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An Extension of the QWERTY Effect: Not Just the Right Hand, Expertise and Typability

Predict Valence Ratings of Words

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Abstract

Typing is a ubiquitous daily action for many individuals; yet, research on how these actions 12 have changed our perception of language is limited. One such influence, deemed the 13 QWERTY effect, is an increase in valence ratings for words typed more with the right hand 14 on a traditional keyboard (Jasmin & Casasanto, 2012). Although this finding is intuitively 15 appealing given both right handed dominance and the smaller number of letters typed with 16 the right hand, extension and replication of the right side advantage is warranted. The 17 present paper reexamined the QWERTY effect within the embodied cognition framework 18 (Barsalou, 1999) and found that the right side advantage is replicable to new valence stimuli, 19 along with findings supporting embodied cognition. Further, when examining expertise, right 20 side advantage interacted with typing speed and typability (i.e., alternating hand key presses 21 or finger switches) portraying that both skill and our procedural actions play a role in 22 judgment of valence on words. 23

24 Keywords: expertise, embodied cognition, valence, QWERTY

25 Word count: 4419

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Predict Valence Ratings of Words

From its creation in 1868, to its appearance in our homes today, the QWERTY 28 keyboard has held the interest of psychologists. The process of typing on a keyboard requires 29 many procedures to function in tandem, which creates a wealth of actions to research (Inhoff 30 & Gordon, 1997). Rumelhart and Norman (1982)'s computer model of skilled typing is still 31 highly influential. They hypothesize that typing results from the activation of three levels of cognition: the word level, the key press level, and the response level. They believe that after 33 word perception, the word level is activated, causing the key press level to initiate a schema of the letters involved in typing the word. This schema includes the optimal position on the 35 keyboard for that specific hand-finger combination to move to at the appropriate time for individual keystrokes. Concurrently, the response system sends feedback information to initiate a key press motion when the finger is in the appropriate space. Their theory proposes that schemata and motion activations occur simultaneously, constantly pulling or pushing the hands and fingers in the right direction. While many studies have focused on errors in typing to investigate response system 41 feedback (Logan, 1999), Logan (2003) argued for parallel activation of key presses. He 42 examined the Simon effect to show more than one letter is activated at the same time, and consequently, the second key press motion is begun before the first key press is done. The Simon effect occurs when congruent stimuli create faster responses than incongruent stimuli, much like the Stroop task (Simon, 1990; Simon & Small, 1969). For example, if we are asked to type the letter f (a left handed letter), we type it faster if the f is presented on the left side of the screen. Similarly, Rieger (2004) reported finger-congruency effects by altering a Stroop task: participants were required to respond to centrally presented letters based on color-key combinations. When the letter and color were congruent (i.e., a right-handed letter

was presented in the designated color for a right response), the skilled typists' responses were

faster than incongruent combinations. Further, this effect was present when participants

responded to items with their hands crossed on the responding device, suggesting the effect was expertise-based rather than experiment-response based. These results imply that automatic actions stimulate motor and imagery representations concurrently and may be linked together in the brain (Hommel, Müsseler, Aschersleben, & Prinz, 2001; Logan & Zbrodoff, 1998; Rieger, 2004). This dual activation of motor and imagined items is the basis for embodied cognition, a rapidly expanding field in psychology (Barsalou, 1999; Salthouse, 1986).

60 Embodied Cognition

While the mind was traditionally considered an abstract symbol processor (Newell & 61 Simon, 1976), newer cognitive psychology theories focus on the interaction between the 62 brain's sensorimotor systems and mental representations of events and objects (Barsalou, 63 1999; Zwaan, 1999). The interplay between these systems has been found in both neurological (Hauk, Johnsrude, & Pulvermüller, 2004; Lyons et al., 2010; Tettamanti et al., 2005) and behavioral research (Cartmill, Goldin-Meadow, & Beilock, 2012; Holt & Beilock, 2006; Zwaan & Taylor, 2006). Motor representations of tasks are activated even when not specifically asked to perform the task, and if the action is well-learned, the task is perceived as pleasant (Beilock & Holt, 2007; Ping, Dhillon, & Beilock, 2009; Yang, Gallo, & Beilock, 2009). For example, Beilock and Holt (2007) asked novice and expert typists to pick which one of two letter dyads they preferred, which were either different hand combinations (CJ)71 or same finger combinations (FV). They found that novices have no preference in selection, while expert typists more reliably picked the combinations that were easier to type. To show that this effect was due to covert motor representation activation, and thus, expanding on findings from van den Bergh, Vrana, and Eelen (1990), participants also made preference selections while repeating a key press combination. When expert motor planning was distracted by remembering the pattern presented, no preference for letter dyads was found, indicating that the simultaneous activation of the motor representation was necessary to

- ⁷⁹ influence their likability ratings. Similar embodied findings have also been portrayed with
- emotionally charged sentences and facial movements (Havas, Glenberg, & Rinck, 2007),
- positive-negative actions, such as head nodding or arm movements (Glenberg, Webster,
- Mouilso, Havas, & Lindeman, 2009; Ping et al., 2009), and perceptuomotor fluency
- 83 (Oppenheimer, 2008; Yang et al., 2009).

84 Body Specificity Hypothesis

Using an embodied framework, Casasanto (2009) has proposed that handedness 85 dictates preference because our representations of actions are grounded in our physical interactions with the environment. In several studies, he portrayed that handedness influenced preference for spatial presentation (i.e., left handed individuals associate "good" with left, while right handed individuals associate "good" with right), which in turn influenced judgments of happiness and intelligence and our decision making in hiring job candidates and shopping. In all these studies, participants reliably selected the 91 hand-dominant side more often, which does not match cultural or neurolinguistic representations of positive-is-right and negative-is-left (Davidson, 1992). These findings 93 imply that our handedness is a motor expertise that causes ease of action on the dominant side to positively influence our perceptions of items presented on that side. Further, Casasanto (2011) compiled a review of body specific actions and their representation in the brain using fMRIs. Handedness interacted with imagining actions, reading action, and 97 perceiving the meanings of action verbs, such that fMRI patterns were mirrored for left and right handed participants matching their dominant side.

The QWERTY Effect

These effects lead Jasmin and Casasanto (2012) to propose the idea that typing, an action that often replaces speaking, has the ability to create semantic changes in how we perceive words. The asymmetrical arrangement of letters on the QWERTY keyboard increases fluency of typing letters on the right side because there are fewer keys, and thus,

less competition for fingers. That arrangement should then cause us to perceive the letters on the right side as more positive and letters on the left side as more negative. Consequently, words that are composed of more letters from the right side (the right side advantage; RSA) should be rated as more positive than those with more letters on the left. They found this preference for RSA over three languages (English, Spanish, and Dutch), and the effect was even stronger on words created after the invention of the QWERTY keyboard (i.e., lol), as well as evident in pseudowords such as plook. However, in contrast to the body specificity hypothesis, left and right handed participants showed the same trend in effects for positive-is-right words.

Current Study

The current study examined the right side advantage's interaction with traditional embodied cognition definitions (expertise, fluency). We analyzed the different implications of the body specificity hypothesis and a more general embodied hypothesis by testing the following:

- 1) Expertise was measured through participant typing speed, and fluency or typability was measured through finger and hand switches that would occur if the word was typed on a QWERTY keyboard (akin to Beilock and Holt (2007)'s different hand preferences). Given that typing involves the procedural action system, we would expect to find that increased hand and finger switches are positively related to ratings of valence because words that are typed on alternating fingers and hands are easier to type. It was unclear if expertise would directly influence overall ratings, as we expected an interaction of the variables (described below). The RSA may still be present when accounting for these variables, as humans are primarily a right side dominant species.
- 2) The interaction between RSA, hand and fingers switches, and expertise was examined to determine if these hypotheses can be combined. This analysis allowed us to explore the nuance of skill and typability on valence ratings and to determine the effects of the

RSA at different levels these variables.

132 Method

33 Participants

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Participants (N = 606) were recruited from the undergraduate human subject pool at 134 a large Midwest university and received course credit for their time. 72433 rows of data were 135 present for these participants, where 564 participants had complete data (i.e., 120 rows, see 136 below), 39 were missing one data point, and 3 were missing many data points. All possible 137 data points were considered and missing data points were usually computer error (i.e., 138 freezing during the experiment) or participant error (i.e., missed key press). 139 Rating data were screened for multivariate outliers, and two participant's ratings were 140 found to have extreme Mahalanobis distance scores (Tabachnick & Fidell, 2012) but were 141 kept in the data set. 11.2 percent of the sample was left-handed, 0.2 percent marked 142 ambidextrous, and 0.3 percent was missing handedness information. The average typing 143 speed was 47.89 words per minute (SD = 13.31), and the average percent accuracy rate for 144 the typing test was 92.69 (SD = 8.36). 145

146 Materials

The English ANEW (Bradley & Lang, 1999) norms were used to create the stimuli for this study, in an effort to replicate Jasmin and Casasanto (2012) experiments, and 2743 words were selected for this experiment. Pseudowords were selected from Appendix E of the supplementary materials presented from the QWERTY publication. These words were coded as described below for RSA, finger and hand switches, word length, and letter frequency. Average word length was 4.75 (SD = 1.47; range = 3 - 13). All materials, data, and the Rmarkdown document that created this manuscript are available at our Open Science Foundation (OSF) page: https://osf.io/zs2qj/.

155 Coding

Each of the words used in this study were coded for control and experimental variables. 156 Control variables included word length and average letter frequency. Average letter frequency 157 was created by averaging the English letter frequency (Lewand, 2000) for each letter in a 158 word. Words with high average letter frequencies contain more commonly used letters (e, t, 159 (z, q, x, j), while words with lower frequencies use more of the less common letters (z, q, x, j). 160 Experimental variables included RSA, number of hand switches, and number of finger 161 switches. Typing manuals were consulted, and letters were coded as left (q, w, e, r, t, a, s, d, s,162 f, g, z, x, c, v, b) or right-handed letters (y, u, i, o, p, h, j, k, l, n, m). Left handed letters 163 were coded with -1 and right handed letters with +1, which created summed scores 164 indicating the overall right side advantage for a word. Words were coded for the number of 165 hand switches within a word using the left-right coding system described above. Finally, the number of finger switches were coded using traditional typing manuals for each finger. Finger switches was highly correlated with word length, r = .89, and therefore, word length 168 was excluded as a control variable due to focus on typing skill in our hypotheses. 169

170 Procedure

Upon consent to participate in the experiment, participants were given a typing test by 171 using a free typing test website (TypingMaster, 2013). Each participant typed Aesop's 172 Fables for one minute before the website would reveal their typing speed and accuracy rate, 173 which was recorded by the experimenter. After this test, participants indicated their 174 dominant writing hand. Participants were then given 120 of the possible stimuli to rate for pleasantness (60 real words, 60 pseudowords). This smaller number of stimuli was used to control fatigue/boredom on participants. These stimuli were counterbalanced across 177 participants, and the order of the stimuli was randomized. Participants were told to rate 178 each word for how pleasant it seemed using a 9 point Likert type scale (1 - very unpleasant, 4 179 - neutral, 9 - very pleasant). The same self-assessment manikin from Jasmin and Casasanto

(2012) was shown to participants at the top of the computer screen to indicate the points on the Likert scale. The words appeared in the middle of the screen in 18 point Arial font. Participants then typed the number of their rating on the computer keyboard. Once they rated all stimuli, participants were debriefed and allowed to leave.

185 Results

Data Analytic Plan

Because each participant constituted multiple data points within the dataset, a 187 multilevel model was used to control for correlated error (Gelman, 2006). Pinheiro, Bates, 188 Debroy, Sarkar, and Team (2017)'s nlme package in R was used to calculate these analyses. 189 A maximum likelihood multilevel model was used to examine hypotheses of interactions 190 between typing speed, hand/finger switching, and RSA while adjusting for letter frequency 191 when predicting item pleasantness ratings. Pseudowords and real words were examined 192 separately in two multilevel model analyses. Participants were included as a random 193 intercept factor, as comparison to a non-random intercept was significant (see Table 1). 194 Typing speed, finger/hand switches, and RSA were mean centered before analyses to control 195 for multicollinearity. 196

197 Main Effects

After setting participants as a random intercept factor, letter frequency was used as an adjustor variable. As seen in Table 1, this variable was not a significant predictor for pseudowords, b = -0.006, but was a significant predictor for real words, b = 0.056. All predictor statistics are provided in an Excel sheet on the OSF page for each step of the model. Next, the main effects of typing speed, hand switches, finger switches, and RSA were added to the models for pseudowords and real words. In both models, the addition of these variables overall was significant, p < .001. For pseudowords, typing speed was not a significant predictor of valence ratings, b = 0.003, t(601) = 0.97, p = .332. Similarly, typing

speed was not a significant predictor for valence ratings on real words, b = 0.000, t(604) =206 0.04, p = .971. In contrast, the measures of typability in hand and finger switching were 207 significant for both pseudowords and real words. For pseudowords, increased hand switching, 208 b = -0.026, t(35535) = -2.84, p = .004, and increased finger switching, b = -0.074, t(35535)209 = -5.85, p < .001, decreased the overall valence ratings. However increased hand switching, b 210 = 0.061, t(35681) = 4.71, p < .001, increased valence ratings for real words, while increased 211 finger switching, b = -0.091, t(35681) = -7.82, p < .001, decreased the overall valence ratings. 212 Even adjusting for these typing style variables, the RSA effect replicated for both 213 pseudowords, b = 0.050, t(35535) = 11.50, p < .001, and real words, b = 0.051, t(35681) = 0.051214 8.35, p < .001. In the next section, we explored the interactions of typability and RSA, to 215 present a more nuanced view of typing's effect on valence ratings. 216

217 Interactions

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Next, the four-way interaction of typing speed, finger switching, hand switching, and 218 RSA was entered into the equation, including all the smaller two- and three-way interactions. 219 We focused on the most complex interaction found, breaking down interaction terms into 220 simple slopes of low (-1SD), average, and high (+1SD) to explore each effect. For example, if 221 the four-way interaction was significant, one variable would be broken into simple slopes, and 222 the next most complex interactions would be examined. This procedure was iterated until 223 the interactions were no longer significant or only main effects were examined. When 224 multiple interactions were present, we choose a common variable to help break down the 225 interactions with the least number of steps. Table 1 portrays that the addition of the 226 interaction components was significant for both pseudoword, p = .003, and real word, p < .003.001, models. 228 **Pseudoword Simple Slopes.** For pseudowords, finger switches by RSA, b = 0.014, 229 t(35524) = 2.66, p < .001, and typing speed by RSA, b = -0.001, t(35524) = -2.00, p = .045230

were the only significant interactions. Low and high simple slopes for RSA were created to

examine the effects of typing speed and finger switches at these levels, and these interactions 232 are displayed in Figure 1. For low RSA (words with more left handed letters), speed 233 positively predicted valence, b = 0.007, t(541) = 2.08, p = .038, and finger switching 234 negatively predicted valence, b = -0.094, t(31984) = -5.29, p < .001. For average RSA, speed 235 no longer predicted valence, b = 0.003, t(601) = 1.07, p = .285, while finger switches still 236 negatively predicted valence, albeit to a lesser extent than at low RSA, b = -0.062, t(35524)237 = -4.76, p < .001. Last, at high RSA (more right handed words), speed did not predict 238 valence, b = 0.002, t(541) = 0.62, p = .536, and neither did finger switches, b = -0.002, 239 t(31984) = -0.09, p = .926. In sum, this interaction indicated that expertise may be seen as 240 positively influencing ratings for more left handed words, but was not a predictor of words 241 that were typed more with the right hand. When words were more left handed, there was a 242 negative influence of finger switching, but as we transition to more right handed words the number of switches did not influence valence ratings. These results seemed to indicate that expertise and typability were influential factors for left handed words, but the RSA washed 245 out these effects when rating right handed pseudowords. 246

Real Word Simple Slopes. For real words, the three-way interactions of finger 247 switch by hand switch by RSA, b = -0.009, t(35670) = -5.00, p < .001, and speed by finger 248 switch by hand switch, b = -0.001, t(35670) = -2.62, p = .009, were the largest significant 249 interaction predictors. Low and high simple slopes for finger switches were created to explore 250 the three-way interaction. For lower finger switches, the hand switching by RSA interaction 251 was significant, b = 0.024, t(32130) = 2.81, p = .005; however, the hand switches by speed 252 interaction was not significant, b = 0.001, t(32130) = 0.43, p = .664. At average finger switching, the hand switches by RSA interaction was not significant, b = -0.002, t(35670) =-0.38, p = .705, and neither was the hand switches by speed interaction, b = -0.001, t(35670)255 = -1.21, p = .228. At a higher number of finger switches the hand switches by RSA interaction was significant, b = -0.016, t(32130) = -2.91, p = .004, along with the hand 257 switches by speed interaction, b = -0.001, t(32130) = -2.64, p = .008. 258

For significant two-way effects of hand switch by RSA and hand switch by speed, we 259 then calculated the low and high simple slopes for hand switches, see Figure 2. Therefore, we 260 explored the low and high finger switch effects that were significant with low and high hand 261 switches for RSA and speed main effects. At low finger switches and low hand switches, RSA 262 was a significant predictor of valence, b = 0.026, t(32130) = 2.80, p = .005. Speed was not 263 examined because the two-way interaction was not significant. At low finger switches and 264 average hand switches, RSA was a stronger predictor of valence, b = 0.060, t(32130) = 4.04, 265 p < .001. Last, at low finger switches and high hand switches, RSA increased in strength, b 266 = 0.094, t(32130) = 3.71, p < .001. Therefore, at low numbers of finger switches, as hand 267 switching increased, the strength of the RSA positivity effect also increased. This result 268 implied that as words required switching hands, words with more right handed letters during 269 these switches were more likely to be rated positively in valence.

At a high number of finger switches, we found both speed and RSA interactions with 271 hand switching, see Figure 2 top right and bottom left panels. When there were low numbers 272 of hand switches for these words, RSA was a positive significant predictor, b = 0.099, 273 t(32130) = 7.48, p < .001, along with speed, b = 0.009, t(544) = 2.32, p = .021. As hand 274 switches increase, the effects of RSA and speed decrease. For high finger switches and 275 average hand switches, RSA was significant, b = 0.075, t(32130) = 8.58, p < .001, while 276 speed was not b = 0.005, t(544) = 1.80, p = .073. With high finger and hand switches, RSA 277 was significant but smaller than low and average, b = 0.052, t(32130) = 4.94, p < .001, and 278 speed was not a significant predictor, b = 0.001, t(544) = 0.54, p = .589. Therefore, at an 279 elevated number of finger switches, and a low number of hand switches, we found that RSA and speed were positive predictors of valence ratings. As hand switching and finger switching increased, the effects of expertise and RSA decreased. This result implied that the 282 coordination of controlling for finger and hand switching decreased the positive valence 283 effects of both RSA and expertise. All interaction statistics are included online in an Excel 284 sheet on our OSF page. 285

286 Discussion

These results replicated and extended the QWERTY effect to portray an interactive 287 view of expertise, typability, and RSA that lead to stronger valence ratings for words. The 288 QWERTY keyboard layout has influenced our perceptions of positivity, as hypothesized by 280 the body specificity hypothesis, but the complexity of typing and action has additionally 290 lead to changing valence ratings for words. This influence was examined in our study by 291 incorporating the work of Beilock and Holt (2007), wherein we measured typing speed as a 292 measure of expertise, as well as embodied fluency or action through coding the way words 293 would be typed with finger and hand switches. For pseudowords, we replicated the RSA 294 effect, and additionally, showed that finger and hand switches predicted valence ratings. 295 However, both switch variables were negative predictors, indicating that we dislike words 296 that switch hands and fingers when adjusting for RSA, speed, and letter frequency. One 297 interpretation of this finding may be that pseudowords are, by definition, not traditionally 298 typed, which may have lead participants to rate words that required hand coordination, 299 along with concentration on the physical letters, as less positive. If we imagine typing a captcha (i.e., a set of letters and/or numbers designed to eliminate spam responses), we may 301 find that we would "peck" at the keyboard to hit the correct letter combination. Therefore, 302 words that would require us to use more hands and fingers may be less desirable. 303

For real words, the RSA effect was replicated, and both switch variables predicted valence ratings. In contrast to the pseudowords, we found that hand switching was a positive predictor of valence, while finger switching was a negative predictor of valence. Hand switching coordination would be easier to manage than finger switches, especially as we consider the flexibility and movement range of the non-index fingers. Therefore, it appeared that we found words on different hands as more positive, replicating Beilock and Holt (2007), but when forced to coordinate switching finger movements, we liked these words less. Many of the most frequent letters on the QWERTY keyboard are on the left side, which may frustrate a typist because of the need to coordinate finger press schemata that involve same

finger muscle movements (Rumelhart & Norman, 1982). Consequently, the number of switches becomes increasingly important to help decrease interference from the need to 314 continue to use the same hand. The ease of action by switching back and forth is then 315 translated as positive feelings for those fluent actions (Oppenheimer, 2008). The complexity 316 of this coordination's effect on valence was found in the multiway interactions unearthed in 317 this study. Globally, typing speed was not a significant predictor for pseudo or real words. 318 Viewing expertise through an embodied framework, it was unclear if speed would directly 319 affect valence, as speed was more likely to affect our interpretations of typing, rather than 320 positivity. Therefore, we examined the interaction of typability and speed to explore how 321 expertise might influence valence through ways that words are typed. 322

Pseudowords showed an interaction of typing speed by RSA and finger switching by 323 RSA when predicting valence. In this interaction, we focused on RSA as the common 324 variable between these interactions. When RSA was low, and thus, the words contained 325 more left-handed letters, we find that speed positively influenced valence, while finger 326 switches negatively predicted valence. For words typed completely on the right hand (high 327 RSA), neither variable influence valence. Therefore, it appears when we are required to use 328 the left hand, and thus, lessened the influence of RSA, typability and expertise play a role in the valence ratings of words. Both Beilock and Holt (2007) and van den Bergh et al. (1990) 330 showed expert preferences for two and three letter combinations that were typed with different fingers. Our results could imply that our embodied actions influence preferences for 332 procedures that are more likely in our environment. While our pseudowords were legal 333 English phoneme combinations, they are extremely unlikely to have been previously practiced or encountered in our daily tasks. Therefore, switching preference will not extend 335 to pseudowords (unpracticed actions) because they are not fluent (Oppenheimer, 2008). 336

Further, three-way interactions of finger switches by hand switches by RSA and finger switches by hand switches by speed were found for real word valence ratings. Finger switches were first separated in low, average, and high numbers of switches to see where the two-way

interactions were present. At low finger switches (less than two finger switches), only the hand switches by RSA interaction was present. This interaction indicated that increasing 341 hand switches also lead to increasing effects of RSA on valence. Therefore, when finger 342 switching competition was low, increased hand switching also lead to increased RSA effects. 343 This effect indicates that right handed words are still preferred, but additionally, we find 344 words that are typed with opposite hands as more positive. At average finger switching, we 345 found no two-way effects. However, at higher finger switching, we find both a speed and 346 RSA interaction with hand switching. For RSA, increasing levels of hand switching lead to lessening the impact of RSA. Therefore, when finger and hand switching needed to both be 348 coordinated, RSA's impact on valence decreased but was still significant. For speed, we 349 found that increasing levels of hand switching also lead to lessened effects of expertise. This 350 result runs counter to the idea that increased levels of hand and finger switching would require the most coordination, and thus, experts should be better at this task. This result instead implies that the effect of focusing on that coordination may dampen the effects of 353 expertise on valence ratings. 354

These embodied results mirror a clever set of studies by Holt and Beilock (2006) 355 wherein they showed participants sentences that matched or did not match a set of pictures 356 (i.e., the umbrella is in the air paired with a picture of an open umbrella). Given dual-coding 357 theory (Paivio, 1991), it was not surprising that participants were faster to indicate 358 picture-sentence matches than non-matches (also see Stanfield & Zwaan, 2001; Zwaan, 359 Stanfield, & Yaxley, 2002). Further, they showed these results extended to an expertise match; hockey and football players were much faster for sentence-picture combinations that matched within their sport than non-matches, while novices showed no difference in speed for matches or non-matches on sports questions. Even more compelling are results that these effects extend to fans of a sport and are consistent neurologically (i.e., motor cortex activation in experts; Beilock, Lyons, Mattarella-Micke, Nusbaum, & Small, 2008). These 365 studies clearly reinforce the idea that expertise and fluency unconsciously affect our choices, 366

even when it comes to perceived pleasantness of words.

This extension of the QWERTY effect illuminates the need to examine how skill and 368 action can influence cognitive processes. Additionally, typing style, while not recorded 369 directly in this experiment, could potentially illuminate differences in ratings across 370 left-handed and right-handed words. Hunt-and-peck typists are often slower than the strict 371 typing manual typists, which may eliminate or change the effects of RSA and switches since 372 typists may not follow left or right hand rules and just switch hands back and forth 373 regardless of key position. The middle of a QWERTY layout also poses interesting problems, 374 as many typists admit to "cheating" the middle letters, such as t, and y or not even knowing 375 which finger should actually type the b key. Further work could also investigate these effects 376 on other keyboard layouts, such as Dvorak, which was designed to predominately type by 377 alternating hands to increase speed and efficiency (Noyes, 1983).

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Table 1

Area under curve model statistics

Word Type	Model	df	AIC	BIC	χ^2	$\Delta \chi^2$	p
Pseudo	Intercept Only	2	144345.73	144362.72	-72170.87	NA	NA
Pseudo	Random Intercept	3	134813.09	134838.57	-67403.54	9534.65	< .001
Pseudo	Adjustor Variable	4	134814.22	134848.20	-67403.11	0.87	.351
Pseudo	Main Effects	8	134577.92	134645.89	-67280.96	244.29	< .001
Pseudo	Interactions	19	134577.46	134738.87	-67269.73	22.47	.021
Real	Intercept Only	2	168169.14	168186.14	-84082.57	NA	NA
Real	Random Intercept	3	166459.55	166485.05	-83226.78	1711.59	< .001
Real	Adjustor Variable	4	166424.46	166458.46	-83208.23	37.09	< .001
Real	Main Effects	8	166281.81	166349.81	-83132.91	150.65	< .001
Real	Interactions	19	166253.65	166415.14	-83107.82	50.16	< .001

Note. AIC: Aikaike Information Criterion, BIC: Bayesian Information Criterion

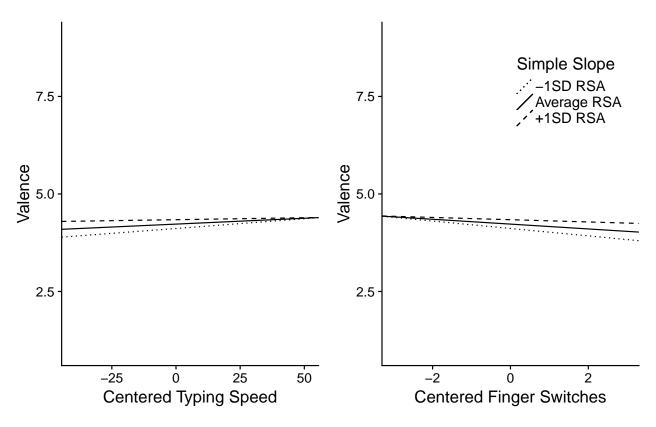


Figure 1. Simple slopes for pseudowords interaction effects. The left plot indicates the speed interaction across simple slopes of RSA, while the right plot indicates the interaction of finger switches and RSA. Speed has positive effects when RSA is low (left handed words), while finger switches have negative effects when RSA is low.

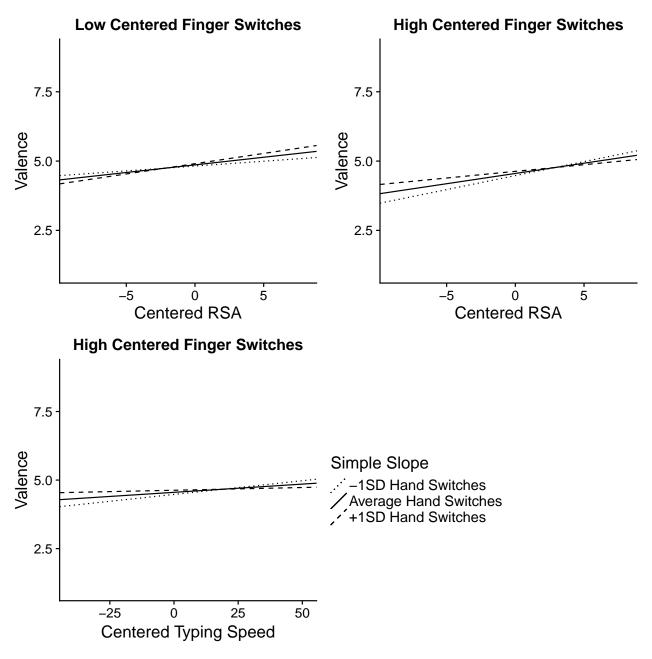


Figure 2. Simple Slopes for real word interactions of finger switchings by RSA by hand switches and finger switches by speed by hand switches. The top left figure indicates the interaction for RSA and hand switches at low finger switches. The average level of finger switches did not show an interaction. The top right panel portrays the interaction of RSA and hand switches at high simple slopes for finger switches. The bottom left figure shows the interaction of typing speed and hand switches at high finger switches. Low and average finger switches did not show this interaction.