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Abstract:

The design science research (DSR) paradigm prescribes building and evaluating an Information Technology (IT) artifact to address organizational problems incorporating and enhancing relevant theories. In this article, we present a large-scale design science project that aims to address problems in the digital-government domain. The COPLINK project started at the University of Arizona with National Institute of Justice (NIJ) funding to integrate data from multiple law-enforcement databases and subsequently broadened in scope and funding to develop methodologies for capturing, searching, analyzing, and visualizing information for law-enforcement, intelligence, and national security applications. The project involves a complete spectrum of DSR activities including problem diagnosis, technology invention, technology evaluation, and theory building, while contributing to appropriate knowledge bases and making a significant impact on the real-world IT problems. Algorithms, methodologies, techniques, and the resulting Information Systems designed as part of the project are being successfully used in over 300 law-enforcement agencies, and have generated more than forty peer-reviewed publications. In this article, we describe these artifacts and the design process behind them, present summaries of evaluative studies, and discuss the factors that we believe were behind the success of this decade-long DSR project.

Keywords: design science research, digital government, law enforcement, COPLINK, knowledge management, data mining
I. INTRODUCTION

The design science research (DSR) paradigm, in its broadest terms, prescribes building and evaluating an Information Technology (IT) artifact to address organizational problems while utilizing and expanding relevant theories [Hevner et al., 2004; Kuechler et al., 2008; Land et al., 2009; March et al., 1995; Nunamaker et al., 1991]. Innovative IT artifacts are constructed to aid individuals within an organization perform their task efficiently and effectively. In order to achieve this, information science (IS) researchers need to identify relevant problems, understand and contribute to appropriate theoretical bases, develop Information Technology (IT) artifacts, and perform rigorous evaluation to pinpoint problems, measure improvement, or pursue alternative solutions.

The thrust to introduce IT at all levels of government has provided an opportunity for IS researchers to develop and evaluate systems for various problems in the digital government domain. IT is being used in the government to provide citizens with access to information using the Web and other avenues, enable transactions with the government (e.g., using websites and kiosks), enable citizen participation (with e-voting and participatory rule-making), and facilitate information sharing and management within government agencies [Marchionini et al., 2003]. Application areas represent a cross-section of government mandates: healthcare and safety; law enforcement, security, and justice; education; land use; and many others [Chen et al., 2008]. The IS-oriented issues faced by government employees and citizens using government services are not unlike those common to businesses and their customers. Such problems provide IS researchers with the opportunity to design systems for knowledge management, information retrieval, information access, data analysis and data mining. The area also offers excellent scope for studies dealing with the evaluation, adaption, use, and effectiveness of IT artifacts. Since government agencies, just like other organizations, face issues with cultural inertia and technology acceptance, behavioral research methodologies can be used to examine perceived usefulness, impact, valuation, and management of IT artifacts in diverse settings. While the digital government domain offers many opportunities, effective system implementation is a common and critical challenge. Inherently, many of the problems faced by government agencies may be categorized as “wicked problems” typically comprised of technical, economic, and political elements that operate together as a system [Hutchinson et al., 2002]. Solutions to such problems must be technically feasible and organizationally viable. According to Vreede and Vogel [2000], such systems need to be collaboratively designed in such a way that they are capable of addressing the human elements (members of the organization and their working methods), technical elements (hardware and software), and informational elements (data).

In this article, we present the COPLINK and associated digital government research projects at the University of Arizona’s Artificial Intelligence (AI) Lab within the context of DSR frameworks. We describe in detail the DSR activities carried out (problem diagnosis, technology invention, technology evaluation, and theory building) and the lessons we learned that might benefit similar DSR projects. The COPLINK project started with National Institute of Justice (NIJ) funding in 1997, with the goal of designing a system for law enforcement personnel to seamlessly share, and effectively analyze, data extracted from multiple sources. Since then, the project has expanded in scope to include information retrieval, identity resolution, social network and association analysis, and spatio-temporal analysis; it has received additional funding from agencies that include the National Science Foundation (NSF) and the Department of Homeland Security (DHS). Reflecting its broadened goals and impacts, the project has also acquired other titles, such as BorderSafe and Regional Information Sharing and Collaboration (RISC). In July 2009, the COPLINK system was acquired by the industry leading company ‘i2.’ The combined company has more than 450 employees and $100 million in sales in 2009 and is a leading provider of security information sharing and crime analytics. Algorithms, methodologies, techniques, and the resulting Information Systems designed as part of the COPLINK project are being successfully used in over 1600 law-enforcement and intelligence agencies and have generated more than forty peer-reviewed publications. The research has also been featured in the New York Times, Newsweek, Los Angeles Times, Washington Post, and on ABC News, among others. The COPLINK project was selected as a finalist by the prestigious International Association of Chiefs of Police (IACP)/Motorola 2003 Weaver Seavey Award for Quality in Law Enforcement in 2003 and won the INFORMS Design Science Award in 2008.

Section II discusses the COPLINK research project by placing it within the DSR activity framework. Section II presents an overview of COPLINK research and systems developed at the University of Arizona along with a brief description and conclusions of some of the major DSR studies conducted. In the COPLINK project, we used design-science and behavioral-science research in a complementary way to further system development effectiveness. Section III presents a summary of the results of various technology acceptance studies of the COPLINK system. Section IV presents the conclusions and lessons learned from the project.
II. COPLINK RESEARCH PROJECT WITHIN THE DSR FRAMEWORKS

Figure 1 presents the primary research areas within the project and their interactions with knowledge bases in Information Systems (IS) and other reference disciplines within the context of the law enforcement problem domain. As shown, we engage in DSR activities (problem diagnosis, technology invention, technology evaluation, and theory building) presented in Venable [2006] and Nunamaker et al. [1991] in the context of the environment-knowledge base and rigor-relevance framework proposed by Hevner et al. [2004].
Problem Diagnosis
The objective of DSR is to acquire knowledge and understanding that enables the design, development, and implementation of technology-based solutions to organizational problems [Hevner et al., 2004]. The COPLINK project was initiated by a collaborative effort between the Tucson Police Department (TPD) and the AI Lab to address the technical challenges of data compatibility, information accessibility, and knowledge sharing in law-enforcement agencies. TPD agreed to participate in investigations of state-of-the-art, near-term, and cost-effective database, Intranet, and multimedia technologies capable of making justice information integration, management, and access more efficient (implying that the task at hand could be achieved faster than comparative techniques) and effective (implying that the task at hand could be performed better than comparative techniques) [Chen et al., 2003b]. As the project expanded in scope, an officer was assigned as a liaison for the project to facilitate the interactions between AI Lab researchers and TPD officers, detectives, and crime analysts. The liaison officer visited the AI Lab for weekly meetings and provided specialized, in-depth domain expertise on ongoing research efforts. Furthermore, this officer was instrumental in our understanding of other IT-related problems faced by police and other law-enforcement agencies. When a problem was identified, the officer arranged meetings with appropriate parties in TPD that allowed us to better understand the problem scope, prior investigations and the methods employed, the current problem-solving strategy or process, and the existing technology solution (if any). We tackled each of the problems with a team-based approach. A team consisted of a research lead, several scientists and doctoral students, a domain expert from TPD, and a group of software developers. In many cases, a team was further decomposed into multiple subgroups, each concentrating on specific aspects of the problem. The team typically met on a weekly basis (usually face-to-face) to exchange ideas, review the progress, analyze and solve problems, define action items, recruit more members and adjust time tables, if necessary. This was in line with the suggestions by Hutchinson, English, and Mughal [2002] for addressing wicked problems. Our team structure and routine meetings provided the communications and collaborative mechanisms necessary for developing and validating a DSR based IT solution.

Technology Invention, Design, Development
DSR aims at the creation of an artifact to address organizational problems [Hevner et al., 2004]. In the COPLINK project, various challenging problems were addressed using innovative artifacts. These included data integration and identity resolution techniques and models [Chen et al., 2003a; Chen et al., 2003b; Kaza et al., 2008; Wang et al., 2004a, 2004b], social network analysis metrics [Kaza et al., 2009; Xu et al., 2005a, 2005b], spatio-temporal event detection models and algorithms [Chang et al., 2008; Chen et al., 2005], and association analysis techniques [Kaza et al., 2007]. Throughout the development of these artifacts, we approached design as a search process [Hevner et al., 2004]. That is, a prototype artifact was constructed in consultation with the liaison officer; alternatives were continually examined against the agency’s requirements in the design process; algorithms, models, and systems were developed on the basis of real data and realistic assumptions, followed by rigorous evaluations and incremental improvements that lead to the final technology solutions.

Denning [1997] and Hevner [2004] suggest that the artifacts should be more than algorithms, prototypes, or Information Systems. Rather, they should be innovations, practices, and products through which the implementation and use of Information Systems can be effectively and efficiently accomplished. During the development of these IT artifacts, we learned about management’s reluctance and administrative hurdles impeding effective technology implementation within agencies. To address these barriers, administrators at TPD and UA designed template documents for information sharing among government agencies and research institutions, critical to information sharing and implementation of Information Systems in law enforcement agencies [Atabakhsh et al., 2004].

Technology Evaluation
The artifacts produced during the research process were instantiated as prototypes necessary for feasibility demonstrations and system evaluations. We adopted evaluation techniques from data mining (e.g., n-fold cross validation, ROC curves [Han et al., 2006]), employed metrics from information retrieval (e.g., recall and precision), and used prevalent theoretical models including the technology acceptance model [Davis, 1989] and unified theory of acceptance and use of technology [Venkatesh et al., 2003]. Following DSR guidelines [Hevner et al., 2004; Nunamaker et al., 1991], we instantiated the novel algorithms and techniques into systems and tested them within the TPD’s existing technical infrastructure. Our evaluations, embracing both functionality and usability, involved student participants [Chung et al., 2005; Xiang et al., 2005] as well as crime analysts and detectives from law enforcement agencies [Hu et al., 2005; Xu et al., 2004; Xu et al., 2005a], and comparison against appropriate gold standards and metrics [Kaza et al., 2007; Marshall et al., 2008; Wang et al., 2006].

Theory Building
We used appropriate theories in IS and reference disciplines to inform the research conducted in COPLINK project. We employed the social network analysis (SNA) theory [Granovetter, 1982; Wasserman et al., 1994] to examine the
structure, formation, growth, and destruction of different criminal networks based on narcotics, gang related, and other crimes. Record linkage [Fellegi et al., 1969] and other theories provided a basis for identity resolution and data integration methods developed for efficient access to data stored in multiple law enforcement jurisdictions. We used measures in the information theory [Fano, 1961] to conduct association analyses that identify suspect vehicles crossing the southern border. Kernel theories [Vapnik, 1999] were used for building innovative spatial scanning methods for spatio-temporal hotspot detection. All the above studies contributed to the appropriate knowledge bases by the application of theories to large-scale real-world datasets and developing new measures that were more appropriate and effective than existing measures in the problem domain. The studies developed and tested utility theories [Venable, 2006] that asserted that our systems or methods outperformed existing solutions. In addition, we anchored in salient behavioral theories to study target officers' acceptance of the resulting technology artifacts. For example, we developed a factor model premised in the theory of planned behavior [Ajzen 1991] and technology acceptance model [Davis, 1989] and empirically tested the model for explaining law enforcement officers' technology acceptance of a system developed as part of the project [Hu et al., 2005]. We also tested a model built upon unified theory of acceptance and use of technology [Venkatesh et al., 2003] to examine officers' acceptance of COPLINK mobile, a handheld device for accessing COPLINK applications remotely (details of this study are presented in Section IV).

III. SYSTEM OVERVIEW AND RESEARCH STUDIES
The COPLINK research project started with NIJ funding to integrate data from multiple law-enforcement databases and subsequently broadened in scope and funding to develop methodologies for capturing, accessing, analyzing, and visualizing for law-enforcement, intelligence, and national security applications. The first comprehensive version of the system consisted of two major components (the current version includes seven): COPLINK Connect and COPLINK Detect [Chen et al., 2003b]. COPLINK Connect is designed to allow police departments to share data seamlessly through an easy-to-use interface that integrates different data sources including legacy record management systems. The intended users of COPLINK Connect are police officers who have pressing, often mission critical, information needs. The design of this system was closely guided by user requirements acquired through multiphase brainstorming sessions, storyboards, mock system demonstrations, focus groups, and more formally structured questionnaires and interviews [Chen et al., 2003b].

COPLINK Detect uncovers various types of criminal associations that exist in law-enforcement databases. It uses a technique called concept space to identify such associations from existing crime data automatically [Chen et al., 2003b]. In general, a concept space is a network of terms and weighted associations that represent the concepts and their links within an underlying information space. COPLINK Detect uses statistical techniques such as co-occurrence analysis and clustering functions to weight relationships between all possible pairs of concepts.

COPLINK Connect and Detect formed the basis for further research projects to improve the efficacy and efficiency of individuals involved in law enforcement. The primary areas include (1) social network analysis and visualization, (2) spatio-temporal analysis and visualization, (3) identity resolution and data integration, and (4) association analysis. We summarize in the following subsections a few studies on these topics and refer the reader to the appropriate publications for details.

Social Network Analysis and Visualization
Knowledge about the structure and organization of criminal networks is important for both crime investigation and the development of effective strategies to prevent crimes. However, except for network visualization, criminal network analysis remains primarily a manual process. We found that existing tools did not provide advanced structural analysis techniques that allow extraction of network knowledge from large volumes of criminal-justice data. Several studies [Hu et al., 2009; Kaza et al., 2009; Marshall et al., 2008; Xu et al., 2004, 2005a, 2005b] were conducted that ranged from studying the structure of criminal networks using and defining metrics in the SNA theory to the development of systems for the analysis and visualization of networks by end-users. Most studies included the entire range of DSR activities from problem diagnosis to technology evaluation. Two of these studies are summarized below.

CrimeNet Explorer: A Framework for Criminal Network Knowledge Discovery
To aid law-enforcement and intelligence agencies in the efficient and effective use of criminal network knowledge, a framework was proposed for automated network analysis and visualization [Xu et al., 2005a]. Based on the framework, we developed a system called CrimeNet Explorer (Figure 2) that incorporated several advanced techniques: a concept space approach, hierarchical clustering, SNA methods, and multidimensional scaling. Results from controlled experiments involving student subjects demonstrated that our system could achieve higher clustering recall and precision than did untrained subjects when detecting subgroups from criminal networks. Thus, the system could perform effectively. Moreover, subjects identified central members and interaction patterns between groups...
significantly faster with the help of structural analysis functionality than with only visualization functionality. Domain experts comprising of detectives and analysts also reported that CrimeNet Explorer could be very useful in crime investigations. Figure 2 shows snapshots of the CrimeNet Explorer system.

Figure 2. Snapshots from the Crimenet Explorer System Showing Interaction Patterns and Sub-Group Identification [Xu et al., 2005a]

Identifying Significant Facilitators of Dark Network Evolution

In this study [Hu et al., 2009], we used dynamic social-network analysis methods to examine several plausible facilitators of co-offending relationships in a large-scale narcotics network consisting of individuals and vehicles extracted from police records. Innovative methods were used to model network evolution. Multivariate Cox regression and a two-proportion z-test on the cyclic and focal closure metrics showed that mutual acquaintance and vehicle affiliations were significant facilitators for the network under study. We also found that homophily with respect to age, race, and gender were not good predictors of future link formation in these networks. Moreover, we examined the social causes and policy implications for the significance and insignificance of various facilitators including common jails on future co-offending. These findings provided important insights into the link-formation processes and the resilience of social networks.

Spatio-temporal Analysis and Visualization

Event visualization holds the promise of alleviating information overload in human analysis and numerous tools and techniques have been developed and evaluated in this domain. However, previous work does not specifically address either the coordination of event dimensions with the types of tasks involved or the way that visualizing different event dimensions can benefit human analysis. Research projects [Chang et al., 2008; Chen et al., 2005; Chung et al., 2005] in the University of Arizona focused on designing and implementing effective visualization tools for spatio-temporal events and proposing innovative algorithms for the identification of event hot-spots to aid in containment efforts.
Evaluating Event Visualization: A Usability Study of the Spatio-Temporal Visualizer

In this study [Chung et al., 2005], we propose a taxonomy of event visualization and present a methodology for evaluating a coordinated event visualization tool called the Spatio-Temporal Visualizer (STV), which was developed at the AI Lab. The STV tool (Figure 3) provides an integrated visualization environment that includes GIS, periodic, and timeline views along with a time slider to allow simultaneous examination of the same data. The GIS view displays a map of Tucson to help locate geographic clusters of crime incidents. The periodic pattern view provides aggregated information of a collection of incidents. A circular chart is used to display how many incidents occurred over a specified period in a selected time unit. The timeline view on a chart shows crime incidents as square boxes arranged in chronological order, with groups of incidents displayed in a hierarchy on the left of the chart, the time slider adjusts the temporal viewing window [Chung et al., 2005]. The evaluation methodology examines different event dimensions and different task types, thus juxtaposing two important elements of evaluating a tool. To achieve both internal and external validity, a laboratory experiment with students and a field study with crime analysis experts were conducted. Findings showed that STV could support crime analysis involving multiple, coordinated event dimensions as effectively as it could analyze individual, uncoordinated event dimensions. STV performed better using various metrics (see Chung et al., 2005, for details) as compared to Microsoft Excel (which was the tool primarily used by crime analysts to perform their tasks). User comments also showed STV to be intuitive, useful, and preferable to existing crime analysis methods.

A Stack-based Prospective Spatio-temporal Data Analysis Approach

In this study [Chang et al., 2008], we developed a new spatio-temporal data analysis approach aimed at discovering abnormal spatio-temporal clustering patterns. We also proposed a quantitative evaluation framework and compared our approach against a widely used space–time scan statistic-based method. Our approach is based on a robust clustering engine using support vector machines and incorporates ideas from existing online surveillance methods to track incremental changes over time. Experimental results using both simulated and crime datasets indicate that our approach is able to detect abnormal areas with irregular shapes more accurately than the space–time scan statistic-based method implemented in the widely used SaTScan tool. This innovative algorithm can be used in the STV tool to aid analysts identify emerging hotspots in crime, disease, and other spatio-temporal data.
Identity Resolution
Identity deception is a serious problem encountered in the law-enforcement and intelligence communities. Most of the existing techniques are experimental and cannot be easily applied to real applications because of problems such as missing values and large data size. In this stream of research, we developed techniques that can automatically detect identity deception.

A Study of Criminal Identity Deception and Deception Detection in Law Enforcement
This study [Wang et al., 2004b] focused on uncovering patterns of criminal identity deception observed through a case study performed at TPD. We defined criminal identity deception based on an understanding of the various theories of deception and interviewed a police detective to refine the definition of criminal identity deception. A novel taxonomy for criminal identity deception was built to represent the various patterns identified in the case study. This taxonomy formed the basis for the development of deception detection models.

Automatically Detecting Criminal Identity Deception: An Adaptive Detection Algorithm
In this study [Wang et al., 2006], we proposed an adaptive detection algorithm that adapts well to incomplete identities with missing values and to large datasets containing millions of records. We described three experiments to show that the algorithm is significantly more efficient than the existing record comparison algorithm with little loss in accuracy. A study conducted at TPD showed that our algorithm was useful in detecting both intentional deception and unintentional data errors. Figure 4 shows a snapshot of the ID resolution tool developed as part of this research. The tool can be used to determine similar identity clusters in databases and provide an interface to merge them. A comparison of the tool using compression ratio, precision, and recall metrics to the industry leading IBM Identity Resolution Software showed that the Arizona ID Matcher obtained better recalls and overall F-measure [Wang et al., 2007].

Figure 4. A Snapshot of the Arizona ID Resolution Tool

Association Analysis
In recent years, border safety has been identified as a critical part of homeland security. Customs and Border Protection (CBP) agents search vehicles entering the country for drugs and other contraband. This process is time consuming, and long waiting times impair the flow of people, vehicles, and commerce. So agents at the border are under pressure to balance security needs with operational efficiency.

CBP agents believe that vehicles involved in smuggling operate and cross the Southern border in groups. When one vehicle approaches the checkpoint, the others wait and join the line only if the vehicle crosses into the U.S. successfully. This ensures that the others can turn back into Mexico if the vehicle before them is inspected and caught. So, if the criminal links of one or more vehicles in a group are known, the group’s crossing patterns may be used to identify other partner vehicles. Law-enforcement data can be used as a good anchor to perform such
User acceptance is vital to the success of COPLINK in law enforcement agencies. It is a critical dependent variable of Information Systems success [DeLone et al., 1992] and has been shown to be a common challenge impeding the proliferation of an advanced Information System in various organizational settings, including law enforcement [Lin et al., 2004]. Law enforcement officers need constant, timely information access and effective knowledge support for crime fighting and investigation activities. However, the deployment of COPLINK by an agency cannot guarantee its
use by the targeted officers. Therefore, our technology evaluation included the examination of key factors influencing individual officers’ technology acceptance decisions. We conducted two empirical studies to investigate law enforcement officers’ acceptance of COPLINK. Both studies addressed the following questions:

- What are the important factors influencing a law enforcement officer’s acceptance of COPLINK or COPLINK Mobile?
- What are the inter-relationships between these determinants that explain an officer’s decision on whether to use COPLINK or COPLINK Mobile?

We took a theory-based approach by developing research models for explaining officers’ technology acceptance on the basis of the theory of planned behavior [Ajzen, 1991], the theory of innovation diffusion [Rogers, 1983], technology acceptance model [Davis, 1989], and the unified theory of acceptance and use of technology [Venkatesh et al., 2003]. As depicted in Figure 1, we anchored appropriate theoretical premises to develop factor models explaining or predicting officers’ technology acceptance decisions. We then empirically tested each model and the associated hypotheses by conducting a survey study involving the targeted officers in their work context. Our investigations responded to the need for further examining user technology acceptance in various professional work settings [Chau et al., 2001]. In the next section, we detail each study in terms of research model, measurements, instrument revalidation, data collection, and key results.

Examining Officers’ Acceptance of COPLINK
Research Model
We posit that an officer’s decision to accept or not accept COPLINK can be jointly explained by important characteristics of the technology, the targeted user group, and the adopting agency context, congruent with the suggestion by Chau and Hu [2001]. According to our model (shown in Figure 6), officers’ acceptance of COPLINK is directly determined by their attitudes toward the technology, perceptions of its usefulness and availability, as well as the agency’s subjective norm toward using the technology. In general, availability refers to an officer’s perception of the availability of the computing equipment necessary for accessing and using COPLINK, and subjective norm is an officer’s assessment or perception of significant referents’ opinions regarding his or her use of COPLINK [Taylor et al., 1995]. Perceived usefulness refers to the extent to which an officer considers COPLINK to be useful for his or her job tasks [Davis et al., 1989]. This perception is influenced by efficiency gain, perceived ease of use, and subjective norm. The efficiency gain refers to the degree to which an officer perceives that his or her task performance efficiency can be improved through the use of COPLINK [Davis, 1989]. The perceived ease of use refers to an officer’s perception of his or her use of COPLINK to be free of effort [Davis et al., 1989]. Attitude also affects user acceptance and refers to an officer’s positive or negative attitudinal beliefs about using COPLINK [Taylor et al., 1995]. According to our model, attitude is influenced by perceived usefulness and perceived ease of use.

![Figure 6. A Factor Model for Explaining Officers’ Acceptance of COPLINK](image-url)
Measurements

We used behavioral intention to measure officers’ acceptance of COPLINK, a dependent variable choice that is theoretically justifiable and has ample empirical support. To measure the constructs in our model, we adapted items from previously validated scales and used a focus group, consisting of law enforcement officers, to verify them at face value. Using their feedback and suggestions, we selected items appropriate for our intended investigation from each scale and made several wording changes appropriate for the targeted law enforcement context. Specifically, we measured perceived usefulness, perceived ease of use, and intention using the respective items adapted from Venkatesh and Davis [1996]. We operationalized attitude and availability with items from Taylor and Todd [1995]. We measured efficiency gain using items from Davis [1989]. All the measurement items employed a 7-point Likert scale ranging from “strongly disagree” to “strongly agree.” To reduce potential ceiling or floor effects that may induce monotonous responses from subjects, we randomized the items in the questionnaire and negated half of the questions.

Instrument Revalidation

We conducted a pilot study to revalidate our instrument [Straub, 1989], involving a total of forty-two officers who had completed the mandatory user training program but had not participated in the development of COPLINK or our survey instrument development. We used their evaluative responses to reexamine the instrument’s reliability as well as its convergent and discriminant validity. According to our results, the instrument exhibited satisfactory reliability; i.e., the Cronbach’s alpha value for the investigated constructs ranged from .73 to .93, exceeding the common .7 threshold [Nunnally, 1978]. To examine the convergent and discriminant validity, we performed a confirmatory factor analysis using a varimax with Kaiser normalization rotation. Our results show that items measuring the same construct have loadings significantly higher than those measuring a different construct. In addition, each component extracted has an eigenvalue greater than 1.0. Taken together, our pilot results show that the instrument exhibits satisfactory convergent and discriminant validity.

Data Collection

Our study agency is a police department of a medium-sized city located in southwestern United States. With the assistance of the administrators, including assistant chiefs and captains, we distributed the survey to 481 officers who had completed the mandatory user training program three months prior. Each officer had two weeks to complete and return the survey. Officers who had not responded in the normal response window were contacted and offered another two weeks to do so. A final one-week window was then offered to officers who had not returned the survey during the extended response window. We received a total of 283 completed surveys, representing a 68.9 percent effective response rate. Most of our subjects were from the Investigative and Field Operations divisions, the two largest in the study agency. The gender distribution of the respondents shows an approximate 4-to-1 ratio, in favor of males. Most of the responding officers had a two-year college degree (41 percent), followed by those with a high school diploma (30 percent) and those holding a bachelor’s degree (29 percent). As a group, our subjects were 38.4 years of age and had accumulated 12.1 years of experience in law enforcement services. Our analysis shows no significant differences between the early and late respondents (i.e., those returning the survey in the normal two-week response window versus those who needed additional time to do so) in home division, gender distribution, education background, age, or prior experience in law enforcement. Thus, the threat of nonrespondent bias was reduced.

Results

We took a structural equation modeling approach to test our model and hypotheses (i.e., causal paths). Specifically, we used LISREL to examine the measurement model and the structural model. Overall, our data show a good fit to the research model, as suggested by common model fit indexes exceeding the respective thresholds; e.g., Comparative Fit Index being .91, Non-norm Fit Index being .89, Standardized Root Mean Square Residual being 0.06. Our model exhibits satisfactory explanatory power, accounting for 58 percent of the variance in officers’ intention to use COPLINK, 66 percent of the variance in their attitudes toward COPLINK, and 60 percent of the variance in perceived usefulness. Judged by both the statistical significance and the magnitude of influence, perceived usefulness seems to be the most important determinant of user acceptance and is significantly influenced by efficiency gain and subjective norm. This finding suggests a pragmatic orientation in officers’ technology acceptance decision-making.

Examining Officers’ Acceptance of COPLINK Mobile

COPLINK Mobile provides field officers with a core set of COPLINK functionalities through a lightweight, handheld device or mobile applications running on a limited bandwidth. It represents a crucial extension of COPLINK because law enforcement activities are becoming increasingly networked and mobile. COPLINK Mobile can be implemented on PDAs, mobile phones, or tablet devices, thereby providing field officers with anytime, anywhere access to core COPLINK applications with a small screen space. With COPLINK Mobile, a field officer can use general packet radio
service (GPRS) communications to query and access, through a secured channel and in a nearly real-time fashion, important information regarding specific suspects, locations, weapons, vehicles, and/or crime events, with few spatial or temporal constraints.

**Research Model and Measurements**
In order to make meaningful comparison with prior study examining officers’ acceptance of COPLINK, we adapted the same research model and empirically tested this model by conducting a survey study for explaining field officers’ voluntary use of COPLINK Mobile. We also used the same measurements and employed a 7-point Likert scale ranging from "strongly disagree" to "strongly agree."

**Data Collection**
We targeted all the officers (153 in total) who had received a custom-made hand-held device for using COPLINK Mobile and completed the mandatory user training program. We obtained important demographic data from each officer during the training program and distributed the survey three months after the training, with assistance of their administrators. All officers had two weeks to complete and return the survey and those who did not return the survey had two additional weeks to do so. A total of forty-two officers returned the survey; one of which was partially completed and thus was removed from our subsequent data analyses. As a result, our sample consists of forty-one responses, showing an effective response rate of 26.8 percent. Our sample is representative of the overall field officers, as suggested by the insignificant between-group differences in age, number of years in law-enforcement services, and computer competence; p-values all greater than 0.10. We further examined the non-respondent bias by comparing officers returning their completed surveys in the normal two-week response window versus those needing additional time to do so. Both groups are highly comparable in age, number of years in law-enforcement services, and computer competence (p-value > 0.10). Together, our results suggest that the non-respondent bias does not appear to be a serious problem in the study.

**Instrument Revalidation**
We first examined our instrument in terms of reliability and convergent and discriminant validity. For item reliability, we examined the loading of each item on its corresponding construct. We adopted the suggested cutoff thresholds and observed almost all the measurement items showing a factor loading higher than 0.7, a threshold suggested by Nunnally [1978]. The loadings of all the remaining items are statistically significant at the 0.001 level. We assessed our instrument’s construct reliability in terms of internal consistency and composite construct reliability. We used Cronbach’s alpha to examine internal consistency and adopted the common threshold of 0.7 [Nunnally, 1978] and our results showed that each construct exhibited a Cronbach’s alpha value greater than 0.7, thus showing satisfactory internal consistency. The composite reliability of each construct also exceeded 0.7, a threshold commonly used in social science research to signify satisfactory construct reliability [Fornell et al., 1981]. Overall, our results suggest the instrument exhibited satisfactory construct reliability. We then examined convergent validity, using average variance extracted (AVE) that denotes the variance captured by the indicators. Each construct had an AVE value greater than 0.5, suggesting our instrument's adequate convergent validity [Fornell et al., 1981]. We further examined the convergent and discriminant validity of our instrument, on the basis of the cross-loadings computed from the correlation between each construct's component score and the manifest indicators of other constructs [Chin, 1998]. According to our results, all items load substantially higher on their own construct than on any other constructs. Furthermore, the square roots of the AVEs were also greater than the correlation among any pair of latent constructs [Chin, 1998]. Taken together, our results suggest the instrument had appropriate convergent and discriminant validity.

**Results**
We tested our model and the associated hypotheses using partial least squares (PLS), which supports factor analysis with linear regressions and requires minimal distribution assumptions [Gefen et al., 2000]. PLS supports both measurement and structural model testing and is advantageous over other alternative data analysis techniques (e.g., LISREL) because our sample is relatively small and the data distribution may not confirm multivariate normality. We tested our research model by examining the variance of each non-endogenous variable; i.e., $R^2$ value. Our model explains a significant portion of the variance in perceived usefulness (77 percent), attitude (71 percent), and behavioral intention (71 percent). Collectively, the model offers significant utility for explaining field officers' acceptance of COPLNK Mobile. Compared to previous study that examines the officers’ intention to use COPLINK, this model explains more variances in the officers’ intention to use COPLINK Mobile, their attitudes toward COPLINK Mobile, and perceived usefulness. Perceived usefulness appears to be a single most critical acceptance determinant and is directly influenced by efficacy gain and subjective norm, judged by the statistical significance and effect magnitude.
Summary of User Acceptance Studies

The results of our studies show that each research model is capable of explaining officers’ decision on whether to accept COPLINK or COPLINK Mobile. We make several important observations from our study results. First, in both of our studies, a consistent, prominent core influence path from efficiency gain to perceived usefulness and then to user acceptance measured by intention. In turn, this suggests a tendency that law enforcement officers anchor their technology acceptance decisions from a utilitarian perspective. Officers are not likely to use a technology offering marginal utilities to job tasks just because it is easy to use or people important to them (e.g., commanders, supervisors, colleagues) are in favor of the technology. Second, availability is also important; providing sufficient equipment necessary for accessing and using COPLINK or COPLINK Mobile can affect officers’ acceptance and actual use of the technology. Third, perceived ease of use contributes to officers’ use of COPLINK, perhaps through its influence on perceived usefulness; however, the perceived ease of use does not seem to have significant effects on officers’ perceptions about COPLINK Mobile’s usefulness or their attitudes. In our context, the influence of perceived ease of use on an officer’s decision whether or not to use COPLINK or COPLINK Mobile seems limited. In addition, law enforcement officers appear to place a relatively small weight on the opinions of their important referents regarding their use of COPLINK or COPLINK Mobile. According to our findings, subjective norm influence officers’ perceptions of a technology’s usefulness, but does not affect their acceptance decisions directly.

As Chau and Hu [2001] noted, people are likely to exhibit subtle differences in their technology acceptance decision making, partially because of their professional context. For example, specialized personnel require specialized training, are accustomed to professional work arrangements, and normally have considerable autonomy in their work practices [Chau et al., 2001]; this implies that law enforcement officers may consider different factors when deciding whether to use a new technology. Our results reveal that compared to prior literature in technology acceptance by ordinary users, police officers exhibit different considerations while making the technology acceptance decisions. According to the reported surveys, neither the perceived ease of use, nor the subjective norm affects the officers’ adoption decision. The most important determinants of the officers’ adoption decision seem to be whether the use of COPLINK or COPLINK Mobile can help to improve job performance in their routine tasks and operations in the field. Our findings, therefore, provide important insights into user acceptance of an advanced technology in mandatory settings and highlight the utilitarian focus of the officers’ decision making.

V. CONCLUSIONS AND LESSONS LEARNED

Algorithms, methodologies, techniques, and the resulting systems developed as part of the COPLINK project have been used successfully within law-enforcement agencies at the local, state, and national levels for several years. Each research project was informed by theories, methodologies, and instruments from IS and its various reference disciplines and has contributed back to the knowledge bases by developing new design techniques, data analysis methodologies, and evaluation methods. Various studies within the projects have aimed to understand the enforcement issues.

As suggested by all DSR researchers [Hevner et al., 2004; Nunamaker et al., 1991], DSR should focus on relevant real-world problems. Our relationship with law-enforcement agencies played a very important role in the identification of problems to be solved. Right from the inception of the project, the research team identified problems based on input from crime analysts and officers at the Tucson Police Department, Customs, and Border Protection, and several other agencies. Most of these problems were time-sensitive in nature and required workable solutions applicable in government agencies. The incentive to produce working systems along with the added satisfaction of aiding law-enforcement encouraged high quality research.

- **Relevant problems**—As suggested by all DSR researchers [Hevner et al., 2004; Nunamaker et al., 1991], DSR should focus on relevant real-world problems. Our relationship with law-enforcement agencies played a very important role in the identification of problems to be solved. Right from the inception of the project, the research team identified problems based on input from crime analysts and officers at the Tucson Police Department, Customs, and Border Protection, and several other agencies. Most of these problems were time-sensitive in nature and required workable solutions applicable in government agencies. The incentive to produce working systems along with the added satisfaction of aiding law-enforcement encouraged high quality research.

- **Application of established theory to new areas**—Hevner et al. [2004] mentioned that success in a DSR project is predicated by the researcher’s skilled selection of appropriate techniques to develop an artifact and the selection of appropriate means to evaluate it. We placed the problems faced by government agencies in the context of theoretical research in the MIS and related disciplines. This not only allowed us to publish research and enhance theories, but also produced creative solutions by drawing on established ideas never applied to this problem domain. Various related disciplines (like sociology and computer science) were searched for techniques like social network analysis and record linkage, both of which had clear applications to law-enforcement issues.

- **Communication of research in academia, industry, and media**—We presented our research in various conferences and published in many tier-1 design science oriented journals. This provided credibility to the
project not just in academia, but also in the eyes of funding and law-enforcement agencies. In addition, COPLINK received notable coverage in the media with articles in the New York Times [Sink, 2002] and the Washington Post [O’Harrow et al., 2008], among others. The media coverage and the presence of academic credibility through the University of Arizona further encouraged outside organizations to trust and implement the artifacts produced [Linder, 2007]. This allowed the research team to test and evaluate algorithms and prototype systems on real data and often in the actual IT infrastructure at the agencies. The feedback obtained from these evaluations could be readily incorporated back into the research to produce systems that had a significant amount of end-user buy-in. Thus, the problem diagnosis–technology invention–theory building–technology evaluation loop shown in Figure 1 went a full circle several times during the duration of the project.

- **External funding and commercialization**—We believe that external funding both from national agencies like NSF, DHS, and NIJ and the industry plays an important role in design science research. External funding allowed police agencies to implement the artifacts, provide salaries for personnel to liaison with the university, and support research staff at the university. Since system prototypes were tested and implemented at agencies with success, this allowed us to leverage a variety of agencies (both research and practice oriented) for funding. Even though many artifacts constructed are rarely full-grown Information Systems that are used in practice [Hevner et al., 2004], the initial systems developed within the research project were commercialized using venture capital funding. The systems were rewritten to make the code robust, reliable, and commercial-grade [Linder, 2007]. This led to the research project’s growth as more organizations approached the AI Lab with newer and relevant problems.

The success of the COPLINK project at the University of Arizona has led to other large-scale projects that plan to use the same model to identify problems and conduct high-impact design science research. These projects include the Dark Web project that aims to computationally analyze information systematically obtained from the Internet, the BioPortal project that aims at developing a cross-jurisdictional information sharing and data analysis environment, and the business intelligence projects that include research on opinion mining, analysis of the Web 2.0, and recommender systems among others.

**ACKNOWLEDGMENTS**


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We greatly appreciate the excellent COPLINK team members and officers at TPD and CBP for their time and contributions to this research project.

**REFERENCES**

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