both firsthand and synthetic, as many as can be found (oral history, diaries, journals, newspapers, photos, economic data, artifacts, etc.)
KEY CONCEPTS AND CONCEPTUAL UNDERSTANDINGS NECESSARY TO ADDRESS QUESTIONS (non-exclusive examples)	Theories of political behavior, rationality, self-interest, political parties, power flow, government, fiscal policy; relationships between the state and markets; constitutional limits on government, debates about those limits; evidence (to make claims)	Application of different types of economic theories to gauge inflation/deflation, labor shrinkage, capital contraction, asset/liability analyses from banking sector, changes in supply and demand; evidence (to make claims)	Theories of human land/resource use; spatial representation, scale, degree of distortion, map symbols, specialized GIS symbolic systems and representations; evidence (to make claims)	Theories of human behavior, thought, perspective, agency, context, historical significance; historical imagination; moral judgment; evidence (to make claims)
KEY STRATEGIES AND SKILLS NEEDED TO ADDRESS QUESTIONS (non-exclusive examples)	Reading statistics from polls, conducting polls and interview research; reading sub-text into policies/pronouncements; reading power flow and blockage, converting such data into evidence to make arguments and claims that answer sub-questions	Capability to read statistics critically, for assessing agendas behind statistical representations; conducting survey research; capability to convert statistics into meaningful arguments and claims that answer the sub-questions	Cartography including using map symbol systems, critical reading and thinking, capability of using statistics to represent spatial change, capability to use statistical and spatial (often digitized) representations to make arguments and claims that address sub-questions	Critical reading and thinking, analysis and synthesis, reading subtext and agency in older sources; statistics; converting verbal, written, photographic, oral, artifactual accounts into evidence to make arguments and claims that answer the sub-questions

Figure 5: Dimension 2 of the C3 Framework showing the kinds of data needed to address disciplinary research questions and the specific reading strategies needed (National Council for the Social Studies 2013). Citation: National Council for the Social Studies (NCSS), The College, Career, and Civic Life (C3) Framework for Social Studies State Standards: Guidance for Enhancing the Rigor of K-12 Civics, Economics, Geography, and History (Silver Spring, MD: NCSS, 2013).

Strategies

- **Teach source literacy within a discipline.** (Murphy 2016). By contextualizing historical data as a type of primary source, data literacy gains recognition beyond math classes.
- **Compare types of data collection methods** (e.g., census vs. poll) to understand the context of a statistic.
- **Plan one-shot library lessons as beads on a necklace.** Use big-picture knowledge of the school curriculum to contextualize one-shot data literacy lessons in long-term learning sequences built on information literacy goals.
- **Teach key data sources as models.** Choose a substantial source of data and articulate why you've chosen it. Your goal is not merely to teach "how to use" a particular data set or a digital tool. Rather students learn that, in every field, certain key data sources can help them develop a deeper understanding of their research topic.

Try this

While some countries collect data about race, ethnicity or religion, it is illegal in others to ask questions about these topics in a census (INSEE 2015). Look for opportunities in language classes or global studies where students could look at the social and cultural contexts of census data in other countries.

Scenario 3: Pandemics: The emotional context

Students' fears about catching Ebola or the Zika virus have spiraled as reports about deaths and birth defects flood the news. The combination of unusual symptoms, impact on infants, and limited prevention information, along with emotionally-charged graphic descriptions of transmission and high death rates, are a sure recipe that an "availability bias" that will color how students respond to data reported in the news. Availability bias is a

psychological term referring to how the mind can give greater weight to the newest or most familiar information. When data is presented as odd (unique or unusual) and memories are recent and filled with anxiety, one is likely to overestimate the likelihood of something bad happening.

A teacher and the librarian decide to do a health unit to help students manage these gut-level responses. They want students to slow down their thinking, so they decide to add moments of “friction” to the lesson (Seroff, Bergson-Michelson, and Abilock 2015). By doing so, they hope students will learn to step back from knee-jerk emotions to perform a more dispassionate evaluation of news about disease outbreaks and quantify their risks analytically. There is particular urgency in learning this because dissemination of information about epidemics in social media has preempted the traditional role of health officials who normally issue warnings on authoritative websites with full explanations of symptoms, risks and prevention plans (see, for example, the World Health Organization (WHO) and its Disease Outbreak News alerts posted at <http://www.who.int/csr/don/en/>). Indeed, a recent study found that tweets, not official WHO Disease Outbreak News (DONs), broke the Ebola story to over 60 million people over a three-day period (Odlum and Yoon 2015).

The teacher and the librarian acknowledge that they, probably like their students, are unclear about the differences between commonly-used terms like epidemics and pandemics, so they look for background information from the Centers for Disease Control and Prevention (CDC), the major source of health data about Americans:

The amount of a particular disease that is usually present in a community is referred to as the baseline or *endemic* level of the disease. This level is not necessarily the desired level, which may in fact be zero, but rather is the observed level. In the absence of intervention and assuming that the level is

not high enough to deplete the pool of susceptible persons, the disease may continue to occur at this level indefinitely. Thus, the baseline level is often regarded as the expected level of the disease (CDC 2012).

Occasionally, the amount of disease in a community rises above the expected level. *Epidemic* refers to an increase, often sudden, in the number of cases of a disease above what is normally expected in that population in that area. *Outbreak* carries the same definition of epidemic, but is often used for a more limited geographic area. *Cluster* refers to an aggregation of cases grouped in place and time that are suspected to be greater than the number expected, even though the expected number may not be known. *Pandemic* refers to an epidemic that has spread over several countries or continents, usually affecting many people (CDC 2012a).

The teachers decide to focus on infectious diseases like Zika and Ebola, rather than including non-contagious epidemics like diabetes and obesity. Their first thought is to prompt students with headlines about epidemics that include or imply statistics, but locating a sufficient number proves to be too laborious. Instead they locate a fear-based infographic that received quite a bit of traction when it was released ([https:// www.good.is/infographics/infographic-the-deadliest-disease-outbreaks-in-history](https://www.good.is/infographics/infographic-the-deadliest-disease-outbreaks-in-history)).

Students pick a disease from the infographic — one that either worries or interests them — and then team up by their chosen disease. Groups will deconstruct the data given in the infographic and compare it to information about the disease from other sources. Knowing that students' free-text searches will return all sorts of random data bits, the teacher and librarian decide to limit students' searches to two sources of reasonably comparable data about pandemics:

- » **The Centers for Disease Control and Prevention** contains a rich range of data with a focus on Americans, ranging from simple statistics in FastStats (<http://www.cdc.gov/nchs/fastats/>) to raw data sets on diseases and conditions (<http://www.cdc.gov/DiseasesConditions>).
- » **The World Health Organization**, the major international source of disease data, publishes Global Health Observatory by country (<http://www.who.int/gho/en>) and by topic (<http://www.who.int/topics/en>).¹

To be sure that all groups collect similar information, the instructors create a spreadsheet-like matrix to contain the data found by each group (<http://noodle.to/pandemic>). To maximize their efficiency while researching and recording data, the teachers add the following links to the matrix:

- » **A link to the infographic**
- » **A source for historical census data**
- » **Links to the CDC and WHO information about infectious diseases**
- » **Definitions of key terms students will encounter: endemic, epidemic and pandemic**

Since they want students to also evaluate the visual display of the data in the infographic, they include a column for a statistic called the Case Fatality Rate (CFR), which quantifies the deaths among cases.

In data literacy, it is essential that students understand how to select comparable numbers. For example, the Zika outbreak in

¹Both the National Center for Health Statistics (NCHS) and the World Health Organization (WHO) are developing easy-to-use online tools that allow users to examine vital statistics data interactively and create their own tables within the tool, as well as export data for use in other formats. Since these major sources for public health data are likely to be valuable for future research projects (e.g., maternal and child health, nutrition, dental care, substance abuse, noninfectious diseases), school librarians might want to explore them now and follow their evolution.

Brazil that began in April 2015 ought to be matched with data from the same period (2015-2016), some of which is likely to be an estimate. They show students a matrix of Brazil's population data collected from various sources (goo.gl/Op4Se1) and remind them to avoid a "precision bias," a cognitive predisposition to assume that the most precise number they find is *de facto* the most accurate.

During a class discussion, they consider other evaluation criteria for Brazilian population statistics. Should they use <http://Worldometers.info>, which seems to update in real time but is clearly an estimate? Should they use some average of the population data from the same year(s)? Perhaps they should use the Brazilian Census figures because they're "official" government numbers?

Ultimately, they decide to use data from The World Bank because it standardizes Brazil's aggregated data from global, national and regional sources so that it's comparable with other countries' data. The online interface allows them to visualize the data on maps or in graphs, view it in tables or download it as data sets. Students also find the Bank's country office contacts in Brazil and Washington D.C. and the names of a specific team focused on combatting Zika in Latin American that they might email with questions. The availability of expert help, the comparable numbers from identifiable sources, and the prospect of experimenting with varied visualizations outweigh the students' initial preference for data sites that offer easier access and simpler results.

As students start to compare U.S.-focused information with global data, they realize that there are differences in the case fatality ratio (CFR), which is the risk of death expressed as a percent based on the proportion of people who die from a disease outbreak out of those who are infected. CFR will differ based on regional or local conditions and health care systems. For example, while measles is currently under control (endemic — and therefore not much of

a worry) in countries like the U.S., it is the leading cause of death among children in India, Nigeria and Pakistan (CDC 2015).

Once the teams complete the matrix, each group presents their findings to the class, comparing them with what was displayed in the infographic. As they present, the teacher makes connections to other measures of morbidity, such as the number of Brazilians who die from other causes like heart disease and cancer. This helps students contextualize the scope of the outbreak (CDC 2012a).

Next the teachers want the class to practice using data selectively from the matrix as evidence in an argument. To help students appreciate how the same data can be used to support very different conclusions, they show a *New York Times* interactive graphic that models using the same job numbers for different arguments (<http://www.nytimes.com/interactive/2012/10/05/business/economy/one-report-diverging-perspectives.html>). The class discusses how drawing different conclusions from the same data can be done with complete honesty; it's not always a signal of intentional manipulation.

The librarian reinforces this idea by explaining one experiment reported in *Nature* in which 29 scientific teams looked at the same information about soccer games and answered the same question: "Are dark-skinned players more likely to be given red cards than light-skinned ones?" The scientists came to widely disparate conclusions; some saw no difference between light- and dark-skinned players while others saw a strong trend toward giving more red cards to dark-skinned players. The study concludes that each team's inferences were contextualized by their expertise and background:

Teams approached the data with a wide array of analytical techniques, and obtained highly varied results. Next, we organized rounds of peer feedback, technique refinement and

joint discussion to see whether the initial variety could be channeled into a joint conclusion. We found that the overall group consensus was much more tentative than would be expected from a single-team analysis (Silberzahn and Uhlmann 2015).

As students become more comfortable with data literacy skills, they approach ambiguities in exercises like these with greater confidence and are open to understanding that data is neither infallible nor arbitrary — human interpretation plays a key role.

Teachers assign the “Infographic Design Matrix” to help students identify their audience (Abilock and Williams 2014). As a final task, each student will select some statistics from the matrix to support a unique visual argument — to display either on a PowerPoint slide or, if they have time, as an infographic — using some of the statistics they have collected. Teachers caution students to resist “anchoring,” another cognitive bias in which students accept the first piece of evidence as “truth” and measure all other information against that first data bit. The teacher and librarian hope that this selection process reinforces students’ understanding that data is being used as evidence rather than as immutable facts.

Wrap-up

In this scenario, the librarian and teacher decide to look at a visual display of decontextualized data — what we commonly call “data in the wild.” Since infographics are so prevalent online, they want students to both deconstruct and, at least partially, to construct their own visual display of data using a typical genre. They explain to students that, to make comparisons easier, they are controlling both the sources of data and the format in which statistics are collected. They offer students a limited choice so that students will be able to compare and discuss comparable results. Initially students work in groups but, to ensure accountability,

they eventually choose data that is relevant to a specific audience in order to craft a compelling argument. Throughout the project, the teacher and librarian continually refer to ways in which data is constructed and contextualized — by cognitive biases, content creators, a purpose and an audience.

Strategies

- **Pick emotionally charged statistics to teach cognitive biases.** Use data about highly charged current topics to raise students' metacognitive awareness of their emotional responses and intuitive but faulty judgments. Then add “friction” to slow down students' thinking and develop the analytical strategies they need to learn in order to manage their emotional reactions.
- **Teach students to read data genres.** Build awareness of the constraints and affordances of infographics and other graphic representations of data by deconstructing and constructing them as genres.
- **Scaffold lessons with manageable data sets to put student attention on thinking with data.** Curation can be informally linked from a collection document, providing controlled exposure to data “in the wild” through manageable sets of statistics. Explain that choices are limited to facilitate comparisons.
- **Teach nuances of data analysis.** There are honest differences among experts about the meaning of data. By encouraging students to bring different deductions and interpretations to the same set of statistics, you resist stereotyping data analysis as a process of always identifying disinformation or discovering propaganda.

Try this

Explore data context. Context can mean both the genre in which numbers are found (e.g., chart, spreadsheet, infographic, or text) as well as the *content described* by the numbers. Expose students

to spreadsheets of data that compare political, social or economic aspects of countries. When seemingly comparable numbers appear in the same spreadsheet, students make superficial comparisons simply because the numbers seem related. In UNICEF's single spreadsheet (<http://data.unicef.org/topic/child-protection/child-labour/>, in "Access the Data" section), Afghanistan and Chile have similar data on percentages of child labor by country. However, the circumstances within each country, as well as the absolute numbers of children, are vastly different. In Afghanistan, most children work in agriculture, in their homes, in forced hazardous brick production or in illicit activities. In Chile, most are in retail businesses or commercial sexual exploitation. In Chile, a prosperous country, there is a significant government push to eliminate the worst forms of child labor, an initiative that Afghanistan, a poor nation, would find much more difficult to implement.

Strategies for teaching context in the wild: Find a problem, build a rule

Both the explicit guidelines that can help novices learn to vet the credibility of new content and the tacit "rules of thumb" that they subconsciously use to evaluate familiar content are part of the "context" we bring to data literacy. Typically, each of us makes unconscious decisions in many, daily situations where we have to make a choice (Gigerenzer 2007). We use mental shortcuts, called heuristics, to speed decision-making. These unarticulated "rules" — which may begin with formally stated recommendations but then transition into tacit, intuitive behaviors — allow us to function efficiently without stopping to think through each choice we make, each action we take, and each detail of a problem we encounter.

These shortcuts are generally good enough. We happily perk along unconsciously using these rules — until they don't work. When we recognize that we are stuck, we will bring the rule

into consciousness (metacognition) and consider revising it. Good instructional design aims to bring these rules to light by putting challenges in front of students so that they reexamine their assumptions, learn from their errors and revise simplistic algorithms.

The general problem with relying on unconscious rules of thumb is that they reinforce cognitive biases. And, while a set of explicit data literacy “rules” may provide guidance for beginners, such lists are not productive in the long run. Students will change rule-based behavior only when cognitive dissonance provokes a shift in their thinking. As Kuhn (2000) asserts, “Strategy training may appear successful, but if nothing has been done to influence the metalevel, the new behavior will quickly disappear once the instructional context is withdrawn and individuals resume meta-level management of their own behavior.”

Another possible teaching strategy is to use discussion and reflection to uncover the useful tacit knowledge within rules of thumb (Polanyi and Sen 2009; André et al. 2002). Initially we can ask students to become aware of their unconscious rules by completing the following sentence:

“When I see...then I do...”

so that they identify and then describe a specific and conditional decision strategy that they employ in a particular situation. In the process of explaining a rule, students may verbalize strategies that their peers have not considered. Or they may be able to convert a vague rule of thumb into a just-in-time checklist, which is what Gawande argues is necessary for critical decision-making in highly charged situations like an operating room or the cockpit of a falling plane (Gawande 2010). One opening instructional move, then, is to have students develop their own checklists targeted to places where it’s essential to make critical decisions about data evaluation or data visualization. For example, we may

teach a novice to look where zero falls on the y-axis of a graph. Over time, however, a student may revise that to a more nuanced checklist:

- » **Don't assume that the y-axis begins at zero.**
- » **Look for labels on the y-axis.**
- » **Look at the increments on the y-axis to help you know if a change is significant or not.**

Of course, experts also have rules of thumb that we can learn from; these are valuable procedures and processes that emerge from their years of experience within a discipline or field. One set of processes specifically related to data literacy is described by a professor of sociology and criminal justice as “statistical benchmarks.” As mentioned throughout this text, these are validated statistics (such as the size of the U.S. population) that can help us judge whether new population statistics we encounter are significant (Best 2013). While we may not have the disciplinary expertise to provide students with benchmark strategies for every topic, we can model a process that involves noticing a problem with odd data, unearthing our tacit assumptions, faulty procedures and unconscious misconceptions and then developing more accurate strategies to evaluate data in the wild. Let's explore what this might look like in the following four short scenarios.

Example 1: Evaluating data in the context of a visualized benchmark

Scenario: *The Internet's Own Boy*

The Internet's Own Boy is a documentary film about programming prodigy and open-access activist Aaron Swartz. In the film, public domain advocate Carl Malamud agrees to work with Swartz to download and provide free access to what are, in fact, public records. Indeed, the Public Access to Court Electronic Re-

cords (PACER) database makes inordinate profits on what should be freely accessible court records (E-Government Act of 2002; *Internet's Own Boy* 2014 31:33 min). As a result of their activism, PACER agrees to provide free access to 17 libraries across the country. Malamud exclaims: "One library for every 22,000 square miles!" Does this make any sense, even in a quick mental check that requires only a few seconds?

Finding a relevant statistical benchmark

When I saw this film, I referenced a quick benchmark from my toolkit: the fact that the 48 contiguous states make a very rough rectangle about 3,000 miles from east to west and perhaps about 1,000 miles from north to south: an area of about 3 million square miles. 3 million square miles divided by 17 locations? It's instantly clear that there are far too few locations for the majority of Americans to reach easily, even without figuring in Alaska and Hawaii. Another way I could have approached this would have been to use the benchmark that we have 50 states. 50 states divided by 17 locations means around one location for every three states: again, not very accessible for most Americans. A third way to approach this would be to have the actual total area of the United States in mind as a statistical benchmark. According to the U.S. Census, the total land and water mass of the United States (including Hawaii and Alaska) is 3,805,927 square miles (U.S. Census 2012). Again, it's very quick to see that 3.8 million divided by 17 is very poor coverage.

Reasoning using the statistical benchmark

So the average area covered by each of those 17 libraries would be about 3.8 million square miles divided by 17 libraries, which is a little over 220,000 square miles. It's beginning to look like that

22,000 figure is off by a full order of magnitude and might be attributable to a decimal point error.

Example 2: Evaluating data in the context of a common misconception

Scenario: *The Martian*

The film *The Martian* (2014) tells the story of a manned mission to Mars that goes awry because the crew leaves one man for dead. NASA realizes he's still alive and pulls out all the stops to bring him home. Under tremendous pressure to launch a rescue ship and worried about the astronaut's mental as well as physical health, Vince Kapoor (Chiwetel Ejiofor) says, "He's 50 million miles away from home, he thinks he's totally alone, he thinks we gave up on him — I mean, what does that do to a man, psychologically? What the hell is he thinking right now?" (2015, 34:10 min). Is that 50 million mile distance from Earth to Mars credible within the context of the film's story?

Uncovering a data misconception with students

From their first picture book about the solar system to endless examples in media, students have been exposed to distorted images of the planets, their orbits and distances. Help students develop a better idea of the actual distances using a video showing a scale model of the solar system on a dry lakebed in Nevada (<https://vimeo.com/139407849>).

Reasoning through a data misconception with students

Many students have already learned that Earth's average distance from the sun is about 93 million miles, and a quick Web check

confirms that Mars' average distance from the sun is about 142 million miles. So it's obvious that the screenwriters probably just subtracted one figure from the other to arrive at 50 million miles. In fact, the two planets orbit the sun at different speeds and come that close to each other only on rare occasions when their positions in their orbits lined up on the same side of the sun. They do not remain the same distance apart throughout their orbits. Indeed, the Mars Mission Director's exclamation would have been more dramatic (and more credible), if he had relegated the stranded astronaut to a position twice as far from home!

Example 3: Answering a data question by making an analogy to a known data context

Scenario: Historical Data on the Black Death

Recently I asked an epidemiologist how we could teach students to figure the case fatality ratio — remember that's the number who die out of the number who get infected — for the Black Death. She acknowledged that historical population numbers, infection rates, and death rates are *very rough estimates*.

Finding a known context and using it as an estimated proxy

She shared a strategy that public health workers use to assess outbreaks of diseases that have a historical trail. *Yersinia pestis*, the bacterium responsible for the Black Death, the Plague of Justinian, and the Third Plague, continues to cause plagues in Africa and Asia today (CDC 2015). Therefore, to estimate a historical disease impact like the Black Death's case fatality ratio she uses the modern case fatality ratio assuming no treatment, since there were no antibiotics during the 14th century.

Reasoning through the data analogy with students

By applying current-day CFR estimates for plagues to Black Death, we can guess that the CFR ranged from about 50% (for the bubonic form) to almost 100% (for the pneumonic form). Of course, CFRs are a moving target. In both historical and modern times, the bubonic plague affects the old and infirm in the first wave, but death rates drop significantly as immunity builds and the weak are wiped out — and so the CFR declines.

Example 4: Questioning a data claim by building background context

Scenario: Unemployment figures

“We have 93 million people out of work. They look for jobs, they give up, and all of a sudden, statistically, they’re considered employed” (Jacobson 2015). This seems like an enormous and very serious problem — and it’s repeated often, more recently upped to 94 million. People assume that “out of work” means “unemployed,” that is, 93 million people *want to work and are looking for a job but can’t find one*.

Finding a context

Two statistical benchmarks for students to remember are the current population of the U.S. (about 325 million) and China (about 1.3 billion). If we go to the Bureau of Labor Statistics (BLS), a good source of government information about jobs, we find that the U.S. unemployment rate is currently about 5% (BLS 2016). If 93 million people represents 5% of the U.S. population, the total U.S. population would have to be almost 2 billion people. That doesn’t make sense — even the population of China is only somewhere over one billion!

Reasoning through the statistical claim with students

The BLS issues a monthly press release on the number of *unemployed* people — currently about 8 million (BLS 2016). In fact, 93 million people are not “out of work” (i.e., unemployed) but rather they are “out of the workforce.” Most of this number consists of people of working age who aren’t looking for jobs — students, disabled people, housewives/househusbands, early retirees — anyone who could theoretically work. Ask students to think about whether the recurring choice of the phrase “out of work” or “out of the labor force” in these claims involves ignorance or intent to deceive.

By gathering examples like these from popular culture, politics, and the media, we can support students as they recognize and wrestle with real-world data challenges.

Conclusion

Throughout this chapter we have modeled teaching strategies to scaffold students’ growing understanding and ability to evaluate data in the wild. By contextual framing, we can address students’ grab-and-go approach to data and create moments of friction (Abilock 2016) at which point they are intrigued enough to reassess their assumptions about numbers as indisputable and fixed. Ambiguity drives inquiry. Investigations of data context result in data insights. As educators, we can choose when our students are ready to tackle this ambiguity and, by doing so, achieve higher levels of data comprehension.

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If you have 5 minutes:

- Point out the quantifiers in an assignment or research question.
- Use one of the scenarios in the final section of this chapter with your students.
- Ask students to bring in printouts or collect screenshots of data to add to a physical or virtual bulletin board.

If you have 15 minutes:

- Choose a student's example (see above) or find a current statistic for discussion in class. Ask questions about the context or brainstorm a statistical benchmark that might make it possible to evaluate the number provided.
- Show students two visualizations of the same statistic and ask which one makes more sense to use for the topic they're researching or argument they're building and why.

If you have 30 minutes:

- Demonstrate how to assess a data source they might use to research their topic. Then ask students to critically annotate a second source from a list you've curated. The resource at <http://www.oercommons.org/courseware/module/11007/overview> provides additional insight and guidance.

If you have a class period or more:

- Ask students to compare two sources of data related to an ongoing topic.
- Integrate a data source into an existing research project.