Assessing the Quality of Electronically Communicated Information at Construction Field Level

Payam R. Zekavat, and Leonhard E. Bernold

Abstract—Current construction practice is suffering from a lack of effective communication. Adopting a ubiquitous smart communication system which provides reliable information will assure operation performance. However the quality of a piece of information exchanged through project network needs to be appraised. An IT-based communication interface, the eCKiosk, has been developed and tested with the goal to improve work performance. This paper presents the Bayesian Belief Network (BBN) procedure to capture how users evaluate the information quality provided by eCKiosk. Poor data quality may lead to variety of expensive process wastes such as idleness, rework, injuries and safety violations. Using such cause-and-effect relationships, this paper will show how a trained BBN is capable of measuring quality of e-communication. Also, a practical example of one quality indicators, ease-of-use, is presented. Learning curve theory is applied to properly model the changing efficiency interacting with proposed IT-based tool.

Index Terms—Bayesian belief network, Ease of use, eCKiosk, Information quality assessment, Learning curve

I. INTRODUCTION

Construction is a large and fragmented industry and hence effective coordination among project participants is essential to achieve project objectives. Of course, key to this is information that is collected, processed and made available for use in a practical format. However, because of the unique conditions, construction suffers from a lack of communication with many islands-of-communication. Construction faces many ill-defined, complex problems with little or unclear information. Indeed, divergent frames of reference, values, and goals generate disagreement, ambiguity and uncertainty. Not surprisingly, [1] predicted drastic improvement in productivity, quality and safety through a modified information supply. Adopting a ubiquitous smart communication system which provides reliable information will improve process performance. However the quality of a piece of information communicated through the project network needs to be evaluated. This paper uses concept of information richness explores compatibility of an information fragment with actual expectation of user at the point of activity. Information richness is associated to the quality of exchanged information. Richer information reduces uncertainty and clarifies ambiguity.

An IT-based communication interface namely Electronic Construction Kiosk (eCKiosk) is introduced in this script. Then following the need to assess the quality of communicated information through the eCKiosk, Bayesian Belief Network (BBN) is used as a tool to measure the information quality (InQ). Finally results of assessing “Ease of Use” as one of the quality attributes is shared.

II. MISCOMMUNICATION

Research has shown that miscommunication in construction is not only responsible for high cost but is associated for many accidents [2], quarrels and even physical fights [3]. Lack of communication may result in faulty job, job tension and/or depression. Fig. 1 depicts a model of bidirectional communication between two persons. Shohet and Friedman [4] addressed importance of effective communication for a construction manager (CM) to aim project objectives. The authors investigated communication pattern between CM and other participants by defining some indexes. They indicate that to overcome huge amount of uncertainty in current complex construction environment, major efforts in coordination and communication between different parties are required. Results introduce interactive communication between CM and design team as the principal channel of communication.

III. The eCKiosk Prototype

This paper aims to offer a model for a site-network consisting of a backbone linked to a tier of hubs that interact flexibly with the next tier of data sinks or sources using smart agents. eCKiosk (Fig. 2), the wireless hub, is designed in a way to be able to be moved wherever the crews are working [5]. Ensuring connectivity at all times requires a robust wireless backbone with secure two-way data interfaces. Although it basically works as information delivery tool but it also provides a data collection platform. Indeed the main advantage of eCKiosk is its interactive design. Although the eCKiosk facilitates interaction among e-communication islands, effects of implementing this new tool on quality of communicated information need to be investigated. Since the conventional paper-based information delivery is changing, recognizing the best way of information presentation which best enhances cognition level is also a matter of importance. In proposed site-network, several different sensors collect data from points of interest which after processing will be retrieved through the eCKiosk upon demand. For example, GPS i.e. one of network components provides invaluable information about real time truck mixer’s position and travel time. GPS data in combination with traffic information records provide a reliable decision making support to dispatch trucks from the batch plant and also adjust on-site crew’s productivity. The live positioning information is communicated to the on-site crew through the eCKiosk. In

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another work package, a set of sensors including but not limited to 360° camera (ladybug), wireless sensors like thermocouples, and Kinect, track and record changes in real time. Actually construction process is fully monitored and each building element’s evolution is recorded “as-it-was-built”. Consider the value to the end user when “as-it-was-built” shows the exact place of the rebar in a concrete wall. The other work package toward smart site is a wireless camera attached to the crane hook assisting the crane driver in blind areas. While works as safety reassurance, inserting this new tool into the work process, decrease required crew in unloading tasks and simultaneously increases crane productivity.

FIG. 1. MODEL OF TWO-WAY COMMUNICATION [6]

Fig. 1. Model of Two-Way Communication [6]

Evaluating the quality of provided information is a way to compare advantage of eCKiosk over paper-based information dissemination approach. To do so BBN was identified as an effective tool to discover the conception of construction laborers about information quality. Using captured users’ belief, a piece of information can be ranked as rich or poor in quality.

IV. BAYESIAN BELIEF NETWORK

A Bayesian network $B = \{G, P\}$, by definition, is a directed acyclic graph (DAG) $G$ and joint probability distribution (JPD) $P$ that together satisfy Markov condition [7]. A DAG is structured by two sets namely nodes and arrows. An arrow from parent node $X_i$ to child node $X_j$ represents a statistical dependency between two nodes. The conditional probability distribution (CPD) at each node depends only on its parents. Accordingly, $B$ defines a unique JPD over $n$ nodes $\{X_1, X_2, \ldots, X_n\}$ as (1).

$$P(X_1, X_2, \ldots, X_n) = \prod_{i=1}^{n} P(X_i | \text{Parents}(X_i))$$ (1)

Predictive and diagnostic are two types of inferences to evaluate unknown JPDs based on evidences about connected nodes. In general $B$ is unknown and finding its structure and parameters is referred to as learning problem.

V. INQ ASSESSMENT

Poor data quality leads to several sorts of waste like idleness, rework, injuries and safety violations. Quality of data plays a critical role in waste reduction and hence productivity and quality improvement. Table I summarizes the attitudes which are used to evaluate information quality quantitatively. Mentioned quality characteristics work as diagnostic tools to infer information quality defects like the symptoms that help to determine diseases. Actually the main purpose of this model is to understand the meaning of rich information in construction labor context. Labor as the end user of this model is to understand the meaning of rich information in construction labor context. Labor as the end user of information was targeted to define information quality and its conditional dependence to the attributes. BBN has been used to represent the casual effect of quality on each trait. Built BBN is a tool to evaluate the quality of information delivered to the worker via IT-based solution.

But before that, training data and prior information are needed to learn the parameters of joint probability distribution (JPD) in the BBN. Training data is required to be collected from a real construction project. Observing the site, a worker is asked to fill a simple questionnaire (Fig. 3) whenever retrieving a single piece of information in a job. This is to figure out to what extent the information consumer believed on quality of the information s/he had already used.

Several different ways of information retrieval were recognized such as a) Using printed documents, b) talking via radio, c) measuring (e.g. by measuring tape), d) estimating/watching, e) discussion with a peer, f) asking a colleague, g) remembering, and h) using own experience. As
shown in the questionnaire, some of aforementioned quality features were combined to make it simpler for the respondents to decide and particularly faster to fill in order not to frustrate the process. The BBN will be trained by obtained data which means calculating and introducing the conditional probability tables (CPTs) to the BBN. Fig. 4 shows how a typical BBN looks like after training. In this case a set of random numbers where used to train the BBN. After training, the resulted BBN will be verified by checking whether it can predict the quality level of available information correctly or not. When model is verified, it can be used to measure the quality level of provided information through the eCKiosk. Before implementing BBN to measure quality of IT-based communicated information, one needs to assign each quality attributes level. These estimates will be feed to BBN to determine the probability of having high quality information. Some of quality characteristics like relevance and interpretability are judged qualitatively while some others like ease of use are decided quantitatively. As an example, the quantitative procedure to evaluate ease of use in retrieving a piece of information from eCKiosk is described below.

<table>
<thead>
<tr>
<th>Data Quality Attitude</th>
<th>Definition</th>
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<tbody>
<tr>
<td>Validity</td>
<td>How much compatible with reality</td>
</tr>
<tr>
<td>Completeness</td>
<td>To be Representative of target population</td>
</tr>
<tr>
<td>Accuracy</td>
<td>Errorless</td>
</tr>
<tr>
<td>Consistency</td>
<td>Identical collection procedure for whole data</td>
</tr>
<tr>
<td>Timeliness</td>
<td>Appropriate Frequency of updating</td>
</tr>
<tr>
<td>Accessibility</td>
<td>Ease of use</td>
</tr>
<tr>
<td>Relevance</td>
<td>The extent to which data matches with needs</td>
</tr>
<tr>
<td>Interpretability</td>
<td>Understandable for the user</td>
</tr>
<tr>
<td>Transparency</td>
<td>Traceable acquisition procedure</td>
</tr>
</tbody>
</table>

TABLE I: DEFINITIONS OF DATA QUALITY ATTITUDES

Fig. 3. The information quality questionnaire

VI. EVALUATING “EASE-OF-USE”

Human brain’s response improves by responding to repeated stimuli. Time Reduction with each trail varies with respect to process complexity [8]. A direct correlation can be established between ease of use and this reduction rate. It means that a larger improvement potential predicted by repeating a simple process while a complex experience will allow only small improvement. Various basic mathematical models for learning curves are: 1) The straight-line power model, 2) the Stanford “B” model, 3) the cubic power model, 4) the piecewise (or stepwise) model, and 5) the exponential model. For its simplicity power law function that assumes constant learning rate through the activity duration was used in form (2).

\[ T_{ea} = T_i x_i^{-\alpha} \]  

where: \( T_{ea} \) is cumulative average time per unit, \( T_i \) is first cycle time, \( x_i \) is cycle number, and \( \alpha \) represents a constant that mirrors the complexity of the work;

Learning rate can be calculated using (3). As described, it can be used as a measure to evaluate ease of use. To have a better terminology from now on progress index will be used instead of learning rate.

\[ 2^{-\alpha} = \text{Progress Index (PI)} \]  

Following paragraph, describes the procedure used to evaluate progress index for eCKiosk information retrieval. On the basis of predicted benefits a scenario was designed with the aim of evaluating information accessibility. An examinee was asked to use the eCKiosk to a) retrieve a piece of information, b) communicate to ask for complementary information and to report progress, c) save modification done and d) register his/her gained knowledge/experience in a text format. In the case of system operation assessment, information flow during a two-way slab reinforcement bar (rebar) placement was simulated. Whole project’s designs including tables and drawings were available in central database. Specifications were documented regarding each concrete component feature [9]. For example required information could be found through Concrete Components Æ Floor system Æ Slabs Æ Two-way Æ Individual Slab’s Designs. In each experiment a slab was selected randomly and the user was asked to focus on that slab. In designed scenario the user needs to find required rebar information including rebar size, length, and number of required pieces, placement pattern and specific prefabricated bundle the rebar should be picked from. During the placement the operator suspected a bundle degree of corrosion, which leads to a
multimodal communication to quality manager to get the approval before placement. At the end of placement the user report progress filling related excel tables. All mentioned information plus rebar sequential placement and weather condition was recorded in predefined as-it-was-built format. Finally valuable gained knowledge regarding different rebar placement plans according to features of rebar and concrete components and subcomponents was recorded. As explained each run consisted of five stages: 1) information retrieval, 2) quality approval communication, 3) progress report, 4) creating as-it-was-built and 5) knowledge recording. Time study results are summarized in Table II.

As shown in Fig. 5 all data points fell between 85 and 90% progress index. The best fitted trendline is:

\[ T_{ca} = 562.6x_{ij}^{-0.214} \]  

Using \( a = 0.214 \), progress index equals 86% which describes the percentage of reduction in total time between two cycle numbers which differ by a factor of 2. Calculated progress index assures fast competence achieving in working with the information through the eCKiosk. Oglesby and Parker [10] indicated that common range of progress index for most of construction tasks falls between 70% and 90%.

VII. APPLICATION EXAMPLE

In order to demonstrate BBN capability in evaluating InQ, a set of random numbers was used to train BBN. Then user estimates of quality attitudes were feed into BBN. With random training probabilities and estimated quality attributes, the information provided by eCKiosk—with 97% probability—is rich in quality (Fig. 6).

VIII. SUMMARY AND CONCLUSION

Communication plays a key role in dynamic and large scale projects. However, conventional paper-based information flow is not capable of coping with complex, uncertain and fast track construction processes. As a consequence, there is a need for easy-to-use, reliable and fast means of communication. Based on its capabilities, field-level ICT application in the construction promises an integrated information dissemination system to add extra value to current practice.

This paper discussed how the well known Bayesian Belief Network provides a solid approach to develop a framework to measure the evolving quality of process related information. Using data retrieval from an eCKiosk, the authors demonstrated how to implement it. In this example, quality was considered as the level of user satisfaction in working with information.

Diagnostic inference implementing trained BBN was able to determine the confidence level about the quality of communicated information. To calculated diagnostic inference, some quality attributes were estimated qualitatively while the rest were calculated quantitatively. Indeed learning theory was applied to quantify the “ease of use” of provided information via eCKiosk. A progress index, quantifying the speed of learning how to operate the eCKiosk, was assessed at 86% which suggests an easy application. In summary, the test of the BBN showed that it could provide a solid foundation for assessing the quality of electronically communicated information at construction filed level.

![Learning curve of system use](image)

![Application of BBN to predict quality of information provided by eCKiosk](image)
REFERENCES


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