Analysing Interdependencies of Complex Engineering Systems using a Digital Twin-Driven Design Structure Matrix

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ABSTRACT

As engineering design process becomes increasingly complex and interconnected, multidisciplinary teams need to work together, integrating diverse expertise across disciplinary models. Design team often finds it difficult to handle the complexity and interdependencies in both engineering systems and collaborative design process. Previous efforts have been made to address this problem through tracking the system interdependencies using Design Structure Matrix (DSM). However, it is still challenging to deal with complex systems with large data sets and multiple disciplines involved. This paper proposes an effective way to simplify complexity and clarify interdependencies in complex systems through generating DSM from a digital twin (e.g., building information modelling). It firstly defined Elements in the digitally generated DSM (Digital-DSM) drawing on the Uniclass 2015 classification system and prior work, of which three levels were classified as system-of-systems level, project level, and asset level. Interdependencies in Digital-DSM were then defined for different levels of Elements correspondingly based on the geometric and semantic information in Industry Foundation Classes (IFC). Finally, a prototype has been constructed to illustrate how the Digital-DSM helps to improve the efficiency of late design change in complex engineering systems. The results show that the proposed Digital-DSM can initiate a leaner design process through analysing interdependencies and further tracking the propagation of design changes clearly and conveniently.

INTRODUCTION

Through a series of infrastructure projects and programmes including those at Heathrow, London 2012 Olympics, Crossrail, Tideway and High-Speed 2 (HS2), the UK is developing significant expertise in the delivery of major infrastructure projects. (Whyte, 2019; Davies, 2017). To address the challenges of infrastructure delivery, multidisciplinary teams need work together, integrating diverse expertise across disciplines. One challenge that arises is that design teams find it difficult to understand the impacts of late design changes in such complex systems (e.g., infrastructure projects), where changes in one sub-system may have implications for others (Whyte et al., 2016). The interdependencies in such complex engineering systems should be clarified to help designers improve work efficiency in design changes.

While data consistency and quality has traditionally been relatively poor, such complex engineering systems and projects are becoming rich sources of data from

multidisciplinary models and systems (Whyte et al., 2016), raising new opportunities to model and understand the interdependencies in the design of new and operation of existing infrastructure projects.

There have been many efforts in identifying system interdependencies from complex infrastructure projects. For instance, Eusgeld et al. (2011) proposed an ‘system-of-systems’ approach to identify and analyse interdependencies between industrial control systems (ICS). Goldbeck et al. (2019) stepped forward in developing a new, effect-based classification of interdependencies between infrastructure systems. With the increasing complexity of infrastructure projects, design structure matrix (DSM) has been further introduced as a simple, compact and visual representation of a complex system to model system decomposition and interdependencies (Browning, 2001). The DSM has been adopted widely to model and analyse the product decomposition and interdependencies (Pimmler and Eppinger, 1994), to show data interdependencies in multidisciplinary design process (Lambe and Martins2012), and to formulate the interdependencies between drawings of physical components at the lowest level of abstraction in construction design (Senthilkumar and Varghese, 2009). However, current methods for identifying and managing these interdependencies are inadequate for increasingly integrated and cyber-physical complex systems, and do not make full use of such rich sources of data and translate well into a digitally enabled built environment.

The developments of information and communication technologies (ICT), especially building information modelling (BIM) techniques, provide the way to transfer complex systems into the digitally enabled built environment. The Industry Foundation Classes (IFC) as a standard representation of BIM knowledge for interoperability has further opened the possibility to integrate DSM with digital data for more convenient identification of system interdependencies and for more effective design management. Inspired by the interoperability of BIM, Pektas (2010) proposed the complementary integration of IFC process modelling (e.g., IDEF0) and parameter-based DSM in modelling collaborative building design process. Jacob and Varghese (2011) proposed a logic of integration of BIM and DSM to improve construction design process through the definition of new classes ‘IfcDSM’ and ‘IfcConnenctByInformation’. Based on that, Jacob and Varghese (2012) further made efforts to address design process complexity derived from the increased sophistication of product models through integrating the process model (e.g., IFC process model and DSM) with the product model (i.e., IFC product model). However, these efforts were mainly established on the hypothesis of potential IFC extensions (e.g., IfcDSM), and were limited to activity/process-based DSM to identify manually identified system interdependencies and improve construction design process. Gopsill et al. (2016) succeeded in automatically generating component-based DSM to deal with component interdependencies by monitoring the changes in the digital models representing the product. Saoud et al. (2017) proposed the integration of parameter-based DSM with BIM to identify interdependencies between design parameter relationships, supporting and managing the visual representation of predicting design change propagation. Although these efforts have succeeded in identifying system interdependencies between components or design parameters, the work of Gopsill et al. (2016) is inadequate for clarifying and classifying different kinds of interdependencies between

components, and the work of Saoud et al. (2017) is still heavily dependent on manually identified interdependencies between design parameters. There is a need to identify and clarify system interdependencies automatically for complex systems through integrating the digital model (e.g., BIM) and DSM.

This paper aims to develop a novel and effective way to simplify complexity and clarify interdependencies in complex systems through generating DSM from a digital twin (e.g., BIM). It firstly reviewed the past research in identifying system interdependencies using DSM, and briefly introduced the theoretical foundation of this work, including digital twin, IFC and Uniclass classification system. Based on that, the Digital-DSM was defined to clarify the elements and their associated interdependencies across levels (i.e., asset level, project level, and system-of-systems level). Finally, the strategy has been provided to generate Digital-DSM from digital twin at the asset level for example. A simple demonstrating case study has been also conducted to illustrate how the asset level Digital-DSM can be generated and helps to improve the efficiency of late design change via identifying and analysing system interdependencies. This research and results can provide a first step toward next-generation complex engineering systems by demonstrating the feasibility of using a digital twin to generate new insight on systems relationships and interdependencies. The use of digital data and analytics can enable engineering decision-makers to rapidly understand interdependencies within projects and across their boundaries in complex engineering systems with growing complexity.

BACKGROUND

System interdependencies and design structure matrix

The infrastructure projects in the UK is subject to ever increasing loads for population growth, where the addition and retrofit of new/existing infrastructure that is constrained by, and should function harmoniously with, the existing infrastructure (Zhang and Peeta, 2011). Infrastructure is regarded as complex engineering system as it often involves many components that, to a large extent, are open and interconnected, and include assets of different ages. As these become increasingly cyber-physical and interacting, an explicit understanding of their interdependencies for design and delivery should be necessary for the effectiveness and efficiency of the individual system and across systems.

The DSM provides a simple, compact, and visual representation of complex systems and interdependencies, which is effective in addressing decomposition and integration problems, and managing iterative tasks (e.g., design process). A DSM is a square matrix (N^2) with identical row and column labels, where an off-diagonal mark ('cross' in Figure 1) signifies the dependency of one element on another. Reading across a row indicates input sources, while reading down a column reveals output sinks. In Figure 1, element B provides something to or is related by element E, and it depends on something from or relates to element A (Browning, 2001).

Figure 1. Related elements and the representation of interdependencies in DSM

There are four main categories of DSM in applications, including (Browning, 1999, 2001): (1) Component-based or architecture DSM, which is used for modelling system architectures based on components and/or subsystems and their relationships; (2) Team-based or organization DSM, which is used for modelling organization structures based on people and/or groups and their interactions; (3) Activity-based or schedule DSM, which is used for modelling processes and activity networks based on activities and their information flow and other dependencies; and (4) Parameter-based or low-level schedule DSM: which is used for modelling low-level relationship between design decisions and parameters, systems of equations, subroutine parameter exchanges, etc.

The use of DSM in both research and industrial practice increased greatly since the 1990s (Browning, 2001), where it has been widely applied in the construction industry for identifying and analysing interdependencies (as shown in Table 1).

However, in these researches, manual effort is often required to clarify (and/or quantify) these interdependencies in the matrix, and researchers that use the DSM often proceed by reviewing architectural diagrams, system schematics and other relevant documents, or interviewing relevant engineers and experts. It is quite labour intensive and time consuming. For example, the 46² NASA pathfinder DSM took five months (Brady, 2002) and the 84² Xerox printing system DSM needs three months (Suh et al., 2010). The design documents (e.g., process model and drawings) may not be available or accurate during the project delivery process (Wilschut et al., 2018). There is a need to generate DSM in a more effective and convenient way.

Table 1. System Interdependencies in Construction using DSM

DSM Types	Authorship	Elements of DSM	Interdependencies of DSM	Data Collection
Component-based	Senthilkumar and Varghese, 2009	Drawings	Design issues of drawings	Interface meeting
	Gopsill et al., 2016	Digital models (e.g., CAD, FEA ¹)	Component dependencies across models	Monitoring of co-occurrence of model edits
Team-based	Senthilkumar and Varghese, 2009	Design teams	Design interfaces between teams	Workshop
Activity-based	Austin et al., 2000	Design activities	Design information dependencies	Design process model
	Prasad and Jacob, 2018	Design activities	Design information dependencies	-

Parameter-based	Pektaş and Pultar, 2006	Design parameters and requirements	Design information flows	Interviews
	Pektas, 2010	Design parameters	Design information flows	Design process model
	Jacob and Varghese, 2011	Design parameters	Sequence of design parameters	Design process model
	Jacob and Varghese, 2012	Design parameters	Sequence of design parameters	Design process model
	Saoud et al., 2017	Design parameters (geometry, position, specification)	Spatial and analytical dependencies (<i>in, inter, self, join, host, s.host, s.sys</i> ²)	Interviews
Multi-domain matrix	Wilschut et al., 2018	Components, functions, and parameters	Spatial, information, energy and material dependencies	Structured textual function specifications

Note: ¹ CAD: Computer-aided design; FEA: Finite element analysis.

² in: elements belong to; inter: intersection between; self: relation within the same item; join: elements are connected to each other; host: being the host element; s.host: two elements have the same host element; s.sys: elements are defined in one system.

Digital twin and IFC standard

Digital twins in infrastructure are defined as: “realistic digital representations of physical things. They unlock value by enabling improved insights that support better decisions, leading to better outcomes in the physical world.” (Bolton, et al. 2018). Attention is drawn to the use of flows of digital information about physical assets to improve decisions, outcomes and performance. While the shift from computer-aided design to Building Information Modelling (BIM) involved the addition of asset information as well as geometry, the shift to a digital twin requires the use of a broader range of sources of data, which may involve geometries, asset information and associated time series data on processes. Such data may be generated through the activities of professionals, in production and operation, and through a range of sensing devices (including photographs, laser scans, and embedded sensors). Developing and using the rich data implied by ‘digital twin’ is relevant to projects to develop new infrastructure in the context of existing infrastructure systems, and to understand systems interdependencies through integration with DSM.

The IFC standard is deemed a promising open data standard for information exchange between digital twin and DSM. It is defined based on the Standard for Exchange of Product model data (STEP) (ISO 10303) and registered as ISO 16739-1:2018 (2018). The IFC provides a large variety of different geometry representations (e.g., Constructive Solid Geometry, Swept Area Representations and boundary representations), which can be adopted to identify spatial interdependencies. The IfcRelationship is the abstract generalization of all objectified relationships in IFC and can be adopted to handle non-spatial interdependencies among objects (buildingSMART, 2019). There are mainly six types of IfcRelationship, including IfcRelAssigns, IfcRelAssociates, IfcRelConnects, IfcRelDeclares, IfcRelDecomposes, and IfcRelDefines.

Uniclass classification system

Uniclass is a unified classification for the UK construction industry, classifying items of all scale from a facility (e.g., airport) down through to products (e.g., a door in airport terminal building) (Delany, 2019). It provides a hierarchical suite of tables describing and classifying all ‘things’ to support the digital plan of work, including seven core tables (i.e., Complexes, Activities, Entities, Spaces/Location, Elements/Functions, Systems, and Products) and several additional tables (e.g., Form of Information, and Project Management). With the aid of Uniclass, information about complex engineering systems can be generated, used and retrieved throughout the life cycle.

DEVELOPMENT OF DIGITAL-DSM MODEL

This research defined the Digital-DSM, which provided a theoretical basis for generating DSM from digital model automatically. Based on extant literature, two essential building blocks of Digital-DSM were classified and identified, i.e., elements and interdependencies.

Definition of Digital-DSM elements

The elements of Digital-DSM at different scales are defined drawing on the Uniclass classification system and prior work (Pimmler and Eppinger, 1994; Senthilkumar and Varghese, 2009; Saoud et al., 2017). As shown in Table 2, there are four levels of elements defined in a top-down approach, where the definition in Digital-DSM is provided from the perspective of digital model rather than physical model in DSM.

Table 2. Definitions of Elements for Digital-DSM

Level	Terminology	Definition in DSM	Definition in Digital-DSM	Example
System -of- systems Project	Complexes	Assembly of physical systems for high-level functions	Main file defined as assembly of systems for high-level functions	<ul style="list-style-type: none">• Airport system;• Railway system.
	Project	Geographically distinct entity created to meet specific high-level functions for the users of the project.	Sub file defined as digital model created for geographical distinct entity.	<ul style="list-style-type: none">• Terminal building project;• Railway station project.
Asset	Sub-system	Physical entity created to meet specific functions for the project	System defined as digital model created for physical entity with specific functions in each sub file.	<ul style="list-style-type: none">• Architecture, structure, MEP, HVAC system etc. for terminal building;• Architecture, structure, MEP system etc. for railway station;
	Component	Physical entity with specific functions in each sub-system.	Family of objects with a common set of properties, called parameters, and a related graphical representation in each sub-system.	<ul style="list-style-type: none">• Duct, beam, column etc. in building.

	Digital component	-	Specific digital object with unique ID in each family.	<ul style="list-style-type: none"> • #1 duct, #2 duct, #1 beam etc. in building
Parameter	Parameter	Input/Output data to/of the design process	Design parameters as the properties of objects.	<ul style="list-style-type: none"> • Depth of footing, cut-out sizes; • Cut-out sizes and duct elevation, etc.

Definition of Digital-DSM interdependencies

Based on literature review, the interdependencies in Digital-DSM are defined in three types, including the *geospatial* (Borrmann and Rank, 2009), *physical* (Borrmann et al., 2015; buildingSMART, 2019) and *logical* (buildingSMART, 2019).

The *geospatial* interdependencies were defined according to the 9-intersection model in 3D space (Borrmann and Rank, 2009) for spatial analyses, while the *physical* and *logical* interdependencies were defined based on the IfcRelationship for non-spatial analyses. Although these interdependencies are defined across scales, this paper will only focus on the asset level interdependencies to provide an example for the generation of Digital-DSM (Table 3).

Table 3. Definitions of Elements for Digital-DSM at Asset Level

Types	Terminology	Details	Definition
Geospatial	Physical adjacency	Disjoint; Contain; Within; Touch; Overlap	The physical adjacency or topological relationship between each asset.
Physical	Material specification	<i>IfcRelAssociatesMaterial</i>	An objectified relationship between a material definition and elements or element types to which this material definition applies ³
Logical	Group	<i>IfcRelAssignsToGroup</i>	An objectified relationship assigning object definitions (IfcObject or IfcTypeObject) to a group (IfcGroup)
	Resource share	<i>IfcRelAssignsToResource</i>	An objectified relationship assigning objects (IfcObject), acting as a resource usage or consumption, to a resource (IfcResource)

Strategy to generate asset level Digital-DSM

According to the elements and interdependencies defined, an Open BIM approach is proposed based on IFC and BIM query methods to generate the Digital-DSM. This research used the Thames Tideway Tunnel in London as a case study to demonstrate how this approach can be applied in analysing system interdependencies in practice. With the help of BIM Views and Query engine in BIMserver, the IFC file of Tideway can be read, visualized (as Figure 2(a)(b)), and decomposed into different layers of components at asset level (as Figure 2(c)(d)).

Among these components, five kinds of them were selected randomly as the elements for constructing DSM, and the interdependencies between such components were identified according to the definition in Table 3 using Query engine. Two kinds of DSMs were thus generated to show the *geospatial* and *physical* interdependencies

between components. Using the DSMs, the design teams will be able to clearly understand the impact of design change. For example, if component B ‘*C-L151-M_Sheet piles*’ is changed, then element E ‘*C-G215-M_Foundations Piles*’ will be affected in both spatial location and material.

The results clearly clarify the complexity of the complex engineering systems, Tideway, and are helpful to understand the design change propagation and impacts.

Figure 2. Generating Digital-DSM for Tideway at asset level

DISCUSSION AND CONCLUSIONS

Late design change can introduce uncertainty and complexity into the design of complex systems, where identifying interdependencies is significant to ensure the accuracy and consistency of data. Compared to the conventional design change process, the digital twin-driven DSM for design change can perform better in processing time, cost, and accuracy (Table 4). There is no need for designers to track design changes and propagation manually. The proposed approach can help to automatically construct the DSM according to identified interdependencies at asset-level and manage the change propagation through following these interdependencies.

This research has taken steps towards developing a novel and effective way to simplify complexity and clarify interdependencies in complex systems through generating DSM from a digital twin (e.g., BIM). To do so, the Digital-DSM has been defined in terms of elements and interdependencies across scales, providing a theoretical understanding for integrating digital twin with DSM in analysing systems interdependencies. A prototype based on Tideway has been provided to illustrate how the Digital-DSM helps to improve the efficiency of late design change in complex engineering systems. The results show that the Digital-DSM can make a practical contribution to the identification, analyses and management of interdependencies, and is promising to improve the design efficiency of complex engineering systems.

Future work will focus on (1) validating and tailoring the proposed approach through more case studies and feedbacks from the industry; (2) the extension of Digital-DSM theoretical model to address the limited kinds of interdependencies defined in Digital-DSM; and (3) integrating the Digital-DSM with design change propagation algorithms to evaluate the impact of design change quantitatively.

Table 4. Comparison between Conventional and Proposed Design Change Process in Analysing Interdependencies

Evaluation Metrics	Conventional Design Change Process	Digital Twin-Driven DSM
Processing time	Time-consuming with manual efforts	High efficiency with automatic identification and analyses
Cost	High labour cost	Low labour cost with digital twin supported
Accuracy	Highly depended on expertise	Based on definition of interdependencies
Level of design change	Asset, project and system-of-systems levels	Asset-level, but can be expanded to others

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