

The role of Graphics in Video Lectures

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Abstract

With the increase in online course use (Allen & Seaman, 2017), there is an increasing need to determine the most effective (i.e., the most conducive for learning) way to present lectures online (e.g., video lectures). Lecture graphics that are interesting but extraneous to the content (e.g., a celebrity), have been shown to impair comprehension of the material, likely resulting from an increase in cognitive load. In this study, the use of graphics on the slides of an online psychology lecture was manipulated to determine the extent to which images can improve (or impair) comprehension as well as the effect it may have on intentional and unintentional mindwandering. Across our two experiments, we demonstrate no differences across conditions (i.e., unnecessary graphics, relevant graphics, no graphics) in overall comprehension and limited differences in mind wandering behavior.

Keywords: unnecessary graphics, video lectures, mind wandering, inattention, affect

Online learning and multimedia learning

There has been a growing use of online courses in post-secondary education in the last 20 years. Colleges in the United States reported an increase in the proportion of students enrolled in an online course from 10% in 2002 to 30% in 2015 (Allen & Seaman, 2017). In Canada, the same pattern of results is apparent with 2/3 of post-secondary institutions reporting an increase in online course enrollment from the 2015-16 school year to the 2017-18 school year, half of which report an increase of 10% or more (Canadian Digital Learning Research Association, 2019). Given the rising use, there is a growing need for researchers to study these environments and determine ways to optimize learning.

Online learning environments rely heavily on video lectures as a means of communicating course content (e.g., Gorissen, van Bruggen & Jochems, 2012). Video lectures can include audio, images and text making them a form of multimedia. Research investigating multimedia learning has a long history (Mayer, Heiser & Lonn, 2001; Mayer, 1997; Mayer & Moreno, 2003; Najjar, 1996; Moreno & Mayer, 1999) and has clearly demonstrated that the decisions one makes about the combinations of media and their structure can have notable effects on learning (Mayer, 2005; Mayer, 2009). For example, Mayer and Anderson (1991) found that when presenting learners with audio of how a bike pump works, followed by an animation showing how the pump works, learners had poorer learning outcomes compared to those who were presented the audio and animation simultaneously. In the present investigation we continue this general line of inquiry by examining, in a postsecondary video lecture, the influence of the presence of graphics in lecture slides. Specifically, we investigate how graphics that are unnecessary with respect to the lecture material might affect comprehension and attentional engagement with the lecture material.

Cognitive load theory and unnecessary graphics

The dominant framework for understanding multimedia learning is cognitive load theory (Mayer, 2009; Sweller, 2010). According to this theory learners have a limited capacity for information processing, and therefore can experience overload, thus impairing learning (Mayer, 2005; Mayer, 2009; Sweller, 2010). Within this theory, learners have specific channels for processing information, one is auditory, and one is visual (Mayer, 2005). When attending to the information presented in either of these channels, the sensory information is sent to the working memory system to be organized into a coherent representation of what is being taught. When in working memory, this information can get further organized and integrated with information in long-term memory, allowing for deeper learning. That said, the limits on working memory capacity restrict how much information can be processed at once. If the load becomes too great, from too much input from the information channels, this prevents further processing, therefore impairing learning. Consequently, decisions about what forms of multimedia to use need to consider how it impacts the load imposed on the learner.

One common combination of media used in post-secondary lectures is merging audio or text with graphics. Research investigating the use of graphics on comprehension has found that in some cases (e.g., when the graphics are unnecessary), the addition of graphics can have a negative impact (Harp & Mayer, 1997; Mayer, Griffith, Jurkowitz & Rothman, 2008; Sanchez & Wiley, 2006; Sung & Mayer, 2012). For example, Harp and Mayer (1997) conducted a series of experiments comparing the addition of “seductive” graphics (i.e., irrelevant to learning but catchy) to conditions without those graphics and found that comprehension was reduced in the former condition even though it increased ratings of interest.

This negative effect on comprehension due to unnecessary graphics has been found across various studies (Garner, Gillingham & White, 1989; Harp & Maslich, 2005; Park, Moreno, Seufert & Brunken, 2010; Sanchez & Wiley 2006), and is thought to be caused by distracting the learner (Harp & Mayer, 1998) thus, consuming some of the limited resources potentially available to learn the material (Mayer, 2009; Sung & Mayer, 2012; Paas, Renkl & Sweller, 2003). That said, the inclusion of unnecessary graphics do not always produce negative effects (Ketzer-Nöltge, Schweppe & Rummer, 2018; Köhl, Moersdorf, Römer & Münzer, 2018; Strobel, Grund & Lindner, 2018), and under low load conditions (e.g., when information is split and presented across the visual and auditory channels, reducing the load in either one) they may actually improve learning outcomes (Park et al., 2010). These inconsistencies in the literature have led some researchers to believe that the negative effect of unnecessary graphics may have been exaggerated (Eitel & Kuhl, 2018).

Mind wandering and unnecessary graphics

While the primary consideration in designing video lectures is comprehension, comprehension is intimately tied up with the video lecture's ability to maintain learner's attentional engagement (Farley, Risko & Kingstone, 2013; Risko, Anderson, Sarwal, Engelhardt, & Kingstone, 2012; Risko, Buchanan, Medimorec, Kingstone, 2013; Seli, Wammes, Risko, Smilek, 2016). Indeed, one can consider the capacity limit noted above to reflect our attentional capacity limits. That is, if a learner is attending to something other than the lecture (e.g., looking at a cellphone), then the learner is not devoting attention to the lecture content, impairing their ability to store in memory and comprehend the material (McVay & Kane, 2012; Unsworth &

McMillan, 2013; Risko et al., 2012; Smallwood & Schooler 2006). This consideration is arguably even more important when learning is online because individuals are engaging with the material on their own schedule (Gorissen et al., 2012), where they can easily disengage through media multi-tasking (e.g., switching to social media; Burak, 2012; Ralph, Thomson, Cheyne & Smilek, 2014), or quit the lecture early (Kim et al., 2014).

A popular means of indexing the extent to which an individual is attending to the lecture is to measure mind wandering – the shifting of attention from a task to thoughts unrelated to the task (Smallwood & Schooler, 2006). Recent studies have suggested that mind wandering occurs often in video lectures (approximately 40% of the time; Risko et al., 2012), which can lead to impaired comprehension of the lecture material (Farley et al., 2013; Risko et al., 2012; Risko, Buchanan, Medimorec & Kingstone, 2013). Given that a lack of attentional engagement in a lecture will typically have negative consequences on comprehension, it is important for researchers to investigate ways to keep learners' attention on task with their lecture designs. For example, some studies suggest that time spent in a lecture can change rates of mind wandering, such that as the time into the lecture increases (i.e., the longer the learner is required to sustain their attention to the task) mind wandering increases (Farley et al., 2013; Risko et al., 2012; Szpunar, Multon & Schater, 2013).

Attentional engagement can also be directed by the learner's motivation to engage with the material (Moreno & Mayer, 2007). Research has suggested that some combinations of media (e.g., audio, text and graphics) may lead to more or less motivation for learners to continue to engage with the learning material (Moreno & Mayer 2007). Specifically, the use of attractive illustrations in a lecture has been demonstrated to be helpful in increasing interest in study material, however, these images impaired comprehension on an immediate test (Harp & Mayer,

1997; Magner, Schwonke, Alevén, Popescu, & Renkl, 2014). If it is the case that these attractive, arguably unnecessary graphics keep learners engaged and interested in the material, then mind wandering might decrease in conditions where these types of graphics are used. Nonetheless, while attractive graphics may keep attention to the lecture through reduced mind wandering, they may still impair (or not benefit) comprehension if they are irrelevant to the lecture material. That is, unnecessary graphics could draw the learner's attention to an irrelevant aspect of the lecture (the graphic) and while their attention might be "on task" resources are being diverted from the to-be-learned material (Harp & Mayer 1998). Thus, a deeper understanding of how graphics influence attention, as indexed by mind wandering, would be valuable in informing instructional design.

Present Investigation

In the present investigation we examined the use of unnecessary graphics in an online video lecture, specifically, how it affects participant's comprehension and mind wandering during the lecture. Participants viewed one of three types of video lecture; 1) unnecessary graphics, 2) relevant graphics, and 3) no graphics, and completed a comprehension test to measure learning. In addition, during the lecture participants were probed to determine if they were mind wandering and if so what distinct type: intentional or unintentional. Intentional mind wandering is considered to be a purposeful shift from the task to a thought, whereas unintentional mind wandering is considered shifting attention away from the task even though the intent was to remain on task (Seli, Carriere & Smilek, 2015; Seli, Risko & Smilek, 2016; Seli, Risko, Smilek & Schacter, 2016). The inclusion of this more detailed reporting of mind wandering will allow us

to determine not just the effect of mind wandering overall, but if the effect is driven by one type more than another.

In addition to mind wandering, we included an exploratory measure, with the intention of providing further insight into disengagement from the lecture. Specifically, we included an “offtask” measure of behaviour, that consisted of recording how often participants clicked away from the task screen showing the lecture (i.e., our participants were completing the task online, as opposed to in a laboratory, and as such could freely disengage in this manner). This provided a behavioral measure of disengagement that complemented the self-report based mind wandering measure.

We were also interested in participant’s perceptions of how much they learned from the lecture and of the graphics used. As such, participants were asked to provide judgments of learning (Wilson, Martinez, Mills, D’Mello, Smilek & Risko, 2018) before the comprehension test, and concluded the study with a rating scale of how helpful and relevant the graphics used were. The latter served as a manipulation check to determine if participants who were in the conditions with graphics, perceived the graphics as we intended them.

Experiment 1

Methods

Participants. A total of 215 individuals participated in this online study through Amazon’s Mechanical Turk. Participants received \$5 in compensation for completion of the study. An a-priori sample size was determined in order to achieve 0.80 power, at alpha .05, to detect a small-to-medium effect size of $\eta^2 = 0.05$ in a one-way omnibus ANOVA.

Stimuli. The video lecture consisted of a 25-minute slideshow presentation paired with audio of the instructor explaining the content of the lecture (i.e., an introduction to reasoning and decision making). The audio of the lecture was recorded before creating the matching slideshow presentation. Participants in all conditions listened to the same audio track. The lecture slides were developed using the audio recording. The text and positioning of the text on all of the lecture slides were matched across conditions, however the images that accompanied the text on the slide varied by condition (see Figure 1). For both conditions containing graphics, there was at minimum one graphic per slide, some contained up to 4 images. In all conditions, there were 71 slides which made up the presentation, meaning that all the content presented in the lecture had graphics associated with them in the graphics conditions. In the unnecessary graphics condition, images that had no additional explanatory value with respect to the material but were engaging (e.g., cat photos) were featured on each slide of the presentation. In the relevant graphics condition, there were images that related to the content of the lecture (e.g., exemplars, infographics) presented on each slide. In the no graphics condition, there were no images of any kind on the slides.

Measures. Participants completed a demographic questionnaire where they reported their age, gender, highest level of education, current student status, as well as the number of online courses and the number of psychology courses they had taken. They were then asked to read a description of the two types of mind wandering we were interested in (intentional and unintentional) and asked a multiple-choice question about how these types of mind wandering were different. This was a check to ensure participants knew how to assess their own thoughts and respond to the probe throughout the lecture appropriately. Participants were then presented with two pre-assessment questions assessing their existing knowledge of the content for the

online lecture (i.e., reasoning and decision making). These questions were open response and participants were told they could respond with “I don’t know” if they did not know the answer and did not want to guess (see Appendix A). These two pre-assessment questions were on the final comprehension test.

During the video lecture participants received 10 mind wandering probes. The probe would appear at the same time for everyone, however they were presented at unequal intervals to prevent participants from anticipating the probe. The unequal intervals were decided using a random number generator, which lead to 6 of the probes occurring in the first half of the video (i.e., the first 12.5 mins) and the remaining 4 were distributed in the last half of the video (see Appendix B for timings). The probe asked if the participant had been mind wandering in the moments before it appeared and participants needed to select one of the following three responses: “yes -I was intentionally mind wandering”, “yes-I was unintentionally mind wandering”, or “no- I was fully focused on the lecture.”

After the video participants provided an estimate of the score they expected on the upcoming test (judgment of learning). The study concluded with a test of the content of the lecture they just watched. There were 15 comprehension questions (10 multiple choice, 5 open response, see Appendix A). These test questions were additionally categorized as being “deep” (i.e., required integration/manipulation of lecture material), or “shallow” (i.e., vocabulary, definitions, or have content verbatim from the lecture). Three of the test questions were categorized as deep learning questions, and the remaining 12 were categorized as shallow learning questions. We calculated Cronbach’s alpha for the items with this test and found an internal reliability of .73, and .74 in Experiments 1 and 2 respectively.

As a manipulation check, we asked participants at the end of the study to rate on a 1-7 scale how helpful they thought the graphics in the lecture were, and how relevant they thought the

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graphics were, where 1 meant completely relevant/helpful and 7 meant completely irrelevant/unhelpful. For easier data interpretation, we reverse scored responses so that a larger number represented greater helpfulness and relevance.

Additionally, throughout the study, if the participant navigated away from the task screen (e.g., minimizing the page featuring our study) and the page with our study was no longer in

focus, the change was coded and timestamped. Participants were not told at the beginning of the study that these data would be recorded. We included this exploratory variable as a measure of “off-task” behaviour.

Procedure. After providing consent, participants were asked to complete the demographic questions and the pre-assessment questions. After the pre-assessment, participants were instructed to read a short description about what mind wandering is and what the different types of mind wandering are (i.e., intentional and unintentional). Participants were asked a comprehension (multiple choice) question about the difference between intentional and unintentional mind wandering to ensure they knew how to respond when probed during the video. The participants were then instructed to watch a 25-minute psychology lecture on reasoning and decision making, where they responded to the mind wandering probes. Participants knew they would be asked questions about the video before the video started. When the video ended, participants completed the judgment of learning (JOL) questions, followed by the comprehension test. Lastly, they were asked how helpful and how relevant they thought the graphics were in the lecture they watched. Once these questions had been completed participants were debriefed and compensated.

[Figure 1 near here]

Results

Twenty-nine participants were excluded from data analysis since they did not complete all necessary aspects of the study (i.e., they quit before the post-assessment questions were completed). A further 11 participants were excluded for failing to correctly answer the

comprehension question on the difference between intentional and unintentional mind wandering. The distribution of the remaining 175 participants for each condition were as follows:

61 participants saw unnecessary graphics, 55 saw relevant graphics, and 59 saw no graphics.

Demographic data was collected for all of the participants. The mean age was 35 years old.

There were 84 males and 91 females. Participants reported their highest level of education with 63 participants reporting having a high school diploma, 28 having a college diploma, 64 having a bachelor's degree, 18 having a master's degree, and 2 having a doctorate degree. A final question asked was if participants were currently university students, 146 reported not being a current student, and the remaining 29 said they were.

Open response comprehension questions (pre and post assessment questions) were scored by a research assistant who was blind to the condition participants were in. Participants could receive one mark for a correct response, half a mark for a partially correct response or zero for an incorrect response, therefore each question was scored out of 1. All correct multiple-choice responses received a score of 1. The questions were then split by when the content was delivered in the video lecture, specifically if it fell within the first half of the video or the second half.

Below we first report the effect of condition on accuracy in the test. We then examine the effect of condition on the proportion of wandering events across the two types of mind wandering.

Mind wandering was also analyzed across time (i.e., split by if the probe occurred at the start or end of the video). Lastly, we report an exploratory analysis of an "off-task" behavioral measure (i.e., deviation from the task screen). We conclude with an analysis for reports of helpfulness and

relevance for the lectures which contained graphics, and conducted bivariate correlations for all of our dependent variables. All data and analysis scripts are available here: <https://osf.io/3q8wa/>.

Comprehension. There was a marginally significant effect of condition on the proportion of correct answers for the pre-assessment questions, $F(2,172) = 2.91, p = .057, \eta_p^2 = .03$ (see Table 1), such that those in the relevant graphic condition had a greater proportion of correct answers compared to those in the no graphics condition, $t(112) = 2.45, p = .015, d = 0.46$. There was no difference between those in the relevant graphics and unnecessary graphics conditions, $t(114) = 0.87, p = .387, d = 0.16$, and no difference between those in the unnecessary graphics and those in the no graphics conditions, $t(118) = 1.64, p = .104, d = 0.30$. Overall, the preassessment means were all low (i.e., below 15% correct) and the majority of participants (80%) had zero correct responses on the pre-assessment.

There was no significant effect of condition on the proportion of correct answers to the comprehension test questions, $F(2,172) = 0.66, p = .518, \eta_p^2 < .01$. There was also no significant effect of condition on shallow learning questions, $F(2, 172) = 0.97, p = .380, \eta_p^2 < .01$, or deep learning questions, $F(2, 172) = 0.17, p = .843, \eta_p^2 < .01$ (see Table 1). When we compared across the different question formats, we also found no effect of condition for either multiple choice, $F(2, 172) = 1.24, p = .293, \eta_p^2 = .01$, or open response questions, $F(2, 172) = 0.04, p = .959, \eta_p^2 < .01$ (see Table 1).

As noted above, there was a small difference across conditions in pre-knowledge of the lecture material and as such, we conducted a linear regression examining test scores across the conditions, controlling for pre-assessment accuracy. There was still no effect of condition on comprehension when we controlled for pre-knowledge, $F(2, 171) = 0.42, p = .661$. As well, when we removed the prior knowledge questions from the comprehension test score and

analyzed the new proportion of correct answers on the remaining 13 questions, we still found no significant effect of condition.

To determine if there was an effect of time on task and if it differed by video condition, we split questions by whether the content occurred during the first half (first 12.5 mins) or the second half of the video, and conducted a condition by time (first half, second half) mixed ANOVA. There was no significant main effect of time, $F(1, 172) = 2.52, p = .114, \eta_p^2 = .01$ and no interaction of time with condition, $F(2, 172) = 0.12, p = .885, \eta_p^2 < .01$ (see Table 1).

Judgments of Learning. Three participants did not respond to this question and therefore are not included in the analysis. There was no significant effect of condition for JOLs, $F(2, 169) = 0.70, p = .499, \eta_p^2 < .01$ (see Table 1). When we compared actual test accuracy with JOLs using a paired t-test, there was a significant difference, $t(171) = 4.14, p < .001, d = 0.31$, such that participants reported higher JOLs ($M = 0.58, SD=0.19$) than their actual test score ($M = 0.51, SD=0.21$; i.e., they were overconfident).

[Table 1 near here]

Mind Wandering. There was no effect of condition on overall mind wandering, $F(2, 172) = 1.34, p = .264, \eta_p^2 = .02$. There was also no effect of condition on intentional mind wandering, $F(2, 172) = 0.37, p = .695, \eta_p^2 < .01$, but there was a significant difference between our conditions in unintentional mind wandering, $F(2, 172) = 3.29, p = .039, \eta_p^2 = .04$, such that those in the no graphics condition reported significantly more unintentional mind wandering than those in both the unnecessary graphics, $t(118) = 2.27, p = .025, d = 0.41$, and the relevant graphics conditions, $t(112) = 2.19, p = .031, d = 0.41$. There was no difference in unintentional

mind wandering across the unnecessary and relevant graphics conditions, $t(114) = 0.13, p = .893$, $d =$

0.02 (see Table 2).

To investigate time on task effects, mind wandering at the start of the video was defined as mind wandering reported on the first two probes, and mind wandering at the end of the video was defined as mind wandering reported on the last two probes. We conducted a condition by time (start, end) mixed ANOVA on the proportion of overall mind wandering. There was a main effect of time, $F(1, 172) = 31.80, p < .001, \eta_p^2 = .16$, such that at the end of the video lecture there was a greater proportion of mind wandering, compared to the start. There was no interaction between time and condition, $F(2, 172) = 0.39, p = .677, \eta_p^2 < .01$. This pattern of results was the same for both types of mind wandering (intentional: time, $F(1, 172) = 15.37, p < .001, \eta_p^2 = .08$, interaction, $F(2, 172) = 0.79, p = .457, \eta_p^2 < .01$; unintentional: time, $F(1, 172) = 13.17, p < .001, \eta_p^2 = .07$, interaction, $F(2, 172) = 0.61, p = .546, \eta_p^2 < .001$; see Table 3).

We additionally conducted a mixed effects logistic regression analysis at the level of individual probe responses, due to the fact that our distributions for mind wandering were nonnormal (i.e., majority of responses were “not mind wandering”). Mind wandering (overall, intentional, unintentional) was the dependent variable, condition and probe number (1-10; continuous) were fixed effects and participant was a random effect. We used a model comparison approach to assess the main effects of condition and probe number and their interaction, such that we compared the main effects models to a null model, and we compared the interaction effects model to the main effects model. These model comparisons revealed a qualitatively similar pattern to that reported above (i.e., effects that were significant or not were the same),

with one exception. The model with unintentional mind wandering as the dependant variable showed a non-significant effect of condition ($p = .051$), which was not what was found in the ANOVA reported above ($p = .039$).

[Table 2 near here]

Off-Task Frequency. Off-task behaviour was the number of times the participants clicked away from the task screen and the study was no longer in focus. We examined the frequency of off-task events which occurred during the video lecture, across the conditions with a one-way ANOVA. Three participants were removed for having extreme scores (z scores above 3). There was a significant main effect of condition, $F(2, 169) = 5.19, p = .006, \eta_p^2 = .06$, such that those in the relevant graphics condition had less off-task events during the video ($M = 3.52, SD = 5.88$), compared to the unnecessary graphics condition ($M = 9.12, SD = 11.55$), $t(111) = 3.20, p = .001, d = 0.60$, and the no graphics condition ($M = 6.68, SD = 9.16$), $t(111) = 2.16, p = .033, d = 0.41$. Those in the unnecessary graphics and no graphics condition did not significantly differ, $t(116) = 1.27, p = .206, d = 0.23$.

Helpfulness and Relevance. Those in the relevant graphics condition ($M = 5.59, SD = 1.60$) reported that the graphics were more helpful than those in the unnecessary graphics condition ($M = 4.52, SD = 1.81$), $t(113) = 3.33, p = .001, d = 0.62$. Those in the relevant graphics condition, also reported that the graphics were more relevant ($M = 5.70, SD = 1.38$) compared to those in the unnecessary graphics condition ($M = 4.97, SD = 1.73$), $t(113) = 2.49, p = .014, d = 0.47$.

[Table 3 near here]

Bivariate Correlations. While not the focus of the present work, we examined the bivariate relations between the various dependent variables, which are displayed in Table 4.

Provided the large number of correlations significance values should be interpreted cautiously.

As expected, test accuracy positively correlated with pre-assessment accuracy, and negatively correlated with overall reports of mind wandering and off-task frequency. Test accuracy also positively correlated with JOLs, suggesting that participants could judge their learning, at least to some extent.

Interestingly, mind wandering did not correlate with off-task frequency, but it did negatively correlate with JOLs, suggesting that participants may have been aware in some respect that mind wandering relates to reduced learning outcomes. The latter is likely driven by reports of unintentional mind wandering, since JOLs significantly and negatively correlated with unintentional mind wandering but not intentional mind wandering. JOLs also correlated positively with ratings of helpfulness and relevance, these ratings positively correlated with each other. Helpfulness also negatively correlated with overall reports of mind wandering. [Table 4 near here]

Discussion

In Experiment 1 there was no effect of lecture condition on comprehension. In addition, there was no effect of lecture condition on the overall amount of mind wandering, but there was a greater proportion of unintentional mind wandering for those in the no graphics condition compared to conditions with graphics (unnecessary and relevant). JOLs did not significantly vary across the conditions, however participants were overconfident in their estimates. We did find differences across conditions in off-task frequency, such that those in the relevant graphics

condition had the lowest rates of being away from the screen compared to the other two lecture conditions. Interestingly, off-task frequency negatively correlated with test accuracy, but did not correlate with mind wandering. Importantly, participants were able to report the graphics used in the unnecessary graphics condition were unhelpful and irrelevant to the lecture material, suggesting our conditions were being perceived by participants the way we intended them to be.

Experiment 1 also replicated some results from previous studies. Comprehension and overall mind wandering correlated negatively supporting findings from other studies (Risko et al., 2012; Wilson et al., 2018). With regards to mind wandering, we found time on task effects for each type of mind wandering, such that there was greater mind wandering reported at the end of the video (Farley et al. 2013; Risko et al., 2012; Risko et al., 2013). Additionally, participants were overconfident in their JOLs, which is common when JOLs are asked immediately after encoding (Koriat & Bjork, 2006). Before discussing these results further we report a replication and extension.

Experiment 2

Rationale

In Experiment 2 we set out to replicate and extend Experiment 1's results. While we did not find differences in comprehension across our lecture conditions in Experiment 1, participant's experiences with the video lecture seemingly varied across conditions. For example, participants reported that the unnecessary graphics were less helpful and relevant than the graphics in the relevant graphics lecture. In Experiment 2, we extended the outcomes of interest to include more non-cognitive factors, in particular, we included a measure of positive and negative affect for the lectures. Some research has suggested that visuals used in a lecture may

increase interest, even if it does not improve learning outcomes (Sung & Mayer, 2012; Wilson et al., 2018). For example, Wilson et al. (2018) demonstrated that the addition of an instructor to a video lecture (compared to one with only audio) changed attitudes towards the lecture without changing performance (e.g., on a comprehension test). Specifically, they found that participants reported being less likely to drop a class with the instructor in the video, even though this addition did not improve learning outcomes. Moreno's (2006) cognitive affective theory of learning with media combines the traditional cognitive load theory with the additional consideration of affect and motivation (Park, Plass, Bruken 2014). Specifically, this model suggests that motivational factors (such as interest and liking) can modulate learning through changes in cognitive engagement (Moreno, 2006). That is, the more an individual is interested in the lecture, the more they will engage with the material cognitively (i.e., reflect on it), thus affecting learning outcomes. Such a mechanism might explain the small benefit of relevant graphics on unintentional mind wandering and off-task events in Experiment 1. In addition to cognition, affect is important to examine given the fact that affect and attention often correlate. Specifically, negative affect (e.g., anger) has been shown to correlate positively with taskirrelevant thinking, but this relation is negative when the affect is positive (e.g., enjoyment; Pekrun, 2011). Positive affect should also facilitate learning of connected materials (e.g., related concepts; Perkrun, 2011). As well, Pekrun (2006) has suggested that some positive affect states like enjoyment will positively correlate with motivation to learn, thus keeping students engaged. With regards to metacognition, positive affective states positively correlate with self-regulation of learning strategies (i.e., students will be better at making goals and assessing their learning; Pekrun, 2011).

Overall, the learner's feelings towards the video lectures will also be important to consider as it might influence learner's engagement with the material at a broader level. As such, even if affect does not modulate cognitive engagement within the lecture, it is nevertheless important that we have a deeper understanding of how various design features influence positive and negative affect experienced during video lecture viewing.

We have included in Experiment 2 the Positive and Negative Affect Schedule (PANAS). Within this scale, high positive affect reflects high energy and engagement measured through items like alert, attentive, and active. In contrast, negative affect is not the opposite of positive affect (i.e., they are not inverse of each other) but a reflection of adverse mood states measured with items like afraid, irritable and nervous (Watson, Clark, Tellegen, 1988). The use of this scale allows us to measure positive and negative affect independently, though it was not designed for assessing affect for video lectures. For example, it is unlikely that our video lectures will evoke strong adverse mood states (e.g., anger). Nevertheless, the scale as a whole offers a unique perspective into the affect experienced in lectures.

Methods

Participants. A total of 226 participants from Amazon's Mechanical Turk participated in this study. Sample size for this study was chosen using the same power analysis as Experiment 1.

Stimuli. The video lectures were identical to those used in Experiment 1.

Measures. Participants completed the same demographic, pre-assessment, mind wandering probes, JOL and test questions as was used in Experiment 1. In addition to the measures used in Experiment 1, participants completed the positive and negative affect schedule

(PANAS; Watson, et al. 1988). The PANAS scale consists of 10 positive word items (e.g., interested) and 10 negative word items (e.g., upset) that get rated on a 1-5 scale. This scale is reliable for both positive and negative items at $\alpha > .84$. Unlike in the original scale which asks how the participants feel about each of the items in the current moment, we asked participants to “indicate to what extent you felt this way while watching the video lecture”.

Procedure. The procedure follows Experiment 1 except that immediately after the lecture participants completed the PANAS. They then completed the JOL and the comprehension test. Lastly, they were asked about the helpfulness and relevance of the graphics in the lecture they watched. Once these questions had been completed participants were debriefed and compensated.

Results

Thirty-four participants were removed for not completing all aspects of the study. An additional 13 participants were removed for failing to correctly answer the question distinguishing the types of mind wandering. The distribution of the remaining 179 participants for each condition were as follows: 61 participants saw unnecessary graphics, 59 saw relevant graphics, and 59 saw no graphics. The mean age was 37 years old. There were 92 males, 81 females, and 1 other. Participants also reported their highest level of education with 63 participants reporting having a high school diploma, 32 having a college diploma, 68 having a bachelor’s degree, 13 having a master’s degree, and 2 having a doctorate degree. A final question asked was if participants were currently university students, 152 reported not being a current student. The analyses follow those in Experiment 1 with the additional analysis of the positive and negative affect scale.

Comprehension. We found no effect of condition for the proportion of correct answers for the pre-assessment questions, $F(2,176) = 0.60, p = .551, \eta_p^2 < .01$. The majority of participants (82%) had a score of zero on the pre- assessment questions. There was also no effect of condition for the proportion of correct answers to the test questions, $F(2,176) = 2.28, p = .106, \eta_p^2 = .03$ (see Table 5). As in Experiment 1, when we removed the prior knowledge questions from the comprehension test score and analyzed the proportion of correct answers on the remaining 13 questions, we still found no significant effect of condition.

Unlike in Experiment 1, when we examined the proportion of correct answers to the shallow test questions there was a significant effect of condition, $F(2, 176) = 3.61, p = .029, \eta_p^2 = .04$, such that those in the unnecessary graphics condition did significantly better on these questions than those in the no graphics condition, $t(118) = 2.68, p = .009, d = 0.49$, but did no better than those in the relevant graphics condition, $t(118) = 1.42, p = 0.159, d = 0.26$. There was no difference between those in the relevant graphics and no graphics conditions $t(116) = 1.25, p = .215, d = 0.22$ (see Table 5). There was no effect of condition for the deep learning questions, $F(2, 176) = 0.00, p = .999, \eta_p^2 < .01$ (see Table 5).

We also found an effect of condition when we analyzed just the multiple choice questions, $F(2, 176) = 3.65, p = .028, \eta_p^2 = .04$, such that those in the unnecessary graphics condition did significantly better on these questions than those in the no graphics condition, $t(118) = 2.66, p = .009, d = 0.49$, but did no better than those in the relevant graphics condition, $t(118) = 1.40, p = 0.165, d = 0.25$. There was no difference between those in the relevant graphics and no graphics conditions, $t(116) = 1.31, p = .193, d = 0.24$ (see Table 5). There was no effect of condition for the open response questions, $F(2,176) = 0.34, p = .714, \eta_p^2 < .01$ (see Table 5).

In the analysis including time as a factor, there was a significant main effect of time, $F(1, 176) = 23.98, p < .001, \eta_p^2 = .12$, such that participants had greater accuracy for questions with content that was from the second half of the video lecture ($M = 0.56, SD=0.24$) compared to questions with content from the first half of the lecture ($M = 0.48, SD=0.24$). There was no interaction between time and lecture condition, $F(2,176) = 0.02, p = .981, \eta_p^2 < .01$ (see Table 5).

[Table 5 near here]

Judgments of Learning. Two participants did not respond to this question and therefore are not included in the analysis. There was no effect of condition for JOLs, $F(2, 174) = 0.65, p = .523, \eta_p^2 < .01$ (see Table 5). There was also no difference between participants' JOLs and their score on the test, $t(176) = 1.02, p = .311, d = 0.07$, thus unlike Experiment 1 participants were not overconfident.

Mind Wandering. Consistent with Experiment 1, there was no effect of condition on proportion of overall mind wandering, $F(2, 176) = 0.74, p = .478, \eta_p^2 < .01$, or intentional mind wandering, $F(2, 176) = 0.23, p = .791, \eta_p^2 < .01$. Unlike in Experiment 1, there was no effect of condition on unintentional mind wandering, $F(2, 176) = 0.36, p = .700, \eta_p^2 < .01$, (see Table 6).

[Table 6 near here]

With respect to time on task, for overall mind wandering there was a main effect of time, $F(1, 176) = 63.27, p < .001, \eta_p^2 = .26$, such that at the end of the video lecture there was more reported mind wandering, compared to the start. There was no interaction between time and condition, $F(2, 176) = 0.47, p = .623, \eta_p^2 < .01$ (see Table 7). This same pattern of results was found for both intentional and unintentional mind wandering types (intentional: time, $F(1, 176) =$

26.84, $p < .001$, $\eta_p^2 = .13$, interaction, $F(2, 176) = 0.44$, $p = .644$, $\eta_p^2 < .01$; unintentional: time, $F(1, 176) = 17.82$, $p < .001$, $\eta_p^2 = .09$, interaction, $F(2, 176) = 0.19$, $p = .830$, $\eta_p^2 < .01$).

We additionally conducted a mixed effects logistic regression analysis at the level of individual probe responses. Mind wandering (overall, intentional, unintentional) was the dependent variable, condition and probe number (1-10) were fixed factors and participant was a random effect. Using a model comparison approach to assess the main effects of condition and probe number and their interaction revealed a qualitatively similar pattern to that reported above (i.e., effects that were significant or not were the same).

[Table 7 near here]

Positive and Negative Affect Scale. Fourteen participants were removed from this analysis for providing an impossible response (a number that was not between 1 and 5) or for leaving a response on the questionnaire blank. For the remaining 165 participants we calculated their positive and negative affect score and conducted a one-way ANOVA to determine if affect scores varied by condition. There were no differences across the conditions for positive affect ratings, $F(2, 162) = 0.39$, $p = .679$, $\eta_p^2 < .01$, or for negative affect ratings, $F(2, 162) = 1.81$, $p = .166$, $\eta_p^2 = .02$ (see Table 8). We examined individual items from the PANAS to determine if ratings for any single item (e.g., interest) varied by condition, however there was no significant effect of condition for any item on the scale (all p 's $> .1$).

[Table 8 near here]

Off-Task Frequency. Seven participants were removed for having extreme scores (z scores above 3). Unlike in Experiment 1, there was no main effect of condition, $F(2, 169) = 1.36$, $p = .258$, $\eta_p^2 = .02$ (unnecessary $M=4.16$, $SD=6.01$; relevant $M=5.72$, $SD=7.56$; no graphics

$M=6.45$, $SD=8.96$).

Helpfulness and Relevance. Those in the relevant graphics condition ($M = 5.38$, $SD=1.51$) reported greater helpfulness of the graphics compared to those in the unnecessary graphics condition ($M = 4.75$, $SD=1.60$), $t(116) = 2.20$, $p = .029$, $d = 0.41$ as well as greater relevance of the graphics ($M = 5.74$, $SD=1.15$) compared to those in the unnecessary graphics condition ($M = 4.97$, $SD=1.64$), $t(117) = 2.97$, $p = .003$, $d = 0.54$.

Bivariate correlations. The bivariate correlations between all our dependent variables are displayed in Table 9. Again, correlations should be interpreted cautiously given the number of correlations. As with Experiment 1, test accuracy positively correlated with pre-assessment accuracy, and negatively correlated with overall reports of mind wandering and off-task frequency. In both Experiment 1 and 2 the correlations between test accuracy and mind wandering type (intentional and unintentional) were low. Past research has shown that these correlations are often higher (Phillips, Mills, D'Mello & Risko, 2016; Seli et al., 2016; Wammes, Seli, Cheyne, Boucher, & Smilek 2016), some finding correlations as low as seen here (Martin, Mills, D'Mello, & Risko, 2018). That said, these small correlations may be due to the limited range of scores on the comprehension test (50% of scores fall within the 40-60% correct range). Both types of mind wandering and off-task frequency negatively correlated with JOLs. Again, test accuracy positively correlated with JOLs. Unlike in Experiment 1, mind wandering did not positively correlate with off-task frequency.

Like in Experiment 1, helpfulness ratings negatively correlated with overall mind wandering, however in this Experiment we also found that relevance ratings showed the same relation. Again, both ratings positively correlated with each other suggesting that if the graphics were perceived as helpful they were also perceived to be relevant.

Interestingly, positive, but not negative, affect negatively correlated with both types of mind wandering and off-task frequency, suggesting that greater behavioural distraction is associated with less positive affect, this is consistent with previous findings in the education literature (Pekrun, 2011). Positive and negative affect were both related to JOLs, though in opposite directions, such that positive affect was positively related and negative affect was negatively related, this again seems consistent with recent findings with affect and selfregulation of learning (Pekrun, 2011). As well, positive affect significantly correlated with participants' preassessment scores, suggesting that greater prior knowledge related to greater positive affect. Neither positive nor negative affect related to test accuracy, which is inconsistent with past studies, however the direction of the non-significant correlations is in the expected direction (i.e., positive affect should correlate positively with memory of material and negative affect negatively). As well, positive affect positively correlated with helpfulness and relevance ratings. Lastly, positive affect correlated positively with negative affect.

[Table 9 near here]

Discussion

As in Experiment 1, in Experiment 2 there were no effect of condition on overall comprehension. That said, those in the unnecessary graphics condition did better on shallow learning questions, in particular multiple-choice questions (these were largely the same items), than did those in the no graphics condition. This was not the case in Experiment 1. Again, there was no effect of condition on the overall amount of mind wandering, but unlike in Experiment 1, there was also no effect of condition on unintentional mind wandering. There was no effect of

condition on JOLs and participants were not overconfident in their estimates as we would have expected given what we found in Experiment 1.

In Experiment 2 there was no effect of condition on our exploratory measure of off task behaviour, though we found an effect in Experiment 1. We also found that off-task frequency positively correlated with overall mind wandering, which it did not in Experiment 1. Consistently, off-task frequency and mind wandering negatively correlated with test accuracy. As well, we continue to find evidence of time on task effects for each type of mind wandering, with mind wandering increasing over time.

While participants were able to report that the graphics used in the unnecessary graphics condition were unhelpful and irrelevant to the lecture material, as they did in Experiment 1, we did not find an effect of condition for ratings of positive and negative affect. That said, positive affect, significantly and positively correlated with helpfulness and relevance ratings, suggesting that the more helpful and relevant participants thought the graphics were, the greater their positive affect was. In addition, positive affect correlated significantly and positively with JOLs. The cognitive affective theory of learning with media (Moreno, 2006), suggests a relation between metacognition and affect in learning, so this correlation shows some support for this framework. Lastly, positive affect significantly and negatively correlated with both mind wandering and off-task frequency. This negative relation is not surprising for a few reasons, first, past research has demonstrated that affect can influence mind wandering behaviour (Killingsworth, & Gilbert, 2010; Smallwood, Fitzgerald, Miles & Phillips, 2009; Stawarczyk, Majerus & D'Argembeau, 2013), and second, the items on the positive portion of the PANAS (e.g., attentive, interested, alert) should logically be negative given that they include attention related items. The relation between mind wandering and positive affect is consistent with the

affective mediation principle in Moreno's (2006) cognitive affective learning model, which suggests that positive affect leads to more cognitive engagement. The overall lack of significant relations between negative affect and the other variables such as attention are likely due to the fact that there was limited range in the data for negative affect, the lowest score that can be received on this scale is 10 and the average score for our participants was around 12.5 (see Table 8). Additionally, it was surprising that positive and negative affect had a strong positive relation since the PANAS was designed for these two subscales to not correlate, however this may also be due to the low variability in negative affect.

General Discussion

The purpose of this study was to examine what role unnecessary graphics played in learning via online video lectures in a post-secondary context. Across our two experiments we examined the effect unnecessary graphics (i.e., graphics which had no additional explanatory value to the lecture material) had on comprehension and mind wandering in a video lecture compared to a lecture with relevant graphics and one with no graphics. In both Experiment 1 and 2, there was no effect of condition on overall comprehension test scores. The results showed inconsistent effects when we examined comprehension further separating questions by depth, response type and timing. Specifically, there was no effect of graphics on question depth (shallow, deep), or response type (multiple-choice, short answer) and timing (first-half, second half) in Experiment 1, but we do find effects in Experiment 2. This might have reflected a Type I error in Experiment 2 or, given the small effect size in Experiment 2 ($\eta_p^2 = .04$), a Type II error in Experiment 1. There were no substantive differences across Experiments 1 and 2 thus the most reasonable explanation of this inconsistency is differences across samples. In the case of mind wandering, there was no effect of condition on overall or intentional mind wandering in both

Experiments, however in Experiment 1 only we found an effect of condition on unintentional mind wandering. The same rationale is true for the inconsistencies in mind wandering between Experiment 1, and 2, such that is likely driven by differences across samples.

In both Experiments, there was no effect of condition on participants' JOLs. When we examined off-task frequency, we did not find a consistent effect of condition. In Experiment 1 the relevant graphics had the lowest frequency of navigating away from the task screen, but this was not replicated in Experiment 2. Similarly, in Experiment 1 there was no correlation between off-task frequency and mind wandering, however there was a positive correlation between them in Experiment 2.

For our manipulation check, we found participants ratings of helpfulness and relevance differed reliably across the conditions as we would expect (i.e., the relevant graphics are rated as more helpful and relevant than the unnecessary graphics). While this suggests participants were sensitive to the graphics used, we did not find that the graphics had any effect on ratings of positive and negative affect in Experiment 2.

Despite not finding an effect of condition for our main dependent variables of comprehension and attention, we did find a number of results that are consistent with existing literature. First, in both experiments there was a negative correlation between overall mind wandering and test accuracy. Additionally, we found that mind wandering occurred often (around 40% of the time), and that reports of mind wandering increased as the time on the task increased.

The Influence of Graphics on Comprehension and Attention

Overall, there was no effect of condition on comprehension. As noted above, despite there being clear empirical demonstrations of negative effects of irrelevant graphics on

comprehension, this effect is likely small (Eitel & Kuhl, 2018). Thus, it might not be surprising that we found no such effect here. That said, it is possible that our video conditions had no effect due to the fact that our graphics were themselves ineffective. This might also explain why, across our measures of mind wandering, off-task frequency and affect, we found no consistent effect of video condition. That is, the unnecessary graphics might not have been sufficiently distracting or the relevant graphics sufficiently informative (or both). That said, participants were attending to the graphics enough to note they were unhelpful and irrelevant in the unnecessary condition.

Additionally, our lecture content did not require graphics (given its abstract content), and as such the instructor did not reference the graphics when they appeared on the screen. While explicit mention of the graphics may have facilitated attention to them, this represents a different, though still interesting, research question. That is, in the case of relevant graphics, comparing conditions where the instructor does vs. does not refer to the graphic on the slide is an interesting (though different) question. In addition, the motivation here to examine the potential negative impact of irrelevant graphics would be difficult to implement in a straightforward manner (i.e., it would require the instructor to have referred to the irrelevant graphics). It is important to note that in this study, our goal was for ecological validity and as such we chose graphics common to what a lecture may actually use (e.g., stock photo of a student reading a book).

Another potential explanation for the lack of effect of video condition on comprehension, could be due to the test questions we asked. Mayer (1999) has suggested that many effects in the multimedia learning literature are specific to “deep” or “transfer” type questions that require learners to engage with the material beyond simple recall. For example, Mayer (1999) summarizes a series of studies that found that “transfer” question performance was affected in

lessons with designs that may cause overload (e.g., the use of additional extraneous text or images). While our deep questions required slightly more than verbatim responses from the lecture, they might not have required participants to transfer their knowledge in a new way (Mayer, 2002). For example, in one of our deeper learning questions we ask what type of reasoning is involved to solve the syllogism, and though the content of this syllogism is different from what was presented in the lecture, it was ordered the same way (i.e., if a then b, a therefore b).

Measuring Attention in Lectures

The present research introduced a novel measure of attentional engagement in online lectures via a behavioral index of whether or not individuals navigate away from the lecture display. Importantly, we consistently found a negative correlation between off-task frequency and test accuracy. This suggests that this behavioral measure of distraction can predict learners' test accuracy. Off-task frequency and mind wandering were not correlated in Experiment 1 but moderately correlated in Experiment 2 suggesting that this measure, at least to some extent, is uniquely measuring a learners' attentional engagement (i.e., off-task frequency may be capturing something beyond what is captured by mind wandering reports). Like mind wandering, off-task frequency was also negatively related to positive affect and JOLs. One interesting possibility is that our off-task measure is capturing learner's media multitasking live, since past research has shown a positive correlation for self-reported media multitasking, and self-reported mind wandering (Ralph et al., 2014).

Theoretical implications

With regards to the unnecessary graphics literature, our results support some recent research that suggests the effect of unnecessary graphics is quite small (Eitel & Kuhl, 2018). In this research we chose stimuli that would best represent those used in a university lecture today and as such, the irrelevant graphics may not have overloaded the cognitive system. Importantly, these results suggest that any putative negative effect of unnecessary graphics is likely small and in a typical university lecture the effect should likely not rank as particularly concerning in instructional design.

In addition, our correlational results with affect in Experiment 2 supports aspects of the cognitive affective theory of learning with media (Moreno, 2006). Specifically, the results partially supported the affective mediation principle, which suggests that affect can motivate a learner to select sensory information to process in working memory, which leads to better integration of knowledge and more mental engagement with the information. Positive affect negatively correlated with mind wandering of all types, suggesting that the greater positive affect experienced during the video lecture the less mind wandering that occurred. Positive affect may have been motivating learners to engage more with the material, thus leaving less resources available for mind wandering. This suggests the need for future research to investigate affect when examining learning environments.

Conclusions

Across two Experiments, we found no effect of unnecessary graphics on both overall comprehension and attentional engagement. These results suggest that unnecessary graphics might not be detrimental to comprehension or if they are the effect is small and difficult to detect, which is consistent with a recent review of this effect (Eitel & Kuhl, 2018). From a practical point of view, our results suggest that instructors producing a video lecture may not

need to be concerned with the incorporation of some unnecessary graphics. That said, the inclusion of unnecessary graphics also had no consistent beneficial effects in terms of comprehension, attention, or affect, thus there also seems to be no pressing reason to include them. Future work further investigating other relevant instructional design issues from an attention and learning perspective would be valuable.

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Tables

Table 1: Means and SDs for the proportion of correct answers to the pre-assessment questions, the test questions (split by question depth, response format, and time occurred) and participants’ JOL (score out of 100) in Experiment 1.

	Unnecessary Graphics	Relevant Graphics	No Graphics
Pre-Assessment	0.10 (0.20)	0.13 (0.23)	0.05 (0.14)
Test	0.48 (0.21)	0.53 (0.23)	0.49 (0.21)

Shallow	0.48 (0.22)	0.53 (0.23)	0.48 (0.20)
Deep	0.51 (0.33)	0.53 (0.33)	0.54 (0.32)
Multiple Choice	0.48 (0.21)	0.54 (0.23)	0.49 (0.20)
Open Response	0.49 (0.29)	0.51 (0.30)	0.50 (0.31)
First Half	0.47 (0.25)	0.51 (0.27)	0.49 (0.23)
Second Half	0.50 (0.23)	0.55 (0.26)	0.50 (0.23)
JOL	57 (18)	61 (19)	56 (20)

Table 2

Means and SDs for the proportion of reported mind wandering during the video lecture for each type of mind wandering by condition in Experiment 1.

	Unnecessary Graphics	Relevant Graphics	No Graphics
Overall MW	0.35 (0.24)	0.33 (0.25)	0.40 (0.23)
Intentional MW	0.09 (0.11)	0.08 (0.10)	0.09 (0.12)
Unintentional MW	0.26 (0.17)	0.24 (0.20)	0.31 (0.18)

Table 3

Means and SDs for the proportion of reported mind wandering during the video lecture across time (start, end) and condition in

Experiment 1.

	Unnecessary Graphics		Relevant Graphics		No Graphics	
	Start	End	Start	End	Start	End
Overall MW	0.25 (0.28)	0.40 (0.41)	0.21 (0.30)	0.43 (0.38)	0.31 (0.36)	0.49 (0.38)
Intentional MW	0.03 (0.12)	0.11 (0.25)	0.03 (0.11)	0.13 (0.24)	0.08 (0.20)	0.12 (0.23)
Unintentional MW	0.22 (0.25)	0.29 (0.36)	0.18 (0.28)	0.30 (0.34)	0.23 (0.28)	0.37 (0.37)

Table 4

Bivariate correlation table for Experiment 1.

Variable	1	2	3	4	5	6	7	8
1. Overall MW								
2. Intentional MW	.57**							
3. Unintentional MW	.79**	.20**						
4. Test Accuracy	-.18*	.00	-.12					
5. Pre-Assessment Accuracy	-.10	-.06	-.07	.20**				
6. Off-Task Frequency	.03	.02	.02	-.22**	-.13			
7. JOL	-.30**	-.12	-.32**	.37**	.11	-.00		
8. Helpfulness	-.16*	-.13	-.15	.03	.09	.04	.17*	
9. Relevance	-.13	-.12	-.12	.06	.13	.03	.18*	.66*

Note. * indicates $p < .05$. ** indicates $p < .01$

Table 5

Means and SDs for the proportion of correct answers to the pre-assessment questions, the test questions (split by question depth, response format, and time occurred), and participants' JOL (score out of 100) in Experiment 2.

	Unnecessary Graphics	Relevant Graphics	No Graphics
Pre-Assessment	0.10 (0.21)	0.06 (0.16)	0.08 (0.19)
Test	0.55 (0.24)	0.51 (0.21)	0.47 (0.19)
Shallow	0.57 (0.24)	0.51 (0.22)	0.47 (0.19)
Deep	0.48(0.32)	0.47 (0.32)	0.48(0.27)
Multiple Choice	0.58 (0.23)	0.52 (0.21)	0.47 (0.19)
Open Response	0.50 (0.31)	0.47 (0.31)	0.46 (0.25)
First Half	0.52 (0.26)	0.48 (0.25)	0.44 (0.20)
Second Half	0.60 (0.24)	0.55 (0.22)	0.52 (0.25)
JOL	56 (21)	51 (23)	52 (24)

Table 6

Means and 95% SDs for the proportion of reported mind wandering during the video lecture for each type of mind wandering by video condition in Experiment 2.

	Unnecessary Graphics	Relevant Graphics	No Graphics
Overall MW	0.41 (0.26)	0.45 (0.26)	0.46 (0.25)
Intentional MW	0.12 (0.17)	0.14 (0.15)	0.13 (0.17)
Unintentional MW	0.29 (0.21)	0.30 (0.19)	0.32 (0.19)

Table 7

Means and SDs for the proportion of reported mind wandering during the video lecture across time (start, end) and condition in

Experiment 2.

	Unnecessary Graphics		Relevant Graphics		No Graphics	
	Start	End	Start	End	Start	End
Overall MW	0.33 (0.34)	0.54 (0.40)	0.28 (0.34)	0.56 (0.43)	0.35 (0.35)	0.63 (0.38)
Intentional MW	0.05 (0.15)	0.16 (0.28)	0.07 (0.21)	0.22 (0.34)	0.06 (0.21)	0.18 (0.29)
Unintentional MW	0.28 (0.31)	0.39 (0.37)	0.21 (0.30)	0.34 (0.36)	0.29 (0.32)	0.44 (0.38)

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Table 8

Means and SDs for the Positive and Negative Affect scale by condition in Experiment 2.

	Unnecessary Graphics	Relevant Graphics	No Graphics
Positive Affect	26.77 (8.32)	25.70 (9.35)	25.36 (8.83)
Negative Affect	12.43 (3.86)	12.08 (4.13)	13.82 (6.73)

Table 9

Bivariate correlation table for Experiment 2.

Variable	1	2	3	4	5	6	7	8	9	10
1. Overall MW										
2. Intentional MW	.60**									
3. Unintentional MW	.73**	.00								
4. Test Accuracy	-.16*	-.18*	-.08							
5. Pre-assessment Accuracy	-.10	-.04	-.07	.24**						
6. Off-Task Frequency	.23**	.22**	.08	-.16*	-.01					
7. JOL	-.40**	-.23**	-.35**	.33**	.20**	-.19**				
8. Helpfulness	-.28**	-.08	-.24**	.03	.12	-.03	.18*			
9. Relevance	-.20**	-.13	-.11	.11	.10	-.03	.14	.68**		
10. Positive Affect	-.36**	-.21**	-.29**	.06	.20**	-.24**	.37**	.32**	.20**	

11. Negative Affect	.11	-.12	-.00	-.13	.07	.05	-.20**	-.06	-.08	.37**
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Note. * indicates $p < .05$. ** indicates $p < .01$.

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Appendix A –*Comprehension Test Questions*

List of the 15 test questions in the order participants received them. Questions 1 and 2 were used as the pre-assessment questions. Questions 1, 3, 4, 5, 6, 7, 8, 14 & 15 contained content in the first half of the lecture, the rest of the questions fell in the last 12.5 minutes. The exact timing for the question's content in the video is noted first. "OR" represents the open response questions, "MC" represents the multiple-choice questions. Questions with a * next to the type of response were considered deep learning questions.

- 1) **07:45- OR**. Name 3 different types of problem solving techniques studied by cognitive psychologists.
- 2) **18:12- OR**. Name 2 different types of reasoning studied by cognitive psychologists.
- 3) **03:30- MC**. When it comes to reasoning, problem solving, and decision making, researchers have relied on a method that involves the detailed, concurrent, and nonjudgmental observation of the contents of your own consciousness as you work on a problem. This method is called:
- 4) **05:00- MC**. Writing a letter is best considered an example of:
- 5) **05:40- MC**. Which of the following is NOT one of the reasons cognitive psychologists have focused on well-defined problems?
- 6) **08:00- OR***. Sally has forgotten her password to log in to her email account. To solve this problem she enters all of her commonly used passwords until she finds the correct

one. This problem solving approach is most similar to which one of the approaches discussed in the lecture (please type your response below):

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- 7) **10:20 & 11:20- MC.** A problem space includes an initial state, intermediate state, and goal state, all of which are important parts of these two problem solving techniques:
- 8) **10:45- MC.** In means ends analysis a permissible move in the problem space is referred to as an:
- 9) **14:30- MC.** In Gick and Holyoak (1980), participants who were told to use the story of the general to help solve the tumor problem were better able to solve it than participants not told to use the story. This was attributed to the utility of what problem solving technique:
- 10) **18:45- MC.** The following is an example of what type of reasoning:

Brian is a university student, Brian lives in a dormitory; therefore all undergrads live in dormitories
- 11) **19:40- MC.** When individuals solve syllogisms quantifiers like all, none, and some _____
- 12) **20:50- MC.** Imagine if I gave you a pattern (e.g., 2-5-9) and asked you to generate the rule that was used to generate it. This task would be examining what type of reasoning?
- 13) **23:30- MC.** The generation of a quasi-pictorial representation of the relationship between the information in the premises and the conclusion when reasoning would be consistent with which approach

14) **04:00- OR***. The lecture discussed both ill-defined and well-defined problems. Provide an example of each that is different from the examples provided in the lecture

15) **18:34- OR***. Aaron argues that all cats are lazy, you have a cat named Boots, and he concludes that Boots is lazy. What kind of reasoning is Aaron using?

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Appendix B- *Mind Wandering Probe Timings*

The timing of each probe for all participants. The first 6 fell within the first 12.5 minutes of the 25 minute video lecture and the last 4 fell at the end of the video lecture.

Probe 1- 01:04

Probe 2- 03:29

Probe 3- 05:22

Probe 4- 07:26

Probe 5- 10:08

Probe 6- 13:25

Probe 7- 16:17

Probe 8- 18:32

Probe 9- 20:11

Probe 10- 24:01