

# Electronic Alerts and Clinician Turnover: The Influence of User Acceptance

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**R**etaining primary care providers (PCPs) is critical to ensuring healthcare access and quality. However, PCPs are moving to other specialty areas or leaving medicine altogether, a significant threat to high-quality care in many US regions.<sup>1-4</sup> Recent healthcare legislation and initiatives such as the Health Information Technology for Economic and Clinical Health (HITECH) Act of 2009, the Affordable Care Act of 2010, and the Patient-Centered Medical Home make explicit resource provisions such as training, additional staff, and resources to implement electronic health records (EHRs), all of which could make primary care more attractive.<sup>5-7</sup> While PCPs' decisions to seek alternate employment might be determined by a multitude of factors,<sup>8</sup> provider dissatisfaction with the implementation and meaningful use of EHRs<sup>9</sup> may pose unique retention challenges despite the HITECH Act's strong incentives for their use.

Implementing a full-service EHR constitutes a major organizational intervention; it presents a significant change in clinician-to-clinician communication and in some instances can require additional skills beyond those that were needed in paper-based systems to deliver care of comparable quality.<sup>10,11</sup> For example, providers spend an average of 49 minutes per day reviewing and responding to electronic alert notifications, yet nearly half of these notifications do not contain messages that providers perceive to contain "high value" information.<sup>12</sup> This volume of information is significantly higher than what would be expected in a paper system (given the additional resources involved with physical mail and messaging), and thus requires a different work strategy for accurate and timely handling. In addition, research has also shown marked differences in workflow efficiency of paper-based versus EHR<sup>13-15</sup>; paper-based clinical information often gets "lost in the shuffle" and becomes untrackable.<sup>16,17</sup> The needed changes in communication and work flow attributable to EHR use are significant enough that despite

## ABSTRACT

### Objectives

Use of certain components of electronic health records (EHRs), such as EHR-based alerting systems (EASs), might reduce provider satisfaction, a strong precursor to turnover. We examined the impact of factors likely to influence providers' acceptance of an alerting system, designed to facilitate electronic communication in outpatient settings, on provider satisfaction, intentions to quit, and turnover.

### Study Design and Methods

We conducted a cross-sectional Web-based survey of EAS-related practices from a nationwide sample of primary care providers (PCPs) practicing at Department of Veterans Affairs (VA) medical facilities. Of 5001 invited VA PCPs, 2590 completed the survey. We relied on Venkatesh's Unified Theory of Acceptance and Use of Technology to create survey measures of 4 factors likely to impact user acceptance of EAS: supportive norms, monitoring/feedback, training, and providers' perceptions of the value (PPOV) of EASs to provider effectiveness. Facility-level PCP turnover was measured via the VA's Service Support Center Human Resources Cube. Hypotheses were tested using structural equation modeling.

### Results

After accounting for intercorrelations among predictors, monitoring/feedback regarding EASs significantly predicted intention to quit ( $\beta = 0.30$ ,  $P < .01$ ), and PPOV of EASs predicted both overall provider satisfaction ( $\beta = 0.58$ ,  $P < .01$ ) and facility-level provider turnover levels ( $\beta = -0.19$ ,  $P < .05$ ), all without relying on any intervening mechanisms.

### Conclusions

Design, implementation, and use of EASs might impact provider satisfaction and retention. Institutions should consider strategies to help providers perceive greater value in these clinical tools.

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their benefits, 12% of pediatric urologists reported they would retire if EHR use were mandated.<sup>18</sup>

Recent research has demonstrated a relationship between the use of health information technology (HIT) and physician satisfaction,<sup>9</sup> although different components of HIT have different effects on physician satisfaction. For example, compared with traditional, paper-based forms of communication, PCPs who communicated electronically with patients and other providers, and who shared their visit notes electronically with patients, were more likely to report higher satisfaction levels<sup>19</sup>; in contrast, PCPs who wrote prescriptions electronically were less likely to report high satisfaction levels.<sup>20</sup> Ensuring physicians' satisfaction with their work is important because poor satisfaction often leads to several undesirable results, including turnover, mental health concerns (eg, anxiety, depression, burnout), poorer relationships with patients, and reduced quality of care.<sup>21</sup>

Within the Accountable Care Organization model, EHRs are expected to facilitate communication and coordination, especially in the outpatient setting.<sup>22</sup> Increasingly, practices are relying on EHR-based alerting systems (EASs) within their EHR to track, route, and communicate clinical information such as test results.<sup>23</sup> This electronic communication may occur through an "asynchronous" alert notification inbox, much like e-mail, where the sender and recipient need not be simultaneously engaged. Notifications transmitted through these systems could include test results, referrals, status updates on patients, and other provider-to-provider communications. Although many commercial EHRs already feature EASs functionality for communication, its use is expected to grow. For example, results management is one of the core EHR functionalities<sup>24</sup> and key criteria for achieving Stage 2 meaningful use<sup>25</sup> due to its potential to reduce lag time in recognition and treatment of medical problems, reduce redundant testing, and improve appropriate and timely follow-up. Unlike other EHR components, however, the impact of EASs on provider satisfaction and turnover is not well documented.

PCPs' utilization of EHR-based EASs has become an integral job characteristic, especially because PCPs utilizing EASs tend to spend significant amounts of time interfacing with them.<sup>12,26-28</sup> Consequently, problems related to EASs could potentially impact PCP job attitudes such as satisfaction and intentions to quit, both known antecedents of turnover.<sup>29</sup> For example, our previous research

### Take-Away Points

How electronic health record (EHR)-based alerting systems are implemented, accepted, and used in real-world clinical settings can impact not just quality of care, but also provider satisfaction and retention.

- Our study is the first to our knowledge to link provider perceptions of the value of EHR-based alert notifications to actual turnover in a large national sample.
- When providers do not perceive EHR features to be of high value to their practice, they can become dissatisfied with their work.
- EHR-based alerting systems could become catalysts for turnover, unless providers clearly understand their value for delivering high-quality care effectively and efficiently.

suggests providers using EASs receive an average of 56 to 63 alert notifications per day<sup>12,22,30</sup>; in addition, providers do little to customize their alerts interface to optimize efficiency and effectiveness. Rather, they employ varying strategies for managing these notifications, with mixed success,<sup>31</sup> leading to information overload.<sup>32</sup>

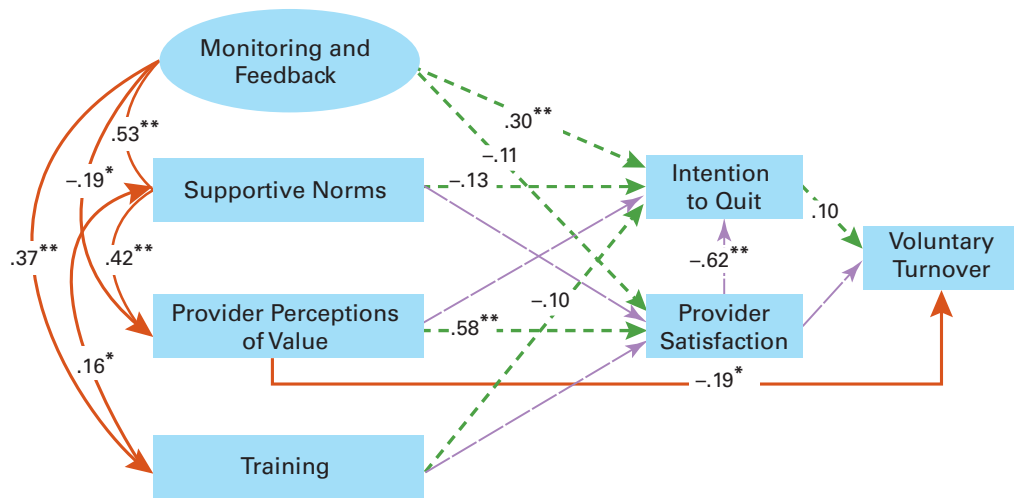
PCPs have requested new visualization tools such as color coding and advanced filtering to address some of these problems.<sup>33</sup> Thus, without modifications, EASs use could lead to a combination of high-volume, low-value work that could function as a driver of turnover for PCPs, rather than a source of retention. As there is no prior research that explores the relationship between EHR variables and provider outcomes, our objective was to conduct an initial examination to identify how providers' perceptions of the use of EASs may impact their satisfaction and intention to quit. This research could inform strategies about how those perceptions could be altered if needed, and can also serve as a stepping-stone for the integration of EHR research with turnover research.

## METHODS

### Conceptual Model

To answer our research questions, we took guidance from 2 theoretical models to ultimately derive the model depicted in **Figure 1**. The Job Demands Resource Model of burnout (JD-RM)<sup>34</sup> posits that job demands (ie, aspects of the job such as, potentially, the use of EASs) that require sustained physical or mental effort and lead to increased workload can lead to negative outcomes such as low satisfaction, intentions to quit one's job, and eventually actual turnover. The model further proposes that *job resources* (ie, aspects of the job that are functional in achieving work goals) that reduce job demands, or stimulate personal growth and development, relate to positive attitudinal outcomes and lower levels of withdrawal.

■ **Figure 1.** Model of Interrelationships Among Study Variables and Standardized Estimates



Curved arrows denote covariances (nondirectional associations among variables). Straight arrows denote regression weights (independent variable–dependent variable relationships). All estimates shown are standardized. Purple arrows indicate relationships originally hypothesized as part of the model that were deleted after model trimming; orange arrows indicate relationships not originally hypothesized as part of the model that were added after model trimming; green arrows indicate relationships originally hypothesized as part of the model that remained in the model after trimming. Monitoring and feedback is an aggregate of 2 variables, and thus is indicated as a latent variable.

With respect to EASs, it is unclear whether PCPs perceive EASs as a demand or a resource. The second model, Venkatesh’s Unified Theory of Acceptance and Use of Technology (UTAUT),<sup>35</sup> sheds some light on what may drive this decision for providers. The UTAUT proposes several factors that could impact the aforementioned outcomes; among these are performance expectancy (the extent to which the user believes the system will help attain gains in performance); social influence (the extent to which the user perceives that important others, such as family and friends, believe the system should be used), and facilitating conditions (the extent to which the user perceives that technical and organizational resources exist to support system use). We thus examined 4 specific examples of the types of factors proposed by Venkatesh and their impact on physician satisfaction, intention to quit, and turnover: (1) EAS-supportive norms, such as the extent to which colleagues use and see value in the notifications (an example of social influence), (2) whether providers receive feedback about their use of EASs, (3) whether providers receive training on the use of notifications (both examples of facilitating conditions), and (4) the perceived contribution of EASs to provider effectiveness (an example of performance expectancy, henceforth referred to as provider perceptions of value [PPOV]). Based on the JDRM and UTAUT, we hypothesized that each of the 4 factors will positively impact provider satisfaction, and inversely relate to intention to quit. Fur-

thermore, intention to quit and provider satisfaction will significantly impact turnover.

**Design**

The present study is part of a larger cross-sectional Web-based survey of EAS practices conducted between June and November 2010 on a nationwide sample of PCPs practicing at Department of Veterans Affairs (VA) medical facilities.<sup>30,32,33</sup>

**Setting**

The VA is the largest integrated healthcare system in the United States and one of the most advanced in terms of fully functional EHR use,<sup>36</sup> with EAS-capable EHRs having been in place at all medical facilities for almost a decade.<sup>37</sup> In addition, the VA provides various national-level resources to support providers in their use of the Computerized Patient Records System (CPRS, the VA’s EHR), including nationally developed training modules, clinical application coordinators whose role is to assist clinicians in using CPRS, and a national performance measurement and quality reporting system. Despite these nationally available resources, however, previous research has shown considerable variability in the implementation of standardized national resources across VA facilities<sup>38</sup> (eg, computerized clinical reminders)<sup>39</sup> and in some cases this local variability has significantly impacted quality and performance.<sup>40</sup> It is this local variation and degree of adap-

■ **Figure 2.** Alert Notification Window in Computerized Patient Records System

**Patient Selection**

Patient List:  Default,  Providers,  Team/Personal,  Specialties,  Clinics,  Wards,  All

Patients: No Appointments, ZZTestpt1, ZZTestpt 2, ZZTestpt 3, ZZTestpt 4, ZZTestpt 5, ZZTestpt 11, ZZTestpt 21, ZZTestpt 31, ZZTestpt 41, ZZTestpt 51

Buttons: OK, Cancel, Save Patient List Settings

**Notifications**

Info	Patient	Location	Urgency	Alert Date/Time	Message
	ZZTestpt 1		Moderate	01/20/2009@13:47	Scheduled Consult: ORTHOPEDICS
	ZZTestpt 9999999999	3B	Moderate	01/22/2009@11:37	Scheduled Consult: NON-INVASI ECHO
	ZZTestpt 333	3C MED	HIGH	01/21/2009@00:21	Medications nearing expiration.
	ZZTestpt 55555		Moderate	01/22/2009@12:39	Imaging request held: KNEE 3 VIEWS, LEFT E
	ZZTestpt 55555		Moderate	01/22/2009@15:14	Imaging Results: KNEE 3 VIEWS, RIGHT
	ZZTestpt 55555		Moderate	01/22/2009@15:10	Imaging Results: KNEE 3 VIEWS, LEFT
	ZZTestpt 7777777	2A REHAB	Moderate	01/16/2009@12:09	Imaging Results: CT THORAX W/CONT
	ZZTestpt 7777777	2A REHAB	Moderate	01/16/2009@09:17	Imaging Results: CHEST SINGLE VIEW
	ZZTestpt 7777777	2A REHAB	Moderate	01/16/2009@12:09	Imaging Results: 3D/SAG/COR/RECONSTR
	ZZTestpt 22		Moderate	01/22/2009@12:43	Forwarded consult PHARMACY HOUSTON OI
	ZZTestpt 666666		Moderate	01/21/2009@10:57	Discontinued consult NON-INVASI ECHO
	ZZTestpt 4444		Moderate	01/21/2009@12:39	Completed Consult NON-INVASI EKG: BEDS
	ZZTestpt 9999999999	3B	Moderate	01/22/2009@13:06	Completed Consult NON-INVASI ECHO
	ZZTestpt 22		Moderate	01/14/2009@09:39	Completed Consult AUDIOLOGY
	ZZTestpt 666666		Moderate	01/21/2009@13:05	Abnormal labs - [TROPONIN I]
	ZZTestpt 55555		Moderate	01/21/2009@11:20	Abnormal labs - [LIPID PROFILE, COMPREHE
	ZZTestpt 888888888		Moderate	01/16/2009@08:18	Abnormal labs - [COMPREHENSIVE METABO
	ZZTestpt 55555		Moderate	01/21/2009@11:08	Abnormal labs - [CBC&PLT] WITH DIFF]

Buttons: Process Info, Process All, Process, Forward, Show Comments, Remove

tation among facilities, despite the availability of national-level resources, which inspired the facility-level analyses in this study.

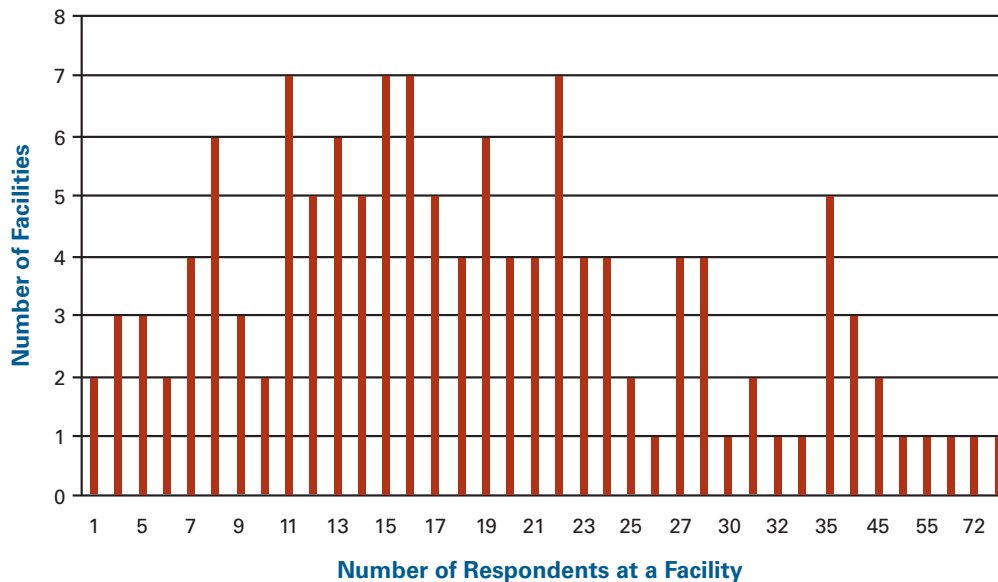
**Electronic alert system features.** CPRS features an inbox style electronic alert notification system (“View Alerts”).<sup>41</sup> The View Alert window is displayed when a provider logs into CPRS, notifying the user of clinically significant events such as abnormal diagnostic test results (see **Figure 2**; a full taxonomy of the available alert/notification types within CPRS has been described elsewhere).<sup>12</sup> The user can customize how alerts are displayed in several ways via features such as sorting and turning off nonmandatory notifications.<sup>31</sup> Alerts stay in the View Alert window for a prespecified time or until the user specifically acknowledges the alert (ie, clicks the alert to read it). Thus, the View Alert system is a system of asynchronous com-

munication used nationally in the VA to facilitate communication among multiple members of the patient’s care team. A core set of CPRS functionalities is determined at the national VA level. In addition, individual facilities have the flexibility to alter some of the CPRS settings. For instance, some facilities opt to have providers receive a larger number of relevant alerts, while other facilities alter settings so providers only view certain types of alerts that are considered “mandatory” at the institution level.

### Participants and Procedure

Details of the survey’s development are reported elsewhere.<sup>32</sup> In brief, using a nationwide VA administrative database (VA Primary Care Management Module), we invited all VA PCPs with a minimum practice panel size of 250 patients (N = 5001). Of 5001 PCPs invited, 2590 (51.8%)

■ **Figure 3.** Histogram of Number of Survey Respondents per Facility



Number of respondents are raw respondents, rather than percentage of potential respondents. We did not have access to the number of potential respondents at each facility, as our original list of potential participants was only identified by geographic network. Survey respondents indicated to which facility they were assigned in response to a question in the survey. Consequently, it was not possible to calculate a within-facility response rate.

responded, representing data from 131 different VA facilities. **Figure 3** displays the distribution of respondents within a facility across all 131 facilities. Respondents were 55.4% female, 31.1% nonwhite, and 31.5% nonphysician providers (eg, physician assistants, nurse practitioners); and 82.1% had 2 or more years in VA practice (**Table 1**). Within VA primary care, nonphysician providers behave largely as physicians do: They have their own patient panels, do at least 85% of the same work as physicians (differences being largely administrative),<sup>42</sup> and use CPRS in largely the same way. Consequently, physicians and nonphysician providers were treated as a single population and identified as PCPs for study purposes.

Our study was reviewed and approved by our local institutional review board. Participants were recruited as follows: We first asked chiefs of primary care at each facility to e-mail information about the project and the upcoming survey to the PCPs at their respective sites. We subsequently invited all participants via a personalized e-mail from the study's principal investigator; this e-mail described the study and provided a link to the Web-based survey. To increase response rates, invitation e-mails and subsequent reminders were followed by telephone attempts to reach nonrespondents.

## Measures

**Table 2** contains a list of constructs, construct defini-

tions, sample items, and response scales used for the current study.

**User Acceptance Factors.** Measures for EAS Supportive Norms, Monitoring/Feedback and Training Infrastructure, and PPOV were developed specifically for this study based on a literature review<sup>43-50</sup> and refined through pilot-testing with PCPs. Descriptive statistics, correlations, and reliability coefficients for these measures appear in **Table 3**. Details on the development of the survey instrument are reported elsewhere.<sup>33</sup>

**Provider Satisfaction and Intention to Quit.** Satisfaction and intention to quit were each measured via a single survey item, as suggested by Cortese and Quaglino.<sup>51</sup>

**Turnover.** Facility-level voluntary turnover rates for 2010 were obtained from the Veterans Health Administration Service Support Center Human Resources Cube, a large administrative database repository maintained centrally by the VA.

## Data Analysis

**Facility-Level Aggregation.** Because our primary outcome, turnover, is a facility-level variable, it was necessary to aggregate all predictors (monitoring/feedback, supportive norms, PPOV, training, intention to quit, and provider satisfaction) to the facility level (n = 131) in order to successfully assess their impact on facility-level turnover. To test whether within-facilities responses were suffi-



■ **Table 1.** Characteristics of Survey Respondents (n = 2590)

Characteristic	Per Facility				Samplewide
	Mean (n)	SD (n)	min	max	n (%)
<b>Age</b>					
20-39 years	2.58	2.43	0	11	338 (13.1)
40-49 years	5.23	4.09	0	21	685 (26.4)
50-59 years	7.34	5.10	1	32	961 (37.1)
60 years and older	3.07	2.66	0	15	402 (15.5)
Not reported	1.56	1.50	0	7	204 (7.9)
<b>Gender</b>					
Male	8.24	5.89	0	36	1080 (41.7)
Female	10.25	7.16	0	41	1343 (51.9)
Not reported	1.27	1.32	0	7	167 (6.4)
<b>Race</b>					
White	12.44	9.37	0	57	1630 (62.9)
Black	0.90	1.42	0	7	118 (4.6)
Asian	3.29	3.44	0	16	431 (16.6)
Other	1.44	1.40	0	6	188 (7.3)
Not reported	1.70	1.70	0	9	223 (8.6)
<b>Job classification</b>					
Physician, academic	3.34	4.04	0	18	438 (16.9)
Physician, nonacademic	9.37	7.61	1	39	1228 (47.4)
Nurse practitioner	4.28	3.99	0	19	561 (21.7)
Physician assistant	1.56	2.19	0	14	204 (7.9)
Not reported	1.21	1.31	0	7	159 (6.1)
<b>Years at VA</b>					
<2	3.34	3.40	0	20	437 (16.9)
2-10	9.31	6.48	0	34	1219 (47.1)
11-20	4.50	3.49	0	14	589 (22.7)
>20	1.53	1.73	0	9	201 (7.8)
Not reported	1.10	1.18	0	7	144 (5.6)
<b>Native language</b>					
English	14.59	9.93	0	58	1911 (73.8)
Other	3.80	3.95	0	19	498 (19.2)
Not reported	1.38	1.38	0	8	181 (7.0)
<b>Overall</b>	19.77	12.62	1	76	2590
VA indicates Department of Veterans Affairs. Per facility statistics indicate the mean number of respondents per facility with the indicated characteristic (mean [n]) and the SD around that mean (SD [n]).					

ciently homogeneous to justify aggregation, we calculated  $r_{wg}$ , a measure of inter-rater agreement,<sup>52</sup> for each relevant variable for each facility. Average  $r_{wg}$  score was 0.71, suggesting sufficient agreement to warrant aggregation.

**Model Test.** We used structural equation modeling to test our hypothesized path model. Figure 1 presents the model tested; purple lines and green lines denote the initial model; red and green lines denote the final, best-fitting

model. We additionally computed simple bivariate correlations to further explore the data. All analyses were conducted using SPSS Amos 17.<sup>53</sup>

## RESULTS

### Descriptive Statistics

Table 3 presents means, standard deviations, and

■ **Table 2.** Construct Definitions and Sample Items

Construct & Definition	Sample Item (number of items)	Response Scale
<b>Monitoring and Feedback:</b> Degree to which employee's performance using EAS is monitored and to which feedback is provided	"I receive feedback from my facility leadership on my performance regarding follow-up of alert notification." (2)	Strongly disagree (1) to Strongly agree (5)
<b>Supportive Norms:</b> Degree to which socially relevant organizational members believe and communicate that EAS is useful and effective	"My supervisor believes alert notifications in CPRS are an essential component of effective primary care." (2)	Strongly disagree (1) to Strongly agree (5)
<b>Training:</b> Degree to which respondents receive adequate and continuous EAS training	"I have received appropriate levels of training in using the CPRS alert notification system." (2)	Strongly disagree (1) to Strongly agree (5)
<b>EAS Provider Perceptions of Value:</b> Degree to which providers perceive EAS to impact their effectiveness and performance as a provider	"Using alert notifications in CPRS enhances my ability to provide safe patient care." (4)	Strongly disagree (1) to Strongly agree (5)
<b>Intention to Quit:</b> Plans to withdraw from one's job	"The frustration I experience with the View Alert system has led me to consider quitting work at the VA." (1)	Strongly disagree (1) to Strongly agree (5)
<b>Physician satisfaction:</b> Overall evaluation of one's job containing cognition and affect-oriented evaluation components	"Overall, I am satisfied with my job." (1)	Strongly disagree (1) to Strongly agree (5)
<b>Voluntary Turnover:</b> Actual percentage of employees who voluntarily quit their jobs at a given facility	Facility-level voluntary turnover rates as recorded in VA-systemwide database (Veterans Support Service Center-Human Resource Data Cube)	Not applicable

EAS indicates electronic health record-based alert system; CPRS, Computerized Patient Records System; VA, Department of Veterans Affairs.

correlations among study variables. Of note, bivariate correlations indicated a significant positive relationship between intention to quit and facility level turnover ( $r = 0.169, P < .05$ ); and a significant negative relationship between provider satisfaction and facility-level turnover ( $r = -0.167, P < .05$ ). Additionally, supportive norms and PPOV each correlated with provider satisfaction ( $r = 0.286, P < .01$  and  $r = 0.495, P < .01$ , respectively) and intention to quit ( $r = -0.170, P < .05$  and  $r = -0.383, P < .01$ , respectively), whereas monitoring/feedback correlated only with intention to quit ( $r = 0.185, P < .05$ ). These significant correlations suggest testing a complete model is warranted.

**Test of Hypothesized Model**

**Initial model fit.** The hypothesized relationships (ie, each factor independently predicts provider satisfaction and intention to quit, both of which intercorrelate and in turn predict turnover, depicted in Figure 1 in the purple and green lines) resulted in poor fit (RMSEA = 0.21, PCLOSE  $< .001$ ) when tested as a cohesive model. Of note, facility-level turnover was unrelated to provider satisfaction or intentions to quit.

**Final model.** An important feature of the original model is that the factors in the model were considered orthogonal, independent predictors of satisfaction, intention to quit, and turnover. Bivariate correlations,

however, suggested this was an incorrect assumption. Consequently, based on the initial model results and the simple bivariate correlations, we trimmed unnecessary relationships from the model, and allowed the predictors to covary. The resulting model showed good fit (RMSEA = 0.04, PCLOSE = 0.47), and is presented in Figure 1 (depicted by the green and red lines). As can be seen from the figure and consistent with the bivariate correlations analyses, the 4 factors are significantly correlated, and thus cannot be treated as independent predictors of provider satisfaction. After accounting for intercorrelations amongst the independent variables, monitoring/feedback significantly predicted intention to quit ( $\beta = 0.30, P < .01$ ), and PPOV predicted both provider satisfaction ( $\beta = 0.58, P < .01$ ) and facility level turnover ( $\beta = -0.19, P < .05$ ), all without relying on either provider satisfaction or intention to quit as intermediary mechanisms. Of note, high levels of monitoring and feedback were associated with *increased* intentions to quit.

**DISCUSSION**

This study sought to examine the impact of user acceptance factors of electronic health record-based alert notification systems on the satisfaction, intentions to quit, and turnover of providers who used them. Contrary to existing theory (both the JDRM and the

**Table 3.** Means, Standard Deviations, and Bivariate Correlations Among Study Variables

	Mean	SD	Monitoring and Feedback	Training	Supportive Norms	PPOV	Intention to Quit	Physician Satisfaction	Facility Turnover Rate
Monitoring and feedback	2.95	0.48	<b>0.75</b>						
Training	3.11	0.40	0.418 <sup>a</sup>	<b>N/A</b>					
Supportive norms	3.36	0.23	0.502 <sup>a</sup>	0.229 <sup>a</sup>	<b>0.72</b>				
PPOV	3.34	0.32	0.298 <sup>a</sup>	0.100	0.840 <sup>a</sup>	<b>0.85</b>			
Intention to quit	2.81	0.51	0.185 <sup>b</sup>	0.106	-0.170 <sup>b</sup>	-0.383 <sup>a</sup>	<b>N/A</b>		
Physician satisfaction	3.55	0.35	-0.022	-0.045	0.286 <sup>a</sup>	0.495 <sup>a</sup>	-0.653 <sup>a</sup>	<b>N/A</b>	
Facility turnover rate	0.07	0.03	0.051	-0.029	-0.024	-0.103	0.169 <sup>b</sup>	-0.167 <sup>b</sup>	<b>N/A</b>

PPOV indicates provider perception of value.

<sup>a</sup>Correlation is significant at the .01 level (2-tailed).

<sup>b</sup>Correlation is significant at the .05 level (2-tailed).

Values in blue along the diagonal represent reliability coefficients for the respective scales. Measures with reliabilities of N/A are either constructs measured via a single item or measured via objective rates, which do not require reliability analysis.

UTAUT), we found that monitoring/feedback on EASs practices, training on the use of EASs, and supportive norms about EAS had little impact on provider satisfaction. However, monitoring/feedback were associated with increased intention to quit.

Our results suggest that EASs, and by extension EHRs, could become catalysts for turnover, unless providers clearly understand their value to delivering high-quality care effectively and efficiently. As evidenced by the non-significant relationships between monitoring/feedback and provider satisfaction, as well as the nonsignificant relationship between training and both satisfaction and intention to quit, our data suggest that the aforementioned facilitating conditions may be insufficient to accomplish this goal, though we have no specific details in our data about the quality of the feedback or training. More importantly, when providers do not perceive the value of these electronic aids to their practice, they might become dissatisfied with their work environment, and potentially seek work elsewhere altogether.

EASs likely represent one of the most frustrating components of EHRs for providers<sup>54,56</sup>—compared with paper communication systems, they are perceived to “increase the number of work items, inflate the time to process each, and divert work previously done by office staff to them.”<sup>57</sup> Other work has shown that providers perceive many of the alerts they receive to be unnecessary,<sup>58</sup> and has documented variable physician acceptance of features like computerized reminders and electronic alerts.<sup>59</sup> Therefore, future work should target the problem from multiple angles, such as content and design of feedback, effectiveness of training, and social influence factors, in addition to already ongoing efforts to optimize EAS design, so that it is inherently perceived as valuable by providers. The United States

already has a shortage of primary care providers,<sup>2</sup> and research shows dissatisfied providers are both leaving primary care for other specialties and/or leaving medicine completely.<sup>3</sup>

Several possible reasons might exist to explain the positive effect of monitoring/feedback on intention to quit. First, participants might have reacted more strongly to the monitoring aspect than to the feedback aspect of this construct. Second, the nature of the feedback provided could minimize feedback's impact on satisfaction. Feedback characteristics can have a significant impact on its effectiveness at changing cognitions and behavior.<sup>60,61</sup> Our ongoing research in another domain has found that feedback is often delivered primarily via written reports providing only numeric scores without correct solution information<sup>62</sup> (one of the most powerful single characteristics of feedback interventions).<sup>60,61</sup> Third, both feedback delivery mechanisms and providers' perceptions of being monitored constantly by the organization could have led to the observed result.

In contrast, PPOV showed a direct positive relationship to provider satisfaction (providers who perceived greater value in electronic notifications were more likely to be satisfied); a direct negative relationship to turnover (providers who perceived greater value in electronic notifications were less likely to quit); and an indirect link to intention to quit via provider satisfaction (providers who perceived greater value in alert notifications were more likely to be satisfied, and in turn less likely to express intentions to quit.). The relationship between provider satisfaction and intention to quit is not surprising, as it has been well documented in the literature.<sup>63,64</sup> The more novel finding in this research is the direct, negative relationship between PPOV and turnover (ie, providers at facilities with higher provider turnover rates



have lower perceptions of value for EASs). We are not aware of any studies directly linking these types of perceptions to *actual* turnover, particularly at the organizational level with a national sample as large as this one: 2590 respondents at 131 facilities. From a scientific perspective, this finding links the JDRM and UTAUT: if users do not perceive EASs to be of value, EASs are more likely to be considered a demand rather than a resource (and thereby less likely to be accepted), thus leading to increased turnover. From a practical perspective knowing that EASs have to be perceived as performance enhancing by physicians in order for them to not negatively affect turnover should signal facility leadership to take care regarding how such systems are designed, marketed within the facility, and supported.

In addition to this important finding, we are also not aware of any studies simultaneously examining the effects of satisfaction, intention to quit, and turnover in the healthcare setting. Understanding the interrelationships among user acceptance of technological tools intended to help providers, factors that impact this acceptance, and provider outcomes can help the design and implementation of HIT tools with which providers will *want* to work.

### Limitations

The study was conducted within the VA system, representing one of the largest and most sophisticated healthcare systems in the United States. Hence, in the spirit of constructive replication and to enhance external validity, we recommend that our findings be replicated in subsequent studies at other facilities without centralized organizational structure. Nevertheless, alerting systems such as the one we studied are being increasingly used across commercial EHRs. Second, the structure of the archival turnover data obtained for this study limits turnover analysis over time and prevents the application of statistical techniques such as survival analysis that lead to the most informative results for turnover-type data. Third, our sample consisted of employees who were all using the same EAS-capable EHR, limiting our ability to generalize results to other commonly used EAS-capable EHR systems (eg, Epic Systems, Verona, Wisconsin) that are officially certified (by an Office of the National Coordinator for Health Information Technology-Authorized Testing and Certification Body). Hence, we recommend that future research focus on more heterogeneous samples, examining different types of EASs and EHRs. Finally, although this study identified a new, very specific source of dissatisfaction and potential turnover among providers, future

studies should examine the incremental contribution of this source in the context of more traditional predictors of provider satisfaction such as supervisory relations, availability of resources, and work environment conditions.<sup>65,66</sup> We further encourage future research to closely investigate how providers' perceptions of EHR variables develop over time, and whether system characteristics or more distal factors (eg, supervisory behavior) impact these perceptions.

### CONCLUSIONS

We conclude that designing and implementing EHR-based notification systems effectively may no longer simply be assumed to be an antecedent to efficiency, safety, or quality of care; how these systems are implemented, accepted, and used in real-world practice, as our research shows, might also impact provider satisfaction and retention. Given the recent HITECH stimulus and the new healthcare law, EHRs will be a reality nationwide in a few short years and will connect members of the healthcare team like never before. In fact, one reason for the heavy emphasis on EHR adoption is to improve communication. Depending on how the EHR is designed and implemented, it can become a source of competitive advantage (or turnover) for clinical practices. In addition, how an organization creates and manages its internal policies can make or break both the safety and efficiency of the clinicians' work. As EHRs become more widespread and providers increasingly communicate clinical information through EASs, institutions should consider strategies to help providers perceive greater value in these vital clinical tools.

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