Implementation of Construction Performance Database Prototype for Curtain Wall Operation in High-Rise Building Construction

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Abstract

Project process planning is implemented based mainly on the experience of engineers, which can lead to critical cost overruns and delays. This suggests a need for precise planning based on reliable data to prevent serious problems in project management. Accurately predicting productivity is one of the most important issues in process planning. In this regard, simulation methodology is considered to be an effective tool for predicting productivity. However, it is limited in terms of applications to actual sites due to modeling and coding difficulties. Although many researchers have developed various methodologies for addressing such difficulties, no study has focused on developing an optimized model for site personnel. We analyze unit cost and productivity by using a simulation technique. In addition, we standardize data by using statistical methods, and develop a database prototype. In this way, we highlight appropriate analytic and predictable methods for the construction field. In addition, the proposed statistical analysis provides objective insights into experimental construction data.

Keywords: curtain wall; productivity; simulation; database

1. Introduction

1.1 Aim and Objectives

Recent years have witnessed an increasing number of high-rise construction projects, and construction processes have become increasingly complicated (Kim and Kim 2010). Curtain wall systems have been widely employed in the construction of high-rise buildings because they are light in weight, and units of them can be produced in a factory. The curtain wall installation process requires precise planning. However, in most construction projects, this planning implementation is based mainly on the experience of site personnel. Appropriate process planning requires the continuous analysis and prediction of construction performance (Lim 2009). Simulation has been recognized as an appropriate methodology for analyzing and predicting construction performance, but is limited in terms of applications to actual sites because of modeling and coding difficulties (Shin et al. 2012). An emerging methodology based on excellent outcomes by simulation application resulting in reliable performance prediction is highly demanded. Many researchers have indicated that database structures would be one of the best solutions to resolve this limitation since a database is capable of providing users with feasible information that was analyzed by specific tools. In this regard, the objective of this study is to develop a prototype database by using simulation to overcome such limitations, and to suggest a more reliable decision-making tool for site personnel. To achieve this objective, we considered several issues: 1) the specific formats of construction simulation implementation, 2) a detailed procedure for finding statistic locations of the selected values, 3) the frameworks of the database structure and a prototype of a screen shot presenting specific categories of output information, and 4) a feasible case showing all procedures and outputs.

1.2 Methodology and Scope

This study was conducted using the following procedures. The curtain wall operation was targeted as the subject of our methodology application, as described in Section 1.1. Then,

1) We performed a literature review of the curtain wall system and fundamental methodologies, such as simulation techniques and various methods for construction productivity predictions.
2) Site visits and observations of activities of curtain wall operations were conducted at three relevant sites.
3) Data collection was achieved through interviews with site engineers, and measurements of cycle times...
and hourly productivity were made.

4) Simulation modeling was performed based on the collected information of the detailed working processes, and on the cycle times for each activity.

5) An analysis of construction performance was performed in terms of unit cost (US Dollars/cycle) and productivity (cycles/hour) through simulation.

6) Data were sorted and stored with information on the number of resources, the values of productivity and costs, and the total times in the database.

7) Data standardization was performed using Z conversion based on statistical methods, with the assumption of a normal distribution for the acquired information in the database.

8) The developed database prototype suggested results (the productivity) of simulation implementation depending on selected scenarios.

2. Literature Review
2.1 High-rise Curtain-wall Operation
Curtain wall systems have been widely used in high-rise construction because of their numerous advantages. For example, curtain wall units are light, and most are manufactured in factories—not at the construction site—thereby reducing the space needed for unit stocking and thus the construction time. We analyzed the procedure for installing a curtain wall system by dividing it into several processes. Table 1. shows these processes, and Fig.1. shows the installation procedure (Kim 2012; Lee et al. 2011).

2.2 Construction Simulation
Simulation involves modeling processes or experiments with actual systems to improve or analyze a given system. Our method of construction simulation consists of collecting data, modeling using CYCLONE (cyclic operation network) software, and conducting and analyzing simulation models (Halpin and Riggs 1992). Various simulation methods have been used for analyzing the productivity of construction operations and predicting construction performance (Kannan 1999; Wang and Halpin 2004; Han et al. 2006; Han and Lee 2008).

Table 1. Curtain Wall Unit System Operation Procedure

<table>
<thead>
<tr>
<th>Event</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>Truck arrival</td>
<td>A winch is positioned in the upper floor to be able to lift units from the installation floor</td>
</tr>
<tr>
<td>Placing unit on a pre-installed unit</td>
<td>After the units are lifted, they are placed on pre-installed units</td>
</tr>
<tr>
<td>Installing sleeve</td>
<td>Sleeve is installed in the upper part by laborers</td>
</tr>
<tr>
<td>Fixing unit by fastener temporarily</td>
<td>The units are fixed temporarily by a fastener to check the level of the unit</td>
</tr>
<tr>
<td>Measuring and adjusting the level of unit</td>
<td>The level of the unit is measured by a laser and tape line and adjusted by the winch</td>
</tr>
<tr>
<td>Fixing unit completely by fastener</td>
<td>The unit is fixed to the slab completely by a fastener and the need unit is ready to be installed</td>
</tr>
</tbody>
</table>

Fig.1. Curtain Wall Unit System Installation Process

CYCLONE, developed by Halpin and Woodhead (1976), is a typical construction simulation tool for repeated processes. A model can be easily developed by considering resources, time, and logistical relationships between tasks (Halpin and Woodhead 1976). In addition, CYCLONE has been used for data generation wherein the raw data are insufficient to function as input data for other prediction methods such as linear regression and artificial neural networks (Han and Halpin 2005; Han et al. 2008; Lee et al. 2011). However, previous studies have highlighted the limitations of CYCLONE in terms of practical applications because of its implementation difficulties.

2.3 Productivity Assessment and Prediction
Construction productivity is considered to be a fundamental factor in determining if the targeted operations are well achieved. Many studies have
suggested various methods for productivity assessments (Adeli 2001; Han and Lee 2008; Han 2010). Also, many studies have been performed to examine productivity using simulation techniques, statistic tools, and artificial neural networks (Adeli 2001; Graham and Smith 2004; Lowe et al. 2006; Ali et al. 2007; Han et al. 2008; Han 2010). Several studies have considered models for overcoming the problem of data shortages for decision-making tool applications. Multiple regression analysis has been widely used for performance predictions in construction. Artificial neural networks alone, or combined with fuzzy approaches, have been used for productivity prediction (Carpenter et al. 1991; Adeli 2001; Lu et al. 2006). Several studies have extended the scope further by considering data generation in order to acquire more feasible results (Han et al. 2008). All of these efforts indicate the limitation of practical applications in that users hardly utilize these models with no professional knowledge of various tools such as statistics, artificial intelligence and so on. Accordingly, the system of a database where the results via these kinds of tools are systemically controlled can provide users with an efficient way to select feasible performance data prior to actual operations.

3. Data Collection and Analysis

3.1 Site Description

Our data were collected from three construction sites employing curtain wall systems. Tables 2.-4. describe these sites.

3.2 Simulation Modeling and Productivity Analysis

Construction productivity was analyzed using the simulation models shown in Figs.2.-4. Fig.2. shows a simulation model of the unloading process, to which this model was applied (Lee et al. 2011). The results indicate that preparation and truck standby times have the greatest impact on productivity, which implies that reducing the truck standby time can improve productivity. We conducted a simulation study by considering various combinations of construction equipment. Table 5. shows the productivity analyzed using the simulation model. In this study, the unloading process was simulated with one to five trucks and forklifts.

Sites A and B use hoists to lift curtain wall units, and Site C uses a crane. Therefore, two simulation models were developed and used as shown in Figs.3. and 4. The model shown in Fig.3. was applied to Sites A and B. The model shown in Fig.4. was applied to Site C. The moving time is the amount of time required for workers to move a unit to the lift area. The results indicate that the moving time has the greatest impact on productivity, thereby implying a need for allocating an appropriate space for stacking units as close as possible to the lift area. The lifting time varies according to the floor. The lifting process was simulated by using one to two hoists and cranes with one to two groups of four workers. Table 6. shows the productivity analyzed using the simulation model.
models based on real process data were developed, and overall data collected reflect 58 cycles: 10 cycles of winches, and two groups of one to five workers. The unit installation process was simulated using one to five productivity analyzed using the simulation model. The level is measured in the up part. Table 7. shows the in the down part, while the fastener is being fixed and depends mainly on the communication between up and down parts. The results of the simulation indicate that the productivity of each process was analyzed. Various combinations of equipment were simulated. The total number of datasets obtained from simulation is 75 for the unloading process, 192 for the lifting process, and 375 for the installation process.

Fig. 5. shows our simulation model of the installation process. The results of the simulation indicate that preparation and fastener-fixing times have the greatest impact on productivity. The installation process depends mainly on the communication between up and down parts. A unit is prepared for its installation in the down part, while the fastener is being fixed and the level is measured in the up part. Table 7. shows the productivity analyzed using the simulation model. The unit installation process was simulated using one to five winches, and two groups of one to five workers. The overall data collected reflect 58 cycles: 10 cycles of the unloading process, 15 cycles of the lifting process, and 33 cycles of the installation process. Simulation models based on real process data were developed, and

Table 5. Unloading Process Productivity Analysis

<table>
<thead>
<tr>
<th>Site</th>
<th>Truck</th>
<th>Forklift</th>
<th>Productivity [cycles/hour]</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>1</td>
<td>1</td>
<td>10.39</td>
</tr>
<tr>
<td>B</td>
<td>1</td>
<td>1</td>
<td>5.14</td>
</tr>
<tr>
<td>C</td>
<td>1</td>
<td>1</td>
<td>5.03</td>
</tr>
<tr>
<td>A</td>
<td>2</td>
<td>2</td>
<td>15.91</td>
</tr>
<tr>
<td>B</td>
<td>2</td>
<td>2</td>
<td>6.56</td>
</tr>
<tr>
<td>C</td>
<td>2</td>
<td>2</td>
<td>6.66</td>
</tr>
</tbody>
</table>

Table 6. Lifting Process Productivity Analysis

<table>
<thead>
<tr>
<th>Site</th>
<th>Forklift</th>
<th>Hoist/ Crane</th>
<th>Crew 1</th>
<th>Crew 2</th>
<th>Productivity [cycles/hour]</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>4.09</td>
</tr>
<tr>
<td>B</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>5.01</td>
</tr>
<tr>
<td>C</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>3.68</td>
</tr>
<tr>
<td>A</td>
<td>2</td>
<td>2</td>
<td>2</td>
<td>2</td>
<td>9.01</td>
</tr>
<tr>
<td>B</td>
<td>2</td>
<td>2</td>
<td>2</td>
<td>2</td>
<td>10.54</td>
</tr>
<tr>
<td>C</td>
<td>2</td>
<td>2</td>
<td>2</td>
<td>2</td>
<td>7.50</td>
</tr>
</tbody>
</table>

Fig. 4. Lifting Process Simulation Model (Hoist)

Fig. 5. Lifting Process Simulation Model (Crane)

Table 7. Installation Process Productivity Analysis

<table>
<thead>
<tr>
<th>Site</th>
<th>Winch</th>
<th>Up</th>
<th>Down</th>
<th>Productivity [cycles/hour]</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>4.18</td>
</tr>
<tr>
<td>B</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>3.66</td>
</tr>
<tr>
<td>C</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>4.66</td>
</tr>
<tr>
<td>A</td>
<td>2</td>
<td>2</td>
<td>2</td>
<td>8.66</td>
</tr>
<tr>
<td>B</td>
<td>2</td>
<td>2</td>
<td>2</td>
<td>8.02</td>
</tr>
<tr>
<td>C</td>
<td>2</td>
<td>2</td>
<td>2</td>
<td>9.55</td>
</tr>
</tbody>
</table>

3.3 Unit Cost Analysis

The unit cost was analyzed using Eq. (1), which is based on productivity data obtained through simulation. The cost of labor and equipment used in this study was obtained through interviews with site personnel, and from the Cost Estimation Standards in Korea.

\[
\text{Unit Cost} = \frac{L.C + E.C}{P} + M.C \quad (1)
\]

where,
- L.C. = Labor cost [US dollars/hour]
- E.C. = Equipment cost [US dollars/hour]
- P = Productivity [cycles/hour]
- M.C. = Material cost [US dollars/hour]

The unloading, lifting, and installation processes were analyzed separately, and the cost of the curtain wall unit was applied only to the installation process.

Table 8. Unloading Process Unit Cost Analysis

<table>
<thead>
<tr>
<th>Site</th>
<th>Truck</th>
<th>Forklift</th>
<th>Unit cost [USD/cycle]</th>
</tr>
</thead>
<tbody>
<tr>
<td>A</td>
<td>1</td>
<td>1</td>
<td>2.63</td>
</tr>
<tr>
<td>B</td>
<td>1</td>
<td>1</td>
<td>5.31</td>
</tr>
<tr>
<td>C</td>
<td>1</td>
<td>1</td>
<td>5.42</td>
</tr>
<tr>
<td>A</td>
<td>2</td>
<td>2</td>
<td>3.43</td>
</tr>
<tr>
<td>B</td>
<td>2</td>
<td>2</td>
<td>8.32</td>
</tr>
<tr>
<td>C</td>
<td>2</td>
<td>2</td>
<td>8.19</td>
</tr>
</tbody>
</table>
to prevent cost overlaps. The results for the unit costs are shown in Tables 8.-10.. The numbers of workers and equipment units shown in the figures are consistent with the results of the productivity analysis.

3.4 Data Standardization

The normal distribution, which is one of the most important distributions in statistics, has numerous mathematical properties, and the results of repeated experiments demonstrate a normal distribution (Ko 1998). In construction, the strength of reinforced bars follows a normal distribution. Equation (2) shows the probability density function of the normal distribution:

\[ f(x; \mu, \sigma) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \]  

where,
- \( \mu \) = Mean
- \( \sigma \) = Standard deviation
- \( \sigma > 0 \) and \( \lim_{\sigma \to 0} \mu < \infty \)

A distribution with a mean of \( \mu \) and a variance of \( \sigma^2 \) is referred to as a normal distribution, and is expressed as X-N(\( \mu, \sigma^2 \)). Here, \( f(x; \mu, \sigma) \) is a normal distribution only if the area under the function is 1. In particular, transforming the distribution into \( Z \) produces \( \mu = 0 \) and \( \sigma = 1 \), which is referred to as the standard normal distribution (Kim and Yoon 2004; Ko 1998). In this study, data on construction performance were standardized using Eq. (3) to allow for a comparison between the unit cost and productivity. In addition, this standardization produced information on ranges of specific data in the distribution. Tables 11.-16. show the standardized data, and Fig.6. provides a sample range for the distribution.

\[ Z = \frac{x-\mu}{\sigma} \]  

4. Database Prototype Development

Unit costs and productivity were analyzed for the three construction sites. The unit cost, labor, equipment, and productivity are linked together as one dataset collected from each site. Fig.7. shows the
proposed database prototype, which was developed as follows (Lee et al. 2012):
1) We focused on curtain wall systems, but this approach is expected to be applicable to other systems. Therefore, in further studies, this option can be changed to the targeted operation which users are interested in.
2) The operation chosen in option 1 is divided into the main processes. In the case of curtain wall systems, option 2 is divided into the unloading, lifting, and installation processes.
3) Option 3 allows the user to choose the number of datasets to be shown in the database. For example, if the user selects 30, the database will show only the 30 datasets.
4) Construction data vary even for the same site and operation. Therefore, in option 4, the user can choose the order of datasets based on the lowest unit cost, the minimum use of resources, and the highest productivity. The use of option 4 enables the user to obtain information on current conditions of the site.
5) Option 5 shows the site name.
6) Option 6 shows the resources.
7) Option 7 shows the productivity expected by using the resources chosen in option 6.
8) Option 8 shows the cycle time expected by using the resources chosen in option 6.
9) Option 9 shows the unit cost expected by using the resources chosen in option 6.
10) Option 10 shows the Z score for the unit cost and productivity.
11) Option 11 shows the distribution of data.
11-1) Option 11-1 shows the overall distribution of productivity.
11-2) Option 11-2 shows the overall distribution of unit cost.
11-3) Option 11-3 shows ranges of productivity for selected data.
11-4) Option 11-4 shows ranges of the unit cost for selected data.

5. Feasible Case
We considered a feasible case with unit cost and productivity predictions based on the proposed methodology. The following assumptions were made:
Unloading process:
1) Four trucks are used.
2) Three forklifts are used.
Lifting process:
1) A hoist is used to lift curtain wall units.
2) Two hoists are used.
3) Three workers are responsible for lifting curtain wall units.
4) Two workers are responsible for receiving curtain wall units on the installation floor.
Installation process:
1) Two winches are used.
2) Two workers are responsible for the up part.
3) Two workers are responsible for the down part.

Table 17 shows data on the unit cost and productivity that can be obtained by the user. Table 18 shows the ideal data (i.e., data for the highest productivity and the lowest unit cost) that can be obtained by the user. In addition, the database prototype shows the overall distribution and ranges of the data. The user can consider site-specific situations and make appropriate decisions based on reliable data. Fig.8 shows the ranges of the unit costs and productivity of the installation process shown in Table 18. The left side of Fig.8 shows the productivity, and the right side shows the unit cost. The user can see that
the ideal range is 55% to 60% for productivity and 10% to 15% for unit cost. For example, in this case, if the construction site is experiencing a delay, then the user can determine that the low range is 10% to 15% and increase it to enhance construction productivity.

Table 17. Productivity and Unit Cost Prediction Example

<table>
<thead>
<tr>
<th>Process</th>
<th>Equipment</th>
<th>No.</th>
<th>Productivity [cycles/hour]</th>
<th>Unit cost [USD/cycle]</th>
</tr>
</thead>
<tbody>
<tr>
<td>Unloading</td>
<td>Truck</td>
<td>4</td>
<td>16.13</td>
<td>5.07</td>
</tr>
<tr>
<td></td>
<td>Forklift</td>
<td>3</td>
<td>6.71</td>
<td>12.19</td>
</tr>
<tr>
<td>Lifting</td>
<td>Hoist</td>
<td>2</td>
<td>12.55</td>
<td>11.05</td>
</tr>
<tr>
<td></td>
<td>Up</td>
<td>2</td>
<td>14.87</td>
<td>9.33</td>
</tr>
<tr>
<td>Installation</td>
<td>Winch</td>
<td>2</td>
<td>8.66</td>
<td>1,377.59</td>
</tr>
<tr>
<td></td>
<td>Up</td>
<td>2</td>
<td>8.02</td>
<td>1,378.72</td>
</tr>
<tr>
<td></td>
<td>Down</td>
<td>2</td>
<td>9.55</td>
<td>1,376.29</td>
</tr>
</tbody>
</table>

![Fig.8. Rank of Productivity and Unit Cost](image)

Table 18. Ideal Results

<table>
<thead>
<tr>
<th>Process</th>
<th>Equipment</th>
<th>No.</th>
<th>Highest productivity [cycles/hour]</th>
<th>Lowest cost [USD/cycle]</th>
</tr>
</thead>
<tbody>
<tr>
<td>Unloading</td>
<td>Truck</td>
<td>4</td>
<td>16.13</td>
<td>5.07</td>
</tr>
<tr>
<td></td>
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<td>Up</td>
<td>2</td>
<td>14.87</td>
<td>9.33</td>
</tr>
<tr>
<td>Installation</td>
<td>Winch</td>
<td>2</td>
<td>9.55</td>
<td>1,376.29</td>
</tr>
</tbody>
</table>

6. Conclusion

6.1 Summary

Our study was motivated by a lack of feasible systems or methodologies for predicting the operational performance of the curtain wall operation, which represents a critical activity in the construction of high-rise buildings. We propose a practical system composed of simulation techniques and database systems. We developed a prototype for testing and investigating the feasibility and availability of the proposed system. Data were collected from interviews with engineers and records of curtain wall operations used in high-rise construction. Simulation models were developed based on the collected data. In addition, the productivity of various equipment combinations was predicted, and the unit cost was analyzed using productivity data. The data are standardized to minimize statistical variance. Based on the statements above, this database can function as a kind of decision-making tool which is used for the establishment of a construction operational plan prior to actual construction operations.

6.2 Significance and Contributions

We propose a methodology that can facilitate appropriate decision making for the unloading, lifting, and installing processes in a curtain wall installation based on productivity and unit cost data. Our results suggest that the proposed approach can make important theoretical and practical contributions. First, methodologies for implementing database systems, such as data generation through simulation techniques and data sorting through specific ranges of unit cost and productivity, can be beneficial to researchers interested in analytic and predictive construction methods. Second, our statistical analysis can provide researchers with objective insights into experimental construction data. Finally, practitioners can use our results for more reliable operational planning, which has generally been based only on the experience of senior personnel. In this regard, this study contributes to process management decision-making based on both the experience of engineers and on objective data.

6.3 Limitations and Future Works

For applications of the proposed methodologies and resulting systems to practical situations, large amounts of raw data collected from construction sites are required. Future researchers should consider practical applications of the proposed database system and investigate the proposed methodology by using updated data. In addition, given this study’s limitations, future researchers should consider a wider range of methods for collecting data and acquiring statistical knowledge, including the Bayesian model.

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