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Fuzzy cognitive modeling with users for design system analysis

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Abstract

This paper outlines a fuzzy cognitive mapping (FCM) approach for engaging users in constructing a model for engineering design system analysis. The model's scope is drawn in reference to a socio-technical system and demonstrated with an assembly production system (a socio-technical system archetype). In particular, this paper focuses on modeling an existing assembly production system that needs to be re-designed, then analyzing the system models to inform the re-design task. The modeling approach engages users as participants (18 in this research) in observation and interviews, and these data are coded into adjacency matrices and fuzzy cognitive maps separately then integrated. The ability to model multiple users and technical entities together in breadth and detail, qualitatively and quantitatively, enables designers to zoom in to see the detail and zoom out to see a holistic perspective. The models are analyzed for overall cause, efect, and central variables. Through the FCM analysis of these variables, the elements of the existing design solution are made explicit, including inputs, external and boundary constraints, design principles, outcomes and outputs, function, and operations and structure. This is particularly useful in re-design, as demonstrated in the industrial re-design project here, where the FCM models make the current system design explicit and their analyses inform re-design intent by being synthesized into re-design foci and tasks.

Keywords System analysis · User design · Socio-technical systems · Fuzzy cognitive mapping

1 Introduction

Modeling is critical to engineering design because the information that models offer may be needed at any point in the engineering design process (Pahl et al. [2007\)](#page-27-0). Specifcally in the system analysis phase, models are commonly used to examine the behavior of solution variants so that a solution can be selected. System analysis can also be performed on an existing system to make the current design solution and its behavior explicit. In the latter use, these models provide insight to refne a design solution and, more deeply, can inform the intent for re-design, where whole or

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part of a previous design solution is re-worked to generate a new design (Dixon and Colton [2000\)](#page-26-0). This paper outlines a modeling approach for system analysis with an industrial re-design project: making a current design solution and its behavior explicit, then harnessing this analysis as feedback to defne new re-design tasks. This purpose makes this modeling approach valuable to design engineers who seek to learn from, and improve upon, a previous design.

In addition to its purpose, a good model must clarify what it represents and how it represents it; the former refers to the model's scope and contents, the latter its mode of communication. The former is addressed here in this introduction, generally then specifcally in the context of the research presented. The latter, the fuzzy cognitive modeling approach, is addressed in Sect. [2](#page-2-0).

In engineering design, a model generally represents an abstraction of reality that serves as a cognitive tool (Goel and Helms [2013\)](#page-27-1). The model's boundary can be drawn at various degrees of scope, impacting what the model's contents will include. When modeling a design and its behavior in system analysis, where the line of scope is drawn can position users within or outside of the model. At a micro level, the model could include a rack and pinion inside a machine

(a mechanical system). At a more macro-level, the model could include the machine and its user(s) (a user-machine system). At an even greater macro-level, the context could include a set of multiple users and their machines operating an assembly production system together (a socio-technical system) and so on.

As the scope for the boundary is drawn wider, the system model involves increasing consideration for users in tandem with additional and diversifed system entities and connections. As this interconnected system design grows, the system behavior becomes more uncertain because users exercise choice and multiple users working together exhibit social relations. As a result, the potential for multiple solutions, under-constrained systems, and ill-defned problems is more likely. At the same time, this interconnected system poses an opportunity to assess design use, behavior, and redesign more comprehensively. Accordingly, the need for a system model to assist in system analysis intensifes as the scope of the system model is drawn wider.

This need is further clarifed here with the industrial redesign project addressed in this paper. This project involved re-designing an assembly production system where two builders at a time manually assemble over 396 fnal product variants with the use of tools and machines in a fxed product layout. The operators (builders) are temporarily assigned to the position, so the teams are consistently changing. Additional users of the assembly production system included the roles of lead hand, supervisor, planner, and manager. It is clear that the multiple users of this system can offer valuable input to the design process towards clarifying frst how they use the system and how it operates and, in turn, how it is not working and can be improved through re-design. Typically, these types of systems are re-designed by addressing one factor at a time and continuously improving; however, the question is—what factor is most critical to address in the re-design task? If a designer asked each user for their perspective, they are likely to difer based on their various roles. Often times the decision is made by management or an engineer who is in the position of designer. Developing a system model with user input (manager and non-manager) positions the engineering designer as a facilitator to analytically defne the re-design task based on analyzing the system model, which harnesses the collective knowledge of all users and provides robust rationale for the decision. In this systems approach, it is also possible to identify synergistic factors that can be re-designed together. Accordingly for this project, system analysis was performed with the system users to create a model of the existing production system in use (characterizing its behavior), then the analysis of this model was further synthesized into re-design tasks.

This focus is consistent with calls for user-centered approaches to engineering design as a whole (Stappers et al. [2007](#page-27-2)) and specifcally integrating users into research on engineering design systems to better understand how people use designs (Piela et al. [1992\)](#page-27-3). There is a broad body of literature on user evaluations in product design, and positive impacts on assembly have been noted (Sundin et al. [2004\)](#page-27-4). In engineering design, users have been involved in participatory design mostly in regard to ergonomics (Sundin [2003](#page-27-5); Laing et al. [2005,](#page-27-6) [2007](#page-27-7)), with demonstrated positive outcomes on worker comfort and productivity (Vink et al. [2006](#page-28-0)). User engagement in modeling has included sketching for design problem analysis (Römer et al. [2000](#page-27-8)) and behavioral modeling for design validation (Malak and Paredis [2007](#page-27-9)). The engineering design literature does not address how users can directly participate in system analysis modeling when re-designing socio-technical systems, which is the contribution of the paper presented here. By addressing this need, this paper also aligns with the emphasis that because engineering design models are cognitive constructs it is critical that a "shared understanding is constructed through social processes of discussion and clarifcation" (Eckert and Stacey [2010](#page-26-1); Chakrabarti and Blessing [2014](#page-26-2)). This takes place here with users in constructing a system model.

The integral role of users in operating systems, such as assembly production systems, is defned as socio-technical systems (STS), which have been examined in the engineering design literature in numerous contexts. They have been applied to product design (Gish and Hansen [2013;](#page-27-10) Fernandes et al. [2014](#page-27-11)) and systems design more broadly (Zhao and Steier [1993](#page-28-1); Naumann et al. [2011](#page-27-12); Jones et al. [2013](#page-27-13)). They have been related to various stages of the engineering design process, such as requirements analysis (Sutclife [2000](#page-27-14)), prototyping (Kember and Murray [1988](#page-27-15)), and implementation (Zhao et al. [1992](#page-28-2); Panebianco and Pahl-Wostl [2006](#page-27-16)). They have been related to additional design methodologies, such as co-design (Clancey [1993](#page-26-3)) and collaborative design (Lu and Cai [2001](#page-27-17); Jing and Lu [2011](#page-27-18)). STSs have also been related to engineering design holistically in themes of sustainability and life-cycle performance (Cardin et al. [2013](#page-26-4); Hu and Cardin [2015\)](#page-27-19), innovation (Legardeur et al. [2010](#page-27-20)), knowledge (Clancey [1993;](#page-26-3) Baxter et al. [2007](#page-26-5)), best practices (Stevenson et al. [1994\)](#page-27-21), and decision making (Brissaud et al. [2003;](#page-26-6) Ostrosi et al. [2011\)](#page-27-22). This paper contributes to this literature with an approach for modeling a STS for system analysis and subsequent re-design.

The research presented here answers the following research questions: how can a socio-technical system model of an existing system be built with users for system analysis? How can this inform engineering re-design, especially the re-design task? To examine these questions, modeling is frst discussed and specifcally fuzzy cognitive modeling (FCM) used in this research. The research methodology is then outlined, including the feld study design context and fuzzy cognitive mapping technique. Next, the results and analysis are presented, featuring the fuzzy cognitive models of the system design. The discussion interprets the specifc fuzzy cognitive models generally in terms of Lin and Zhang [\(2004\)](#page-27-23) and Zhang et al.'s ([2011\)](#page-28-3) 'unifed' engineering system design model, so that the specifc results of the industrial re-design project can be understood more broadly and related to re-design considerations, especially to re-design tasks. The re-design tasks from the industrial feld study are then aligned with participants' re-designs. Trustworthiness and validation are then discussed with limitations, benefts, and contributions followed by conclusions. This fow is illustrated in Fig. [1](#page-2-1).

2 Modeling with fuzzy cognitive mapping

Models used in engineering design can be descriptive or predictive; deterministic or non-deterministic; iconic, analog, or symbolic (Dieter and Schmidt [2008\)](#page-26-7). This paper focuses on creating a descriptive, non-deterministic, and empirically symbolic model to address the aforementioned challenges. Descriptive models explain a real-world phenomenon as a system, providing critical insight into the design behavior. Non-deterministic models describe a system with ill-defned problems, managing "incomplete task descriptions and nondeterministic solution paths" (Hayes [1989](#page-27-24)). A symbolic model built on empirical evidence is critical to making the many system entities and connections explicit, especially the human aspects in a socio-technical system that cannot be derived from scientifc rules as they are based on personal experience and localized social norms.

Existing engineering design approaches for system modeling and analysis with multiple stakeholder inputs are restricted in this respect. Value-driven design aims to "convey the true preferences of the stakeholder using mathematical formulations like value models", such as Topcu and Mesmer's value models ([2018](#page-27-25)); this value model example is strong on analysis but has limited fexibility and breadth of stakeholder input since the majority direct their preferences towards a primary stakeholder's preference. Other approaches to design system modeling and analysis emphasize fexibility in stakeholder input, such as the KJ method, which clusters ideas; this method is strong on fexible input but has limited system analysis because it does not examine how clusters relate to one another. More broadly, analytical engineering design methods such as those summarized by Wynn and Clarkson ([2018](#page-28-4)), could be adapted for participant input but require structured inputs needing translation between design and stakeholder languages (e.g., system dynamics) and have limited system scope (e.g., process fowcharts). What is needed is a means of fexible stakeholder input coupled with rigorous system analysis with the required breadth (macro) and depth (micro) for modeling and analyzing a socio-technical system for engineering re-design. To unite these qualities, the authors considered logic as a common language and starting point. In searching further, Fuzzy Cognitive Mapping (FCM) was identifed as a potential tool since it identifes concepts and their relationships, capable of capturing a broad variety of user inputs while providing a logical foundation for system analysis. To support maximum fexibility in user input, these relationships and concepts were extrapolated from interviews and observations with users, providing freedom for users to say (designers to listen) and users to do (designers to observe) relative to the system under investigation. This fexibility in input and analysis structure also offers the potential to discover things that the users and designers did not expect, a potential "ah ha". What has resulted is a broadly applicable method for modeling system analysis with users that provides design engineers with the rigorous analysis they need to understand socio-technical concepts and relationships while providing users with freedom of expression,

Fig. 1 Research process fow

Evaluation

of Results

thus contributing to the model's ability to represent reality and discover emergent insights into a system and its re-design.

Current FCM progress in engineering has been summarized by Papageorgiou ([2014](#page-27-26)). Papageorgiou and Salmeron ([2014\)](#page-27-27) establish that fuzzy cognitive mapping has a wide scope of applicability, particularly useful in modeling complex systems with existing knowledge and human experience in a flexible, adaptable, and easy to use approach. Thus far, the FCM approach has had limited use in engineering design, applied in the context of environmental planning (Borri et al. [2015\)](#page-26-8), manufacturability analysis (Gavankar and Rao [1995](#page-27-28)), virtual modeling in the industrial design of automobiles (Silva [1996](#page-27-29)), failure analysis (Augustine et al. [2012](#page-26-9)), new product development (Achiche et al. [2013](#page-26-10)), consensus in the design process (Ostrosi et al. [2011\)](#page-27-22), and assembly design decision making (Kim et al. [2008\)](#page-27-30). The application of fuzzy cognitive mapping to modeling socio-technical systems in system analysis made in the research presented here relates this advancement in knowledge engineering to engineering design and does so with the following methodology.

A cognitive map consists of causal relationships (linkages) between concepts (causes and efects). Visually, the map is drawn as a di-graph, with linkages depicted as vectors and concepts as nodes (as presented in Fig. [2](#page-3-0)). Fuzzy cognitive maps allocate fuzzy logic values to the vectors so that the importance of the node can be analyzed. FCM is advantageous in organizing big data and representing knowledge depth and connectivity (Papageorgiou [2014](#page-27-26)). The ability to model multiple users and technical entities together in breadth and detail makes FCM quite useful for modeling in system analysis.

3 Methodology

This research was conducted in an industrial feld study, where an industrial assembly production system design was modeled with its users. This type of system was selected because it is a socio-technical system archetype (Cherns [1989;](#page-26-11) Emery [1989\)](#page-26-12), involving multiple users working in tandem with technical entities (machines, tools) and one another to operate the system (Vermaas et al. [2011\)](#page-28-5). As such, it provides a rich proof of concept for the proposed approach of system modeling with users that can be simplifed for systems of smaller scope (e.g., one user and a product).

The assembly production system studied was an assemble-to-order system. The re-design problem was to improve the performance of the system design to:

- Address growing product variety (396+final assembly variants);
- Address growing orders (25.5% year-to-year order growth from 2003 to 2012);
- Maintain quality standards;
- Accommodate for a high turnover of temporary employees; and
- Reduce cycle time.

This problem is increasingly more common in the growing paradigm of mass customization (Koren [2010\)](#page-27-31). The primary outcome of the system analysis modeling here is to clarify the re-design tasks.

In this assemble-to-order system, each final assembly consisted of 27–79 total components of 5 main types. The final assembly was manually assembled in batch production using a fixed product layout. Two builders, builder A and builder B, assembled the final products with rigid and flexible components. The builder position was assigned

Fig. 2 A simple fuzzy cognitive map (modeling a social media platform socio-technical system with user involvement in mind)

on a temporary basis to temporary, part-time, or full-time employees. The production environment was unionized. The users of the system involved as participants in the study pertained to the roles of builder, lead hand, planner, supervisor, and manager. In total, 18 participants in these management and non-management roles took part in the research presented here. Since temporary workers were included, recruitment was ongoing; thus, 8 participants (of all 5 roles) took part in the first data collection method (interview) and an additional 10 participants took part in the second data collection method (observation). For further details on the participant demographics and technical context, please refer to Townsend ([2015](#page-27-32)). All participants were recruited in line with research ethics protocols. Once the participants were recruited, the fuzzy cognitive mapping process was applied.

The general process for creating a fuzzy cognitive map consists of collecting data; coding the data into cause, linkage, and effect relationships; arranging the coding into an adjacency matrix; and then plotting the adjacency matrix as a di-graph where cause and effects are nodes (ellipses) and the linkages are a vector. A simple fuzzy cognitive map is illustrated in Fig. [2](#page-3-0) for modeling a social media platform socio-technical system with user involvement in mind. The nodes (ellipses) indicate overall cause concepts (gray filled ellipse), overall effect concepts (black filled ellipses), and central concepts that act as intermediary cause and effect (white filled ellipses). The vectors indicate negative linkages (red dashed arrow) and positive linkages (black solid arrow).

An overview of the fuzzy cognitive mapping process that was utilized in this research is illustrated in Fig. [3.](#page-4-0) What is unique about the process used here, compared to the general FCM process, is that both interview and observation data were utilized and then integrated into one fuzzy cognitive map.

The steps in Fig. [3](#page-4-0) are described in more detail in the subsequent sections.

3.1 Data collection

The participants were asked to participate in semi-structured interviews and observations. Semi-structured interviews were conducted with participants, then recorded, transcribed, and verifed. Participants were asked (1) how would you describe the current assembly process? (2) How would you describe your work with the current assembly process? (3) How would you describe an ideal assembly process? (4) How would you describe your ideal work with the ideal assembly process? These questions are open ended and general to enable the unique perspectives of users to be shared. Any follow-up questions to a participant response generally asked the participant to clarify a concept that s/he mentioned or paraphrased the question if it was unclear to a participant. An example of a clarifcation is a participant mentioned "mistakes" and the interviewer asked "and the mistakes are what you're fnding when you're doing those checks?" An example of paraphrasing is when the interviewer asked the participant, "How would you describe your work with the current assembly process?" and the participant said "good". The interviewer further inquired by asking, "How would you describe the things that you do in order to build [an assembly]?"

Observations of the participants operating the production system were conducted at random time intervals. Ten unique datasets (production runs) of 226 unique data members (assembly cycles) were collected, which provided information on the products, process steps, and layout. These data were integrated into an observation operational model using statistical tests (Welch's ANOVA and regression). The observation operational model is further described and defned by equations in the results Sect. [4.2](#page-9-0). After data collection, the data were coded.

3.2 Data coding

Adjacency matrices FCM plots and **Data collection** analysis Adjacency matrices are created for: Design users Each adjacency **Data** Each interview; are recruited coding matrix is plotted All interviews together and data is as a digraph, (collective interviews); Based on collected via: then analyzed The observation • Interviews Wrightson by classifying operational model; and Observation (1976) concepts and The integrated model (operational characterizing (the collective interviews model) the complexity with the observation of the FCM plot. operational model)

In fuzzy cognitive mapping, data can be coded from questionnaires (Roberts [1976](#page-27-33)), by participants in an interview (Carley and Palmquist [1992](#page-26-13)), from interview texts

Fig. 3 The fuzzy cognitive mapping process used in this research

(Wrightson [1976\)](#page-28-6) or through data (Schneider et al. [1998](#page-27-34)). The last two methods were utilized in this paper.

The main goal of the coding is to identify concepts and their relationships in the form of cause concept/linkage/ efect concept. A 'cause' is defned as a concept that precedes or leads to the efect concept. Correspondingly, the 'effect' is a concept that proceeds from or follows the cause concept. The coding is made "fuzzy" by giving the linkage a fuzzy value between −1 and 1, making fuzzy cognitive maps both qualitative and quantitative in nature.

Each interview transcript was coded one sentence at a time, and each observation was coded one equation at a time. The coding process outlined by Wrightson ([1976\)](#page-28-6) is summarized in Fig. [4,](#page-5-0) with further details and special considerations for data in the form of interview text and an observation operational model specifed in Table [1.](#page-5-1)

Additional considerations for the coding are available in Wrightson's ([1976\)](#page-28-6) coding summary; for further details in relation to the industrial re-design project here, please see Townsend [\(2015\)](#page-27-32).

Fig. 4 Summary of Wrightson's [\(1976](#page-28-6)) FCM coding process

After each interview text was coded, as well as each equation in the observation operational model, the codes between them were compared and merged to create the integrated interview and observation model. This was done with the following rules adapted from Table [1:](#page-5-1)

- Does the participant make a distinction between the two concepts in the interview versus observation? If so, keep the concepts separate. If not, merge the concepts.
- Would the participant believe that the logic had been distorted if the concepts were merged? If so, keep the concepts separate. If not, merge the concepts.
- Are antonyms used? There should be one concept for a synonym and antonym.

If one concept from the interview described several concepts in the observation model, or vice versa, they were related with a bi-directional linkage since they were related in meaning but not the same, and one is not necessarily the cause of the other. The same is true if a concept in one partially, but not fully, described a concept in another. The results of this integration are presented later in Table [3.](#page-11-0) After the data were coded, they were organized into adjacency matrices.

3.3 Adjacency matrices

The coding was translated into a square adjacency matrix (*A*) of size *N*, where *N* is the total number of concepts. Each linkage fuzzy logic value was placed in the adjacency matrix (a_{ii}) according to its cause concept (row) and effect concept (column).

For the interview data, an adjacency matrix was created for each interview $(A_{interview})$. In this study, there were eight interviews $(A_{interview1}...A_{interview8})$. These individual interview matrices were then integrated into one matrix (*A*interview(all)). Since one interview can contain the same concept that is expressed in another, the two matrices were integrated (rather than summed) interview-to-interview. The integration process started with the frst interview adjacency matrix $(A_{interview1})$. Concepts in $A_{interview2}$ that differed from $A_{\text{interview1}}$ were added as rows and columns to $A_{\text{interview1}}$ to form *A*interview1–2. Linkage values from *A*interview1 remained and linkage values from $A_{interview2}$ were placed in their corresponding locations in *A*interview1–2. Redundant linkage values were compared (not added). This checked for consistency in the coding process and across participant data; if there were any discrepancies, the interview transcripts were reviewed to see if there was a coding error or to analyze the participant data. In this study, interview1 was integrated with interview2 into $A_{\text{interview1}-2}$, which was then integrated with interview3 into $A_{\text{interview1-3}}$, and so on until $A_{\text{interview1-8}}$ (which is $A_{\text{interview(all)}}$).

For the observation operational model, an adjacency matrix was created for each equation. In this study, there were eight equations presented in Sect. [4.2.](#page-9-0) These eight adjacency matrices were then integrated into one matrix $(A_{observation})$. The integration process started with the frst equation adjacency matrix. Concepts in the second equation adjacency matrix that difered from the frst were added as rows and columns, and so on for all seven matrices, creating the fnal integrated matrix (*A*observation).

For the integrated collective interviews and observation operational model, an adjacency matrix was created. Concepts in the observation operational model (*A*observation) that difered from the collective interviews $(A_{interview(all})$ were added as rows and columns, creating the integrated adjacency matrix (*A*interview(all)&observation). After the adjacency matrices were created, they were plotted.

3.4 FCM plots and analysis

The adjacency matrices (A) were plotted as di-graphs using social network visualization software (Pajek [2018](#page-27-35)) to create the related fuzzy cognitive maps. In the plots, the vector direction travels from the cause node to the effect node. Here, a negative vector was defned with a red dashed line; a positive vector was defned with a black solid line. Nodes were defned as ellipses with diameters relative to their centrality value, which is defned in Eq. [3](#page-6-0).

The fuzzy cognitive maps were then analyzed in terms of their structure by understanding each node as a transmitter, receiver, or ordinary variable (*v*) based on in-degree (id) and out-degree (od) values. Using the FCMapper software (FCMapper [2018](#page-26-14)) helped to automate this process. A receiver variable (overall effect) has an out-degree of 0 and a non-zero in-degree. A transmitter variable (overall cause) has an indegree of 0 and a non-zero out-degree. An ordinary variable has a non-zero in- and out-degree. In-degree and out-degree were calculated in Eqs. [1](#page-6-1) and [2](#page-6-2) from (Özesmi and Özesmi [2004\)](#page-27-36). Based on the in-degree and out-degree values, the centrality (*c*) for each variable was also calculated (Eq. [3](#page-6-0)). Additionally, the number of linkages (L, Eq. [4\)](#page-6-3) was also calculated:

$$
od(v_i) = \sum_{k=1}^{N} |a_{ik}|,
$$
\n(1)

$$
id(v_j) = \sum_{k=1}^{N} |a_{kj}|,
$$
\n(2)

$$
c(v_i) = od(v_i) + id(v_i),
$$
\n(3)

$$
L = \sum_{i=1}^{N} \sum_{j=1}^{N} |a_{ij}|.
$$
 (4)

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The results for the data collection, coding, adjacency matrices, and FCM plots are presented in Sect. [4](#page-7-0). The FCM analysis is shown in Sect. [5](#page-12-0).

4 Results

4.1 Interview results

Eight interviews were coded resulting in 247 codes before merging and 120 unique codes after merging, numbered 1–120. Since the interview texts were detailed, this variety of codes (concepts) is in line with Wrightson's ([1976\)](#page-28-6) statement: "If a document is broad in scope, it follows that mergers of concepts are more likely to be appropriate than if the text is highly specifc" (p. 323). The pre-interview text was specifc here since workers were explaining their work in an assembly production system where work was specialized and the participant pool represented a range of roles. A sample excerpt of an interview transcript and coding is presented in ["Appendix A"](#page-22-0), with the corresponding adjacency matrix in Fig. [5](#page-7-1) and plot in Fig. [6.](#page-8-0)

As shown in Fig. [5](#page-7-1), FCM matrices are not symmetrical since some but not all cause and efect coding relationships

are reciprocal. For example, consider the first coding relationship in the frst row of the table in ["Appendix A](#page-22-0)". When the order size grows it has a negative association with the current amount of space because the existing space becomes more crowded. The amount of space does not dictate the order size.

Additional fuzzy cognitive maps from the interviews are shown in ["Appendix B"](#page-23-0). All interview codes are in ["Appendix C"](#page-24-0). Each interview contributes perspectives to the system model. For example, two code summaries are shown in Fig. [7](#page-8-1), with their centrality, in-degree, and outdegree values for each interview.

In Fig. [7,](#page-8-1) code 16 (idealness of the assembly process) is present in each interview as an overall effect (no outdegree) with diferent codes leading to it. In other words, each participant can think of diferent concepts to efect improvement. Code 17 (permanency of the workforce) is present in only 4 of the 8 interviews, meaning it is relevant to some users but not others, and it acts as an overall cause (no in-degree) infuencing the system. Figure [7](#page-8-1) also highlights that the centrality of a code in an interview is the sum of the in-degree and out-degree—only in-degree for an overall efect (code 16) and only out-degree for an overall cause (code 17).

Fig. 5 Interview adjacency matrix $(A_{interview1})$

 \mathcal{L} Springer

After each interview was mapped individually, they were merged interview-to-interview. The mergence of Interviews 1–3 into a FCM plot is illustrated in ["Appendix B"](#page-23-0) (Fig. [16](#page-24-1)), which also highlights code details. Figure 8 illustrates the mergence of Interviews 1–8 (collective interviews) into a FCM plot. This model highlights how succinctly the FCM method synthesizes information from multiple user perspectives, versus a code-by-code analysis (Fig. [7](#page-8-1)), making the critical areas visibly obvious from node size and vector clusters. In Fig. [8](#page-8-2), the in- and out-degree are visible as vectors and centrality is visible as node size.

Figure [8](#page-8-2) highlights why modeling is critical in re-designing the socio-technical system, because there are many concepts interacting. Without modeling, and without the insight of the user, it would be very difficult to identify all these concepts and their relations. These concepts and relations are critical to re-designing socio-technical systems since they are living systems and re-design involves transforming the system with existing concepts and relations. The observation results provide additional data to consider.

4.2 Observation results

Equations [5](#page-9-1)[–12](#page-9-2) outline the observation operational model, which was previously derived from statistical tests (Welch's ANOVA and regression) on the same observation dataset in the same case study. For example, the correlation between mean cycle time and an observation complexity ratio (*r*) was tested with linear regression, performed with a 95% degree of confidence (R -sq value = 0.81217). The result is shown in Eq. [5.](#page-9-1) The normality of the mean cycle times was confrmed with a probability plot and fat pencil test. For information on the derivation of the formulas, along with the statistical tests, results, and validation please see Townsend and Urbanic [\(2015\)](#page-27-37). The focus of the research presented here is to show how a numerical model and its constituent equations that were derived from observation can be used in fuzzy cognitive mapping and related to the interview data.

mean cycle time = $1.2099r + 1.0333$, (5)

$$
r = V + PC + |DR| + |TR| + |PR| + |AR|,
$$
\n(6)

$$
V = 1 - \frac{n}{2\text{Vol}_{\text{max}}},\tag{7}
$$

$$
V = \frac{n}{2\text{Vol}_{\text{max}}},\tag{8}
$$

$$
|DR| = \frac{|DA - DB|}{DT},\tag{9}
$$

$$
|TR| = \frac{|TA - TB|}{TT},\tag{10}
$$

$$
|PR| = \frac{|PA - PB|}{PT},\tag{11}
$$

$$
|AR| = \frac{|AA - AB|}{AT},\tag{12}
$$

where

r: observation complexity ratio

PC: pallet count accounts for the number of fnished assemblies that will ft on one pallet (relative size of the fnished assembly)

V: production phase variable. The observations relate to a production phase (*V*) in terms of the number of observations taken in the production run (*n*) and the position of the observations relative to the start of a production run (Eq. [7\)](#page-9-3) or end of a production run (Eq. [8\)](#page-9-4). The logic of using these equations is included in the FCM using the variable '*S*' to indicate "start of a production run" (and correspondingly, −*S* to indicate the "end of a production run") and so *S* has relevance to the FCM but does not have a corresponding calculation.

Volmax: maximum total number of final assemblies required to complete the order (i.e., production run volume)

n: observation sample size

DR: distribution of work ratio related to the number of diferent components

DA: number of diferent components that builder A handles

DB: number of diferent components that builder B handles

DT: number of diferent components in the fnal product assembly that refers to the number of distinct component types in an assembly. The number of diferent components is distributed between builder A and builder B $(DA + DB)$.

TR: distribution of work ratio related to the total number of components

TA: number of total components that builder A handles (for one fnal product assembly)

TB: number of total components that builder B handles (for one fnal product assembly)

TT: number of total components in the fnal product assembly. The number of total components is distributed between builder A and builder B $(TA + TB)$.

- PR: distribution of work ratio related to the number of picking tasks
- PA: number of picking tasks that builder A performs PB: number of picking tasks that builder B performs PT: number of picking tasks for the fnal product assembly that refers to selecting the components. The number

of picking tasks is distributed between builder A and builder B ($PA + PB$).

AR: distribution of work ratio related to the number of assembling tasks

AA: number of assembling tasks that builder A performs AB: number of assembling tasks that builder B performs AT: number of assembling tasks for the fnal product assembly that refers to combining and positioning the selected assembly components. The number of assembling tasks is distributed between builder A and builder $B (AA+AB)$.

The main purpose of these equations is to understand how the assembly system organizes division of work (DR, TR, PR, AR) relative to builder A (DA, TA, PA, AA) and builder B (DB, TB, PB, AB) to assemble a fnal product (DT, TT, PT, AT, PC) for a given order and production run $(\text{Vol}_{\text{max}})$

Table 2 Observation operational model sample coding

Link Cause concept	Effect concept					
DT $+1$	$(DA+DB)$					
IDRI $(DA+DB)$ -1						
$ DA - DB $ IDRI $+1$						

relative to an observation set (*n*). The relationship between these variables is synthesized into a complexity ratio (*r*) that correlates with the mean cycle time—when complexity increases so does mean cycle time. To lower mean cycle time, the assembly system needs to be re-designed to lower the complexity ratio (*r*).

A sample of the coding for variable DR from this operational model (Eq. [9](#page-9-5)) is presented in Table [2](#page-10-0). Note that DA and DB are grouped together because the terms are interdependent, e.g., if DA increases then DB decreases. As a logic check, the coding in Table [2](#page-10-0) shows that the total number of components afects how many components builder A and builder B will have to work with. When this division of work is more even, |DA−DB| will decrease, so will DR, then *r*, then mean cycle time. This makes logical sense.

Figure [9](#page-10-1) presents the adjacency matrix for the observation model, which is plotted as a fuzzy cognitive map in Fig. [10.](#page-11-1)

The fuzzy cognitive map from the observation (Fig. [10\)](#page-11-1) contains far fewer nodes than the fuzzy cognitive map from the collective interviews (Fig. [8\)](#page-8-2), nearly 1/5. The observation map had 23 unique codes while the collective interview map had 120 unique codes. To assess if the maps contain the same codes, or diferent, the coding is compared and then merged where appropriate so that the maps can be integrated.

Code	⊨	$TA+TB$	TA-TB	E	5	DA+DB	DA-DB	\overline{p}	눈	PA+PB	PA-PB	PR	$\overline{\mathbf{z}}$	AA+AB	$ AA-AB $	AR	n/(2Volmax)	$1-(n/(2V)$ omax $))$	>	S	ပွ	$\overline{}$	Cycle Time Mean
ТT	Ω	$\mathbf{1}$	0	0	0	Ω	Ω	Ω	0	0	0	0	Ω	Ω	Ω	0	0	Ω	Ω	0	Ω	0	Ω
TA+TB	0	0	0	-1	0	0	0	0	0	0	0	0	0	0	0	0	$\mathbf 0$	0	0	Ω	$\mathbf 0$	0	0
TA-TB	Ω	0	0	$\mathbf{1}$	0	0	Ω	Ω	0	0	0	0	0	Ω	0	0	0	0	0	0	Ω	0	0
TR	0	0	0	0	Ω	Ω	Ω	Ω	0	0	0	0	0	0	0	0	0	0	0	0	Ω	$\mathbf{1}$	0
DT	0	Ω	Ω	0	Ω	$\mathbf{1}$	Ω	Ω	0	0	0	0	0	0	Ω	0	Ω	0	0	Ω	Ω	0	0
DA+DB	0	0	0	0	Ω	Ω	Ω	-1	0	0	0	0	0	0	0	0	Ω	0	0	0	Ω	0	0
DA-DB	0	0	0	0	Ω	0	0	$\mathbf{1}$	0	Ω	0	0	0	0	0	Ω	0	0	0	Ω	Ω	0	0
DR	0	0	0	0	Ω	Ω	Ω	0	0	Ω	0	0	0	Ω	0	0	0	0	0	0	Ω	$\mathbf{1}$	0
PT	$\mathbf 0$	0	0	0	0	0	0	0	0	$\mathbf{1}$	0	0	0	0	0	0	0	0	0	0	0	0	0
PA+PB	$\mathbf 0$	0	0	0	0	Ω	0	0	0	Ω	0	-1	0	Ω	0	0	0	0	0	0	Ω	0	0
IPA-PB	$\mathbf 0$	0	0	0	0	0	0	0	0	0	0	$\mathbf{1}$	0	0	0	0	0	0	0	0	0	0	0
PR	0	0	0	0	0	Ω	Ω	Ω	0	Ω	Ω	Ω	0	Ω	Ω	Ω	Ω	0	Ω	Ω	Ω	$\mathbf{1}$	0
AT	0	0	0	0	0	Ω	Ω	0	0	0	0	0	0	$\mathbf{1}$	0	0	0	0	0	0	Ω	0	0
AA+AB	0	$\mathbf 0$	0	0	Ω	Ω	0	0	0	Ω	0	0	0	0	0	-1	0	0	0	Ω	Ω	0	0
AA-AB	$\mathbf 0$	0	0	0	0	0	0	0	0	$\mathbf 0$	0	0	0	0	0	$\mathbf{1}$	0	0	0	0	0	0	0
AR	0	0	0	0	0	0	Ω	0	0	Ω	0	0	0	0	0	0	0	0	0	Ω	Ω	$\mathbf{1}$	0
n/(2Volmax)	Ω	0	Ω	0	0	Ω	Ω	Ω	0	0	0	Ω	0	Ω	0	0	0	0	$\mathbf{1}$	0	Ω	0	0
1-(n/(2Volmax))	0	Ω	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	$\mathbf{1}$	0	Ω	0	0
V	0	0	0	0	Ω	Ω	Ω	Ω	0	Ω	0	0	0	Ω	0	Ω	0	0	0	Ω	Ω	$\mathbf{1}$	0
Ś	0	0	0	0	0	0	Ω	0	0	0	0	0	0	0	0	0	-1	$\mathbf{1}$	0	0	Ω	0	0
PC	Ω	0	0	0	0	Ω	Ω	Ω	0	0	0	Ω	0	Ω	Ω	Ω	Ω	Ω	Ω	Ω	Ω	$\mathbf{1}$	0
r	Ω	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	$\mathbf{1}$
Mean Cycle Time	0	0	0	0	0	Ω	0	Ω	0	0	0	0	0	Ω	0	0	0	0	0	0	Ω	0	0

Fig. 9 Observation operational model adjacency matrix (A_{observation})

Fig. 10 Observation operational model fuzzy cognitive map

4.3 Integrated (interview with observation) results

Table [3](#page-11-0) outlines how the collective interview codes were merged with links to the observation codes.

Figure [11](#page-12-1) illustrates the linkages and merging from Table [3](#page-11-0). All the concepts from the observation adjacency matrix (A _{observation}) are shown in aqua (dark shading) and the concepts from the collective interview adjacency matrix

Fig. 11 FCM plot of the observation operational model with linked and merged concepts from the collective interview

 $(A_{intervised})$ from Table [3](#page-11-0) are shown in yellow (light shading). The relation between the 23 observation concept codes and 11 collective interview concept codes in Table [3](#page-11-0) highlight that the observation data are focused on physical entities and actions within the system that are directly observable from an observation perspective, which makes sense because the assembly process was directly observed in this case. In addition, 8 of the collective interview concepts relate to 20 observation concepts, showing how the observation accounts for very detailed information related to physical entities. The numerous interview codes is contributable to the breadth of information collected from interviews that often cannot be directly observed at the assembly process, or would require multiple and broader system observation points.

As Fig. [11](#page-12-1) shows, of the 23 codes in the observation dataset, there were only 3 codes that were identical to codes in the collective interview dataset. There were an additional 12 codes in the observation dataset that were linked to the collective interview dataset. There were 8 codes that were unique to the observation dataset. Of the 123 codes in the collective interview dataset, 105 were unique. It is important to note that diferent data collection methods collect different information and that the observation here is focused on a quantitative model while the interviews are focused on qualitative information. A signifcant value of the proposed modeling approach is that it integrates qualitative and quantitative data from the two methods, which contribute uniquely to the system model and relate to one another. The fully integrated observation and interview adjacency matrix

(*A*interview(all)&observation) is plotted in Fig. [12](#page-13-0). All the collective interview adjacency matrix $(A_{interview(all})$ concepts are shown in yellow (light shading). All the concepts form the observation adjacency matrix $(A_{\text{observation}})$ are shown in aqua (dark shading).

Figure [12](#page-13-0) illustrates quite clearly that the interview data contributes a breadth of concepts to the integrated model, while the observation data contribute to a narrower scope of concepts. The integrated model's ability to show both is a signifcant strength of the FCM modeling approach taken in this paper. In Fig. [12](#page-13-0), the size of the nodes indicates the most central concepts (highest centrality value) in the integrated adjacency matrix. In order of centrality value, these are: being able to position materials for the assembly process (code 7); permanency of workforce (builders) (code 17); having a designated position for materials around the table (code 45); builders dividing work evenly (code 20); and idealness of assembly process (code 16). Determining what are the most critical concepts integral to the re-design task, and classifying these variables, is the purpose of the analysis (Sect. [5\)](#page-12-0).

5 Analysis for variable classifcation

Using Eqs. $1-3$, the concepts (codes or nodes) represented in each FCM plot are classifed as receiver (overall efect concept), transmitter (overall cause concept), or ordinary (overall central concept) variables. This analysis is organized and highlighted per six categories:

- 1. The codes with the top 3 highest out-degree values >1 ();
- 2. The codes with the top 3 highest in-degree values >1 (\bigcirc);
- 3. The codes with the top 3 highest centrality values >1 ();
- 4. The top 3 overall cause concepts $(id = 0, and od = highest$ 3 values) (**bold C**);
- 5. The top 3 overall effect concepts $(id = highest 3 values,$ and $od = 0$) (**bold** E);
- 6. The top 3 overall central concepts $(c=$ highest 3 values, id≠0, od≠0) (**bold N**).

In the case of a tie within a category, all the tied codes within that category are included. The highlighted codes unique to one data collection method are indicated with *, and the highlighted codes common with another or integrated method are indicated with $+$. The collective interview analysis is presented in Table [4;](#page-13-1) the observation analysis is presented in Table [5;](#page-14-0) the integrated collective interview with observation analysis is presented in Table [6.](#page-15-0)

Note: in the case of the observation adjacency matrix and plot analyzed in Table [5](#page-14-0), there is only one overall efect (receiver variable) and only one code with an out-degree value > 1. Also, $|TA - TB|$, $|DA - DB|$, $|PA - PB|$, and |AA − AB| do not remain overall causes in the integrated FCM analysis.

The analysis in Tables $4, 5$ $4, 5$ $4, 5$, and 6 shows that the fuzzy cognitive models for the collective interview and observation highlight diferent overall causes, efects, and central concepts. By synthesizing the two together, the integrated model fnds additional considerations. Code 7 arises with a high out-degree, in-degree, and centrality and as a top 3 overall central concept. Code 20 and 45 also arise as top 3

Code	Code description	Out- degree $od(v_i)$	In- degree $id(v_i)$	Centrality $C(V_i)$
17	Permanency of workforce (builders) +	12C		12
43	Like an assembly line +	6		
63	Forecast accuracy *	5		6
64	Order accuracy *	5		
68	Size of customer account +	5C	0	5
92	Current location of the assembly area (versus past location) +	5C		5
16	Idealness of assembly process +	0	29.5E	9.5
5	Assembly process efficiency +	Ω	4E	4
96	Lead hand availability/utilization +	Ω	5E	5
40	Ease of flow of materials *	1.5		7.5N
51	Material handlers pick up finished pallet +	2		9N
27	Assembly components missing *	4		8N

Table 4 Code classifcation for the collective interview FCM

overall central concepts. The overall cause, efect, and central concepts in the fuzzy cognitive models (Tables [4,](#page-13-1) [5,](#page-14-0) [6](#page-15-0)) are interpreted in terms of a general engineering system design model in the following discussion.

6 Discussion

Creating a system model of an engineering design for system analysis requires an understanding of what is being modeled (meta-cognition) to relate the specifc model (the FCM models) to the general model. The engineering system design model, and its variations in four regional schools, is integrated into a 'unifed' model by Lin and Zhang ([2004\)](#page-27-23) and Zhang et al.'s ([2011\)](#page-28-3). In the following paraphrased summary of their model, the elements of their model are related to the variable classifcations in the fuzzy cognitive model (*in italics*). In Lin and Zhang[\(2004](#page-27-23)) and Zhang et al. ([2011\)](#page-28-3) model, the context is the boundary between the system and the environment. In this case, since users are integral to the operation of the socio-technical system and are thus a part of the system model, they are inside the system boundary.

Lin and Zhang (2004) and Zhang et al. (2011) (2011) (2011) model of the design system consists of physical entities that are meaningfully connected and directly related to the fuzzy cognitive mapping model analysis overall cause, efect, and central concepts from Sect. [5](#page-12-0) as follows. When the system is in operation, these entities are perceived as states that refect the system structure (*FCM analysis overall central concepts*). Internal constraints relate to the connection between entities that form the system at time t (*FCM analysis overall central concepts*); the external constraints are imposed from the environment to the entities in the system (*FCM analysis overall cause concepts*). Information, energy, and materials

Code	Code description	Out- degree $od(v_i)$	-In degree $id(v_i)$	Centrality $C(V_i)$
	Being able to position materials for the assembly process [*]	6		14N
17	Permanency of workforce (builders) +	12C		12
43	Like an assembly line +	6		
68	Size of customer account +	5C	n	5
92	Current location of the assembly area (versus past location) $+$	5C		5
16	Idealness of assembly process +	0	29.5E	9.5
51	Material handlers pick up finished pallet +	$\overline{2}$		9
5	Assembly process efficiency +	Ω	4E	4
96	Lead hand availability/utilization +	Ω	5E	5
20	Builders dividing work evenly *	5	5	10N
45	Having a designated position for materials around the table *	5	6	11N

Table 6 Code classifcation for the integrated collective interview with observation FCM

are transmitted from the outside environment into the system as independent state variables, or inputs (*FCM analysis overall cause concepts*). Information, energy, and materials are transmitted from the system to the outside environment as dependent state variables, or outputs (*FCM analysis overall efect concepts*). The system behaves according to how the inputs relate through the entities into outputs; principles govern this behavior (*a principle is a FCM analysis overall cause concept*). The function of the system relates the behavior to the context (*a function could be measured as an outcome or FCM analysis overall efect concept*).

The relation between Lin and Zhang([2004\)](#page-27-23) and Zhang et al. ([2011\)](#page-28-3) design system model and the fuzzy cognitive

model is illustrated in Fig. [13](#page-15-1). In Fig. [13](#page-15-1), the box represents the system with a context boundary and internal structure. Cause concepts are shown as gray ellipses; efect concepts are shown as black ellipses; white ellipses and text are shown as central concepts. As stated earlier, a 'cause' is defned as a concept that precedes or leads to the efect concept. Correspondingly, the 'efect' is a concept that proceeds from or follows the cause concept.

Based on this analogous relationship between the general system design model and the FCM models, the FCM variable classifcations from Tables [4](#page-13-1), [5,](#page-14-0) and [6](#page-15-0) (overall cause, efect, and central concepts) are related the standard terminology for the engineering system design model.

efect, and central concepts

- Inputs: inputs appear in the FCM model as overall causes that move from the outside environment to afect entities in the system.
- External and boundary constraints: external and boundary constraints appear in the FCM model as overall causes that move from the system boundary to afect entities in the system.
- Design principles: design principles appear in the FCM model as overall causes that move from outside or within the system boundary to affect the behavior of the system.
- Outcomes and outputs: outcomes and outputs appear in the FCM model as overall effects that move from entities within the system to the outside environment.
- Functions: functions appear in the FCM model as overall efects that move from the behavior in the system to relate to the context.
- Operations and structure: operations (events) and structure (organization) appear in the FCM model as central concepts that exist within the system boundary.

Relating the FCM variable classifcations to the engineering system design model standard terminology is made even more useful to designers by synthesizing this insight into re-design foci. These re-design foci are themes related to the FCM variables (overall causes, effects, and central concepts), determined by frst grouping the structure aspects—process, layout, and training in this case. Process and layout are common re-design foci for manufacturing systems that are informed and specifed by the FCM analysis variable classifcation per the case study. The other identifed elements of the system (operations, function, outputs, outcomes, design principles, external and boundary constraints and inputs) are then integrated into these re-design foci and where they are not clearly in alignment warrant new re-design foci (in this case diferentiated design was a new insight, so it warranted emphasizing in the re-design foci). Throughout this merging and comparison process attention is paid to ensure that the participant meaning of the codes is preserved. In doing so, the analysis of the behavior and use of the existing design solution (system model) is harnessed to inform the re-design process. The result of this FCM variable classifcation into standard system model terminology and re-design foci is outlined in Table [7](#page-17-0).

The classifcations of the variables in Table [7](#page-17-0) are further outlined relative to their standard system model terminology and re-design foci as follows.

6.1 FCM overall cause concepts: system inputs, constraints, and design principles

The inputs are most readily identifable from the observation FCM model (Table [5](#page-14-0)). The inputs identify what should be monitored in the re-design because changes in them will have

a signifcant impact on the system. In subsequent re-design phases of the case study, these inputs remained the same and the participant users and designers designed with these specifcations in mind (e.g., the overall assembly product designs were specified to the production environment). Note: the number of picking (PT) and assembling tasks (AT) represent materials since the number of tasks are determined by the packaging of the components, which is performed in a system external to the immediate one being studied.

The external and boundary constraints are most readily identifable from the interview and integrated FCM models (Tables [4](#page-13-1) and [6\)](#page-15-0). The external and boundary constraints represent constraints in the re-design that a designer must bear in mind. In subsequent phases of the case study, the participant users and designers designed within these limitations (e.g., the re-design had to optimize space within the space that was currently designated for the assembly process). The permanency of the workforce and size of the customer account are determined outside of the immediate system in the broader business context; the current location of the assembly area is a space constraint defned at the current system boundary.

The FCM model analyses in Tables [4](#page-13-1) and [6](#page-15-0) included the concept "like an assembly line," which was driven by the overall cause concept "applies to us" as a design principle. Upon further inspection of this node, the participants cite that this principle will afect the amount of work for builders, lead hands, and material handling; the ease of fow of materials; a new machine; and the ease of assembly work (the entitles in the system and their internal constraints). In other words, the intent of the re-design principle "like an assembly line" is not simply to create an assembly line; the intent is to adapt the assembly line paradigm in light of the concerns about it. This is a signifcant insight into the redesign task and design thinking—a shift in intention from a universal to a diferentiated solution that "applies to us" ("us" being the design users). The FCM model analysis in Table [5](#page-14-0) also includes overall causes being the diference in the total and diferent components between builders A and B as well as the diference in picking and assembling tasks between builders A and B. These concepts are driven by the design principle of "diference"—in other words, how division of labor is interpreted and applied. The design principles are identifed in all the FCM models (Tables [4](#page-13-1), [5](#page-14-0), [6](#page-15-0)). The design principles are critical to re-design because they represent organizational principles of the system; these are signifcant insights to consider in the conceptual phases of the re-design.

6.2 FCM overall efect concepts: system outputs, outcomes, and function

In the FCM model analyses (Tables [4,](#page-13-1) [6](#page-15-0)), "idealness of the assembly process" is viewed as an overall outcome. Upon

closer inspection of the cause and efect chain, this "idealness" is achieved from such concepts as builders making decisions about work as partners, working conditions that one participant described as "chivalry," and meeting constraints (such as the current location of the assembly area and permanency of the workforce) and leads to assemble fnal assemblies without any components missing (the output). Lead hand availability/utilization is also identifed as an overall outcome. The outcomes and outputs are identifed in the interview and integrated FCM models (Tables [4,](#page-13-1) [6](#page-15-0)). The outcomes and outputs identify what should be monitored in the re-design because changes in them will have a signifcant impact on the value of the system and its impact on the broader system. In subsequent re-design phases of the case study, these participant users and designers designed with these outcomes and outputs in mind (e.g., one of the re-designs was a "double-check system" to ensure that fnal assemblies had no components missing (the output)).

In the FCM model analyses (Tables $4, 5, 6$ $4, 5, 6$ $4, 5, 6$ $4, 5, 6$), the mean cycle time value and assembly process efficiency are measurements of the overall function. In this system model of an assembly production system, the function is to transform inputs into outputs; cycle time and efficiency measure this rate of transformation. The function is measured in the interview and integrated FCM models more generally as efficiency (Tables [4](#page-13-1), [6\)](#page-15-0) and more specifcally as cycle time in the observation FCM model (Table [5](#page-14-0)). The function identifes how the impact of the re-design can be measured. In the case study, the before and after conditions of the system were compared with respect to the cycle time to quantify the system function improvement.

6.3 FCM central concepts: system operations and structure

The operation and structure are identifed in all the FCM models (Tables [4,](#page-13-1) [5](#page-14-0), [6](#page-15-0)). Not all these operations and structure are desirable, and so this analysis highlights areas where the re-design can target corrective actions with improved operations and structure. The operations and structure identify specifcs about what should be re-designed in the system, which are synthesized together and with the previous system elements identifed into re-design foci and tasks as follows.

6.4 Re‑design foci and tasks

Re-*design foci and task #1: Process* The re-design seeks to engage builders to transform the input signals/information (start of production run, orders) and materials (components) into fnal product assemblies with no components missing. This is accomplished through central human–object operations (e.g., the distribution of work, utilization of workers including lead hands, etc.). Together, this behavior performs the system function, as observed in cycle time (a measure of efficiency). The division of work between builders affects, and is afected by, the distribution of the components and picking and assembling tasks, which also afects and is afected by the positioning of the components. The re-design aims to address this inter-relationship between layout and process to accomplish tasks #1 and #2 synergistically.

Re-*design foci and task #2: Layout* Existing space (current location of the assembly area) is a constraint. In this situation, the re-design seeks to better utilize the existing space to position materials for better fow while addressing that it is challenging to have designated positions for materials due to the variety of assembly components and fnal product assemblies and the distribution of the components and picking and assembling tasks. The layout must accommodate pallet positioning and removal.

Re-*design foci and task #3: Training* The temporary builder position is a constraint. In this situation, the redesign seeks to improve the existing builder training practices with consideration for the training time to ensure that builders know what to do in the assembly process especially in regard to quality (ensuring that assembly components are not missing in the fnal product assembly).

Re-*design foci and task #4: Diferentiated design* To accomplish re-design tasks #1–3, the re-design process is intent on working with users to ensure that the re-designed assembly production system "applies to us" (diferentiated design) and accommodates diferences. This user-oriented re-design process begins with these four re-design tasks, which inform the social contract between the designer and users to resolve the tasks together. In doing so, the re-design process is also committed to the continued work culture, described by one participant as "chivalry."

6.5 Participants' re‑designs

In the case study, the re-design foci and tasks were used to plan for and conduct subsequent participatory design events where users were invited to take part in developing concept and detail re-designs through refection-in-action. The frst two events focused on process and layout and the third event focused on training; diferentiated design was a theme throughout. The participants' re-designs (Table [8\)](#page-19-0) are shown here to illustrate their alignment with the four re-design foci and tasks. The re-designs are discussed in relation to the assembly components and products (Table [9\)](#page-20-0).

These re-designs were implemented and then tested with users, specifically using the operational model presented here to compare the before and after system models to assess the impact of the re-design implementation. The function of the system improved—the mean cycle time was found to be lower after the re-design versus before with a statistically significant mean difference of − 0.72 min/assembly (Townsend and Urbanic [2015](#page-27-37)). For further details on the re-designs and their impact, please see Townsend ([2015\)](#page-27-32).

Table 9 Final assembly component variants and precedence (from pre- and post-observation)

Table 10 Task division for the new picker and assembler builder roles

Fig. 14 The re-designed layout

6.6 Trustworthiness and validation of the proposed approach

It is critical to note the epistemological orientation of the qualitative research approach taken herein with respect to the interviews, which considers that participants reveal knowledge to the researcher who in turn strives to see if s/he can see what the participants see, to then help others to see it. This contrasts the positivist epistemological position often taken in engineering, taken herein with respect to the observations, which aims for objectivity. The proposed approach integrates both research orientations because when re-designing a socio-technical system it is critical to see the system as the users see it because ultimately their view of the system *and* the system itself need to be transformed. Krefting ([1991](#page-27-38)) and Shenton [\(2004](#page-27-39)) present a framework that aligns qualitative trustworthiness and quantitative validation; the following provides evidence of the establishment of truth value, consistency, applicability, and neutrality in the research presented.

Krefting [\(1991](#page-27-38)) and Shenton ([2004](#page-27-39)) establish truth value as the confdence in the fndings and context in which the study was undertaken (credibility and internal validity) as well as consistency as the ability to generate the same fndings if the inquiry was replicated (dependability and reliability). In the proposed approach, a detailed methodology and case study context was presented. Eight interviews were conducted with participants from fve diferent roles. The observation random sampling consisted of 226 samples, collected over several months. Integrating both methods provided both breadth and depth of information. Reliability in the coding here was established using multiple data collection methods (interview and observation), keeping a code book during the coding process for consistency, and through the coding redundancy and merging. There were 218 linkages in the initial interview coding with 161 unique linkages; there were 247 codes in the initial interview coding with 120 codes after merging. One person coded and fve academics with PhDs and several interested participants reviewed the models and analysis. The system model was analyzed in terms of overall variable classifcations, which informed the re-design task with ongoing member checking in the next phases of the larger study [further information in Townsend [\(2015\)](#page-27-32)]. By utilizing research ethics protocols, the researchers/designers established trust with the user participants, which was critical to establishing data reliability in both the interview and observation data collection methods.

Krefting [\(1991](#page-27-38)) and Shenton [\(2004](#page-27-39)) establish applicability as the degree to which the fndings can be applied to other contexts (transferability and external validity). The methodology provides background data on the industrial context with detailed methodology; further information can be found in Townsend ([2015](#page-27-32)), which enables context comparison. Assumptions of the proposed modeling approach include the availability of participants; that the process is observable; the process transforms inputs into outputs; the transformation can be measured (timed); statistical test assumptions are not violated; there is a relationship between the process inputs, outputs, and the workers that varies in relation to a work strategy; and there is time to perform the interviews, observations, coding and analysis. To fully understand the transferability of the re-design approach developed here to other contexts, it needs to be tested with other types of sociotechnical systems.

Krefting ([1991](#page-27-38)) and Shenton ([2004](#page-27-39)) establish neutrality as the freedom from bias (confrmability and objectivity). The proposed methodology is detailed and demonstrated with clear examples. A substantial amount of evidence is analyzed with this detailed methodology.

6.7 Limitations and future research for the proposed approach

The limitations of the proposed approach for FCM with users include the following. First, in an industrial, nonresearch application of the proposed approach the designers will need to consider how trust between the designers and users will be established outside of research ethics protocols.

Next, developing questions for the interviews requires consideration. For the interviews, it was critical to ask questions that were broad enough to allow for openness yet within an appropriate scope to manage the number of codes. In future applications of the proposed approach, designers will need to carefully select questions adapting the ones established here.

In addition, it is critical to address model reliability and validation with the time required to conduct the FCM approach. For the observations, random sampling across several months promoted data reliability but required signifcant time. This time enables the designer to understand the data and design context more thoroughly, which benefts data analysis and interpretation, but there is a time and quality trade-of. Similarly, the coding itself takes time and the designers will need to determine an appropriate degree of detail.

Finally, it is critical to note that the fuzzy cognitive modeling approach presented here was conducted in one design context. Future research will need to test this approach in other applications and design contexts to test its transferability, adaptation requirements, and limitations. Future applications to test the proposed approach in engineering design could consider its usefulness in assessing use in product design as well as interactive design between multiple users and technology (e.g., user platforms, collaborative robots, user involvement in industry 4.0, etc.). The latter trends in engineering design emphasize growing inter-dependencies between users and technology, where user insight will increasingly be needed by engineering designers to create robust and sustainable designs.

6.8 Contributions and benefts of the proposed approach

This proposed approach provides an opportunity for users to directly participate in system analysis and modeling when re-designing a socio-technical system. A broad spectrum of users are invited to participate in interview and observation to develop a system model that integrates users' collective knowledge and identifes synergistic re-design foci, as described in this work. Benefts of the proposed approach include the following. Few modeling methods integrate qualitative and quantitative information, which lend unique insights. The proposed approach also balances social and technical aspects, which is needed for modeling a sociotechnical system comprehensively. In the approach, users defne the system model relative to their collective experience in their own words (interview) and actions (observation). In this regard, modeling is performed relative to the user experience versus the user experience being suited to the modeling method. This may seem subtle but it is important to the users to be heard and to the ability of the model to refect reality. Suiting the modeling to the user experience enables users to share information freely and designers to see (observe) and listen (interview); it enables users to guide the level of detail that is relevant; it means that what is important has not already been determined by the modeling method and, in turn, provides an opportunity for unexpected insights (e.g., how would the user idea of "diferentiated design" ft in with another method?). Further studies can directly compare the proposed method with other system modeling techniques (e.g., simulation, system dynamics, etc.) to identify additional benefts and limitations.

7 Conclusion

This research outlines the following approach for building a socio-technical system model with users for system analysis in engineering design. The proposed fuzzy cognitive approach for system modeling engages users as participants (18 here) in holistically modeling the system design; here, an assembly production system (a socio-technical system archetype) was modeled. Data on the participants' use of the existing system are collected in situ via interview and observation. The observation sets are synthesized with statistical analyses into an operational model. With fuzzy cognitive mapping techniques, the interview

data and operational model are coded into cause, linkage, and efect relationships. The coding is arranged into an adjacency matrix and then plotted as a di-graph where cause and efect concepts are nodes and the linkages are vectors. Each interview is coded individually then merged to form a collective interview adjacency matrix and FCM model. The collective interview adjacency matrix and FCM model is then synthesized with the observation operational adjacency matrix and FCM model into an integrated FCM model. A signifcant value of the proposed modeling approach is that it integrates qualitative and quantitative data from interview and observation, which contribute uniquely to the system model and relate to one another. With FCM analysis techniques, the models are analyzed for overall cause, efect, and central concepts.

This system model benefts the design process by directly integrating the multiple perspectives of design users into one system design model, which enables designers to zoom in to see the detail and zoom out to see a holistic perspective. Through the FCM analysis, the design system elements of the existing socio-technical system design solution are made explicit, including inputs, external and boundary constraints, design principles, outcomes and outputs, function, and operations and structure. This is particularly useful in re-design, as is demonstrated in the industrial re-design project here, where the FCM analyses are synthesized into re-design foci and tasks that inform re-design.

The limitations of this approach include the importance of trust between the designer(s) and users and the time required to conduct interviews, observations, and the FCM process. Future research needs to test this approach in additional socio-technical system types and design contexts, and in direct comparison to other modeling methods, to characterize its transferability and further defne its benefts and limitations.

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Appendix A: Transcript excerpt and FCM coding

Interviewer: How would you describe the current assembly process?

Participant: Umm… simplifed.

Interviewer: Ok

Participant: You know, we've, over the years we've gone from doing a few hundred now to probably [XX] …so, and, um, we're currently struggling with space and personnel. So we've, it's grown into its own type of

Table 11 Interview excerpt sample coding

department within a department now and we're struggling to staff it accordingly and be efficient at it. Now that we don't have a lot of room in our building we're unable to bring everything down and lay it all out ahead of the run. Usually what you'd like to do is bring everything down, let's say you are going to do 200 [assemblies], you want to bring every piece of that [assembly]…all the [X], all the [Y], all the componentry, bring it down, make sure it's all accurate, check it over, sign off on it, and then say "ok...

let's build." That's the efficient way. Now, we currently have to bring it in pieces, so at multiple times our material handlers are going to gather the pieces rather than doing it once due to our lack of space…. [interview continues]

See Table [11](#page-23-1).

Appendix B: Integrated interview plots (interviews 1, 2 and 3)

See Figs. [15](#page-24-2) and [16.](#page-24-1)

Fig. 15 FCM plot for interviews 2 and 3

Appendix C: Codes from participant interviews

See Table [12.](#page-25-0)

Table 12 Interview codes

Table 12 (continued)

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