LEARNING TO COLLABORATE DURING TEAM DESIGNING: SOME PRELIMINARY RESULTS FROM MEASUREMENT-BASED TOOLS

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This paper reports on preliminary results of an explorative study of a protocol analysis of team learning while designing using in-situ data. Two measurement-based frameworks are proposed. One framework is based on a sequential analysis of communication utterances and the other on codes using the Function-Behavior-Structure ontology and linked using Linkography. Only the former is reported in this paper. Results indicate that team learning occurs and can be discerned from an analysis of the utterances only but only at a high level of granularity.

Keywords: Team Learning, Design Collaboration, Protocol Analysis, Research Method.

1. BACKGROUND

New design ideas and products are increasingly the result of teamwork. Often a team is put together for a single project and the members of the team need to learn to collaborate. Is collaboration an innate quality of humans or do they have to learn to collaborate? We hypothesise that in order for a team to increase performance, members need to adjust their actions and behaviour through learning. Learning is difficult to study because of the paucity of data and the lack of adequate tools. Most data comes from laboratory (in-vitro) experiments with little coming from in-situ sources. Much in-situ sourced data is too noisy and unstructured for current analysis tools. This study uses in-situ data of a team of cross-disciplinary professional designers collaborating on a novel product design.

1.1. Learning Research

Depending on the field of study, for example education or organization behaviour, the focus of exploring team learning engages with a different set of issues. Bransford et al. summarized the development of learning research as “from speculation to science”. Initially, behaviourists conceptualized learning as a process of connecting stimuli and responses. They assumed the motivation to learn is determined primarily by drives, such as hunger, and the availability of external forces, such as rewards and punishments. This view is rather restrictive in studying phenomena such as understanding, reasoning, and thinking. Cognitive science, which emerged in the late 1950s, provides new tools and methods to examine learning in a scientific manner (e.g. Ref. 2-4). Meanwhile rigorous qualitative research methodologies have provided perspectives on learning that complement and enrich experimental research (e.g. Ref. 5 and Ref. 6).

The new science of learning stresses learning with understanding which leads to the focus on the processes of knowing (e.g. Ref. 7, and Ref. 8). Barab considered the core of traditional pedagogy is a polarization of the learner and the learning context that he called the “Cartesian paradigm” and
he proposed an “ecological paradigm”. The Cartesian model separates the individual (knower) from the environment (known) while the ecological model situates the learner within the learning context. Lifelong learning is an example of situated learning and is difficult to study because of the paucity of data and the lack of adequate tools. A recent analysis of tools found that:

“discussion of 15 instruments results in a number of critical conclusions. There are questions about the coherence between the theoretical base and the operational translation of the theory in the instruments. Instruments are hardly compared or contrasted with one another. As a consequence the empirical base of the validity of the instruments is limited. The analysis is rather critical when it comes to the issue of reliability.”

Most data comes from laboratory experiments with little coming from in-situ sources. Much in-situ sourced data is too noisy for many of the direct statistical tools to be used successfully and requires filtering.

1.2. Learning to Design Collaboratively
In the area of design learning, there are studies that explore cognitive style and design learning. There are also studies that showed the differences between novice and experts designers in terms of knowledge, problem solving behaviour, and cognitive activities. These studies focus on individual designers learning designing.

Team interaction remains one of the most important elements in developing ideas. Systematic studies of designers started in the 1970s. Compared to individual designing, studies of teams have shown that there is a multiplicity of factors that contribute to or affect team designing. Some of these factors are role and relationship, trust, social skills, common ground, organisational context and socio-technical conditions. How team members learn these issues has not been well studied.

“Dewey described learning as an iterative process of designing, carrying out, reflecting upon, and modifying actions, in contrast to what he saw as the human tendency to rely excessively on habitual or automatic behaviour.”

Edmondson treated learning in a team as a process that attempts to articulate the behaviours through which such outcomes were considered as: adaptation to change, greater understanding, or improved performance in teams. We assume team members need to learn: 1) to acquire skills (social and communicational); 2) to establish their role and trust; and 3) to gain knowledge of common ground for collaboration, all with the result of producing performance as a team that is more than the sum of the performances of the individuals. We propose, initially, to study this kind of learning by examining their communication patterns to see if these patterns exhibit behavioural changes — both as individuals and as a group. Then, we further propose to investigate the semantics of the communication — verbal, written words and drawings — by coding and linking the communication content.

2. THE DATA
Data was obtained from the Design Thinking Research Symposium. The source data was a video of design meetings taking place in a product design practice. The data is made up of a 4-camera video recording, Figure 1, and the transcripts of the voice communication. The team consisted of a business consultant, who acted as the moderator (Allan), three mechanical engineers (Jack, Chad and Todd), an electronics business consultant (Tommy), an ergonomist (Sandra), and an industrial design student (Rodney). They were all from the same company and the student, Rodney, was on an internship with the company.

In this brainstorming session, the team was asked to provide ideas for solving technical issues of a working demonstrator of a thermal printing pen, Figure 2. Two main issues were: 1) keeping the print head in contact and optimum angle with the media despite wobbly arm moment; and 2) protecting the print head from abusive use and overheating. Observing the protocol, it can be divided into two episodes corresponding to these two issues.
Research into Design: Supporting Multiple Facets of Product Development

Figure 1. A frame from the 4-camera video recording of a team of 7 designers collaborating.

Figure 2. Illustrations of the function and behaviour of the design object: thermal printing pen.

Table 1. Measuring change of participations as a behaviour change of individuals that indicates learning.

<table>
<thead>
<tr>
<th>Observations</th>
<th>Superficial behaviour</th>
<th>Pure quantitative measures</th>
</tr>
</thead>
<tbody>
<tr>
<td>Change in participation</td>
<td>Number of words</td>
<td>Number of turn-taking</td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Underlying contributions that change performance of team</td>
<td>FBS contributions</td>
<td>Quantitative measures via coding and linking</td>
</tr>
<tr>
<td></td>
<td>Number of ideas</td>
<td>via coding and linking</td>
</tr>
<tr>
<td></td>
<td>Strength of ideas (no. links)</td>
<td>via coding and linking</td>
</tr>
<tr>
<td></td>
<td>Spread of ideas (link distance)</td>
<td>via coding and linking</td>
</tr>
</tbody>
</table>

3. PROPOSED STUDY FRAMEWORK

The units of analysis of this study will be the individual and the team. Learning is treated as a process that results in the improvement of the performance of the team. The performance is measured by the number of ideas and their impact within a specified timeframe. Behaviour is measured by a number of factors: contribution according to turn-taking; contribution according to word utterances; contribution according to semantic content in terms of quantity, strength or importance (number of links), inference or spread (link distance) of ideas, and the reformulation processes of the Function-Behavior-Structure ontology. The transcript was segmented based on ideas and turn taking. Segments form the basis of a linkograph of the protocol. A linkograph is constructed by linking the semantic contents of related segments. The framework is shown in Table 1. This paper only examines and measures the number of words and turn-taking. Markov analysis is used to examine the pattern of turn-taking.

4. RESULTS

4.1. Word Count

A participant was deemed to dominate a time interval if the word count of that individual was over 30% in that interval. Active participation can be observed when turn taking was over 15%, the average participation percentage for 7 participants. Table 2 presents the total number of words and turns in five
Table 2. Total number of words and turns in a five minutes interval.

<table>
<thead>
<tr>
<th>Interval</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
<th>8</th>
<th>9</th>
<th>10</th>
<th>11</th>
<th>12</th>
<th>13</th>
<th>14</th>
<th>15</th>
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<th>17</th>
<th>18</th>
<th>19</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>Total</td>
<td>907</td>
<td>863</td>
<td>828</td>
<td>761</td>
<td>830</td>
<td>700</td>
<td>804</td>
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<td>786</td>
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<td>717</td>
<td>700</td>
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<td>810</td>
<td>563</td>
</tr>
<tr>
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<td>21</td>
<td>61</td>
<td>78</td>
<td>63</td>
<td>74</td>
<td>45</td>
<td>68</td>
<td>77</td>
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<td>94</td>
<td>88</td>
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<td>46</td>
<td>63</td>
<td>94</td>
<td>109</td>
<td>563</td>
</tr>
</tbody>
</table>

Figure 3. (a) The percentages of total words and total turns in a five minutes interval, and (b) the percentages of word count of the three mechanical engineers. "Poly" is the polynomial line of best fit.

Figure 4. (a) Allan’s word count, and (b) the percentages of word count of Todd and Tommy. "Poly" is the polynomial line of best fit.

The group of mechanical engineers was the biggest group with the same background. Since they are of the same design profession, one may expect they share similar mental models, hence their conversation may display a common pattern. However, looking at the curves in Figure 3(b) did not suggest any obvious correlating patterns. Allan, the moderator, controlled the meeting at the beginning. His word contribution decreased after 15 minutes and stayed fairly consistent until the last 10 minutes, Figure 4(a). He dominated over 40% of the time during this session. Todd was the most active participant in this group according to word count and he dominated about 15% of the time. Figure 4(b) shows that the curves of Todd and Tommy are of similar shape from the 2nd interval until the 17th interval; they formed a cross-discipline sub-team which is confirmed in the Markov transition table, Table 3 in the next section.

Our hypothesis is that an individual will exhibit behavioural changes when s/he learns to collaborate with others. Looking at the team as a whole their percentages of interactions increased but their number
of words remained fairly constant. This may indicate they have learned by producing a shared mental model, which is implied by the increased interaction with each interaction requiring fewer words to communicate.

4.2. Turn Taking: First Order Markov Chain

In order to investigate the transition trends between turns we modelled the turn taking behaviour in term of first order Markov chains. The session was re-divided into six sections (15 minutes each), with the last section being 19 minutes. These longer durations provides a larger set on which to base the transitions. Sandra left during the fifth section. Figure 5 shows the transition probabilities of response to/from Allan, the leader, in the six sections. Tommy, Jack, and Todd were engaged with Allan throughout all the six sections but Sandra and Rodney were less engaged in the beginning and the end of the session. In the first section Allan received responses from the entire group except for Rodney, and he interacted with Jack most frequently. In the second section Allan formed a partnership with Todd and Tommy. In the third section Allan’s responses (to and from) were mostly evenly distributed. In the fourth section his partnership with Todd’s surfaced and this sub-team was joined by Tommy and Jack until the end.

Figures 6 and 7 present the transition probabilities of Tommy and Todd respectively. Indicated by the tallest columns in the figures, we can observe their partnership was strong except for the third section; the probability of their interactions was closer to or over 30 percent. Allan, the moderator, also interacts frequently with this sub-team.
In the first section Rodney did not interact with anyone and in the second section he began to join the conversation of the dominant sub-team (Tommy, Todd and Allan), as seen in Figure 8. Allan took notice of him and actively responded to him in the second, third and fourth sections. There is an increase interaction between Jack and Rodney in the last three sections.

Table 3 shows the transition probabilities from one member to another. Figure 9 is a graphical representation of Table 3, the x and y-axis are the team members and the z-axis represents the transition probability from a member in the y-axis to a member in the x-axis. The rows in the y-axis add up to one; and the rows in the x-axis present the absolute quantity of participation. As can be seen from Table 3, Rodney is very responsive to Tommy, Todd, and Allan. Rodney is not very responsive to Jack and Sandra. However, Jack is very responsive to Rodney. The transition probability between Allan and Tommy is the highest, followed by Tommy and Todd and Todd and Allan. The transition probability between Jack and Sandra is the lowest.
be observed from this chart Allan, Tommy, Todd, and Jack were the most active persons picking up conversations (tall columns) whereas Robert, Sarah, and Chad were the quietest (short columns).

5. FINDINGS AND DISCUSSION
In this preliminary study, by simply counting the words and the turn-taking we were able to discern some behaviour patterns changes that indicate the formation of sub-team. The number of words remained fairly constant (the trend was slightly decreasing) throughout the session but the number of turns varied with an increasing trend. This indicates they used fewer words to communicate for each interaction as the team depended more on tacit knowing as they learned to collaborate with each other. First order Markov analysis allowed us to study the relationship of members in terms of intermediate transitions of utterances. This helped us to confirm the formation of sub-teams and examine individual interaction patterns. These preliminary results indicate that this group of individuals exhibited behaviour associated with learning to collaborate.

The next step is to study the content with the FBS contributions and idea links – linkography. With the same design problem within a design session, the FBS contribution pattern of an expert is different from a novice. Therefore, we propose to measure the FBS contributions as an indicator of performance changes and learning. Kan et al. introduced the use of Shannon entropy as a potential measure of the progress of a protocol through measuring the entropy changes in a linkograph. Shannon entropy measures the information carried by a message or symbol and, here, can be related to the amount of collaboration and hence to learning to collaborate. The rate of change of entropy across sequential segments will be explored to determine whether it matches the learning rate.

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REFERENCES