

Original Paper

Making Air Pollution Visible: A Tool for Promoting Environmental Health Literacy

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Abstract

Background: Digital maps are instrumental in conveying information about environmental hazards geographically. For laypersons, computer-based maps can serve as tools to promote environmental health literacy about invisible traffic-related air pollution and ultrafine particles. Concentrations of these pollutants are higher near major roadways and increasingly linked to adverse health effects. Interactive computer maps provide visualizations that can allow users to build mental models of the spatial distribution of ultrafine particles in a community and learn about the risk of exposure in a geographic context.

Objective: The objective of this work was to develop a new software tool appropriate for educating members of the Boston Chinatown community (Boston, MA, USA) about the nature and potential health risks of traffic-related air pollution. The tool, the Interactive Map of Chinatown Traffic Pollution (“Air Pollution Map” hereafter), is a prototype that can be adapted for the purpose of educating community members across a range of socioeconomic contexts.

Methods: We built the educational visualization tool on the open source Weave software platform. We designed the tool as the centerpiece of a multimodal and intergenerational educational intervention about the health risk of traffic-related air pollution. We used a previously published fine resolution (20 m) hourly land-use regression model of ultrafine particles as the algorithm for predicting pollution levels and applied it to one neighborhood, Boston Chinatown. In designing the map, we consulted community experts to help customize the user interface to communication styles prevalent in the target community.

Results: The product is a map that displays ultrafine particulate concentrations averaged across census blocks using a color gradation from white to dark red. The interactive features allow users to explore and learn how changing meteorological conditions and traffic volume influence ultrafine particle concentrations. Users can also select from multiple map layers, such as a street

map or satellite view. The map legends and labels are available in both Chinese and English, and are thus accessible to immigrants and residents with proficiency in either language. The map can be either Web or desktop based.

Conclusions: The Air Pollution Map incorporates relevant language and landmarks to make complex scientific information about ultrafine particles accessible to members of the Boston Chinatown community. In future work, we will test the map in an educational intervention that features intergenerational colearning and the use of supplementary multimedia presentations.

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KEYWORDS

computer visualization; digital cartography; environmental health literacy; health communication; environmental health; computer-based education; air pollution; ultrafine particles; immigrant education

Introduction

There is a pressing need for new tools to promote environmental health literacy [1]. As research reveals new knowledge about the public health effects of environmental pollutants, it remains challenging to communicate complex scientific information to laypersons. Environmental health literacy will be significantly improved when people not only learn factual knowledge but also understand what tactics can reduce their exposure to risk [1]. Recent advances in the technology of computer mapping hold the potential to increase environmental health literacy by helping users visualize information about environmental hazards in spatial terms and within a geographic area. By forming mental representations of this information, users may be able to make more informed decisions and have more well-defined discussions when considering health-promoting measures [2].

This work aimed to develop a computer visualization tool to promote environmental health literacy about traffic-related pollution in Boston Chinatown, Massachusetts, USA. One component of traffic pollution consists of particulate matter. Particulate matter in ambient air is a leading cause of mortality and morbidity worldwide. It is estimated that ambient particulate matter, which consists of particles 2.5 μm or less in diameter (PM_{2.5}), causes 100,000 to 200,000 deaths in the United States and 3.2 million deaths globally each year [3-5]. Ultrafine particles, a subfraction of PM_{2.5} that are less than 0.1 μm in diameter, tend to be higher locally near heavy traffic. Thus, a community's proximity to highways and major roadways presents a health risk to its inhabitants [6,7]. In Chinatown, where poverty levels are higher than those in most other parts of the city, many residents have low levels of educational attainment and limited English proficiency [8]. For this reason, it is important to improve residents' access to health information.

People vary in their capacity to access, learn, and use information [9]. Efforts to disseminate health information often result in differential learning, with the poor and socially disadvantaged benefiting the least [10,11]. Messages about environmental health need to be tailored to audiences with varying socioeconomic and ethnic backgrounds [1]. Extensive research points to the importance of making computer tools accessible to individuals with low levels of English proficiency and limited computer skills [12]. Development of accessible tools for improving environmental health literacy would help educators cross these linguistic and technological barriers.

Pictures can aid the accessibility of health communication messages [13]. Studies have shown that visual information is better retained than verbal information by users in the cognitive process of building mental models [14-17]. Because of this, visualization technology is particularly well-suited to education about ultrafine particles, as the small size of these particles makes them invisible to the naked eye, and people are often unaware of their presence. Consequently, visualization of such ultrafine particle concentrations on a computer screen provides a means of increasing community awareness.

Visualization tools can also aid community health education. For example, computer maps can offer multiple views of a geographic area and have the potential to improve cognitive performance. The main bottleneck in cognitive processing and visual thinking is often the limited capacity of working memory, which can retain only about 3 to 5 objects [18]. Having access to more than one visualization can allow users to store information in working memory while thinking about, analyzing, and processing that information, and can improve learning and decision making [19]. In a map of pollution, for example, one map layer might show selected features of the built environment, such as roadways and high-rise buildings. Another layer can depict the location and extent of open green space or vegetation. In addition, interactivity of digital maps can further facilitate learning [20]. Users can manipulate the display through visual animations that convey different types of information. This would allow them to explore the effect of temperature and wind severity, and thereby increase the potential for interest and engagement.

Visualization additionally helps overcome the limitations of health literacy promotion reliant on textual representation of information, which often simplifies concepts for readability [21]. Simplification can erase nuances important to comprehension of complex scientific information in general and, in our case, about ultrafine particles. A map showing the spatial distribution of the particles in a community can improve awareness of many factors in the physical environment that affect community residents' daily lives, such as roadways, green space, and residential buildings. As environmental researchers discover how changes in concentrations of ultrafine particles in the air are affected by factors such as temperature and wind, it becomes possible to depict these changing concentrations by means of animated visualizations.

The objective of this work was to develop a visualization tool that would educate members of the Boston Chinatown

community about one form of traffic-related pollution, ultrafine particles. This paper focuses on development of the Interactive Map of Chinatown Traffic Pollution (or Air Pollution Map). Its use in educational interventions will be discussed in a future publication.

Methods

We designed the Air Pollution Map as the first step of a community educational intervention to meet the challenges of disseminating complex scientific knowledge about environmental hazards to laypersons. Users can explore spatial patterns of ultrafine particle concentrations, measured in particle number concentration (PNC) units, and see how concentrations vary with changing conditions, including temperature, wind severity, wind direction, and traffic volume. The map we created provides 4 visual controls as sliders, which allow a user to modify conditions of temperature, wind direction, wind speed, and traffic volume (model explained below). We created bilingual English-Chinese Web-based and desktop-based versions of the visualization.

We created the Air Pollution Map using Weave v1.9.38, a Web-based analysis and visualization environment [22]. Weave was developed at the University of Massachusetts Lowell, Lowell, MA, USA and is now supported as an open source tool. Weave provides the ability to integrate, analyze, and visualize data using a suite of over 20 different interactive tools. Completed visualizations can be disseminated via a website.

Weave [23] is freely available to the public and is simple to install on personal computers. Its accessibility facilitates its use by educators in community settings such as schools, community-based organizations, libraries, and homes. The interactivity of the tool that we developed in Weave is designed to be used as a launching point for group-based conversation about the problems posed by air pollution and tactics that communities can pursue to reduce exposure.

The data used to construct the Air Pollution Map were collected and analyzed by an investigatory team of the Community

Assessment of Freeway Exposure and Health Study (CAFEH). CAFEH measured ultrafine particle concentrations under a variety of weather and traffic conditions in Boston Chinatown [24] and built a statistical model that describes how the ambient conditions were related to the ultrafine particle concentrations [25]. We applied this model as an algorithm and incorporated it into the interactive map in ways that we hoped would allow users to interactively explore hypothetical scenarios that affect ultrafine particle concentrations across the community.

The variables included in our algorithm were wind speed, temperature, traffic volume, and wind direction. These were the primary predictive variables in the ultrafine particles model (Table 1). The model predicts lower ultrafine particle concentrations when winds are stronger, which blow the particles away more effectively. Higher concentrations are expected when there are more cars traveling on Interstate 93, a major north-south highway near Boston Chinatown, resulting in increased emissions. Concentrations are also higher with lower temperatures due to (1) more formation of particles from less-efficient engines, (2) less dispersion of particles in the air, and (3) lower particle removal due to slower evaporation and particles sticking together. The model predicts that when the wind comes from the east, concentrations are highest near Interstate 93, the major highway, and a major train station on the far side of the highway (see Figure 1). Under these conditions, concentrations decrease toward the west side of Chinatown. We developed and ran the model in the R-Project v3.1.1 programming language (R Core Team, R Foundation for Statistical Computing), and we averaged PNC to census blocks for use in our map.

We consulted community experts to assist in developing the user interface for the map. Two community health promoters who had prior experience in educational projects concerning traffic pollution in Chinatown recommended Chinese wording for menu labels and selected community landmarks recognizable to residents.

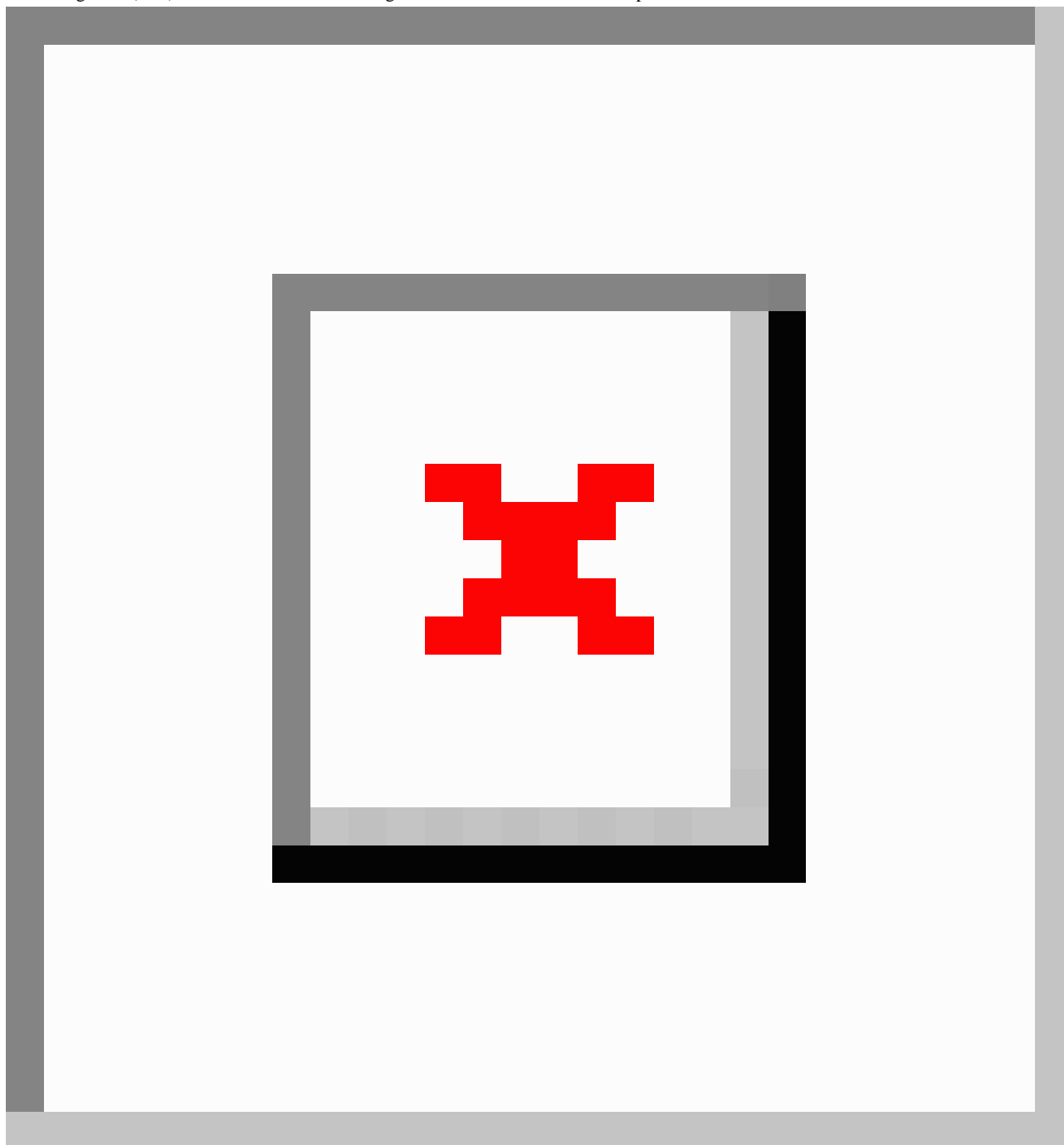
The University of Massachusetts Institutional Review Board approved this study (#2014161).

Table 1. Variables of interest and their effects on particle number concentration (PNC).

Variable	Effect of increase in variable	Selectable values in Weave
Wind speed	Percentage decrease in PNC	Calm, breezy, windy
Temperature	Percentage decrease in PNC	0 to 100°F or -20 to 40°C
Traffic volume	Percentage increase in PNC	Light, medium, heavy
Wind direction	More complicated: highest for east winds, and with larger effect near Interstate 93	N, NE, E, SE, S, SW, W, NW ^a

^aWind direction variables: north (N), northeast (NE), east (E), southeast (SE), south (S), southwest (SW), west (W), northwest (NW).

Figure 1. Several views of the Interactive Map of Chinatown Traffic Pollution: (a) Full aerial Google maps view showing sliders, map layers, scenario slider, and color legend; (b) scenario of the lowest ultrafine particle concentrations; (c) scenario of the highest ultrafine particle concentrations; (d) Community Assessment of Freeway Exposure and Health Study mobile laboratory route (solid line) taken to measure ultrafine particles and locations of surrounding trees (dots). Visualizations created using Weave software v1.9.38. PNC: particle number concentration.



Results

The Air Pollution Map (Figure 1) allows users to see how meteorological and traffic conditions can affect changes in ultrafine particle concentration by moving the 4 sliders for temperature, wind direction, wind severity, and traffic volume (Figure 1 a). PNC at each census block is shown through choropleth mapping, in which brighter color intensity indicates higher pollution levels (key in Figure 1 a). We chose the wording used for the labels on the sliders to conform to popular language conventions among members of the Chinatown

community. For example, “calm,” “breezy,” and “windy” are commonly understood terms describing wind severity. Icons for commonly recognized landmarks are used to orient the user within the neighborhood.

There are several layers to the map that can be made visible or hidden. These include a base map showing the street level or a satellite view of the study area with prominent landmarks (Figure 1 a), the route taken by the CAFEH mobile laboratory when measuring ultrafine particles (Figure 1 d), the location of trees (Figure 1 d), and the ultrafine particle concentration in particles/cm³ (Figure 1 b and 1c). These features allow users to

learn experientially by changing values for the variables, such as setting the weather conditions to current values. In doing so, they can discover which combination of conditions leads to the highest (Figure 1 c) and lowest (Figure 1 b) concentrations of ultrafine particles. They can also explore effects of each of the input parameters while holding others constant or in interaction with one another. We included a separate “scenario” slider to quickly navigate to the conditions producing the best and worst levels of pollution (Figure 1 a). Additional exploratory map features include the ability to drill down to Chinatown street levels or zoom out to see adjacent neighborhoods.

Advanced toolbar features of Weave are purposefully hidden, so that buttons not vital to the educational process are not accessible to the novice user. The educational version of the map also hides the scenario slider, so that students may independently arrive at the contributors to the best and worst pollution levels by exploration. Weave’s session state-based architecture allows all significant interactions made with the system to be recorded (eg, selecting, subsetting, and toggling). By analyzing or even simply replaying the actions each student took while exploring the map, educators may find patterns that are informative about the learning process. For example, teachers can analyze steps students take to identify what combinations of temperature and wind conditions are associated with high levels of pollution.

The bilingual feature and the dual temperature scale (Celsius/Fahrenheit) were necessary to accommodate the needs of the different segments of the community, including longtime residents and newly arrived immigrants. The visualization tool was used in educational workshops targeted to high school youth, English-language adult learners, and seniors in Boston Chinatown. We developed a plan for evaluation of the educational intervention. In future work, we will analyze data collected on the effectiveness of the intervention and produce a report of findings.

A stable version of the Interactive Map of Chinatown Traffic Pollution is available [26]. The bilingual desktop version is available free of charge from the first or last author upon request.

Discussion

Our Air Pollution Map aimed to make complex information about the spatial distribution of ultrafine particulate pollution in a community accessible and usable for residents of Boston Chinatown and to aid them in engaging in health-promoting behaviors. Health educators could, for example, use the visualizations to explain why altering certain choices, such as time and place of outdoor exercise, can reduce exposure [20]. Interactive maps are commonly used by professionals for environmental analysis and planning. However, user interfaces are rarely linguistically tailored to persons with limited English proficiency. One study has shown that limited English proficiency is associated with low health literacy across diverse racial groups. Among a population-based sample of limited English-proficiency adults, a higher percentage of Chinese respondents self-reported lower health literacy (68.3%)

compared with their white counterparts (18.8%) [27]. Increasing accessibility of information is an important part of reducing the digital divide in health literacy [28].

We sought to develop the Air Pollution Map with several features not offered in existing mapping tools. Unlike other air pollution maps, which focus on other pollutants, our tool makes ultrafine particles visible to the user. Moreover, concentrations of ultrafine particles are displayed at the census block level, an appropriate granularity of geographic unit for education about how concentrations of ultrafine particles vary across a neighborhood. We created the user interface for ease of use by individuals with limited English and technical skills. Specifically, the design approach emphasized accessibility by members of the Chinatown community, most of whom are immigrants. Map labels enable users of the tool who live and work in this neighborhood to recognize familiar streets or areas, including highways and local landmarks.

These features of the Air Pollution Map add new capabilities to an evolving technology of mapping air pollutants. Two other tools, Plume Air Report [29] and AirNow [30], display real-time information on pollution and provide recommendations for outdoor activities. The Community LINE Source Model (C-LINE) maps contributions of mobile sources to local air pollution under user-defined traffic emissions and meteorology conditions [31]. Each of these existing tools gives users access to important information about types of air pollution in their community, but not exposure to ultrafine particulate pollution.

In designing an educational intervention using the Air Pollution Map, it is important to consider not only the educational potential of the map but also its limitations. The present version requires users to enter values for temperature, wind severity and direction, and traffic volume by manipulating sliders. In future work, we will add a new capability for the user to automatically import the actual wind and temperature pattern for the current day and time. In addition, the user interface currently uses only English and Chinese. Translations to other languages such as Spanish would be helpful in order to widen accessibility to residents of other neighborhoods. Alterations of the user interface have not yet been implemented for persons who are visually impaired.

The next phase of work entails development and evaluation of an educational intervention using the Air Pollution Map. We have developed multimedia educational presentations, which have been shown to complement visual learning [32,33]. We also devised small-group learning activities and walking tours in the community to aid learning about sources of traffic-related pollution and features of the built environment. These multimodal learning activities are intended to take place in a supportive social environment, which helps motivate learning [34]. In particular, the educational approach emphasizes intergenerational colearning; specifically, teenagers who are familiar with computer technology are first trained to use the interactive map; in turn, the youth teach those who are less familiar with computer technology, such as elders.

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Conflicts of Interest

None declared.

References

1. Finn S, O'Fallon L. The emergence of environmental health literacy—from its roots to its future potential. *Environ Health Perspect* 2015 Jun 30 [FREE Full text] [doi: [10.1289/ehp.1409337](https://doi.org/10.1289/ehp.1409337)] [Medline: [26126293](https://pubmed.ncbi.nlm.nih.gov/26126293/)]
2. World Bank. Thinking with mental models. In: *World Development Report 2015: Mind, Society, and Behavior*. Washington, DC: World Bank; 2013:62-75.
3. Lim SS, Vos T, Flaxman AD, Danaei G, Shibuya K, Adair-Rohani H, et al. A comparative risk assessment of burden of disease and injury attributable to 67 risk factors and risk factor clusters in 21 regions, 1990–2010: a systematic analysis for the Global Burden of Disease Study 2010. *Lancet* 2012 Dec 15;380(9859):2224–2260 [FREE Full text] [doi: [10.1016/S0140-6736\(12\)61766-8](https://doi.org/10.1016/S0140-6736(12)61766-8)] [Medline: [23245609](https://pubmed.ncbi.nlm.nih.gov/23245609/)]
4. Fann N, Lamson AD, Anenberg SC, Wesson K, Risley D, Hubbell BJ. Estimating the national public health burden associated with exposure to ambient PM_{2.5} and ozone. *Risk Anal* 2012 Jan;32(1):81–95. [doi: [10.1111/j.1539-6924.2011.01630.x](https://doi.org/10.1111/j.1539-6924.2011.01630.x)] [Medline: [21627672](https://pubmed.ncbi.nlm.nih.gov/21627672/)]
5. Pope CA, Dockery DW. Health effects of fine particulate air pollution: lines that connect. *J Air Waste Manag Assoc* 2006 Jun;56(6):709–742. [Medline: [16805397](https://pubmed.ncbi.nlm.nih.gov/16805397/)]
6. Ghosh R, Lurmann F, Perez L, Penfold B, Brandt S, Wilson J, et al. Near-roadway air pollution and coronary heart disease: burden of disease and potential impact of a greenhouse gas reduction strategy in southern California. *Environ Health Perspect* 2016 Feb;124(2):193–200 [FREE Full text] [doi: [10.1289/ehp.1408865](https://doi.org/10.1289/ehp.1408865)] [Medline: [26149207](https://pubmed.ncbi.nlm.nih.gov/26149207/)]
7. HEI Review Panel on Ultrafine Particles. *HEI Perspectives 3: understanding the health effects of ambient ultrafine particles*. Boston, MA: Health Effects Institute; 2013 Jan. URL: <https://www.healtheffects.org/system/files/Perspectives3.pdf> [accessed 2017-03-06] [WebCite Cache ID 6oldkjpNhh]
8. Boston Public Health Commission. *Health of Boston 2012–2013: a neighborhood focus*. Boston, MA: Boston Public Health Commission Research Office; 2013. URL: http://www.bphc.org/healthdata/health-of-boston-report/Documents/HOB-2012-2013/HOB12-13_FullReport.pdf [accessed 2017-03-06] [WebCite Cache ID 6olmlwoHI]
9. Viswanath K. Public communications and its role in reducing and eliminating health disparities. In: Thomson GE, Mitchell F, Williams MB, editors. *Examining the Health Disparities Research Plan of the National Institutes of Health: Unfinished Business*. Washington, DC: National Academies Press; 2006:215–253.
10. Hornik R. *Public Health Communication: Evidence for Behavior Change*. Mahwah, NJ: L Erlbaum Associates; 2002.
11. Viswanath K, Finnegan JR. The knowledge gap hypothesis: twenty-five years later. *Ann Int Commun Assoc* 1996;19(1):187–228. [doi: [10.1080/23808985.1996.11678931](https://doi.org/10.1080/23808985.1996.11678931)]
12. Sarkar U, Karter AJ, Liu JY, Adler NE, Nguyen R, Lopez A, et al. The literacy divide: health literacy and the use of an internet-based patient portal in an integrated health system—results from the diabetes study of northern California (DISTANCE). *J Health Commun* 2010;15 Suppl 2:183–196 [FREE Full text] [doi: [10.1080/10810730.2010.499988](https://doi.org/10.1080/10810730.2010.499988)] [Medline: [20845203](https://pubmed.ncbi.nlm.nih.gov/20845203/)]
13. Houts PS, Doak CC, Doak LG, Loscalzo MJ. The role of pictures in improving health communication: a review of research on attention, comprehension, recall, and adherence. *Patient Educ Couns* 2006 May;61(2):173–190. [doi: [10.1016/j.pec.2005.05.004](https://doi.org/10.1016/j.pec.2005.05.004)] [Medline: [16122896](https://pubmed.ncbi.nlm.nih.gov/16122896/)]
14. Standing L, Conezio J, Haber RN. Perception and memory for pictures: single-trial learning of 2500 visual stimuli. *Psychon Sci* 1970;19(2):73–74. [doi: [10.3758/bf03337426](https://doi.org/10.3758/bf03337426)]
15. Bower GH, Karlin MB, Dueck A. Comprehension and memory for pictures. *Mem Cognit* 1975 Mar;3(2):216–220. [doi: [10.3758/BF03212900](https://doi.org/10.3758/BF03212900)] [Medline: [21287062](https://pubmed.ncbi.nlm.nih.gov/21287062/)]
16. Haring MJ, Fry MA. Effect of pictures on children's comprehension of written text. *Educ Commun Technol J* 1979;27(3):185–190 [FREE Full text] [doi: [10.1007/BF02765450](https://doi.org/10.1007/BF02765450)]
17. Bartram DJ. Comprehending spatial information: the relative efficiency of different methods of presenting information about bus routes. *J Appl Psychol* 1980 Feb;65(1):103–110. [Medline: [7364703](https://pubmed.ncbi.nlm.nih.gov/7364703/)]
18. Vogel EK, Woodman GF, Luck SJ. Storage of features, conjunctions, and objects in visual working memory. *J Exp Psychol* 2001;27(1):92–114. [doi: [10.1037//0096-1523.27.1.92](https://doi.org/10.1037//0096-1523.27.1.92)]
19. Ware C. *Information Visualization: Perception for Design (Interactive Technologies)*. 3rd edition. Waltham, MA: Morgan Kaufmann; 2012.

20. Domagk S, Schwartz RN, Plass JL. Interactivity in multimedia learning: an integrated model. *Comput Hum Behav* 2010 Sep;26(5):1024-1033. [doi: [10.1016/j.chb.2010.03.003](https://doi.org/10.1016/j.chb.2010.03.003)]
21. Zarcadoolas C. The simplicity complex: exploring simplified health messages in a complex world. *Health Promot Int* 2011 Sep;26(3):338-350. [doi: [10.1093/heapro/daq075](https://doi.org/10.1093/heapro/daq075)] [Medline: [21149317](https://pubmed.ncbi.nlm.nih.gov/21149317/)]
22. Weave. Visualize your data. Lowell, MA: Weave Visual Analytics URL: <http://iweave.com> [accessed 2017-03-07] [WebCite Cache ID 6on2NC8uu]
23. Dufilie A, Fallon J, Stickney P, Grinstein G. Weave: a web-based architecture supporting asynchronous and real-time collaboration. 2012 Presented at: AVI 2012 Advanced Visual Interfaces International Working Conference; May 22-26, 2012; Capri, Italy URL: <https://pdfs.semanticscholar.org/38c4/115e0d11c5530c6128cfa2d87ad7a98479f7.pdf>
24. Patton AP, Perkins J, Zamore W, Levy JI, Brugge D, Durant JL. Spatial and temporal differences in traffic-related air pollution in three urban neighborhoods near an interstate highway. *Atmos Environ* 2014 Dec 01;99:309-321 [FREE Full text] [doi: [10.1016/j.atmosenv.2014.09.072](https://doi.org/10.1016/j.atmosenv.2014.09.072)] [Medline: [25364295](https://pubmed.ncbi.nlm.nih.gov/25364295/)]
25. Patton AP, Zamore W, Naumova EN, Levy JI, Brugge D, Durant JL. Transferability and generalizability of regression models of ultrafine particles in urban neighborhoods in the Boston area. *Environ Sci Technol* 2015 May 19;49(10):6051-6060 [FREE Full text] [doi: [10.1021/es5061676](https://doi.org/10.1021/es5061676)] [Medline: [25867675](https://pubmed.ncbi.nlm.nih.gov/25867675/)]
26. An interactive map of traffic pollution for Boston Chinatown. URL: <http://chinatown.oicweave.com> [accessed 2017-08-14] [WebCite Cache ID 6siTaGLNk]
27. Sentell T, Braun KL. Low health literacy, limited English proficiency, and health status in Asians, Latinos, and other racial/ethnic groups in California. *J Health Commun* 2012;17(suppl 3):82-99 [FREE Full text] [doi: [10.1080/10810730.2012.712621](https://doi.org/10.1080/10810730.2012.712621)] [Medline: [23030563](https://pubmed.ncbi.nlm.nih.gov/23030563/)]
28. Mackert M, Mabry-Flynn A, Champlin S, Donovan EE, Pounders K. Health literacy and health information technology adoption: the potential for a new digital divide. *J Med Internet Res* 2016 Oct 04;18(10):e264 [FREE Full text] [doi: [10.2196/jmir.6349](https://doi.org/10.2196/jmir.6349)] [Medline: [27702738](https://pubmed.ncbi.nlm.nih.gov/27702738/)]
29. Plume Labs. Plume air report: the air pollution forecast app in your city. Paris, France: Plume Labs URL: <https://plumelabs.com/en/products/air-report> [accessed 2017-03-21] [WebCite Cache ID 6p8QT52I0]
30. U.S. Environmental Protection Agency, Office of Air Quality Planning Standards. AirNow. Research Triangle Park, NC: U.S. Environmental Protection Agency, National Service Center for Environmental Publications; 2017. URL: <https://www.airnow.gov/> [accessed 2017-03-20] [WebCite Cache ID 6p6y1plY4]
31. Barzyk TM, Isakov V, Arunachalam S, Venkatram A, Cook R, Naess B. A near-road modeling system for community-scale assessments of traffic-related air pollution in the United States. *Environ Modelling Software* 2015 Apr;66:46-56. [doi: [10.1016/j.envsoft.2014.12.004](https://doi.org/10.1016/j.envsoft.2014.12.004)]
32. Shiffer MJ. Spatial multimedia representations to support community participation. In: Craig WJ, Harris TM, Weiner D, editors. *Community Participation and Geographic Information Systems*. London, UK: Taylor and Francis; 2002:309-319.
33. Meppelink CS, van Weert JC, Haven CJ, Smit EG. The effectiveness of health animations in audiences with different health literacy levels: an experimental study. *J Med Internet Res* 2015;17(1):e11 [FREE Full text] [doi: [10.2196/jmir.3979](https://doi.org/10.2196/jmir.3979)] [Medline: [25586711](https://pubmed.ncbi.nlm.nih.gov/25586711/)]
34. Boratto L, Carta S, Fenu G, Manca M, Mulas F, Pilloni P. The role of social interaction on users motivation to exercise: a persuasive web framework to enhance the self-management of a healthy lifestyle. *Pervasive Mobile Comput* 2017;35:98-114. [doi: [10.1016/j.pmcj.2016.08.009](https://doi.org/10.1016/j.pmcj.2016.08.009)]

Abbreviations

CAFEH: Community Assessment of Freeway Exposure and Health Study

PM_{2.5}: particulate matter ≤ 2.5 μm in diameter

PNC: particle number concentration

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