

Goal Configurations and Processing Strategies as Moderators Between Instructional Design and Cognitive Load: Evidence From Hypertext-Based Instruction

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In this article, we propose some augmentations of cognitive load theory (CLT) inspired by evidence from hypertext-based instruction. In particular, we focus on the role of goal configurations and processing strategies as moderators between instructional design and cognitive load. First, we describe pivotal assumptions of CLT and conceptual shortcomings related to these assumptions. Second, we review evidence from our own hypertext-based research that gives reasons for integrating configurations of teacher goals, learner goals, as well as processing strategies into CLT. These augmentations of CLT are necessary to account for the rather weak relation between instructional design and pattern of cognitive load that shows up in the context of self-controlled learning. CLT, augmented with the notion of learner control, can now better serve as a theoretical foundation for the design of hypertext-based instruction.

Cognitive load theory (CLT) claims to be concerned with “the manner in which cognitive resources are focused and used during learning and problem solving” (Chandler & Sweller, 1991, p. 294). The theory’s main goal is to guide instructional design decisions.

COGNITIVE LOAD THEORY: PIVOTAL ASSUMPTIONS AND CONCEPTUAL SHORTCOMINGS

Pivotal Assumptions of Cognitive Load Theory

CLT is based on a distinction between a potentially unlimited long-term memory and a working memory that is severely limited with regard to the number of cognitive elements that can be simultaneously held active (Baddeley, 1992). Due to this restriction, working memory is considered to be the bot-

tleneck of learning in CLT because all information that is to be stored in long-term memory has to be processed in working memory first.

In accordance with schema theory, it is assumed that knowledge is stored in long-term memory in the form of *schemas*: “A schema categorizes elements of information according to the manner in which they will be used” (Sweller, van Merriënboer, & Paas, 1998, p. 255). With regard to problem solving, a schema is defined “as a construct that allows problem solvers to group problems into categories in which the problems in each category require similar solutions” (Cooper & Sweller, 1987, p. 348). Schemas can vary in their degree of automaticity that determines the amount of conscious effort necessary for their application. The construction and automation of schemas are considered to be the two main prerequisites of problem-solving expertise (VanLehn, 1996) and are, therefore, the most important goals of instruction according to CLT.

The relation between schemas and working memory is twofold: On the one hand, the construction of a new schema demands working memory resources as it requires one to simultaneously process all information units that are to be integrated into that schema. On the other hand, once a schema exists, it allows one to “increase the amount of information

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that can be held in working memory by chunking individual elements into a single element” (Sweller, 1994, p. 299). Therefore, schemas enable learners to treat arbitrarily complex or sophisticated information as a single working memory unit. Furthermore, schema automation frees additional cognitive resources by allowing the application of schemas without conscious effort. The availability of a large number of highly automated schemas enables domain experts to engage in complex and demanding cognitive processes compared with novices who do not possess these schemas.

According to CLT, different types of cognitive load imposed on working memory can be distinguished in instructional settings.

Intrinsic cognitive load. The first type of cognitive load refers to the number of elements that are to be integrated into a to-be-learned schema and, therefore, have to be processed in working memory simultaneously. How many cognitive elements have to be processed simultaneously for schema construction (i.e., *element interactivity*) depends on the relational complexity of the to-be-learned content and the learner’s degree of expertise (i.e., schema availability): “Intrinsic cognitive load through element interactivity is determined by an interaction between the nature of the material to-be-learned and the expertise of the learner” (Sweller et al., 1998, p. 262). An important feature of intrinsic cognitive load is that it cannot be altered by instructional design.

Other than intrinsic cognitive load that is associated with the intrinsic nature of the learning material, there may be additional cognitive load that is imposed onto working memory due to the instructional presentation of the material and the activities in which the learner is engaged. This load can be further subdivided according to whether it is beneficial for schema construction and automation or not.

Extraneous cognitive load. Extraneous or ineffective cognitive load is the result of implementing “instructional techniques that require students to engage in activities that are not directed at schema acquisition or automation” (Sweller, 1994, p. 299). A large amount of work in CLT has shown that many commonly used instructional tasks involve cognitive processes that are not helpful (or even hindering) for learning (e.g., search for a problem solution or search for referents in an explanation). With regard to hypertext-based instruction, several factors can be identified that impose extraneous cognitive load onto the learner: “The use of hypertext features requires making decisions about what to read and the order for reading information. Many learners may not be proficient computer users and must, therefore, use cognitive resources to operate the computer” (Niederhauser, Reynolds, Salmen, & Skolmoski, 2000, p. 251). Extraneous cognitive load may impede learning as it requires cognitive resources that may exceed the limits of working memory capacity. Furthermore, cognitive resources required by extrane-

ous cognitive load can no longer be devoted to mindful cognitive processes that are associated with a third type of cognitive load, namely germane cognitive load.

Germane cognitive load. When intrinsic task demands (resulting in intrinsic cognitive load) leave sufficient cognitive resources available, learners may “invest extra effort in processes that are directly relevant to learning, such as schema construction. These processes also increase cognitive load, but it is germane cognitive load that contributes to, rather than interferes with, learning” (Sweller et al., 1998, p. 264). Germane or effective cognitive load is due to beneficial cognitive processes like elaborations, abstractions, comparisons, and inferences that are encouraged by the instructional presentation. Germane cognitive load may further support schema construction by adding higher level cognitive processes to the mere simultaneous activation of elements in working memory that are to be integrated into a schema.

According to CLT’s analysis of cognitive demands in learning, two important objectives of instructional design can be derived (Sweller et al., 1998, p. 264): (a) Instructional design must aim at reducing extraneous cognitive load, and (b) it must encourage learners to invest unused resources in higher level cognitive processes that are associated with germane cognitive load. However, until now most work in CLT has focused on instructional designs intended to decrease extraneous cognitive load without considering how learners’ attention may be directed toward processes that are relevant to learning. Whereas this is a practical shortcoming of CLT research that is being tackled in more recent work (cf. Sweller et al., 1998), there are theoretical issues that concern the conceptual foundation of CLT. These conceptual shortcomings of CLT are discussed in the next section and illustrated with evidence from hypertext-based instruction.

Conceptual Shortcomings of Cognitive Load Theory

CLT research typically assumes a one-to-one mapping between instructional design and a resulting pattern of extraneous and germane cognitive load without taking into account moderating variables that interfere with this direct mapping. This assumption results from two premises: First, it is postulated that there is a predefined goal of instruction which is to foster schema construction and automation. Second, it is presupposed that a specific instructional design is strongly associated with a specific type of learner activity in the sense that it elicits, encourages, or induces this activity. No variability of learner activities as a reaction toward a specific instructional design is expected. As a result, the relation between instructional design and pattern of extraneous and germane cognitive load is seen as more or less deterministic (cf. Figure 1).

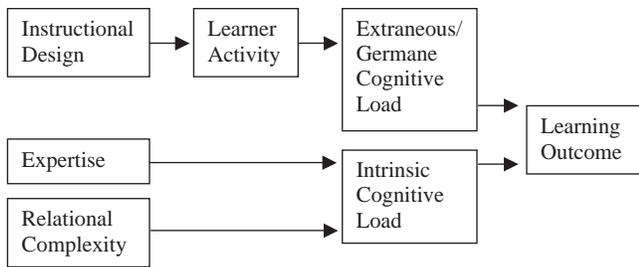


FIGURE 1 Pivotal assumptions of CLT.

In our view the relation between instructional design and cognitive load is far less deterministic than suggested by CLT because the relation is moderated by the configuration of instructional or teacher goals and by the activities of the learner. The latter can be analyzed in terms of the configuration of learner goals and the processing strategies that are deployed to accomplish these goals.

Configuration of teacher goals. As Goldman (1991) already pointed out, the same instructional design may vary with regard to its suitability for the accomplishment of different instructional goals. Instructional goals are usually introduced by a teacher who determines the criteria according to which learning outcomes are evaluated. For instance, teacher goals may include enabling literal recall (i.e., rote learning), acquisition of declarative knowledge, or transfer of procedural knowledge to isomorphic or novel problem-solving situations (i.e., near or far transfer). In addition, teacher goals may specify constraints on how much time may be invested to achieve a particular learning outcome. Because different teacher goals are not mutually exclusive, we use the term *configuration of teacher goals* to refer to relevant instructional goals in a given situation in the remainder of this article.

As a consequence of the multiplicity of instructional goals, an instructional design that has been proven useful for achieving one instructional goal, for example, the induction of well-defined problem categories, may be less suited to achieve a different instructional goal, such as the development of flexible problem-solving knowledge that easily transfers to novel problems. CLT's narrow focus on schema acquisition as the predominant instructional goal leads to the instructional recommendation to prevent learner activities that are not directly related to the acquisition of problem categories. However, these activities may be useful when the learner has to face more challenging tasks, such as solving far transfer problems.

In this line of reasoning Goldman (1991) criticized CLT for being too narrow in that it prescribes instructional procedures that are useful only for near transfer. She claimed that the "dependent measures concentrate too heavily on learning as defined by the ability to master what is presented" (p. 336). If a broader approach to learning and instruction is taken, different teacher goals (e.g., near and far transfer) have to be

considered. As a consequence, the relation between instructional design and the resulting pattern of germane and extraneous cognitive load becomes more ambiguous, especially with regard to "the specification a priori of what is extraneous, or at least less critical, and what is central to the learning task" (Goldman, 1991, p. 336).

As the configuration of teacher goals and the respective criteria for evaluating learning outcomes define the degree to which a specific learner activity imposes germane and extraneous load onto working memory, we propose to integrate the configuration of teacher goals as a theoretical variable into CLT. As a consequence, CLT would result in conditional instructional recommendations as follows: Given the Configuration *X* of teacher goals, Instructional Design *A* will yield a better performance than Instructional Design *B*.

Activities of the learner. From our point of view, even if a configuration of teacher goals has been defined, it still cannot be determined without ambiguity which pattern of cognitive load will result from a specific instructional design. Learners who are confronted with the same instructional environment (including a proposed learning task) and the same configuration of teacher goals may nevertheless differ with regard to their learner activities. This is because the configuration of learner goals and the processing strategies that are deployed to accomplish these goals only partially depend on instructional design and teacher goal configuration:

Configuration of learner goals. Learners are supposed to adopt the teacher goals conveyed through instruction. However, there may also be substantial differences between learner and teacher goals for at least two reasons. First, learner and teacher goals may differ with regard to goal parameters, such as the level of aspiration or the goal orientation (e.g., mastery vs. performance goals). Second, goals that go beyond the teacher goals conveyed through instruction may be relevant for guiding activities of the learner. In particular, their goal configuration may include transient goals that emerge during the navigation in instructional environments (cf. Hirashima, Hachiya, Kashihara, & Toyoda, 1997) or goals that are related to prospective tasks (Heise, Gerjets, & Westermann, 1997) as well as to personal interests (cf. research on seductive details, Garner, Gillingham, & White, 1989; Harp & Mayer, 1998). These additional goals suggest that there is not necessarily a direct mapping between the configuration of teacher goals and the configuration of learner goals.

Processing strategies. Furthermore, learners' processing strategies add variability to the relation between instructional design and cognitive load. A *strategy* for performing a task can be defined as "a procedure or set of procedures for achieving a higher-level goal or task. These procedures do not require conscious awareness to be called a strategy"

(Lemaire & Reder, 1999, p. 365). Bisanz and LeFevre (1990) emphasized that the term strategy should be exclusively used if a goal or task can be tackled by means of different procedures. The selection of processing strategies within the constraints provided by a specific instructional design should depend on learners' goal configurations, their levels of expertise (i.e., declarative and procedural knowledge available), and other factors like available processing resources and cost—benefit ratios for different strategies under specific circumstances (Anderson & Lebiere, 1998; Christensen-Szalanski, 1998). Thus, for a given configuration of learner goals, different processing strategies may result—depending on the selection policy—which, in turn, are associated with a specific pattern of cognitive load.

To summarize, up to now, variability with regard to the way in which learners process instructional materials has not been taken seriously into account in CLT. Instead, learner activities have been taken as a rather stable consequence of a specific instructional design. In this article, we analyze learner activities in terms of goals and strategies to account for the variability with which to-be-learned materials can be handled. According to our view, a learner who is confronted with a specific instructional design has to decide which goals he or she wants to accomplish during learning and which strategies are to be deployed to implement these goals.

Our proposals lead to an augmentation of CLT theory that is displayed in Figure 2. This augmented version of CLT is more complex than the original version; however, this increased complexity allows CLT to be extended to a larger range of instructional settings that are characterized by a high level of learner control (e.g., self-controlled, computer-based learning environments). In the next section we substantiate this claim with evidence from hypertext-based instruction.

EVIDENCE FROM HYPERTEXT-BASED INSTRUCTION

In the following section we draw on several findings from our own research. Details of the experiments discussed are reported in Gerjets, Scheiter, and Heise (2002), Gerjets, Scheiter, and Tack (2000), Scheiter, Gerjets, and Heise

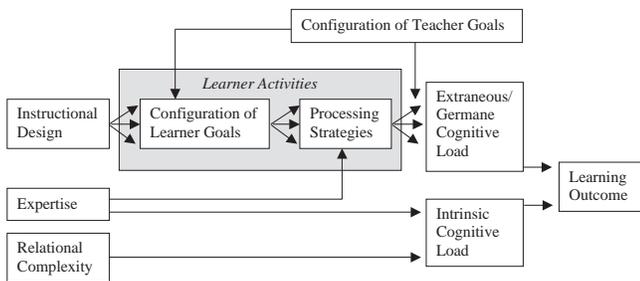


FIGURE 2 Goals and strategies as moderators between instructional design and cognitive load.

(2000), and Schorr, Gerjets, Scheiter, and Laouris (in press). Although this research did not initially aim at testing the augmented version of CLT, the findings obtained indicate that—at least in the case of hypertext-based instruction—the pattern of cognitive load resulting from a specific instructional design can be determined only when taking into account the configuration of teacher and learner goals as well as the processing strategies deployed by the learner.

Hypertext environments are network-like information structures in which fragments of information are stored in “nodes” that are interconnected by electronic hyperlinks (Conklin, 1987; Rouet & Levonen, 1996). To distinguish hypertext structures from conventional text forms, hypertext is often described as nonlinear in contrast to the linear representation and access to information in traditional text. Hypertext is “capable of being explored in different ways, with the different exploration paths producing what are essentially *multiple texts* for the same topic” (Spiro & Jehng, 1990, p. 166). Thus, the user can select information units as well as choose the point of time, the pacing, and sequence of their presentation according to his or her goals. This adaptivity implies that the same hypertext may serve a multitude of different goals that guide information utilization.

Configurations of Teacher Goals Determine the Pattern of Cognitive Load (Study 1)

To demonstrate that configurations of teacher goals determine the resulting pattern of cognitive load, it is useful to evaluate instructional designs with regard to different measures of learning outcomes. Based on an instructional manipulation introduced by Quilici and Mayer (1996), we compared elementary school children who either studied structure-emphasizing or surface-emphasizing sets of worked-out examples to acquire knowledge on different problem categories in the domain of basic arithmetic (Schorr et al., in press). Participants used a hypertext environment that contained 16 worked-out examples that illustrated four different problem categories and that were embedded in four different cover stories. In the surface-emphasizing condition, all four examples that illustrated a particular problem category shared the same cover story so that each problem category was associated with a specific cover story. Conversely, in the structure-emphasizing condition, each cover story was used to illustrate each problem category so that no problem category was associated with a specific cover story. To measure learning outcomes, 12 test problems were used that belonged to the same problem categories as the instructional examples. According to Reed (1999), two types of test problems were designed—equivalent and isomorphic problems. For each equivalent test problem there was at least one instructional example that shared the same problem category and the same cover story. Conversely, a test problem was classified

as isomorphic if its cover story was different from all instructional examples of the same problem category.

The following pattern of results was obtained: First, learners in the surface-emphasizing condition needed less time studying worked-out examples than learners in the structure-emphasizing condition. Second, participants who studied surface-emphasizing sets of worked-out examples performed better on equivalent test problems than participants who studied structure-emphasizing sets of worked-out examples. This pattern was reversed for the performance on isomorphic problems. That is, for isomorphic problems, the structure-emphasizing condition outperformed the surface-emphasizing condition.

From a cognitive load perspective, these results can be interpreted as follows. Longer learning times in the structure-emphasizing condition indicate that this condition elicits more demanding cognitive processes of identifying structural similarities among examples that differ with regard to their cover stories. These processes may be related to abstraction and schema construction and are not elicited in the surface-emphasizing condition in which cover stories are held constant within problem categories. These cognitive processes of abstraction impose additional cognitive load onto the learner which can, however, only be judged as being germane or extraneous in the light of a specific configuration of teacher goals. If the goal of instruction is to learn very fast how to solve problems equivalent to the instructional examples, these additional processes impose extra load onto the learner that is unnecessary or even harmful to accomplish this specific goal of instruction, meaning it is extraneous cognitive load. If, however, the goal of instruction is to solve isomorphic transfer problems, these additional processes are critical to ensure a good performance, meaning they impose germane cognitive load.

To put it more generally, this type of cross-interaction between instructional design conditions and different measures of learning outcomes cannot be explained by CLT without augmentations—like distinguishing different configurations of teacher goals—as proposed in our extended model.

Configurations of Learner Goals Determine the Pattern of Cognitive Load (Study 2)

One way of demonstrating that configurations of learner goals determine the resulting pattern of cognitive load is to manipulate learners' goal configuration during the interaction with a learning environment while keeping the goal of instruction constant. To investigate the impact of learners' goal configuration on extraneous and germane cognitive load, we conducted a hypertext-based experiment in which task difficulty and presence of a prospective second task were manipulated (Gerjets et al., 2002; Scheiter et al., 2000). As a starting point, all participants received three word problems to solve from the domain of combinatorics. During problem solving they could browse an example-based hypertext environment

that contained all information necessary to solve the word problems (i.e., information on problem categories, structural task features, and formulas needed). In addition, task-irrelevant but interesting information pages that were related to the cover stories of the instructional examples (i.e., topics related to attractiveness and mate choice) could be retrieved by the learner. Processing these additional information pages should result in extraneous cognitive load with regard to the given goal of instruction (i.e., solving word problems). On the contrary, processing the information pages necessary to solve the word problems should result in germane cognitive load. Participants were instructed to work as fast and as accurately as possible and could decide by themselves which information pages to retrieve.

To manipulate task difficulty participants either received easy or difficult word problems at the beginning of the experiment. Easy problems should be associated with low levels of intrinsic cognitive load, whereas difficult word problems should impose high levels of intrinsic cognitive load.

As a second independent variable we manipulated learners' goal configuration by inducing a prospective second task. Participants without a prospective second task had to solve only the three word problems, whereas participants in the condition with a prospective second task were additionally informed that they would have to work on a second task within the same hypertext environment after having finished the mathematical problem-solving task. This second task consisted of answering three questions about attractiveness and mate choice that were likewise presented briefly at the beginning of the experiment in this condition. The answers to these questions could be found in the additional information pages related to the cover stories of the instructional examples. Participants were instructed to work on the problem-solving task first and to postpone thinking about the question-answering task until they had finished the word problems. They were assured that they would have enough time afterwards to browse the hypertext environment for information relevant to this second task.

We assumed that the announcement of a to-be-postponed second task might change learners' goal configuration. That is, although the conditions with and without a prospective second task were equivalent with regard to the learning environment and the goal of instruction announced to the learner (i.e., solving the three mathematical word problems), we expected differences in extraneous cognitive load and, therefore, in problem-solving performance between these conditions. These differences are supposed to be due to learners' goal configuration during their interaction with the learning environment. The fact that the prospective second task is part of learners' goal configuration might induce goal competition if information relevant to the second task is available in the learning environment. Processing this information might result in extraneous cognitive load and cause performance impairments with respect to the goal of instruction. Furthermore, from a CLT-perspective these effects of a prospective

second task were expected to be a function of the current task difficulty: Extraneous cognitive load should be especially harmful for high levels of intrinsic cognitive load (i.e., for difficult problem-solving tasks), whereas there should be a smaller effect for less demanding problem-solving tasks.

The results showed that the presence of a prospective second task indeed impaired problem-solving performance, which yields evidence for the importance of learners' goal configuration as a determining factor for the resulting pattern of germane and extraneous cognitive load. In addition, performance impairments due to a prospective second task depended on the difficulty of the problem-solving task. However, the resulting pattern of difficulty-related performance impairments was just the opposite to what was expected under a CLT perspective: For easy word problems, the presence of a prospective second task increased the number of problem-solving errors, whereas for difficult word problems there were no differences with regard to problem-solving performance due to the prospective second task.

Although these findings (as well as Heise et al.'s, 1997, evidence for this pattern of difficulty-related performance impairments due to goal competition) contradict the predictions derived from CLT, the pattern of results is compatible with predictions derived from theories of action control that postulate a difficulty-related effort investment (Heise, Gerjets, & Westermann, 1994). These theories assume that an enhanced task difficulty leads to an increase in effort invested into the current task that, in turn, becomes less vulnerable to distractive effects. To account for difficulty-related performance impairments in CLT, the theory has to be augmented by assumptions on how cognitive resources are allocated to different learner goals (effort allocation). Up to now CLT seems to assume that there is a single learning goal and that all cognitive resources are devoted to this goal to accommodate its task demands. There may, however, be situations in which learners avoid investing all of their available cognitive resources in task accomplishment because they possess more complex goal configurations and try to accomplish different goals at the same time.

To account for our experimental findings we suggest augmenting CLT by a first-come-first-serve principle of working memory allocation: Learners working on easy word problems should be characterized by low initial levels of intrinsic cognitive load and, thus, have remaining working memory resources at their disposal. These spare resources can be claimed by either extraneous or germane cognitive load during the course of learning and problem solving. Participants in the conditions without a prospective second task may use these resources to implement rather sophisticated germane processing strategies. Conversely, participants in the conditions with a prospective second task are presented with information related to their to-be-postponed task that may impose extraneous cognitive load, and that prevents participants from using sophisticated processing strategies for the learning task. This can account for the shallow example-process-

ing strategies and the performance impairments observed in this condition. However, participants working on difficult word problems should be characterized by high initial levels of intrinsic cognitive load and, thus, may not have enough working memory resources available for either implementing sophisticated germane processing strategies or being distracted.

Note that this first-come-first-serve principle of working memory allocation is not part of CLT but rather an additional assumption that is tied to the issue of learners' goal configurations and their impact on the pattern of cognitive load. Without including these augmentations—that concern the configuration of learner goals and the allocation of effort—findings on difficulty-related performance impairments due to the presence of a prospective second task would be hard to explain by CLT.

Processing Strategies Determine Pattern of Cognitive Load (Studies 3 and 4)

In the preceding sections we argued that configurations of teacher goals and learner goals are important variables that determine the pattern of cognitive load resulting from a specific instructional design. In this section we discuss evidence showing that the pattern of cognitive load is additionally determined by the information-processing strategies deployed by learners when interacting with a specific instructional design. In CLT it is usually assumed that a specific instructional design elicits or induces a specific type of learner activity. No variability of learner activities as a reaction toward a specific instructional design is expected, thus yielding a rather deterministic relation between instructional design and cognitive load pattern. We review two studies from our own hyper-text-based research to show that processing strategies may act as important moderators between instructional design and cognitive load—especially in self-controlled learning situations. In these studies we compared different instructional designs and different strategies of using the information provided in these designs concurrently, yielding that processing strategies may be better predictors of learning outcomes than are features of the learning environments.

Provision versus utilization of worked-out examples (Study 3). Research over the last 15 years in the domain of learning and problem solving has demonstrated that worked-out examples play an important role in knowledge acquisition (cf. Atkinson, Derry, Renkl, & Wortham, 2000). Although many studies have focused on contrasting the exposure to worked-out examples with alternative instructional devices—like practice problems or abstract instruction—there is also work demonstrating that students' strategies of processing worked-out examples are a critical factor with regard to learning outcome. That is, the mere availability

of instructional examples is not sufficient to guarantee the acquisition of schemas that are helpful for later problem solving. Rather, a profitable utilization of worked-out examples has to be ensured, which may comprise example comparisons and example elaborations.

Example comparisons: Learners have to compare different examples to notice structural features that differ among problem categories and that are shared by all problems within a category. Contrasting examples within and among problem categories with regard to their differences and similarities may allow the learner to identify the relevant features of worked-out examples and to avoid confusions by examples' surface features (Cummins, 1992; Quilici & Mayer, 1996).

Example elaborations: Learners have to avoid shallow and frugal example processing that may result from illusions of understanding when learning from worked-out examples (Renkl, 1999). Instead, learners have to make inferences about the structure and solution of worked-out examples that go beyond the information presented in the example (e.g., by relating examples to more abstract information; Chi, Bassok, Lewis, Reimann, & Glaser, 1989; Pirolli & Recker, 1994; Renkl, 1997).

Several suggestions have been made regarding how to design example-based instructional settings to foster a profitable utilization of worked-out examples, including structuring examples according to their subgoals (Catrambone, 1998), presenting multiple examples with different surface features (Quilici & Mayer, 1996), or presenting incomplete examples that have to be completed by learners (van Merriënboer & de Croock, 1992).

The aspect of adopting suitable strategies of example utilization gains increasing importance the more the control of the learning process is left up to the learner. To determine the relative impact of example-processing strategies and instructional design in the context of self-controlled learning, we conducted a hypertext-based study in the domain of combinatorics (Gerjets et al., 2000). On the one hand, it can be expected that hypertext-environments are especially suited to foster profitable processes of comparing examples within and among problem categories and of relating examples to abstract information. This should be the case because the linking capabilities of nonlinear hypertext allow the learner to flexibly select examples and abstract information pages as well as to determine the sequence of information presentation. On the other hand, learning with nonlinear hypertext may impose additional extraneous load onto the learner due to the decisions necessary for selecting and sequencing information (Niederhauser et al., 2000).

In the hypertext environment used for experimentation a short introduction to the domain of combinatorics was first presented, and participants were instructed that they would have to solve three probability word problems following a self-paced learning phase. In this learning phase partici-

pants could at least retrieve abstract explanations on six problem categories (with their associated formulas). In the test phase instructional information of the learning phase was no longer available.

As a first independent variable, the availability of worked-out examples that illustrated the abstract information on the problem categories and their associated solution procedure during the learning phase was manipulated. Participants had the opportunity to either receive no worked-out examples, one worked-out example, or three worked-out examples with different surface features for each problem category. All examples in the one-example condition were couched in the same cover story. As a second independent variable, participants with low and high domain-specific prior knowledge were distinguished by means of a multiple-choice questionnaire.

The results showed that prior knowledge, but not instructional design, had a significant impact on later problem-solving performance. That is, the participants did not differ in performance as a function of whether they could retrieve zero, one, or three examples. Thus, at first sight it seems that the instructional examples used in this learning environment were not helpful to foster learning outcomes. However, although the mere provision of one or three examples was obviously not sufficient to improve learning, more detailed analyses of example-processing strategies revealed that mindfully processing these examples was strongly predictive for problem-solving performance. In particular, participants who processed each example more than once in the one-example condition and participants who retrieved more than one example per problem category in the three-examples condition showed better performance than participants who displayed a less intensive example-processing behavior. These two example-processing strategies could be demonstrated to be equally effective in improving performance. There was no relation between example-processing strategies and prior knowledge.

One implication of our findings that has straightforward consequences for CLT is that learners' strategies of using instructional environments may be more important predictors of learning outcomes than instructional design itself. Learners provided with a superior instructional environment (e.g., with multiple worked-out examples) may perform better (by increasing germane cognitive load) or even worse (by suffering from extraneous cognitive load) compared with learners provided with a more inferior instructional environment (e.g., without worked-out examples). Which outcome can be expected depends on whether learners make use of the opportunities to engage in germane processing provided by the instructional design (e.g., by comparing examples within and between problem categories) or not. In case they do not take a chance on these hypertext capabilities, the disadvantages of additional control demands for handling the enriched instructional environment may outweigh the benefits (e.g., decisions related to the selection and sequencing of informa-

tion pages). These additional demands mainly increase extraneous cognitive load and, therefore, impair schema acquisition.

Strategic adaptation to time pressure (Study 4). The aforementioned study demonstrates that improving instructional design (e.g., by providing multiple examples) does not automatically result in improved performance unless it is accompanied by profitable strategies of information utilization. Moreover, as the following study indicates, providing unfavorable instructional conditions from a CLT perspective—such as time pressure—can be rather successfully compensated for by strategic adaptations.

To study the effects of time pressure on performance we compared the six aforementioned instructional conditions of Study 3 (Prior Knowledge \times Number of Worked-Out Examples per Problem Category) with six equivalent conditions in which participants' learning time was severely restricted to about 60% of the mean learning time in the respective condition without temporal limitations (Gerjets et al., 2000). During problem solving there were no time limits. From a CLT perspective it can be postulated that time pressure mainly decreases germane cognitive load by hindering learners from engaging in elaborate processing strategies that make use of helpful additional instructional materials like worked-out examples. The theory would predict that participants learning under time pressure need to skip instructional materials that would under normal circumstances allow for a mindful cognitive processing with a high degree of germane cognitive load. Skipping these materials under time pressure should, therefore, result in a reduced learning outcome.

Contrary to the aforementioned prediction, our findings showed that even a strong reduction of learning time did not necessarily impede learning outcomes. This indicates that participants may have been able to adapt strategically to time pressure. Analyzing strategies of information utilization under time pressure demonstrated that these learners refrained from elaborate example processing and invested their time in processing abstract information without showing deterioration in performance. From a CLT perspective there are two possible explanations to account for these results: Learners under time pressure either process abstract information more thoroughly (i.e., allocate more cognitive resources on processing this information) and thereby increase germane cognitive load, or they may reduce extraneous cognitive load by ignoring most of the additional information provided in the hypertext environment and thereby avoiding extraneous decision processes. Thus, although time pressure should be a very unfavorable instructional condition from a cognitive load perspective, strategic adaptation resulting in an improved pattern of cognitive load may enable learners to cope even with strong time pressure.

SUMMARY

CLT provides a useful framework for analyzing instructional design features with regard to their suitability for supporting processes of schema construction and automation. This analysis is based on determining the pattern of intrinsic, germane, and extraneous cognitive load associated with a specific instructional design at a particular level of expertise. However, to extend the range of successful applications of CLT to instructional settings that are characterized by a high level of learner control (e.g., self-controlled learning in hypertext environments), we recommend that CLT be augmented by several variables that moderate the relation between instructional design and the resulting pattern of cognitive load. We outlined an extended CLT model that specifies these variables and their interrelations. Evidence from our own hypertext-based research indicates that configurations of teacher goals, configurations of learner goals, and learners' processing strategies may be important moderating variables that should be incorporated into CLT.

ACKNOWLEDGMENT

We thank Richard Catrambone, Elke Heise, and Friedrich Hesse for helpful comments on an earlier version of this article.

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