

The Laptop and the Lecture: The Effects of Multitasking in Learning Environments

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ABSTRACT

THE EFFECTS OF MULTITASKING IN THE CLASSROOM were investigated in students in an upper level Communications course. Two groups of students heard the same exact lecture and tested immediately following the lecture. One group of students was allowed to use their laptops to engage in browsing, search, and/or social computing behaviors during the lecture. Students in the second condition were asked to keep their laptops closed for the duration of the lecture. Students in the open laptop condition suffered decrements on traditional measures of memory for lecture content. A second experiment replicated the results of the first. Data were further analyzed by “browsing style.” Results are discussed from Lang’s Limited Process Capacity model in an attempt to better understand the mechanisms involved in the decrement. (*Keywords: multitasking, divided attention, technology, education, limited capacity model*)

INTRODUCTION

THE UBIQUITY, PERVASIVENESS AND MOBILITY of new technologies encourage a simultaneity of activities that goes beyond anything our culture has heretofore ever known. Indeed, the ability to engage in multiple tasks concurrently seems to be the very essence or core motivation for the development of such technologies. Yet there is a long tradition of psychological and media communication research that indicates that our ability to engage in simultaneous task is, at best, limited (Fisch, 2000; Lang, 2000), and at worst, virtually impossible (James, 1890; Woodsworth, 1921; Broadbent, 1958).

In the context of learning, the implementation of wireless technology, in whatever form that might take, introduces additional visual and/or auditory information, above and beyond the visual and auditory information presented by the instructor. Depending on the boundaries and intent for using these tools in the classroom setting, a student might engage in myriad computing activities, from synchronous and asynchronous social computing, to note taking, to Web browsing. Part, all, or none of the activity may be related to the lecture topic at hand. Of course, distraction in the lecture hall or classroom is nothing new; note passing, doodling, talking, completing other class assignments, and even taking notes on the current lecture are all familiar forms of low-tech distraction. However, mobile devices and wireless access in the classroom have the potential to bring distraction to new heights; especially as the study of their effects and benefits is in its relative infancy and schools and Universities grapple with issues concerning boundary setting and high-tech classroom etiquette.

The current study was part of a larger program of research designed to investigate the effects of wireless computing in collaborative learning environments. Students in two very differently structure classroom environments were given laptops with wireless LAN access to use throughout the course of the semester. With their consent, all computing activity was tracked around the clock including E-mail, Web browsing, CHAT, and Instant Messaging (Note: The *content* of

social communication was not recorded). Both direct (proxy logs and focus group interviews) and indirect (classroom observation) evidence indicated to us that students were engaged in computing activities that were often unrelated to the immediate class lecture and tasks. Thus we decided to investigate this multitasking behavior more systematically.

Our hypotheses were developed from the extant literature in both cognitive psychology and mediated learning research. In cognitive psychology there is a long tradition of research that has focused on dividing attention between simultaneously occurring tasks. These experiments have formed the basis for methodological and theoretical developments in nearly every sub field of cognitive psychology including learning, memory, perception, and of course, attention. Perhaps most notable is Broadbent's theory of selective attention (1958, 1970). Based on his dichotic listening experiments that required subjects to shadow speech messages in one ear, while ignoring the messages in the other ear, Broadbent concluded that little, if any, content from the nonattended ear is remembered. From these observations, Broadbent proposed that there is a limited processing channel that information is filtered through from a sensory processing stage on its way to a short-term memory store or buffer. From here, information may be processed further before being transmitted into a longer-term memory store. When this channel becomes overloaded, such as in dichotic listening experiments, some of the information is filtered out, while other information is selected for further processing. The filtering mechanism selects inputs based on different physical cues from the stimulus input, such as location in space, and/or frequency. Since then, psychologists interested in information processing issues such as learning, memory, skill development and retention, processing limitations, and human factors have been investigation the effects of multitasking.

The classic paradigm involves having subjects perform a primary task in which some response is required, while simultaneously monitoring a secondary task for specific information or changes. For example, participants might be asked to learn a list of words pre-

sented visually, while listening for the occurrence of certain digit strings presented through an auditory channel. Participants might then be tested for their memory of the word list. Many different variations have been investigated including different modalities, the same modalities, task difficulty, the effects of practice, the effects of either the primary or secondary task on performance, and testing during encoding or retrieval (Baddeley, Lewis, Eldridge, & Thomson, 1984; Naveh-Benjamin, Craik, Perretta, & Toney, 2000). Johnson, Greenberg, Fisher, & Martin (1970); Spelke, Hirst, & Neisser (1976). Almost without exception performance on one or both task suffers a decrement as a direct result of having to perform the two tasks simultaneously.

Explanations for the performance decrement most typically involve some discussion of limitations in the amount of information that can either be selectively attended to, processed, or encoded such that there no longer exist enough overlap at the time of retrieval for the subject to recognize or recall the to-be-remembered information. (Craik & Lockhart, 1976; Broadbent, 1958, 1970; Tulving & Thompson, 1973). Often the specific impact of the variable(s) of interest in a particular study, for example, the automaticity of the processes studied and/or the depth qualify these interpretations or the processing required by the tasks performed. However, the finding of a performance decrement under divided attention conditions is so robust as to consider it a guiding theoretical principle in these various fields of attention, learning and memory.

Research in another area, which also builds upon the notion that there is a fixed amount of cognitive resources upon which the processor may draw, focuses on delineating the specific sub processes involved and how each may be compromised at every step in the process. Annie Lang's Limited Capacity Model (LCM) (2000), relies heavily upon cognitive constructs as she applies it to mediated learning contexts. The model's predictive power under various conditions of encoding, storage and retrieval is nothing short of elegant and provides an outlet for the discussion of some of the more interesting findings from the current study. Thus, while the classic experiments on dichotic listening tasks and selective attention theory provided the

theoretical (and to some extent methodological) rational for this work, LCM enables us to explore our findings that go beyond the basic predictions of divided attention theories, and to postulate what mechanisms may be involved.

LIMITED CAPACITY MODEL

The model outlines the stepwise progression of the cognitive processes involved in processing information. Conscious and unconscious mechanisms determine what information is selected for encoding. Once in short term, or working memory, previous knowledge is activated and linked with relevant aspects of the new incoming information. Memory for the new information is created through associations between the new and existing knowledge, or by recurrently activating and linking the bits of new information over time. The number of related associations between new and existing information determines memory strength.

Proof by disproof is the work of science, and this models' strength comes not in its ability to prove the existence of these separate sub processes, but in its power to predict failures of memory. Beginning first with the assumption that there is a limited pool of resources from which the individual-as-processor may channel into these sub processes, memory may be compromised as a result of a breakdown at any point in this process. Maintaining a balance between what is required by the message, and the distribution of already limited resources to process that information thoroughly, is the juggling act of the information processor. All breakdowns reflect some misattribution of resources to the task at hand. Disproportionate allocation of resources may result from conscious and intentional mechanisms inherent to the individual, or attributes intrinsic to the information or message.

Lang's model has been applied most typically to the television viewing situation, predicting what viewers will remember following various manipulations in structure and content changes in the medium as well as inducing different motivation or expertise in the viewer. The extension of this model to hypertext media seems natural given

the primary receptive modality; the continual flow of structural and content changes, and the similarly varied needs and goals that the user brings to the situation. The multitasking context of the present work has ecological and theoretical relevance; burdening the user with additional information increases the cognitive load requirement. Successful information retrieval/learning in this kind of situation depends, then, on extremely efficacious resource allocation to both mediums. The results we report here indicate that the effects of multitasking are not simply main effects of condition. They must be qualified by other mediating factors. We believe Lang's model provides explanatory power for these interactive effects and helps lay the groundwork for the systematic investigation of them.

Our hypotheses were as follows:

H1: Students in the open laptop condition would perform significantly poorer on immediate measures of memory for the lecture material.

H2: Similar to other findings from our lab, (Grace-Martin & Gay, 2001), the memory decrement observed would not result from the relatedness of the content viewed in the secondary task (Laptop use) to the primary task (lecture information). In other words, content relevance would not contribute significantly to the variance observed in the main effect above.

METHOD

SUBJECTS

PARTICIPANTS IN THE INITIAL STUDY were 44 students in an upper level Communications course at a prominent university in the Northeast. As part of another project funded by Intel, these students were issued Dell laptops (Dell Latitude Cpt) to use throughout the semester in an effort to study the impact of wireless network technology in collaborative learning spaces. Across campus a series of wireless transmitters (access points) provided the infrastructure for the wireless LAN network. Students were encouraged to conduct their computing activities through a proxy server. In this way we were able to capture and store all tool use (E-mail, discussion board participation, URL

visits, etc) throughout the semester. All students were fully informed as to what their participation would entail and could chose at any time to not go through our proxy server.

The majority of students were Communication majors, although there were also a few Computer Science and Arts and Design students enrolled. 22 males and 22 females were enrolled in the course.

PROCEDURE

Throughout the course students were encouraged to use their laptops in the class as a supplement to the lecture, discussion and lab activities. Generally speaking, students were given the responsibility to monitor their own computing activities during the lecture. It became apparent that while students used their laptops during lecture to explore lecture topics in greater detail on the Internet as well as the library databases, they were also engaging in other forms of computing, such as unrelated browsing, e-mail, and synchronous forms of communication such as CHAT and IM. As stated earlier, it was this high-tech “doodling” that provided the impetus for the current study.

During one lecture, students were randomly assigned to one of two conditions; half of the students left the classroom to take part in a lab exercise in a neighboring classroom. The remaining half of the students were exposed to a typical lecture, and encouraged to use their laptops as usual during the lecture. When the lecture ended, these students switched classrooms with the other half of the class, who then went into the lecture hall and heard the identical lecture. The only difference between the two groups was that during the second repetition of the lecture, students were told to close their laptops. Both groups of students were tested immediately following the lecture. The lecture this day did not differ perceptibly in structure or delivery from any other lecture thus far.

The surprise quiz consisted of 20 questions on the lecture content. Half of the questions were multiple choice (recognition) questions, while the other half were short answer (recall) questions. All students finished the test in approximately ten minutes, after which they were debriefed and thanked for their participation.

The replication study took place two months later. Students who served as controls in the first experiment (Closed laptop condition) participated as experimental subjects in the replication study (Open laptop condition) and vice versa.

RESULTS

TESTS WERE CORRECTED and given three scores; a total score which included the percent correct out of the total number of questions, a recall score and a recognition score, which were the proportion correct out of the total of ten like questions. The data are first reported for the initial study, and then the replication study.

The initial data were first analyzed using a one-way ANOVA with condition entered as a between subjects variable. The results of this analysis revealed a significant effect of condition on total and recall test score measures, with students in the open laptop condition performing significantly poorer than those in the closed laptop condition ($F(1,43) = 4.42, p < .04$; $F(1,43) = 5.00, p < .03$, respectively). Differences between the two groups on recognition scores approached significance, $F(1,43) = 3.45, p < .07$. Figure 1 graphically represents the differences between these two conditions.

The replication data were scored and analyzed in exactly the same manner. Again, the results of this analysis revealed a significant effect of condition on total and recall test scores in the same direction ($F(1,20) = 10.70, p < .004$; $F(1,20) = 6.13, p < .02$ respectively). Again the difference between the open and closed laptop conditions neared significance on recognition scores, $F(1,20) = 2.80, p < .11$.

To ensure that having participated in the experiment twice (even though subjects were unaware of their participation and group assignment was counterbalanced between the initial and replication study), an additional analyses with Experiment (Initial or Replication) was entered into the model. No significant Condition X Experiment interaction was found.

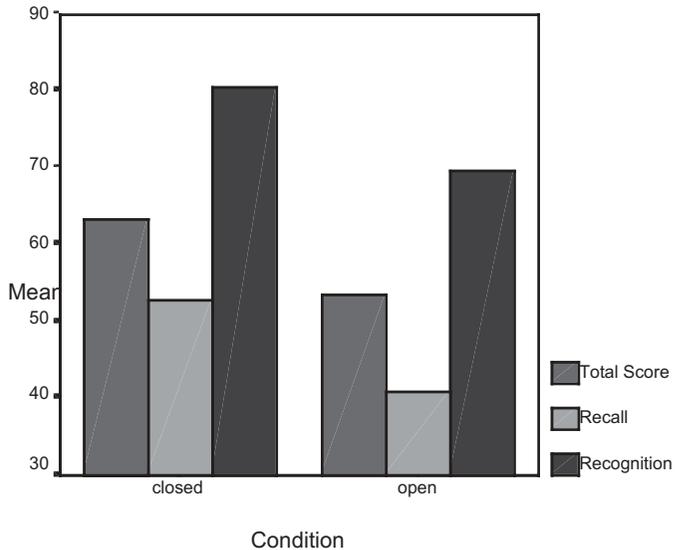


Figure 1. Mean total, recall, and recognition scores for open and closed laptop conditions

While these data are interesting and extend the historical theoretical precedence to a more applied setting such as the classroom, they are hardly surprising. Our intent here was to explore these more predictable results at a deeper level, and perhaps suggest possible mechanisms for their occurrence in this context. The proxy log data allowed us to examine the content of what students were browsing during the lectures and how that might have impacted their performance on the subsequent memory test. We were able to extract their on-line browsing behavior for those class periods and code them according to whether the content was class related or unrelated depending on the lecture content. Content was coded as related if the URL was related to the lecture topic, or was 1 of 3 URL's recommended during the lecture. Non-related content was all other URL's, including Web mail, entertainment sites, E-commerce sites, News sites, and business sites. From this log data we were also able to calculate the amount of time spent on class related and unrelated pages, the overall amount of time spent on-line, and the amount of time spent

per page. Here, we are obviously only interested in those students serving in the open laptop condition.

We first calculated the proportion of time each student spent on class related and unrelated sites. This was simply their overall times spent online during those class periods divided by the amount of time spent browsing related and unrelated pages. When these totals were correlated with later test scores, an interesting, and surprising inverse relationship emerged; for recall scores, the more time spent browsing class related pages resulted in lower recall scores ($r = -.516, p. < .02$), and conversely, when students spent more time browsing class unrelated pages they did better on recall questions ($r = .510, p. < .03$).

In other work we have found relationships between browsing efficiency and class performance, (Grace-Martin & Gay, 2001). We thought that the above results might reflect some inherent differences in browsing “styles.” To explore this relationship further, we classified students as primarily “ontaskers” or “offtaskers” if the proportion of their time spent online was at least 50% on or off task. Thus, a person who spent 65% of the class period browsing class related pages he would be classified as an “ontasker.” If, on the other hand a student spent only 48% of their time on class related pages, she was classified as an “offtasker.” Analysis revealed significant differences in the time spent on and off task by these two groups, with ontaskers spending significantly more time on task than offtaskers, and conversely, offtaskers spending significantly more time off task than ontaskers ($F(1,17) = 6.64, p < .02$; $F(1,17) = 23.44, p < .01$, respectively). Thus it appears that this subject classification scheme did accurately discriminate between the two groups. By classifying subjects in this way we could investigate differences between these two groups when engaged in both class related and unrelated activities.

The mean number of minutes spent on class related and unrelated pages were calculated for each student by dividing the number of related and unrelated pages by the number of minutes spent on each kind of page. An ANOVA yielded a significant main effect of task classification for the mean number of minutes spent on class unrelated pages $F(1,14) = 11.17, p < .005$. Students classified as “ontaskers”

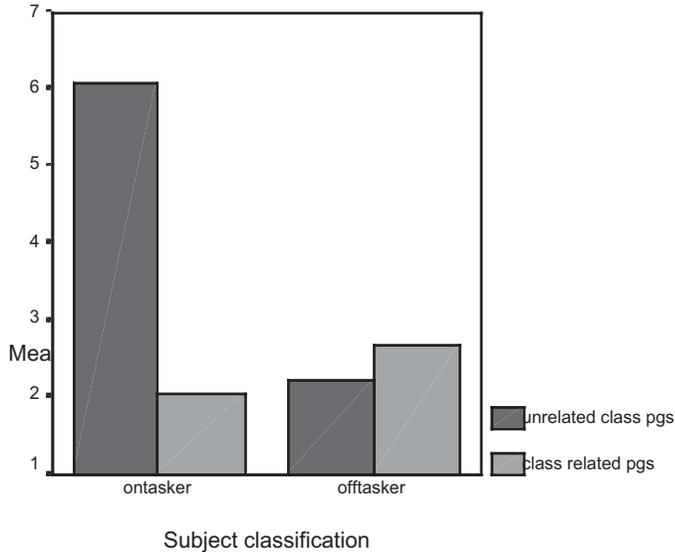


Figure 2. Average number of minutes spent on class related and class unrelated pages by “browsers” and “seekers”

spent significantly more time on class unrelated pages than “offtaskers.” When students that spent the majority of their time *on* task, went *off* task they spent an inordinate amount of time on those class unrelated pages. Those students that spent the majority of their time off task during the entire lecture spent an equivalent amount of time on class related and unrelated pages. (see Figure 2). As it turns out, “ontaskers” did worse on all three measures of performance, and significantly so on their total scores $F(1,17) = 4.85, p < .04$. Figure 3 graphically depicts these differences. Thus, it appears that the negative correlation between the proportion of time spent on class related sites and performance is mediated by one’s ability to monitor or balance their browsing behaviors.

Since performance appears to be based not on relevance but the proportion of time spent away from the primary task, “ontaskers” and “offtaskers” as labels, seem a misnomer and potentially confusing. Henceforth these groups will be distinguished as “browsers” (formally offtaskers), and “seekers” (formally ontaskers).

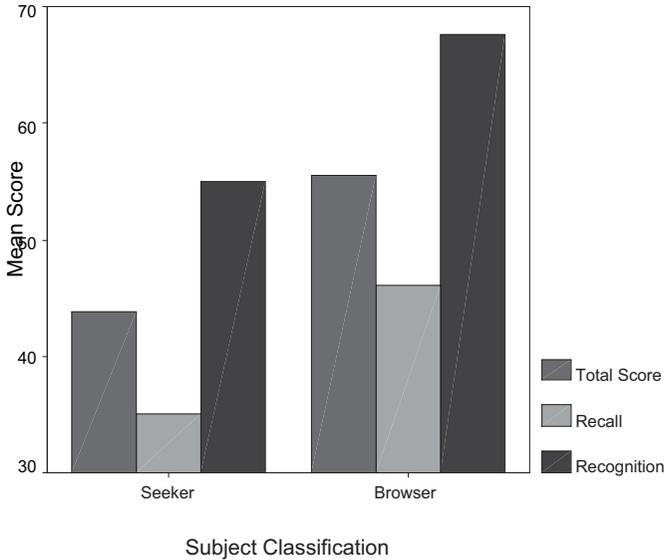


Figure 3. Mean total, recall, and recognition scores for “browsers” and “seekers”

This analysis led us to wonder if students in the open laptop condition who engaged in more superficial browsing might still be able to process the multiple inputs in much the same way as students in the closed laptop condition. Post hoc analyses of ten randomly chosen subjects from the closed laptop condition, browsers (in the open laptop condition), and seekers (recall, that browsers and seekers are both in the open laptop condition), indicated no significant differences between browsers and the randomly chosen control subjects on any of the memory measures. However, post hoc tests revealed that seekers differed significantly from controls as well as their counterparts in the open laptop condition on total score, ($F(1,17) = 2.88$, $p < .04$), and approached significant differences on recognition scores ($F(1,17) = 1.98$, $p < .13$). Partial eta squared analysis revealed a moderate effect size of .21 for total score and .14 for recognition. Figure 4 illustrates these differences. Thus, although these tests were not considered a priori, it appears that the differences between the two groups is moderately related to the level at which they were able to

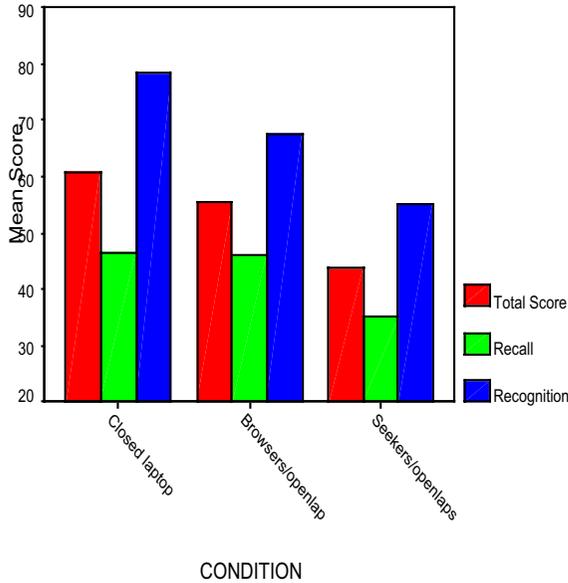


Figure 4. Mean total, recall, and recognition scores for “browsers,” “seekers,” and ten randomly chosen subjects in the closed laptop condition

process the information, as a direct result of their browsing tendencies. However, given the size of the sample, it is still important that these variables be considered systematically.

DISCUSSION

THE WORK HERE explored the effects of engaging in multiple tasks simultaneously on traditional outcome measures of performance. While methodologically the procedures employed in the present study differ somewhat from those of the classic divided attention paradigm, the essence of those procedures has been preserved, and the resulting performance decrement obtained. In two studies, students performing multiple tasks performed significantly poorer on immediate measures of memory for the to-be-learned content.

In follow-up analysis we discovered that page-relevant content did not predict better performance, and spending the majority of class time on class related content did not result in better test performance. This suggests strongly that memory decrement in multitasking situations is the result of the proportion of time drawn off task. Grace-Martin & Gay, (2001) similarly found that longer browsing sessions throughout the course of the semester resulted in lower overall class performance, and that many and shorter browsing sessions during a class period, irrespective of content, led to higher class grades. The sustained distraction, regardless of content relevance appears to be the nemesis of the multitasker; if one is adroit at staccato-like browsing, processing multiple inputs simultaneously may not suffer to the same extent

While resource allocation offers an explanation at a global level, at a mechanistic level the question becomes: What about this browsing style contributes to its' benefit? But first, perhaps the better question is what variable(s) predict this kind of browsing? One candidate we suspect is the page content. Although we no longer have access to these files (for ethical and privacy reasons), content is really the only variable that could not be controlled for, and thus the only thing that could have differed for browsers and seekers. Content would seem a likely culprit for introducing differences in the conscious mechanisms involved in selective attention. It may be that different interest levels motivated the online information sought by seekers. A recent finding in our lab (Lee, Stefanone, & Gay, 2002) indicated that topic interest is directly related to the number of sources and page changes that a user will attempt to access. Increased interest in a site might increase the number of page changes, novelty, and hence cognitive load for the time spent on class unrelated sites. Browsers appear not to have been 'pulled in' by the sites they visited, allowing them to allocate their resources more equally between the two inputs, minimizing the potentially distracting effects of the laptop.

These results are also somewhat reminiscent of the findings from incidental learning paradigms. In these experiments participants are oriented to different memory tasks by instructions to either process

the information with the expectation of a later memory test, or to simply attend to and monitor stimulus information in different ways (e.g., crossing out vowels, or copying words, Tressalt & Mayznor, 1960). Compared with intentional instructions to learn, participants given no instructions to remember do significantly poorer on the follow-up memory test (See Postman, 1964 for a full review of the literature). The explanation for the better performance of the intentional learning group is again related to the kinds of processing activities that subjects engage in during learning; with explicit instructions to learn subjects process information in more elaborative, semantically relevant ways. However, incidental learning performance can be enhanced without explicit orienting instructions to learn if subjects are required to categorize words (Mandler, 1967), make judgments about word meanings ((Tressalt & Mayznor, 1960), or are presented with associative or other-wise meaningful word pairs (Johnson & Jenkins, 1971), (or in other words, process the information semantically). Orienting instructions produce a shift in what Lang and others refer to as controlled selection processes. Essentially these are the conscious goals or intentions of the subject. Since students in the class had been encouraged to use their laptops during the lecture throughout the semester, instructions to close them may have produced a shift in these controlled selection processes and hence resource allocation. However, since the post hoc analyses revealed that most of the variance is explained by only a proportion of the students in the open laptop condition (seekers), this conjecture seems less tenable.

Though not without qualification the resulting poorer performance of some students in the multitasking condition begs the question if technology in the classroom is the panacea that some purport. While students were obviously distracted by having access to the Internet, e-mail, IM, and browsing as evidenced by their performance on traditional tests of memory, their performance in the class overall does not reflect this same disruption. The average final grade for the class was a strong B+, and students had been multitasking in class since the beginning of the semester. The structure of the class was nontra-

ditional, highly interactive and dynamic, and students were encouraged to use their laptops in class to supplement the course lecture. Had the class been more traditional and grades determined by conventional tests of memory the outcome may have been different. Thus, as others have argued (Soloway, Guzdial, & Hay, 1994; Soloway, Krajcik, Blumenfield, & Marx, 1996), statements regarding the advantages or disadvantages of technologies in learning environments need to be qualified by such things as class structure, dynamics, measure of learning, and the like.

While class structure may have been responsible in part for the lack of an observed overall effect on grades, there is also the possibility that over time, students became increasingly adept at multitasking in the classroom setting. Early evidence for practice effects on simultaneous directed activity has been demonstrated (Spelke, Hirst, & Neisser, 1976). The results of the current study suggest that enhanced browsing efficiency might be used as an index of a facilitation effect of time or practice. If students can become “better browsers”, or at the very least become more facile at self-monitoring their browsing behavior, the typical decrement found under multitasking conditions might be negated. The task before us is to develop a taxonomy of the situational and individual variables associated with performance decrement, and identify the cognitive/behavioral indicia that contribute to the effect. Such information could contribute to the development of curriculum and measures of learning perhaps more appropriate in technology enhanced classrooms, as well inform development of training or exercises, and/or devices that help the user to redirect or monitor attention in multitasking situations.

Finally, these results clearly indicate the need for setting boundaries and establishing “tech-etiquette” for using wireless technologies in the classroom. High-tech doodling for some students can defeat the purpose of using them in the first place.

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