Factors affecting willingness to share electronic health data among California consumers

Katherine K. Kim 1*, Pamela Sankar 2, Machelle D. Wilson 3 and Sarah C. Haynes 1

Abstract

Background: Robust technology infrastructure is needed to enable learning health care systems to improve quality, access, and cost. Such infrastructure relies on the trust and confidence of individuals to share their health data for healthcare and research. Few studies have addressed consumers’ views on electronic data sharing and fewer still have explored the dual purposes of healthcare and research together. The objective of the study is to explore factors that affect consumers’ willingness to share electronic health information for healthcare and research.

Methods: This study involved a random-digit dial telephone survey of 800 adult Californians conducted in English and Spanish. Logistic regression was performed using backward selection to test for significant ($p$-value $\leq 0.05$) associations of each explanatory variable with the outcome variable.

Results: The odds of consent for electronic data sharing for healthcare decreased as Likert scale ratings for EHR impact on privacy worsened, odds ratio (OR) $= 0.74$, 95% CI [0.60, 0.90]; security, OR $= 0.80$, 95% CI [0.66, 0.98]; and quality, OR $= 0.59$, 95% CI [0.46–0.75]. The odds of consent for sharing for research was greater for those who think EHR will improve research quality, OR $= 11.26$, 95% CI [4.13, 30.73]; those who value research benefit over privacy OR $= 2.72$, 95% CI [1.55, 4.78]; and those who value control over research benefit OR $= 0.49$, 95% CI [0.26, 0.94].

Conclusions: Consumers’ choices about electronically sharing health information are affected by their attitudes toward EHRs as well as beliefs about research benefit and individual control. Design of person-centered interventions utilizing electronically collected health information, and policies regarding data sharing should address these values of importance to people. Understanding of these perspectives is critical for leveraging health data to support learning health care systems.

Keywords: Consent, Ethics, Electronic health records, Health information exchange, Learning healthcare systems, Distributed research network

Background

Over the last decade, the expansion and improvement of electronic health records (EHRs) has become a national priority and is described as “foundational to achieve the nation’s health and wellness goals [1]”. To promote this goal the 2009 Health Information Technology for Economic and Clinical Health act (HITECH) allocated $35$ billion dollars to incentivize the development and implementation of electronic health record (EHR) systems for all healthcare providers [2–4]. On many counts the effort has succeeded. By 2015 $83.8\%$ of hospitals had basic EHR systems, up from $15.6\%$ in 2010, and physicians with a certified EHR system had increased from $30\%$ to $75\%$ [5].

HITECH’s goals however extend beyond improving the country’s fragmented EHRs. Nationwide adoption of EHRs is meant to enable development of a new model for healthcare, describes as a learning health care system (LHCS), in which patient information, captured at the point of care, is analyzed to assess treatment efficacy, safety, and quality of care, and fed back to improve patient care. The strategy of grounding research in patients’ usual care, intended to overcome the challenge biomedical
research faces producing knowledge that transfers successfully from experimental protocol to daily life, has motivated many new approaches to research including pragmatic clinical trials, comparative effectiveness research (CER), and research on medical practices (ROMP) [6]. Each seeks to contribute to an LHCS model in different ways, but all share a defining feature, which is the need to multiply the uses of patient data. In addition to informing patient care, this data must now support quality improvement and research.

Few dispute the potential benefits of a LHCS model. However, implementing it presents more than a technical challenge. Transforming the patient encounter into research data also requires accepting changes to privacy and confidentiality, which are important foundations of patient trust and effective health care. Studies demonstrate broad support among patients and consumers for approaches to improving health care that rely on using patient information [7, 8]. They also have consistently found privacy to be a major concern [9-14]. The disjuncture between the technical demands of creating an information infrastructure to support LHCS models and respecting patient concerns about the privacy and control of their health information poses a key ethical challenge to this model of health care.

EHRs in themselves worry some patients who believe they offer fewer privacy protections than paper records [13]. A 2015 survey of a random sample of 1000 US residents suggests that this belief is waning, but still found 41.2% believed that EHRs compromised patient privacy [15]. Another recent study suggests that patients withhold more information from practitioners using EHRs than from those using paper records [16]. EHR privacy concerns have been confirmed even among subjects who agree that EHRs are more efficient than paper records [17] and are not necessarily limited to individuals who are less familiar with computers, [18] although other studies contradict this [12, 19]. Higher education has been associated with the belief that EHRs will improve quality of health care and security [15] but also with a decreased willingness to provide access to personal health information [20, 21].

The LHCS model requires EHRs that are fully integrated or linked into regional networks so that information about a patient’s care can be shared across the sites where it is delivered. Health information exchanges (HIEs) are one type of regional network that fulfills this need. To enter a patient’s identifiable health information into an HIE, the research proceeds without further contact or involvement with the patient. For some, this lack of connection signaled a lack of control, and created concerns such as for what future purposes their data might be used [21].

Trust—in the health system, study staff, or a provider’s recommendation—has been shown to increase the likelihood that patients will participate in research [36, 37, 42-44], and distrust to decrease it [38, 39, 45, 46]. Sometimes, in EHR-based research, once a patients’ consent is obtained, the research proceeds without further contact or involvement with the patient. For some, this lack of connection signaled a lack of control, and created concerns such as for what future purposes their data might be used [33].

There is some evidence that patients’ willingness to share data for research may differ from willingness to share data for healthcare, although few studies have addressed both purposes together. A recent study on consent for data sharing in California found that patients are more likely to share de-identified health information for research purposes than to share identified information for

To realize the vision of a LHCS model patient data such as that collected through HIEs or otherwise aggregated through EHRs, also needs to be available for quality improvement and research. There is growing debate over whether and how patients should consent to this use of their health care information, often shared in de-identified form [6, 28, 29]. Some ethicists have argued that refusing to contribute to the public health in this way is shortsighted and that contributing one’s health care information should be as an ethical obligation, not a choice [6, 28]. The fact that patient information is already routinely shared with researchers in de-identified form is not necessarily understood by patients [9, 30, 31] and studies about patient interactions with electronic health information systems do not always distinguish the different uses to which health data might be put [19].

Studies report broad support of the idea of using treatment-generated data to conduct research to improve medical care [32-34]. Respondents however also express concern that such research will jeopardize the privacy of their medical information [19, 27, 32-34], and that these concerns are sufficient to lead people to refuse to share health information or to strictly limit its use [33, 35].

There is a body of research examining attitudes that affect willingness to consent to participate in research studies. Many factors identified are associated with the protocol itself, such as side effects or the stringency and complexity of the protocol [36]. In addition to practical considerations, research participants also cite altruism [37-41] as influencing decisions to participate in research, which studies of willingness to participate in EHR-based research also cite [20, 27, 32].
healthcare purposes [47]. Minority identity has been shown to be linked with reduced willingness to share information for research purposes, [19, 48, 49] but one study linked Asian identity with a greater willingness to share for purposes of an HIE [26].

A better understanding of how consumer characteristics, attitudes, and beliefs affect willingness to consent to electronic data sharing may support the development of strategies to enhance the public’s trust in networks, and hence their participation in networked models that support the development of LHCS models to improve patient care. This study explores the factors that may affect a patient’s willingness to share electronic health data for both healthcare and research purposes.

Objectives
The primary research questions addressed by the study were:

1. What are the factors associated with willingness to share health information electronically for healthcare?
2. What are the factors associated with willingness to share health information electronically for research?

Methods
A random digit dialed, anonymous survey of California residents was conducted in both Spanish and English in early 2013. The survey was developed from themes that arose from patient focus groups and previously published surveys including items related to privacy, security, HIE, EHR, and health research and underwent cognitive testing before being fielded. The study was approved by the institutional review board at San Francisco State University. A detailed description of the study methodology and the complete survey is published elsewhere [47].

Two outcome variables were defined, one for HIE consent and one for research consent, each with four point scales (very likely, somewhat likely, somewhat unlikely, very unlikely) that were condensed to binary response categories of likely and unlikely. Health information exchange (HIE) was described in the survey as medical information shared electronically between the places where a patient receives medical care. HIE consent was addressed with the question “If you were offered the same response categories of improve, no effect, slightly worsen, greatly worsen) that were condensed to three categories of improve, no effect, and worsen. The concepts of privacy and security were described as: Having a say in who can collect, use and share your medical information has to do with the privacy of your records. Having safeguards (including the use of technology) in place has to do with the security of your medical records. The three questions related to HIE attitudes were: “If doctors use electronic medical records, instead of paper records, how do you think that would affect the quality of medical care?” “Sometimes patients may need to share medical information with different places where they receive medical care. If medical information could be shared electronically between the places where a patient receives medical care, how do you think that would affect the privacy of medical information?” and “If medical information could be shared electronically between the places where a patient receives medical care, how do you think that would affect the security of medical records?”

Three questions addressed research attitudes. The first question, “How do you think that making unidentified data from electronic medical records available for research would affect the quality of medical research studies?” was similar to the question asked of HIE and also offered the same response categories of improve, no effect, and worsen. The second question which assessed the trade-off of the benefit of research over individual privacy was “How much do you agree or disagree with this statement: Research that could be beneficial to people’s health is more important than protecting people’s privacy.” The third question which assessed the trade-off of the value of individual control over research benefit was “How much do you agree or disagree with this statement: An individual’s right to control use of their medical information is more important than the possible benefits of medical research.”

Summary statistics were calculated for each variable – frequencies for the categorical variables and means and standard deviation for the continuous. Univariable tests were performed to assess the association of each explanatory variable with the binary outcome variable (likely or not likely to consent). Univariable logistic regressions were fit to estimate unadjusted odds ratios and their 95% confidence intervals. A logistic regression was performed using backward selection to test for significant associations of each explanatory variable with the outcome variable, while adjusting for the other explanatory variables. Nagelkerke’s generalized R squared was calculated to estimate the
improvement over the null model for each multivariable logistic regression model. A p-value of .05 for a covariate in the final model was considered significant. All analyses were performed using SAS® software version 9.3.

**Results**

There were 800 adult respondents with a response rate of 14.0%, a rate similar to recent national, random digit dialed surveys [50]. As previously reported, the survey respondents were more likely to be 65 years or older and college educated than the general California or US populations; less ethnically/racially diverse than the state, but more diverse than the US population. The California sample shows similar technology use characteristics to respondents in recent national Pew surveys [47]. Respondent characteristics are presented in Table 1. The mean age was 53 years ($m = 53.0, SD = 17.29, range 18–99$). The

| Table 1 Sample characteristics | Representativeness of sample |
|-------------------------------|-----------------------------|
| Variable % (N)                | Survey sample | 95% CI | California<sup>a</sup> | US<sup>a</sup> |
| Total responses               | 100.0 (800)   |       |                       |               |
| Gender                        |               |       |                       |               |
| Female                        | 53.0 (424)    | [49.5, 56.5] | 50.3 | 50.8 |
| Age, years                    |               |       |                       |               |
| 18-64                         | 72.0 (576)    | [68.9, 75.1] | 83.0 | 85.8 |
| 65 and older                  | 25.0 (200)    | [22.0, 28.0] | 17.0 | 14.2 |
| Race/Ethnicity                |               |       |                       |               |
| White (not Hispanic/Latino)   | 56.0 (448)    | [52.6, 59.4] | 39.7 | 63.0 |
| Hispanic/Latino               | 22.9 (183)    | [20.0, 25.8] | 38.1 | 16.9 |
| Asian/Pacific Islander        | 8.0 (64)      | [6.1, 9.9] | 6.6 | 5.3 |
| Black                         | 4.6 (37)      | [3.1, 6.1] | 14.1 | 13.1 |
| Mixed/other                   | 4.9 (39)      | [3.4, 6.4] | 3.6 | 2.4 |
| Native American               | 1.1 (9)       | [0.4, 1.8] | 1.7 | 1.2 |
| Education                     |               |       |                       |               |
| Up to high school             | 23.9 (191)    | [20.9, 26.9] |       |       |
| Technical training/some college | 27.6 (216)    | [24.5, 30.7] |       |       |
| College degree or higher      | 48.0 (384)    | [44.5, 51.5] | 30.3<sup>b</sup> | 28.2<sup>b</sup> |
| Geography                     |               |       |                       |               |
| Urban                         | 90.6 (673)    | [88.6, 92.6] |       |       |
| Veteran                       | 11.8 (94)     | [9.6, 14.0] | 7.8 | 0.1 |
| Income                        |               |       |                       |               |
| Median household income in (dollars) $ | 50,000-60,000 | 61,632 | 52,762 |
| Online Technology Use (%)     | Survey Sample | 95% CI | US        |       |
| Internet                      | 86.9 (695)    | [84.6, 89.2] | 85<sup>c</sup> |       |
| email                         | 95.1 (661)    | [93.6, 96.6] | 92<sup>d</sup> |       |
| Ever used email to contact your doctor or nurse | 44.5 (294) | [41.1, 47.9] |       |       |
| Used the internet to connect with other patients | 13.2 (92) | [10.9, 15.5] | 15.8<sup>e</sup> |       |
| Ever participated in an online patient community | 10.9 (76) | [8.7, 13.1] | 8<sup>e</sup> |       |
| Ever shared your own health information for a research project via online patient community (of those in online community) | 9 (11.8) | [9.6, 14.0] |       |       |
| Have an account for a personal health record | 21.0 (168) | [18.2, 23.8] |       |       |

<sup>a</sup>2007-2011 American Community Survey 5-year Estimates. Available at https://www.census.gov/quickfacts/table/PST045216/00, accessed 8/15/2013
<sup>b</sup>Population 25 years of older
<sup>c</sup>http://www.pewinternet.org/fact-sheet/internet-broadband/ Spring 2013 Survey
<sup>d</sup>http://www.pewinternet.org/Reports/2011/Search-and-email/Report.aspx 2011 data
<sup>e</sup>http://www.pewinternet.org/2013/01/15/health-online-2013/
The majority of respondents are likely to consent to sharing data for healthcare and research purposes. Table 2 provides a summary of attitude and consent outcome variables used in the following results.

**Research question 1. What are the factors associated with willingness to electronically share health information for healthcare?**

For the univariable tests, health status and all privacy and security attitudes were significant at the .05 level, while race was significant at the .10 level. See Table 3. No other covariates were significant.

For the logistic regression, the backward selection procedure eliminated the covariates in the following order with their respective p-values: gender (p = 0.99), urban (p = 0.64), personal health record (PHR) (p = 0.61), health status (p = 0.58), regular provider (p = 0.40), and income (p = 0.10). After controlling for the other included variables, the following covariates were significant: education (p = 0.013), ever emailed provider (p = 0.001), EHR effect on privacy (p < 0.001), HIE effect on privacy (p = 0.003), and HIE effect on security (p = 0.028).

The odds of consenting to HIE were lower for those with low education levels compared to high, i.e., high school compared to college graduate OR 0.52, 95% CI [0.11, 2.43] and some college vs. college graduate, OR 0.51, 95% CI [0.33, 0.80]. The odds for those who had ever emailed their provider were only 47% of those who had not, 95% CI [0.30, 0.74], or equivalently, the odds of those who had not emailed their provider were almost twice that of those who had. Attitudes also affect HIE consent. For every unit increase in the Likert scale response for EHR effect on medical quality, i.e. from improve to no effect to worsen, the odds of consent decreased by a factor of 0.59. For every unit increase in HIE effect on privacy, the odds of consent decreased by a factor of 0.74. And for every unit increase in HIE effect on security, the odds of consent decreased by a factor of 0.80.

**Research question 2: What are the factors associated with willingness to electronically share health information for research?**

For the univariable tests, race/ethnicity, income, education, having a regular provider, emailing provider, and PHR characteristics were significant at the .05 level. The three research attitudes were also significant. See Table 5.

For the logistic regression, the backward selection procedure eliminated the covariates in the following order with their respective p-values: gender (p = 0.75), health status (p = 0.65), regular provider (p = 0.48), ever emailed provider (p = 0.31), geography (p = 0.15), income (p = 0.13), and age (p = 0.11). After controlling for the other included variables, the following covariates were significant: race (p = 0.01), education (p = 0.01), PHR (p = 0.05), EHR effect on research quality (p < .0001), research benefit over privacy (p = 0.0005), and control over research benefit (p = 0.03). The odds of consent for those with a PHR are twice that of those who don't have PHR, OR = 1.99, 95% CI [1.01, 3.92].

The odds ratio of consent for those who think EHR will improve research quality compared to those who think it will worsen is 11.26, 95% CI [4.13, 30.73] while the odds ratio for those who are neutral is 1.74 compared to those who think it will worsen quality, 95% CI [0.57, 5.29]. The odds of those who value research benefit over privacy are 2.72 times those who don't, 95% CI [1.55, 4.78] while the odds of those who value individual control over research benefit are 0.49 those who don't, 95% CI [0.26, 0.94]. Nagelkerke's adjusted R squared was 0.26, indicating a 26% improvement over the null model.

| Table 2 | Summary of attitude and consent variables |
|---------|------------------------------------------|
| Attitudes | Percentage |
| | Improve | No effect | Worsen |
| EHR effect on medical quality | 73.5 | 15.0 | 6.8 |
| EHR effect on privacy | 22.0 | 25.6 | 52.4 |
| EHR effect on security | 37.0 | 20.3 | 42.7 |
| HIE effect on privacy | 28.9 | 30.7 | 40.3 |
| HIE effect on security | 30.6 | 27.0 | 42.5 |
| EHR effect on research quality | 74.3 | 12.8 | 8.0 |
| Values | | Agree | Disagree |
| Research benefit over privacy | 50.6 | 46.8 |
| Control over research benefit | 69.8 | 26.9 |
| Consent | | Likely | Unlikely |
| Share for healthcare (HIE) | 56.4 | 42.0 |
| Share for research | 74.8 | 23.4 |
### Table 3  Unadjusted relationships between characteristics and attitudes and likelihood of HIE consent

| Characteristics                                      | N     | Odds ratio | 95% confidence interval | P   |
|------------------------------------------------------|-------|------------|-------------------------|-----|
| Gender: Female vs. Male                              | 784   | 0.88       | (0.67, 1.20)             | 0.39|
| Race                                                 | 767   |            |                         | 0.08b|
| Hispanic/Latino                                      | 1.45  |            | (1.01, 2.08)             |     |
| Black                                                | 0.65  |            | (0.33, 1.27)             |     |
| Asian/Pacific Islander                               | 0.95  |            | (0.56, 1.62)             |     |
| Other                                                | 0.73  |            | (0.40, 1.33)             |     |
| White, not Hispanic/Latino (ref)                     | –     | –          |                         |     |
| Income                                               | 619   |            |                         | 0.11|
| Less than $40,000                                    | 1.40  |            | (0.95, 2.00)             |     |
| $40 K to < $80 K                                     | 1.50  |            | (0.97, 2.20)             |     |
| $80 k or greater (ref)                               | –     | –          |                         |     |
| Education                                            | 778   |            |                         | 0.21|
| Up to High School Diploma                            | 1.50  |            | (0.80, 2.90)             |     |
| HS diploma to Bachelor's Degree                      | 0.88  |            | (0.65, 1.20)             |     |
| More than a Bachelor's Degree (ref)                  | –     | –          |                         |     |
| Geography: Urban vs. Rural                           | 732   | 1.30       | (0.79, 2.20)             | 0.29|
| Health Status                                        | 781   |            |                         | 0.02a|
| Very good                                            | 0.66  |            | (0.49, 0.89)             |     |
| Fair                                                 | 0.57  |            | (0.29, 1.10)             |     |
| Poor (ref)                                           | –     | –          |                         |     |
| Has regular provider (yes vs. no)                    | 779   | 1.10       | (0.78, 1.40)             | 0.72|
| Has used email to contact provider (yes vs. no)      | 654   | 0.87       | (0.64, 1.20)             | 0.39|
| Currently has personal health record (yes vs. no)    | 774   | 0.92       | (0.65, 1.30)             | 0.64|

Attitudes (Likert scale)

| EHR effect on medical quality                        | 752   | 0.59       | (0.51, 0.69)             | <0.001a|
| EHR effect on privacy                                | 730   | 0.66       | (0.58, 0.74)             | <0.001a|
| HIE effect on privacy                                | 752   | 0.69       | (0.61, 0.78)             | <0.001a|
| EHR effect on security                               | 751   | 0.71       | (0.64, 0.80)             | <0.001a|
| HIE effect on security                               | 741   | 0.66       | (0.59, 0.75)             | <0.001a|

*Significant at the .05 level
bSignificant at the .10 level

### Table 4  Adjusted odds ratios for effects on likelihood of HIE consent

| Effect                                              | Odds ratio estimates |
|------------------------------------------------------|----------------------|
|                                                      | Point estimate       | 95% Wald confidence limits |
| Education                                            |                      |
| Up to High School diploma                           | 0.52                 | 0.11                         | 2.43|
| HS diploma to Bachelor's Degree                     | 0.51                 | 0.33                         | 0.80|
| More than a Bachelor's Degree (ref)                  | –                    |                             |     |
| Has PHR (yes vs no)                                  | 0.47                 | 0.30                         | 0.74|
| EHR effect on medical quality (Likert from improve to worsen) | 0.59                 | 0.46                         | 0.75|
| EHR effect on privacy (Likert from improve to worsen) | 0.74                 | 0.60                         | 0.90|
| EHR effect on security (Likert from improve to worsen) | 0.80                 | 0.66                         | 0.98|
Discussion

Demographic factors
While we did not find an association between ethnic or racial identity and willingness to consent to electronic sharing of health information for healthcare, we did find a significant association with respect to sharing for research. This is in line with other studies that have shown that minority identity was linked with reduced willingness to share information (although not specifically electronic information) for research purposes [19, 48, 49] and Asian identity linked with a greater willingness to share for purposes of HIE [26]. However, in the context of consent to participate in research, a previous systematic review that found no difference in consent rates between minority–African Americans, Hispanics– and non-Hispanic whites [51]. While we did not find health status to be a predictor, other studies have found that patients with poorer health are more likely to consent to a research study [52, 53].

Experience with technology
The mere presence of experience with technology for consumers is not indicative of acceptance of electronic sharing of health information. For example, those who used email to contact providers had lower odds of consenting to HIE but those who used a PHR had greater odds of consenting to research exchange. There may be

Table 5 Unadjusted relationships between characteristics and attitudes and likelihood of research consent

| Characteristics                   | N   | Odds ratio | 95% confidence interval | P     |
|-----------------------------------|-----|------------|-------------------------|-------|
| Gender: Female vs Male            | 782 | 0.77       | (0.55, 1.10)            | 0.11  |
| Age (years)                       | 763 | 1.01       | (0.99, 1.02)            |       |
| Race/Ethnicity                    | 766 |            |                         | <0.001*|
| Hispanic/Latino                   |     | 0.44       | (0.30, 0.66)            |       |
| Black                             |     | 0.35       | (0.17, 0.71)            |       |
| Asian/Pacific Islander            |     | 0.52       | (0.29, 0.94)            |       |
| Other                             |     | 0.72       | (0.35, 1.48)            |       |
| White, not Hispanic/Latino (ref)  |     | –          | –                       |       |
| Income                            | 620 |            |                         | 0.048*|
| Less than $40,000                 |     | –          | –                       |       |
| $40K to < $80K                    |     | 0.55       | (0.33, 0.91)            |       |
| $80K or greater (ref)             |     | 0.80       | (0.34, 0.68)            |       |
| Education                         | 772 |            |                         | <0.001*|
| Up to High School diploma         |     | 0.34       | (0.14, 0.66)            |       |
| HS diploma to Bachelor’s Degree   |     | 0.48       | (0.34, 0.68)            |       |
| More than a Bachelor’s Degree (ref)|    | –          | –                       |       |
| Geography: Urban vs Rural         | 731 | 1.36       | (0.77, 2.38)            | 0.29  |
| Health status                     | 778 |            |                         | 0.66  |
| Very good                         |     | 1.38       | (0.68, 2.81)            |       |
| Fair                              |     | 1.38       | (0.67, 2.83)            |       |
| Poor (ref)                        |     | –          | –                       |       |
| Have regular provider (yes vs no) | 777 | 2.03       | (1.44, 2.85)            | <0.001*|
| Ever emailed provider (yes vs no) | 655 | 1.92       | (1.29, 2.85)            | 0.001*|
| Have PHR (yes vs no)              | 772 | 2.25       | (1.40, 3.62)            | <0.001|
| Attitudes                         | 749 |            |                         | <0.001*|
| EHR effect on research quality (Likert) | 749 |   |             |       |
| Improve                           |     | 11.68      | (6.56, 20.80)           |       |
| No effect                         |     | 1.99       | (1.02, 3.85)            |       |
| Worsen (ref)                      |     | –          | –                       |       |
| Research benefit over privacy (agree vs disagree) | 765 | 1.65 | (1.18, 2.31) | 0.003*|
| Control over research benefit (agree vs disagree) | 759 | 0.61 | (0.41, 0.92) | 0.02*|

*Significant at the .05 level
unintended consequences of greater use of technology such as HIE including lack of confidence among patients that appropriate safeguards are in place to protect the privacy and security of their health information [54]. Deeper exploration of past health technology experience may elucidate these relationships.

### Table 6 Adjusted odds ratios for effects on likelihood of sharing for research

| Effect                                           | Odds ratio estimates |
|--------------------------------------------------|----------------------|
| Race/Ethnicity                                   | Point estimate | 95% Wald confidence limits |
| Hispanic/Latino                                  | 0.37          | 0.20 | 0.71 |
| Black                                            | 0.83          | 0.25 | 2.77 |
| Asian/Pacific Islander                           | 0.28          | 0.12 | 0.67 |
| Other                                            | 0.58          | 0.20 | 1.72 |
| White, not Hispanic/Latino (ref)                 | –             | –    | –    |
| Education                                        | Point estimate | 95% Wald confidence limits |
| Up to High School Diploma                        | 0.80          | 0.16 | 4.05 |
| HS diploma to Bachelor’s Degree                  | 0.42          | 0.25 | 0.73 |
| More than a Bachelor’s Degree (ref)              | –             | –    | –    |
| Has EMR (yes vs no)                              | 1.99          | 1.01 | 3.92 |
| Research affects research                        | Point estimate | 95% Wald confidence limits |
| Improve                                          | 11.26         | 4.13 | 30.73 |
| Neutral                                          | 1.74          | 0.57 | 5.29 |
| Worsen (ref)                                     | –             | –    | –    |
| Research benefits more important that privacy (agree vs disagree) | 2.72          | 1.55 | 4.78 |
| Individual rights more important than benefits of research (agree vs disagree) | 0.49          | 0.26 | 0.94 |

Tierney et al. found that patients often prefer restricting provider access to their EHR data; and providers accessed EHR for those patients who had requested restricted access far less often when they knew the patient’s preference [56]. An area of future inquiry would be whether and how provider communication regarding their own beliefs about electronic data sharing, use of EHRs, privacy and security affects patients’ willingness and subsequent participation in electronic data sharing.

In addition, tradeoffs between values related to societal benefits of research, individual control, and privacy impact willingness to consent in research. Individuals may make trade-offs when considering whether to share their health information for research.

These findings taken together suggest that attitudes about information technology might be more complex than previous research has recognized. Our findings add new depth of understanding that attitudes about technology’s role in healthcare and research are important dimensions in individuals’ willingness to participate in electronic data sharing. Simplistic characterizations of personal preferences based on demographic factors or previous technology experience may not be adequate for designing person-centered networks. They add a new dimension to questions concerning trust and research participation, particularly the need to examine the interaction between the electronic basis of LHCS research, trust, and willingness to participate. They also suggest a need to examine the basis of attitudes toward electronic data sharing, and in particular to disambiguate the influence of experience with computer technology generally (for example in banking or shopping) and experience specific to health care. This information might help to inform strategies to address patient concerns about LHCS research demands. This need is particularly critical in light of ethics proposals to reframe this type of research participation as obligatory rather than voluntary [28].

The logistic regression models for factors affecting willingness to share electronic data indicated modest improvement in explanation of variability over the null model (R-squared of 13% for healthcare and 17% and 26% for research). There is much room for future research to deepen our understanding of the role of values and beliefs in decisions about electronic health information sharing.

### Limitations

This study was designed as a statewide survey to help inform California policy making and the stakeholder-informed design of the Scalable National Network for Effectiveness Research distributed research network (DRN) [11]. Although the demographics of the survey sample are similar to the general population there are important differences which may limit the generalizability...
of the findings to either California or the US. The sample is older and has a relatively high education level compared to the general public. Thus respondents may have different experiences related to healthcare policy and technology which lead to varying attitudes. Another limitation is the response rate of 14%. Although comparable to other random-digit dial national surveys, the low response rate tempests inferences made from the results. In order to better understand the relationship of attitudes and experience with EHRs/HIEs and consent behavior, prospective or natural experiments would be illuminating.

**Conclusions**

Learning healthcare systems inextricably link the domains of practice and research. They depend on electronically networked partners, including the person and their caregivers, in order to collect relevant data, analyze information, and understand how to move health system levers that lead to improvements in health in alignment with each person’s needs, values, and preference which is the hallmark of person-centeredness in healthcare. Consequently, understanding of the intricacies of individuals’ attitudes about electronic data sharing, whether those data were collected for healthcare purposes, or are being reused for research, is critical to developing networked systems that embody person-centeredness.

**Abbreviations**

DRN: Distributed research network; EHR: Electronic health record; HIE: Health information exchange; HITECH: 2009 Health Information Technology for Economic and Clinical Health act; LHCS: Learning healthcare system; PHR: Personal health record

**Acknowledgements**

Not applicable.

**Funding**

The authors were partially funded by grants from AHRQ R01HS019913 (KK), Gordon and Betty Moore Foundation grant to the Betty Irene Moore School of Nursing at UC Davis (KK), and National Center for Advancing Translational Sciences, National Institutes of Health, through grant #UL1TR000002 (MDW). Funding for the survey was provided by Award Number 90HT0029 to California Health and Human Services Agency from Office of the National Coordinator for Health Information Technology (ONC), U.S. Department of Health and Human Services. Its contents are solely the responsibility of the authors and do not necessarily represent the official views of ONC or the State of California. The funders had no role in decisions regarding the data or manuscript.

**Availability of data and materials**

The anonymous datasets from the current study are available from the corresponding author on reasonable request. No identifying/confidential patient data was collected.

**Authors’ contributions**

KK oversaw the study, developed the survey, collected and analyzed data, and led the writing of the manuscript. KK was employed at San Francisco State University at the time the research was conducted. MDW conducted the statistical analysis, wrote major sections of the manuscript and revised it critically. PS made substantial contributions to interpretation of the data in the context of biomedical ethics, wrote major sections of the manuscript and revised it critically. SH contributed to interpretation of the data in the context of data sharing and health information exchange, wrote major sections of the manuscript and revised it critically. All authors read and approved the final manuscript.

**Competing interests**

The authors declare that they have no competing interests.

**Consent for publication**

The manuscript does not contain any individual person’s data in any form.

**Ethics approval and consent to participate**

The study was conducted with approval of the Institutional Review Board of San Francisco State University (KK’s institution at the time of the study). This study was an anonymous telephone survey of adults over the age of 18 for which no personal information was collected. Verbal consent was requested prior to commencing the survey.

**Sources of support**

The authors were partially funded by grants from AHRQ R01HS019913 (KK), Gordon and Betty Moore Foundation grant to the Betty Irene Moore School of Nursing at UC Davis (KK, SH), National Center for Advancing Translational Sciences, National Institutes of Health, through grant #UL1TR000002 (MDW). Funding for the survey was provided by Award Number 90HT0029 to California Health and Human Services Agency from Office of the National Coordinator for Health Information Technology (ONC), U.S. Department of Health and Human Services. Its contents are solely the responsibility of the authors and do not necessarily represent the official views of ONC or the State of California. The funders had no role in decisions regarding the data or manuscript.

**Publisher’s Note**

Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.

**Author details**

1. University of California Davis, Betty Irene Moore School of Nursing, 2450 48th Street, Suite 2600, Sacramento, CA 95817, USA. 2. Department of Medical Ethics and Health Policy, University of Pennsylvania, 433 Guardian Drive, Blochley, 14, Philadelphia PA19104-4884, USA. 3. Department of Public Health Sciences, Division of Biostatistics, Clinical and Translational Science Center, University of California Davis, 2921 Stockton Blvd, suite 1400, Sacramento, CA 95817, USA.

**Received: 29 October 2016 Accepted: 24 March 2017**

**Published online: 04 April 2017**

**References**

1. The Office of the National Coordinator for Health Information Technology (ONC). Report to congress: update on the adoption of health information technology and related efforts to facilitate the electronic use and exchange of health information. 2016.
2. Adler-Milstein J, DesRoches CM, Kolbe C, Foster G, Worzala C, Charles D, et al. Electronic health record adoption in US hospitals: progress continues, but challenges persist. Health Aff 2015(Epub before print November).
3. Herman B. EHRs and health IT projects: Are they battering Hospitals’ financial profiles? Becker’s hospital CFO. 2014.
4. 11th Congress of the United State of America. American recovery and reinvestment Act of 2009, 2009.
5. Robeznieks A. Growing EHR Adoption Despite Persistent Problems. Healthcare Financial Management Association. 2016. https://www.hfma.org/Content.aspx?id=49631. Accessed 1 Aug 2016.
6. Cho MK, Magnus D, Constantine M, Lee SS-J, Kelley M, Alexi S, et al. Attitudes towards risk and informed consent for research on medical practices: a cross-sectional survey. Ann Intern Med. 2015;162(10):690–24.
7. Hill K. Building a methodology for monitoring and measuring civic engagement. 2013.
8. Kaplan SH, Gombosev A, Fireman S, Sabin J, Heim L, Shemelman L, et al. The patient’s perspective on the need for informed consent for minimal risk studies: development of a survey-based measure. Empir Bioeth. 2016(12):116–24.
9. Medical Research Council. The use of personal health information in medical research: general public consultation. 2007.
10. Agaku IT, Adisa AO, Ayo-Yusuf OA, Connolly GN. Concern about security and privacy, and perceived control over collection and use of health information are related to withholding of health information from healthcare providers. J Am Med Inform Assoc. 2014;21(2):374–8. doi:10.1136/amiajnl-2013-002079.
11. Kim KK, Browe DK, Logan HC, Holm R, Hack L, Ohno-Machado L. Data governance requirements for distributed clinical research networks: triangulating perspectives of diverse stakeholders. J Am Med Inform Assoc. 2014;21(4):714–9.

12. Ancker JS, Silver M, Miller MC, Kaufal R. Consumer experience with and attitudes toward health information technology: a nationwide survey. J Am Med Inform Assoc. 2013;20(1):152–6. doi:10.1136/amiajnl-2012-001062.

13. Pereira G, Holbrook A, Thabane L, Foster G, Willison DJ. Views on health information sharing and privacy from primary care practices using electronic medical records. Int J Med Inform. 2011;80(2):94–101.

14. Willison DJ, Steeves V, Chesnes C, Schwartz L, Ranford J, Agarwal G, et al. Consent for use of personal information for health research: Do people with potentially stigmatizing health conditions and the general public differ in their opinions? BMC Medical Ethics. 2009;10(1):10. doi:10.1186/1472-6939-10-10.

15. Ancker JS, Brennan S, Richardson JE, Silver M, Kaufal R. Trends in public perceptions of electronic health records during early years of meaningful use. Am J Manag Care. 2015;21(8):e487–93.

16. Campos-Castillo C, Anthony DL. The double-edged sword of electronic health records: implications for patient disclosure. J Am Med Inform Assoc. 2015;22(1):e130–40.

17. Choy A, Hudson Z, Pritts J, Goldman J. Exposed Online: Why the new attitudes towards electronic data sharing for healthcare and research. J Am Med Inform Assoc. 2015;22(4):821–30.

18. McGraw D, Dempsey JX, Harris L, Goldman J. Privacy as an enabler, Not an obstacle. HealthIT.gov. Meaningful Consent Overview. 2014. http://www.healthit.gov/providers-professionals/meaningful-consent-overview. Accessed 27 July 2015.

19. Weitzman ER, Kaci L, Mandl KD. Sharing medical data for health research: the patients’ perspective. J Med Internet Res. 2005;7:e95.

20. Page SA, Manhas KP, Muruve DA. A survey of patient perspectives on the research use of health information and biospecimens. BMC Med Ethics. 2016;17:48.

21. HealthIT.gov. Meaningful Consent Overview. 2014. http://www.healthit.gov/providers-professionals/meaningful-consent-overview. Accessed 27 July 2015.

22. Simon SR, Evans JS, Benjamin A, Delano D, Bates DW. Patients’ attitudes toward electronic health information exchange: qualitative study. J Med Int Res. 2009;11(3):e30.

23. Medford-Davis LN, Chang L, Rhodes KV. Health Information Exchange: What do patients want? Health Informatics J. 2016;Epub before print May 31.

24. Weitzman ER, Kaci L, Mandl KD. Sharing medical data for health research: the early-personal health record experience. J Med Int Res. 2012;12(2):e14. doi:10.2196/jmir.1356.

25. Faden RR, Kass NE, Goodman SN, Pronovost P, Tunis S, Beauchamp TL. An ethical framework for dynamic consent and research feedback: a qualitative study. J Med Internet Res. 2015;17(1):e18.

26. HealthIT.gov. Meaningful Consent Overview. 2014. http://www.healthit.gov/providers-professionals/meaningful-consent-overview. Accessed 27 July 2015.

27. Shavers-Hornaday V, Lynch C, Burmeister L. Factors that influence African-Amercians’ willingness to participate in medical research studies. Cancer. 2001;91(5):1233–6.

28. Galea S, Tracy M. Participation rates in epidemiologic studies. Ann Epidemiol. 2000;10(3):205–15.

29. Tait A, Voepel-Lewis T, Makiya S. Participation of children in clinical research: factors that influence a parent’s decision to consent. Anesthesiology. 2003;99(4):819–25.

30. Lieberman B, Meana M, Stewart D. Cardiac rehabilitation: gender differences in factors influencing participation. J Women’s Health. 1998;7(6):717–23.

31. Llewellyn-Thomas H, McGreal M, Thiel E, Fine S, Erlichman C. Patients’ willingness to enter clinical trials: measuring the association with perceived benefit and preference for decision participation. Soc Sci Med. 1991;32(3):35–42.

32. Shavers-Vornhady V, Lynch C, Burmeister L, Tomer J. Why are African Americans underrepresented in medical research studies? Impediments to participation. Ethnicity and Health. 1997;2(1-2):31.

33. Fox S. The Social Life of Health Information. Washington, DC: Pew Internet & American Life Project; 2011. http://www.pewinternet.org/2011/05/12/the-social-life-of-health-information-2011/. Accessed 2 Apr 2017.

34. Kim KK, Joseph JG, Ohno-Machado L. Comparison of consumers’ views on electronic data sharing for healthcare and research. J Am Med Inform Assoc. 2015;22(4):821–30. doi:10.1093/jamia/ocu014.

35. Grande D, Mitra N, Shah A, Wan F, Asch DA. Public preferences about secondary uses of electronic health information. JAMA Intern Med. 2013; 173(19):1798–806.

36. Dimitropoulos L, Patel V, Scheffler SA, Posnack S. Public attitudes toward health information exchange: perceived benefits and concerns. Am J Manag Care. 2011;17(12 Spec No):SP11–56.

37. Fox S. The Social Life of Health Information. Washington, DC: Pew Internet & American Life Project; 2011. http://www.pewinternet.org/2011/05/12/the-social-life-of-health-information-2011/. Accessed 2 Apr 2017.

38. Wendler D, Kington R, Madans J, Van Wye G, Christ-Schmidt H, Pratt LA, et al. Are racial and ethnic minorities less willing to participate in health research? PLoS Med. 2006;3(2):e201.

39. Friedman DB, Foster C, Bergeron CD, Tanner A, Kim SH. A qualitative study of racial ethnic minorities’ experiences of the cancer clinical research: perspectives of adult patients and parents of pediatric patients undergoing hematopoietic stem cell transplantation. Biol Blood Marrow Transplant. 2012;18(10):1604–11.

40. Steele F, Lloyd N, Harrington L, Wallace P. Use of electronic patient records for medical research: conflicts and contradictions. BMC Health Serv Res. 2015;15:124.