

Guideline-Based Evaluation and Design Opportunities for Mobile Video-based Learning

ANONYMOUS AUTHOR(S)*

Learners consume video-based learning content on various mobile devices due to their mobility and accessibility. However, most video-based learning content is originally designed for desktop without consideration of constraints in mobile learning environments. We focus on readability and visibility problems caused by visual design elements such as text and images considering varying screen sizes. To reveal design issues of current content, we examined mobile learning adequacy of content with 681 video frames from 108 lectures. The content analysis revealed a distribution and guideline compliance rate of visual design elements. We also conducted semi-structured interviews with six video production engineers to investigate current practices and challenges in content design for mobile devices. Based on the interview results, we present a prototype that supports a guideline-based design. Our findings can inform engineers and design tool makers on challenges of editing mobile video-based learning content for accessible and adaptive design across devices.

CCS Concepts: • **Human-centered computing** → **Heuristic evaluations**.

Additional Key Words and Phrases: Design Guidelines, Content Analysis, Video-based Learning, Mobile Devices

ACM Reference Format:

Anonymous Author(s). 2018. Guideline-Based Evaluation and Design Opportunities for Mobile Video-based Learning. In *Woodstock '18: ACM Symposium on Neural Gaze Detection, June 03–05, 2018, Woodstock, NY*. ACM, New York, NY, USA, 9 pages. <https://doi.org/10.1145/1122445.1122456>

1 INTRODUCTION

With the increasing ubiquity of mobile devices, learners access video-based learning material at both a time and place convenient for them [16]. The lockdowns and school closures caused by the global pandemic accelerated an increase in learners on video learning platforms such as MOOCs (e.g., edX, Coursera, Udacity, and FutureLearn) due to their openness and easy accessibility [2, 39, 48]. In addition, prior research proves the synergistic characteristics of mobile learning and MOOCs [15, 17].

However, one of the main limitations of mobile learning is the small screen size, which deteriorates the learning experience and decreases the effectiveness of learning with too small font size, content-heavy lecture slides, and complex graphics to digest in a mobile environment. The existing learning framework also highlighted the importance of such visual design factors in learning. Inappropriate font sizes of learning material impose unnecessary cognitive load [31] and lower judgments of learning (JOLs) [21, 38]. Excessive amount of words is another factor that increases the cognitive load [31, 43, 44] and information overload [4]. Image elements can also increase cognitive load by splitting learners' attention [30, 31]. However, most of the existing video learning content is originally designed for desktops with a widescreen. Moreover, existing studies on mobile video-based learning lack consideration for video production

Permission to make digital or hard copies of all or part of this work for personal or classroom use is granted without fee provided that copies are not made or distributed for profit or commercial advantage and that copies bear this notice and the full citation on the first page. Copyrights for components of this work owned by others than ACM must be honored. Abstracting with credit is permitted. To copy otherwise, or republish, to post on servers or to redistribute to lists, requires prior specific permission and/or a fee. Request permissions from permissions@acm.org.

© 2018 Association for Computing Machinery.

Manuscript submitted to ACM

53 engineers and designers who are direct stakeholders of lecture video design, being involved in the content design
54 process.

55 To better understand this challenge, we examined 681 video frames from 108 MOOC video lectures. The lectures
56 are selected from the top MOOC courses list in 2019 released by Class Central [40] including MOOCs from Coursera,
57 edX, and FutureLearn. To thoroughly understand the common design patterns of current video lectures, we examined
58 the guideline compliance rate for four design elements: font size, the number of words, proportional area (% area) of
59 images, and the number of images for each video frame. The analysis result shows that the current video lectures are
60 not suitable for mobile learning environments with too small font size and dense text, which are not readable and
61 digestible on small screens. 86-98% of the collected video frames had too small font sizes and 60-82% had too dense text,
62 violating the design guidelines. This analysis result reveals the distribution and issues of the current lecture design.
63

64 We also conducted formative interviews with six video production engineers to investigate current practices and
65 challenges in content design for mobile devices. The engineers consider how the content they created will be displayed
66 on mobile devices. They mentioned an increasing number of learners watching a video lecture using various portable
67 devices and different learning environments. For this reason, they try to adapt the content to fit mobile devices by
68 resizing, repositioning, and segmenting the content. The main design factors they consider include font sizes and the
69 amount of information on mobile screens.
70

71 However, editing and tailoring video content remains time-consuming [11, 24, 32]. While the video production
72 engineers kept the mobile version of the video lecture in mind throughout the editing process, designing content that
73 fits the desktop and mobile environments at the same time can be challenging and sometimes requires tedious tasks.
74

75 To address this problem, we explored the feasibility of providing guideline-based feedback for the video content
76 design process. We present a prototype of a computer-guided video content design tool which allows engineers to view
77 the analytics and feedback for the content they have created.
78

79 In summary, the primary contributions of this research are:
80

- 81 • A content analysis that reveals a landscape of design elements of current video-based learning content and
82 automated inspection for design guideline compliance
- 83 • An exploration of challenges and current practices of video content design process for mobile learning environ-
84 ments through interviews with video production engineers
- 85 • A prototype system that supports video production engineers with computer-guided video content design
86
87
88

89 2 VIDEO-BASED LEARNING CONTENT ANALYSIS

90 To investigate the design practices of the current video designs and the suitability for mobile devices, we examined 681
91 video frames from 108 MOOC video lectures, which we selected from top MOOCs released by Class Central [40]. Class
92 Central is a review website for MOOCs and they release a list of the top MOOCs based on user reviews. The selected
93 MOOCs are hosted on Coursera, edX, and FutureLearn and are from 25 universities in 11 countries. We first collected
94 36 MOOCs offered in English with no talking head production. Talking-head style videos contain no design elements
95 which determine the suitability for mobile devices (e.g., images, text), so we excluded them. Except for the talking-head
96 style, the sampled set covered various common types of video production: presentation style, picture-in-picture style,
97 voice-over presentation style, Khan style tutorial, video screencast of the instructor [20, 35]. We then randomly selected
98 three video lectures from each course. The average length of 108 video lectures is 9.07 minutes. After the extraction of
99 video lectures, we sampled 681 video frames from the selected video lectures. They were extracted using edge-based
100
101
102
103
104

105
106
107
108
109
110
111
112
113
114
115
116
117
118
119
120
121
122
123
124
125
126
127
128
129
130
131
132
133
134
135
136
137
138
139
140
141
142
143
144
145
146
147
148
149
150
151
152
153
154
155
156

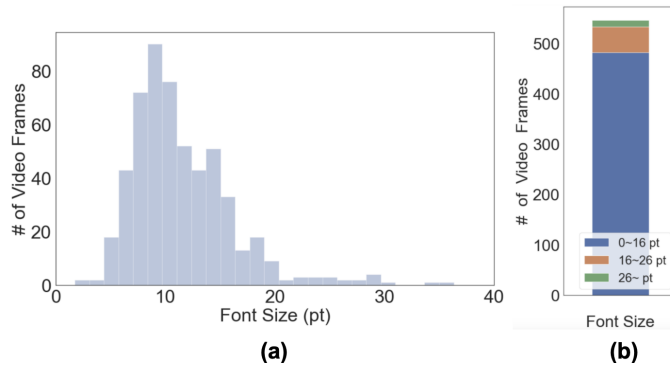


Fig. 1. (a) Distribution of font size in current video lecture content. The average font size of 681 video frames is 11.6 pt. (b) Compliance rate with existing design guidelines for font size. 86% of the video frames have font sizes smaller than 16 pt (Google Material Design Guidelines [19]) and 98% of them have font size smaller than 26 pt (presentation design guidelines [12, 23, 36]).

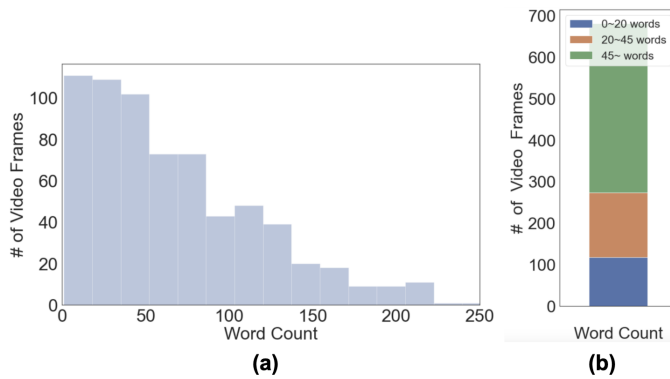


Fig. 2. (a) Distribution of number of words in current video lecture content. The video frames contain 73 words on average. (b) Compliance rate with existing design guidelines for number of words. 82% of the sampled video frames have more than 20 words (guidelines from [9]) and 60% of them contain more than 45 words (guidelines from [5]).

Design Element		Design Guidelines in Literature	Current MOOC Videos	Proportion of Inappropriate Design
Text Element	Avg Font Size	above 16-17 pt (mobile)[19, 26] above 24-26 pt (presentation)[12, 23, 36]	11.6 pt (all text) 11.2 pt (body text)	86-98 % (all text) 88-98 % (body text)
	Avg Number of Words	below 20-45 words[5, 9, 42]	73 words	60-82 %
Image Element	Avg Proportional Area of Images	as large as possible especially for complex images[14, 29]	27 % (talking heads not included)	-
	Avg Number of Images	maximum 2 images[14, 29]	0.88 images (talking heads not included)	8 % (talking heads not included)

Table 1. We examined 681 video frames from 108 MOOC video lectures to inform our exploration of current video lecture design. The proportional range of inappropriate design is estimated by a comparison of sampled video lecture design with existing design guidelines. For the list of video lectures analyzed in this paper, refer to the supplementary material.

157 frame difference which is used by previous research to measure the level of visual change in video lectures [13, 28, 47].
158 The number of video frames per video was 5.6 frames on average (from 1 frame to 31 frames per video lecture). We also
159 used the pytesseract OCR engine [37] for text detection, which shows reliable accuracy in existing work [25, 45, 47].
160

161 To understand the common design patterns of video lectures, we investigated four design features: font size, the
162 number of words, proportional area (% area) of images, and the number of images per video frame. We selected these
163 features based on literature [3, 18, 41] and interviews with video production engineers. According to the interviews,
164 they considered size and amount of text and image content as major factors that determine readability on mobile
165 screens.
166

167 We investigated the four design features of current lecture content in comparison with the design guidelines in
168 literature (Table 1). For comparison, we normalized the font size in video lectures since they are displayed in different
169 sizes depending on screen size and resolution of mobile devices. We used the most common mobile screen size: 5.5-inch
170 diagonal size with 1080 x 1920 screen resolution [1, 6] for normalization.
171

172 2.1 Font Size

173 The average font size of 681 video frames is 11.6 pt. The average font size of the body text, excluding the title text, is 11.2
174 pt. To estimate the compliance rate for design guidelines of current video content, we combined measures of font size
175 that are in different units (e.g., px, sp) to point (pt) for the comparison. Apple's Human Interface Guidelines adopt 17 pt
176 as a default body text size [26] and Google Material Design Guidelines suggest 16 pt as body text size [19], whereas the
177 guidelines for presentation slides encourage using font size above 24 or 26 pt in the body of the slide [12, 23, 36]. Based
178 on mobile design guidelines of Apple and Google, 86% of the video frames have font sizes smaller than 16 pt and 88% of
179 body text has font sizes smaller than 16 pt. Adopting the guidelines for presentation slides, over 98% of the video frames
180 have font size smaller than 26 pt. The distribution of font size and compliance rate for the design guidelines is shown in
181 Fig. 1. Regarding the font size guidelines for mobile environment and presentation slides, the analysis result of the
182 current video lectures implies that they might not be readable enough on small screens. Furthermore, the temporal and
183 transitional dynamics of video content can exacerbate the readability problem.
184
185
186
187
188

189 2.2 Number of Words

190 The video frames contain 73 words on average, which is in accordance with previous work that demonstrates the
191 average number of words per video frame is 69 [46]. The appropriate amount of text in multimedia learning content
192 and presentation slides is suggested by a body of previous work. The redundancy principle and modality principle of
193 multimedia learning theory suggests that learning efficiency increases with less text in multimedia learning content.
194 [27, 34, 41]. More specifically, using no more than 45 words per presentation slide is recommended [5] and more strict
195 guidelines advocate using less than 20 words per slide [9]. Another work advocates that the maximum number of words
196 per slide should be 25 [42]. In our video set, 82% of the sampled video frames have more than 20 words and 60% of
197 them contain more than 45 words, which suggests room for reducing the amount of content in accordance with the
198 suggested guidelines. The distribution of word count and compliance rate for the design guidelines is shown in Fig. 2.
199
200
201
202

203 2.3 Image

204 To analyze image elements in video lectures, we excluded 68 video frames with screencast on code editor or website since
205 we aim to examine the image elements which are added and adjusted deliberately by content engineers or instructors.
206 On the other hand, some of the video lectures display the instructor's talking head in a picture-in-picture mode. Since
207
208

209 talking heads are special types of visuals that are different from the static images, we excluded the picture-in-picture
210 talking heads from image analysis. The analysis result shows that the video frames contain 0.88 images with 27% of
211 image area not including the talking heads. The most common layout was the one with half text area and half image
212 area. With regard to the effects of images in learning material, Mayer's Multimedia Learning Theory demonstrates
213 that people learn better from words and pictures than from words alone [10, 33]. On the other hand, existing work on
214 lecture slide design for radiology recommends lecture slides to contain maximum two images in a single slide [14, 29].
215 In our video set, 8% of video frames contain more than two images per slide. The analysis result for images shows a
216 lower violation rate for the guidelines compared to that of text element from the perspective of legibility.
217

218 The analysis result indicates that the current video lectures with too small font size and dense text need enhancement
219 to be digestible and readable in mobile environments. The result motivated us to develop a design tool to support the
220 design process of mobile video-based learning content based on the known guidelines.
221
222

223 3 SEMI-STRUCTURED INTERVIEWS

224 We conducted semi-structured interviews with six video production engineers to investigate engineers' current practices
225 and challenges in mobile video-based learning content design.
226
227

228 3.1 Participants and Recruitment

229 We recruited six participants (5 male, 1 female) from the U.S. and South Korea via campus mailing lists. All participants
230 have more than 4 years of experience in educational video design (from 4 to 30 years). We selected the interviewees
231 based on the following criteria: the interviewee (1) is responsible for editing visual design elements such as adjusting
232 font size and the amount of information in a video lecture and (2) has experience in using design tools in the working
233 field. Of these, five participants are university staff and have design experience on MOOC content, and one is an
234 independent engineer for editing and publishing video-based learning content. We used a saturation method [7] to
235 determine the number of participants.
236
237
238

239 3.2 Interview Method

240 We asked them about the general design process and the main editing techniques they use for video lecture design and
241 editing. The interviews lasted about one hour. We audio-recorded the semi-structured interviews with permission and
242 they lasted about one hour. The full interview questions are included as supplementary material.
243
244
245

246 3.3 Analysis

247 The audio-recorded interviews were transcribed using either human or machine transcription. One of the authors then
248 manually corrected transcription errors. We analyzed them using thematic analysis methods [8]. We open-coded the
249 data and identified emergent patterns in a similar way to previous research [22] that investigated the design practices
250 of professional content designers. Here we summarize the results.
251
252

253 3.4 Mobile-First or Desktop-First Design

254 The engineers said they keep the mobile version of the content in mind throughout the design process. A major concern
255 regarding mobile devices was the small screen size. They modified font size, image size, and the amount of information
256 displayed on the screen to fit the small screen size. One participant sent internal design guidelines to instructors. This
257 participant noted that "I made this guideline to make my work more efficient. The instructors sometimes do not consider
258
259
260

261 *how the content is displayed in mobile environments. Viewers may not watch the lecture on a large desktop screen. They*
262 *might use smartphones or tablet PCs, so we try to use close-up images, large text with good contrast, and uncluttered slides*
263 *to be readable in small screens." (P2). Another participant also mentioned that his team has internal design guidelines for*
264 *minimum font size in mobile environments. The rest of them, however, did not have specific guidelines for design and*
265 *they explained that the quality of video design varies depending on engineers' style.*

267 One participant explained that *"We always test our content in mobile environments with small screens such as smart-*
268 *phones. All the design decisions on font size and the number of images are determined by legibility and readability in*
269 *mobile devices." (P1). "We have more users on mobile devices than on the desktop, so we enlarge and segment almost every*
270 *content." (P6).*

272 3.5 Main Editing Techniques

274 We asked the main editing technique to fit mobile devices to inform the design decisions of a content adaptation system.
275 The video production engineers pointed out that the instructors bring the lecture slides which are used in classroom
276 settings equipped with large screens which are not appropriate for learning video most of the time due to lack of
277 consideration for various devices. To fit the provided material to the mobile learning environment, the main techniques
278 they use were resizing, repositioning, and segmenting.

281 3.5.1 *Resizing.* They enlarge text and images to make them more visible in various screen sizes. One participant
282 mentioned 20 pt as the minimum font size and they were also using editing software such as Adobe Photoshop and
283 AfterEffects to zoom in complex graphics. The participant noted that *"We make sure that the fonts and graphics are large*
284 *enough to be seen in small screen size." (P2).*

287 3.5.2 *Reposition.* All participants said they reposition text and images by considering the visual flow of the presentation.
288 They move graphics and text to find the proper alignment and composition after resizing content to suit small screens.

290 3.5.3 *Segmenting.* The amount of information in a single frame was another important factor they consider while
291 designing content. One of our participants explained that *"I always need to segment a provided lecture slide into multiple*
292 *slides with less information in them and that is a very tedious task." (P1). Another participant also mentioned that "Too*
293 *much content on a slide turns people off. In such a case, we segment a single slide into two or three slides." (P2). They rarely*
294 *summarize or reformat the content from a lengthy paragraph to a list of bullet points since they cannot learn and*
295 *understand every content they edit. They also try to preserve the instructor's original intent without aggressive editing.*

299 3.6 Takeaways from the Interviews

300 Participants in the semi-structured interviews consider how the content they created will be displayed on mobile
301 devices. For this reason, they adapt the content to fit mobile devices by resizing, repositioning, and segmenting the
302 content. However, designing content that fits the various screen sizes is challenging and sometimes requires tedious
303 tasks.

307 4 COMPUTER-GUIDED VIDEO CONTENT DESIGN

308 We designed an initial prototype system to support engineers' design process of mobile video-based learning content.
309 The prototype system allows engineers to view the analytics and feedback for the content they have created. Fig. 6
310 shows the system which provides guideline-based content editing. The system provides analysis results for the slides
311

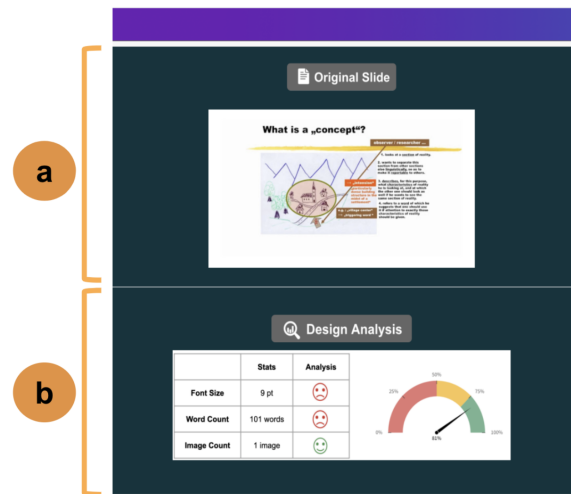


Fig. 3. (a) original slide: slide provided by instructors is displayed. (b) design analysis: analysis result of design based on existing guidelines is displayed.

provided by the instructors (Fig. 6(a)) based on the existing design guidelines (Fig. 6(b)). It provides statistics for three design features (font size, amount of text, amount of images) and whether the given slide complies with the guidelines or not. It displays smiling faces for the criteria that meet the guidelines and frowning faces for the criteria that do not. The gauge on the right of the statistics table displays the overall appropriateness of the design. The gauge has three parts - good, fair, bad.

5 EVALUATION

We conducted a formative evaluation of the prototype with four video production engineers from South Korea and the U.S. We recruited them via campus mailing lists. All participants have more than 2 years of experience in educational video design (from 2 to 30 years). They are responsible for editing video content such as adjusting font size and the amount of information in a video lecture. We asked for feedback on our prototype system and the interviews lasted about one hour. Here we summarize key findings drawn from the interviews.

Three participants mentioned that the provided design analysis for their content can increase the consistency of design. They noted that different design standards and preferences across engineers result in video content that varies in quality. Some participants mentioned that they want to customize the standards used in the design analysis as needed. *"The design guidelines would be helpful as a consistent indicator, but at the same time, I want to modify the metric considering the characteristics and context of the content."* (P2). One participant noted that the analysis can help them detect a flawed slide. *"Given dozens of lecture slides from an instructor, we sometimes miss out on slides with improper design. The system can function as an alert system."* (P4). These findings reveal design implications for a design tool for video-based learning content across devices.

6 CONCLUSION AND FUTURE WORK

We discuss findings and possible extensions of this work.

In this paper, we examined the mobile learning adequacy of current video-based learning material with 681 video frames from 108 lectures. The content analysis revealed a distribution of visual design elements and guideline compliance rate of the main visual design elements. We also conducted semi-structured interviews with six video production engineers to investigate engineers' current practices and challenges in mobile video-based learning content design. Finally, we present a prototype system for engineers and discuss design implications for the design of an authoring tool for engineers.

In future work, we first plan to conduct a content analysis including extended design factors such as font styles, font colors, visual complexities of graphics. Second, we can develop a design tool for engineers or plug-in software that can be integrated into the existing design tools being used by engineers. Third, we can extend the research to accessibility evaluation protocols and systems for the visually impaired, older adults, or dyslexics since readability and visibility of the content are the main factors of accessibility. We expect that the quality and accessibility of video-based learning content can be improved based on our research.

REFERENCES

- [1] Afiliastech.com. 2019 (accessed September 10, 2020). *Viewport, resolution, diagonal screen size and DPI for the most popular smartphones*. <https://deviceatlas.com/blog/viewport-resolution-diagonal-screen-size-and-dpi-most-popular-smartphones>
- [2] Ahmed Alamri, Zhongtian Sun, Alexandra I Cristea, Gautham Senthilnathan, Lei Shi, and Craig Stewart. 2020. Is MOOC Learning Different for Dropouts? A Visually-Driven, Multi-granularity Explanatory ML Approach. In *International Conference on Intelligent Tutoring Systems*. Springer, 353–363.
- [3] Michael Alley and Kathryn A Neeley. 2005. Rethinking the design of presentation slides: A case for sentence headlines and visual evidence. *Technical communication* 52, 4 (2005), 417–426.
- [4] Mohamed Ally. 2005. Using learning theories to design instruction for mobile learning devices. *Mobile learning anytime everywhere* (2005), 5–8.
- [5] Gerald J Alred, Charles T Brusaw, and E Oliu Walter. [n.d.]. *Handbook of Technical Writing*. Bedford/St. Martins, Boston, MA, USA, 2006. *paperback*, 0-312-35267-0 (*hardcover*). xxiv ([n. d.]).
- [6] Shaun Anderson. 2020 (accessed September 10, 2020). *What Are The Best Screen Sizes For Responsive Web Design?* <https://www.hobo-web.co.uk/best-screen-size/>
- [7] H Russell Bernard and Harvey Russell Bernard. 2013. *Social research methods: Qualitative and quantitative approaches*. Sage.
- [8] Virginia Braun and Victoria Clarke. 2006. Using thematic analysis in psychology. *Qualitative research in psychology* 3, 2 (2006), 77–101.
- [9] Sabra Brock and Yogini Joglekar. 2011. Empowering PowerPoint: Slides and teaching effectiveness. *Interdisciplinary Journal of Information, Knowledge, and Management* 6, 1 (2011), 85–94.
- [10] Kirsten R Butcher. 2014. The multimedia principle. *The Cambridge handbook of multimedia learning* 2 (2014), 174–205.
- [11] Juan Casares, A Chris Long, Brad A Myers, Rishi Bhatnagar, Scott M Stevens, Laura Dabbish, Dan Yocum, and Albert Corbett. 2002. Simplifying video editing using metadata. In *Proceedings of the 4th conference on Designing interactive systems: processes, practices, methods, and techniques*. 157–166.
- [12] Terence Cavanaugh and Catherine Cavanaugh. 2000. Interactive PowerPoint for teachers and students. In *Society for Information Technology & Teacher Education International Conference*. Association for the Advancement of Computing in Education (AACE), 496–499.
- [13] Dipesh Chand and Hasan Ogul. 2020. Content-Based Search in Lecture Video: A Systematic Literature Review. In *2020 3rd International Conference on Information and Computer Technologies (ICICT)*. IEEE, 169–176.
- [14] H Christian Davidson and Richard H Wiggins. 2003. Radiology teaching presentation tools. In *Seminars in ultrasound, CT, and MR*, Vol. 24. 420–427.
- [15] Inge De Waard, Apostolos Koutropoulos, Rebecca J Hogue, Sean C Abajian, Nilgün Özdamar Keskin, C Osvaldo Rodriguez, and Michael Sean Gallagher. 2012. Merging MOOC and mLearning for increased learner interactions. *International Journal of Mobile and Blended Learning (IJMBL)* 4, 4 (2012), 34–46.
- [16] Inge DeWaard, Sean Abajian, Michael Sean Gallagher, Rebecca Hogue, Nilgün Keskin, Apostolos Koutropoulos, and Osvaldo C Rodriguez. 2011. Using mLearning and MOOCs to understand chaos, emergence, and complexity in education. *International Review of Research in Open and Distributed Learning* 12, 7 (2011), 94–115.
- [17] Inge deWaard, Apostolos Koutropoulos, N Keskin, Sean C Abajian, Rebecca Hogue, C Osvaldo Rodriguez, and Michael Sean Gallagher. 2011. Exploring the MOOC format as a pedagogical approach for mLearning. In *Proceedings of 10th World Conference on Mobile and Contextual Learning*. 138–145.
- [18] Francis T Durso, Vlad L Pop, John S Burnett, and Eric J Stearman. 2011. Evidence-based human factors guidelines for PowerPoint presentations. *Ergonomics in Design* 19, 3 (2011), 4–8.

- [19] GoogleLLC. 2020 (accessed September 10, 2020). *The type system (Material Design)*. <https://material.io/design/typography/the-type-system.html#type-scale>
- [20] Philip J Guo, Juho Kim, and Rob Rubin. 2014. How video production affects student engagement: An empirical study of MOOC videos. In *Proceedings of the first ACM conference on Learning@ scale conference*. 41–50.
- [21] Vered Halamish. 2018. Can very small font size enhance memory? *Memory & cognition* 46, 6 (2018), 979–993.
- [22] Jane Hoffswell, Wilmot Li, and Zhicheng Liu. 2020. Techniques for Flexible Responsive Visualization Design. In *Proceedings of the 2020 CHI Conference on Human Factors in Computing Systems*. 1–13.
- [23] J Holzl. 1997. Twelve tips for effective PowerPoint presentations for the technologically challenged. *Medical Teacher* 19, 3 (1997), 175–179.
- [24] Xian-Sheng Hua, Zengzhi Wang, and Shipeng Li. 2005. LazyCut: content-aware template-based video authoring. In *Proceedings of the 13th annual ACM international conference on Multimedia*. 792–793.
- [25] Moula Husain, SM Meena, Akash K Sabarad, Harish Hebballi, Shiddu M Nagaralli, and Sonal Shetty. 2015. Counting occurrences of textual words in lecture video frames using apache hadoop framework. In *2015 IEEE International Advance Computing Conference (IACC)*. IEEE, 1144–1147.
- [26] Apple Inc. 2020 (accessed September 10, 2020). *Typography (Human Interface Guidelines)*. <https://developer.apple.com/design/human-interface-guidelines/ios/visual-design/typography/>
- [27] Nabil Issa, Mary Schuller, Susan Santacaterina, Michael Shapiro, Edward Wang, Richard E Mayer, and Debra A DaRosa. 2011. Applying multimedia design principles enhances learning in medical education. *Medical education* 45, 8 (2011), 818–826.
- [28] Hyeungshik Jung, Hijung Valentina Shin, and Juho Kim. 2018. DynamicSlide: Exploring the Design Space of Reference-based Interaction Techniques for Slide-based Lecture Videos. In *Proceedings of the 2018 Workshop on Multimedia for Accessible Human Computer Interface*. 33–41.
- [29] Natasha Larocque, Stephanie Kenny, and Matthew DF McInnes. 2015. Medical school radiology lectures: what are determinants of lecture satisfaction? *American Journal of Roentgenology* 204, 5 (2015), 913–918.
- [30] Hyunjeong Lee, Jan L Plass, and Bruce D Homer. 2006. Optimizing cognitive load for learning from computer-based science simulations. *Journal of educational psychology* 98, 4 (2006), 902.
- [31] Petra J Lewis. 2016. Brain friendly teaching—reducing learner’s cognitive load. *Academic radiology* 23, 7 (2016), 877–880.
- [32] A Chris Long, Brad Myers, Juan Casares, Scott Stevens, and Albert Corbett. 2004. Video Editing Using Lenses and Semantic Zooming. (2004).
- [33] Richard Mayer and Richard E Mayer. 2005. *The Cambridge handbook of multimedia learning*. Cambridge university press.
- [34] Richard E Mayer, Julie Heiser, and Steve Lonn. 2001. Cognitive constraints on multimedia learning: When presenting more material results in less understanding. *Journal of educational psychology* 93, 1 (2001), 187.
- [35] Ozlem Ozan and Yasin Ozarslan. 2016. Video lecture watching behaviors of learners in online courses. *Educational Media International* 53, 1 (2016), 27–41.
- [36] Lesley Pugsley. 2010. How To... Design an effective power point presentation. *Education for Primary Care* 21, 1 (2010), 51–53.
- [37] PythonSoftwareFoundation. 2020 (accessed September 10, 2020). *pytesseract 0.3.6*. <https://pypi.org/project/pytesseract/>
- [38] Matthew G Rhodes and Alan D Castel. 2008. Memory predictions are influenced by perceptual information: evidence for metacognitive illusions. *Journal of experimental psychology: General* 137, 4 (2008), 615.
- [39] Anna C Seale, Maryirene Ibeto, Josie Gallo, Olivier le Polain de Waroux, Judith R Glynn, and Jenny Fogarty. 2020. Learning from each other in the COVID-19 pandemic. *Wellcome Open Research* 5, 105 (2020), 105.
- [40] Dhawal Shah. 2019 (accessed September 10, 2020). *Class Central’s Top 100 MOOCs of All Time (2019 edition)*. <https://www.classcentral.com/report/top-moocs-2019-edition/>
- [41] Dom Shibli. 2019. Using Cognitive Load Theory to improve the use of slideshow presentations and support a more efficient learning process. *Blended Learning in Practice* (2019), 50.
- [42] Karen Stein. 2006. The dos and don’ts of PowerPoint presentations. *Journal of the American Dietetic Association* 106, 11 (2006), 1745–1748.
- [43] John Sweller. 1994. Cognitive load theory, learning difficulty, and instructional design. *Learning and instruction* 4, 4 (1994), 295–312.
- [44] John Sweller, Jeroen JG Van Merriënboer, and Fred GWC Paas. 1998. Cognitive architecture and instructional design. *Educational psychology review* 10, 3 (1998), 251–296.
- [45] Haojin Yang, Maria Siebert, Patrick Luhne, Harald Sack, and Christoph Meinel. 2011. Automatic lecture video indexing using video OCR technology. In *2011 IEEE International Symposium on Multimedia*. IEEE, 111–116.
- [46] Haojin Yang, Maria Siebert, Patrick Luhne, Harald Sack, and Christoph Meinel. 2011. Lecture video indexing and analysis using video ocr technology. In *2011 Seventh International Conference on Signal Image Technology & Internet-Based Systems*. IEEE, 54–61.
- [47] Baoquan Zhao, Songhua Xu, Shujin Lin, Ruomei Wang, and Xiaonan Luo. 2019. A New Visual Interface for Searching and Navigating Slide-Based Lecture Videos. In *2019 IEEE International Conference on Multimedia and Expo (ICME)*. IEEE, 928–933.
- [48] Ting Zhou, Sufang Huang, Jing Cheng, and Yaru Xiao. 2020. The Distance Teaching Practice of Combined Mode of Massive Open Online Course Micro-Video for Interns in Emergency Department During the COVID-19 Epidemic Period. *Telemedicine and e-Health* 26, 5 (2020), 584–588.