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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. The QWERTY effect is an increase in 13 valence ratings for words typed more with the right hand on a traditional keyboard (Jasmin 14 & Casasanto, 2012). Although this finding is intuitively appealing given both right handed 15 dominance and the smaller number of letters typed with the right hand, extension and 16 replication of the right side advantage is warranted. The present paper reexamined the 17 QWERTY effect within the embodied cognition framework (Barsalou, 1999) and found that 18 the right side advantage is replicable to new valence stimuli, as well as experimental 19 manipulation. Further, when examining expertise, right side advantage interacted with 20 typing speed and typability (i.e., alternating hand keypresses or finger switches) portraying 21 that both skill and our procedural actions play a role in judgment of valence on words. 22

Keywords: keyboard, valence, QWERTY, word norms

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

Predict Valence Ratings of Words

From its creation in 1868, to its appearance in our homes today, the QWERTY

keyboard has held the interest of psychologists. The process of typing on a keyboard requires 27 many procedures to function in tandem, which creates a wealth of actions to research (Inhoff 28 & Gordon, 1997). Rumelhart and Norman (1982)'s computer model of skilled typing is still 29 highly influential. They hypothesize that typing results from the activation of three levels of cognition: the word level, the keypress level, and the response level. They believe that after 31 word perception, the word level is activated, causing the keypress level to initiate a schema 32 of the letters involved in typing the word. This schema includes the optimal position on the 33 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 keypress 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 39 feedback (Logan, 1999), Logan (2003) argued for parallel activation of keypresses. He examined the Simon effect to show more than one letter is activated at the same time, and consequently, the second keypress motion is begun before the first keypress 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).

## 58 Embodied Cognition

While the mind was traditionally considered an abstract symbol processor (Newell & 59 Simon, 1976), newer cognitive psychology theories focus on the interaction between the 60 brain's sensorimotor systems and mental representations of events and objects (Barsalou, 61 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; Holt2006; 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) 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 keypress combination. When expert motor planning was distracted by remembering the pattern presented, no preference for letter dyads was found, 75 indicating that the simultaneous activation of the motor representation was necessary to

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 (Oppenheimer, 2008; Yang et al., 2009).

# 82 Body Specificity Hypothesis

Using an embodied framework, Casasanto (2009) has proposed that handedness 83 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 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 91 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 perceiving the meanings of action verbs, such that fMRI patterns were mirrored for left and right handed participants matching their dominant side.

# 98 The QWERTY Effect

These effects inspired 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 103 on the right side as more positive and letters on the left side as more negative. Consequently, 104 words that are composed of more letters from the right side (the right side advantage; RSA) 105 should be rated as more positive than those with more letters on the left. They found this 106 preference for RSA over three languages (English, Spanish, and Dutch), and the effect was 107 even stronger on words created after the invention of the QWERTY keyboard (i.e.,lol), as 108 well as evident in pseudowords such as plook. However, in contrast to the body specificity 109 hypothesis, left and right handed participants showed the same trend in effects for 110 positive-is-right words. 111

## Current Study

The current study examined the right side advantage's interaction with traditional 113 embodied cognition definitions (expertise, fluency). We analyzed the different implications of 114 the body specificity hypothesis and a more general embodied hypothesis by testing the 115 following: 1) To examine embodied cognition, we coded each word for number of hand 116 alternations (akin to Beilock and Holt (2007)'s different hand preferences). Given that 117 typing involves the procedural action system, we would also expect to find that increased hand switches are positively related to ratings of valence because words that are typed on 119 alternating hands are easier to type. 2) The interaction between RSA and switches was 120 examined to determine if these hypotheses can be combined (i.e., we only like right handed 121 words because we have to switch back and forth to type the more commonly used letters, 122 such as e or a). 123

### Experiment 1

Method 125

## **Participants**

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Participants (N = 546) were recruited from the university undergraduate human subject pool and received course credit for their time. 65233 rows of data were present for 128 these participants, where 504 participants included complete data (i.e., 120 rows, see below), 129 39 were missing one data point, and 3 were missing many data points. All data points were 130 included, and missing data points were usually computer error (i.e., freezing during the 131 experiment) or participant error (i.e., missed key press). 132 Rating data were screened for multivariate outliers, and one participant's ratings were 133 found to have extreme Mahalanobis distance scores (???) but were kept in the data set. 11.5

134 percent of the sample was left-handed, 0.2 marked ambigdextrious, and 0.4 was missing 135 handedness information. The average typing speed was 48.20 (SD = 13.45, and the average 136 percent accuracy rate for the typing test was 92.59 (SD = 8.63).

#### Materials

Both Experiment 1 and Experiment 2 use the English ANEW (???) norms to create 139 stimuli for this study, in an effort to replicate Jasmin and Casasanto (2012) experiments, and 140 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 142 coded as described below for RSA, switches, word length, and letter frequency. Average word 143 length was 4.85 (SD = 1.51; range = 3 - 13). All materials, data, and the Rmarkdown document that created this manuscript are avaliable at our Open Science Foundation (OSF) 145 page: https://osf.io/zs2qj/.

### 147 Coding

Each of the words used in this experiment and Experiment 2 were coded for control 148 and experimental variables. Control variables included word length and average letter 149 frequency. Average letter frequency was created by averaging the English letter frequency 150 (???) for each letter in a word. Words with high average letter frequencies contain more 151 commonly used letters (e, t, a, o); while words with lower frequencies use more of the less 152 common letters (z, q, x, j). Experimental variables included RSA, number of hand switches, 153 and number of finger switches. Typing manuals were consulted, and letters were coded as 154 left (q, w, e, r, t, a, s, d, f, g, z, x, c, v, b) or right-handed letters (y, u, i, o, p, h, j, k, l, n, r, left)155 m). Left handed letters were coded with -1 and right handed letters with +1, which created 156 summed scores indicating the overall right side advantage for a word. Words were coded for 157 the number of 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 160 therefore, word length was excluded as a control variable due the interest in typing skill for 161 experimental hypotheses. 162

## 163 Procedure

Upon consent to participate in the experiment, participants were given a typing test by 164 using a free typing test website (???). Each participant typed Aesop's Fables for one minute 165 before the website would reveal their typing speed and accuracy rate, which was recorded by 166 the experimenter. After this test, participants indicated their dominant writing hand. 167 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 169 fatigue/boredom on participants. These stimuli were counterbalanced across participants, 170 and the order of the stimuli was randomized. Participants were told to rate each word for 171 how pleasant it seemed using a 9 point Likert type scale (1 - very unpleasant, 4 - 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.

178 Results

## Data Analytic Plan

Because each participant constituted multiple data points within the dataset, a 180 multilevel model was used to control for correlated error (???). (???)'s nlme package in R 181 was used to calculate these analyses. A maximum likelihood multilevel model was used to 182 examine hypotheses of interactions between typing speed, hand/finger switching, and RSA 183 while adjusting for letter frequency when predicting item pleasantness ratings. Pseudowords 184 and real words were examined separately in two multilevel model analyses. Participants were 185 included as a random intercept factor, as comparison to a non-random intercept was 186 significant (see Table 1). Typing speed, finger/hand switches, and RSA were mean centered 187 before analyses to control for multicollinearity. 188

### 189 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 = -0.005, but was a significant predictor for real words, b = 0.047. 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 = 0.0044, t(541) = 1.45, p = .148. Similarly, typing

speed was not a significant predictor for valence ratings on real words, b = -0.00049, t(544)198 = -0.22, p = .826. In contrast, the measures of typability in hand and finger switching were 199 significant for both pseudowords and real words. For pseudowords, increased hand switching, 200 b = -0.033, t(31995) = -3.35, p = .001, and increased finger switching, b = -0.065, t(31995)201 = -4.78, p < .001, decreased the overall valence ratings. However increased hand switching, b 202 = 0.038, t(32141) = 2.77, p = .006, increased valence ratings for real words, while increased 203 finger switching, b = -0.076, t(32141) = -6.29, p < .001, decreased the overall valence ratings. 204 Even adjusting for these typing style variables, the RSA effect replicated for both 205 pseudowords, b = 0.05, t(31995) = 10.63, p < .001, and real words, b = 0.048, t(32141) = 0.048206 7.39, p < .001. In the next section, we explored the interactions of typability and RSA, to 207 present a more nuanced view of typing's effect on valence ratings. 208

## 209 Interactions

Next, the four-way interaction of typing speed, finger switching, hand switching, and 210 RSA was entered into the equation, including all the smaller two- and three-way interactions. 211 We focused on the most complext interaction found, breaking down interaction terms into 212 simple slopes of low (-1SD), average, and high (+1SD) to explore each effect. For example, if 213 the four-way interaction was significant, one variable would be broken into simple slopes, and 214 the next most complex interactions would be examined. This procedure was iterated until 215 the interactions were no longer significant or only main effects were examined. When 216 multiple interactions were present, we choose a common variable to help break down the 217 interactions with the least number of steps. Table 1 portrays that the addition of the 218 interaction components was significant for both pseudoword, p = .003, and real word, p < .003.001, models. 220 **Pseudoword Simple Slopes.** For pseudowords, finger switches X RSA, b = 0.02, 221 t(31984) = 3.49, p < .001,and typing speed X RSA, b = -0.0011, t(31984) = -2.15, p = .031222 were the only significant interactions. 223

HERE I THINK WE SHOULD BREAK THIS DOWN BY RSA SINCE IT'S THE
SIMILAR ONE BETWEEN THE TWO.

Real Word Simple Slopes. For real words, the three-way interactions of finger switch X hand switch X RSA, b = -0.011, t(32130) = -5.88, p < .001, and speed X finger switch X hand switch, b = -0.00095, t(32130) = -2.64, p = .008, were the largest significant interaction predictors.

BREAK THIS DOWN BY FINGER, THEN HAND, SEE WHAT HAPPENS WITH
SPEED AND RSA

All interaction statistics are included online in an Excel sheet at our OSF page.

### Experiment 2

234 Method

## 235 Participants

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Similar to Experiment 1, 60 participants were recruited from the university undergraduate human subject pool and received course credit for their time. 7200 rows of data were present for these participants, and no data was missing. Rating data were screened for multivariate outliers. Again, part of one participant's ratings were found to have extreme Mahalanobis distance scores (???). However, this individual's ratings were left in the data set. Approximately 8.3 percent of the sample was left-handed. The average typing speed was 45.12 (SD = 11.65), and the average percent accuracy rate for the typing test was 93.58 (SD = 5.31).

#### $_{^{244}}$ Materials

In this experiment, a smaller subset of words (120) from Experiment 1 were used,
which were split evenly between pseudowords and real words. Average word length was 3.80(SD = 0.40; range = 3 - 4).

### Procedure Procedure

In this study, when participants were shown the word (or pseudoword) on the screen, they were first asked to type the word on the keyboard in front of them. After they had typed the word, they were then asked to rate the word for pleasantness using the scale and self-assessment manikin discussed previously.

253 Results

254 Main Effects

55 Interactions

256 Discussion

YADA SCHMADA CHANGE THIS SECTION These results imply that the 257 QWERTY keyboard has influenced our perceptions of words, in a more complex way than a 258 simple body specificity hypothesis. In the overall normed database analyses, the original 259 QWERTY effect was replicable across a large body of various types of stimuli (verbs, 260 Twitter, category norms), with much the same size of effect as Jasmin and Casasanto (2012) 261 published. Word length was often negatively related to valence ratings, which indicated that 262 we like shorter words to type. Average letter frequency was usually a positive predictor of 263 valence ratings wherein ratings are higher for words with more frequent letters; however, 264 these effects were inconsistent. Our measure of fluency (switches) varied across stimulus sets 265 but it appears, by analyzing multiple sources of ratings for words at the same time, that 266 there might have been an interaction between RSA and number of switches. This interaction portrayed that we find words that switch off of left-handed keypresses as more pleasant, while right-handed keypresses are preferable by switching hands less often. These effects were examined in more detail in Experiment 2, which incorporated Beilock and Holt (2007) study by including typing speed as a measure of expertise. Word ratings turned out to be 271 quite complex with a four-way interaction between real/pseudowords, switches, RSA, and

typing speed. All analyses showed a positive effect of right-side words, as well as if they were shorter and used more frequent letters. However, for pseudowords, no other effects were 274 significant. Both Beilock and Holt (2007) and Van den Bergh et al. (1990) showed expert 275 preferences for two and three letter combinations that were typed with different fingers. Our 276 results could imply that our embodied actions influence preferences for procedures that are 277 more likely in our environment. While our pseudowords were legal English phoneme 278 combinations, they are extremely unlikely to have been previously practiced or encountered 279 in our daily tasks. Therefore, switching preference will not extend to pseudowords 280 (unpracticed actions) because they are not fluent (Oppenheimer, 2008). 281

The effect of expertise was shown on real words, where the three-way interaction 282 between RSA, switches, and typing speed was examined by separating out right, equal, and 283 left-handed words. For right-handed words, typing speed (or the interaction) was not a 284 significant predictor of valence, and while not significant, number of switches was negatively 285 related to valence ratings. For equally right-left and left-handed words, pleasantness ratings 286 increase by switching back and forth to the right hand. Further, left-handed words showed 287 an interaction between our two embodied cognition variables, where the number of switches 288 increases valence ratings as the typing speed of the participant decreases. Therefore, it appears that as participants gain fluency through increased typing speed, the number of switches back and forth for left-handed words matters less for pleasantness ratings. Many of the most frequent letters on the QWERTY keyboard are on the left side, which may frustrate a slow typist because of the need to coordinate finger press schemata that involve 293 same finger muscle movements (Rumelhart & Norman, 1982). Consequently, the number of switches becomes increasingly important to help decrease interference from the need to 295 continue to use the same hand. The ease of action by switching back and forth is then 296 translated as positive feelings for those fluent actions (Oppenheimer, 2008). 297

These embodied results mirror a clever set of studies by Holt and Beilock (2006)
wherein they showed participants sentences that matched or did not match a set of pictures

(i.e., the umbrella is in the air paired with a picture of an open umbrella). Given dual-coding 300 theory (???), it was not surprising that participants were faster to indicate picture-sentence 301 matches than non-matches (also see ???, ???). Further, they showed these results extended 302 to an expertise match; hockey and football players were much faster for sentence-picture 303 combinations that matched within their sport than non-matches, while novices showed no 304 difference in speed for matches or non-matches on sports questions. Even more compelling 305 are results that these effects extend to fans of a sport and are consistent neurologically 306 (i.e., motor cortex activation in experts; ???). These studies clearly reinforce the idea that 307 expertise and fluency unconsciously affect our choices, even when it comes to perceived 308 pleasantness of words. 309

This extension of the QWERTY effect illuminates the need to examine how skill can 310 influence cognitive processes. Additionally, typing style, while not recorded in this experiment, could potentially illuminate differences in ratings across left-handed and 312 right-handed words. Hunt-and-peck typists are often slower than the strict typing manual 313 typists, which may eliminate or change the effects of RSA and switches since typists may not 314 follow left or right hand rules and just switch hands back and forth regardless of key position. 315 The middle of a QWERTY layout also poses interesting problems, as many typists admit to 316 "cheating" the middle letters, such as t, and y or not even knowing which finger should 317 actually type the b key. Further work could also investigate these effects on other keyboard 318 layouts, such as Dvorak, which was designed to predominately type by alternating hands to 319 increase speed and efficiency (???). 320

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Table 1  $Area\ under\ curve\ model\ statistics$ 

Word Type	Model	df	AIC	BIC	$\chi^2$	$\Delta \chi^2$	p
Pseudo	Intercept Only	2	130416.47	130433.25	-65206.24	NA	NA
Pseudo	Random Intercept	3	122403.95	122429.12	-61198.97	8014.52	< .001
Pseudo	Adjustor Variable	4	122405.44	122439.00	-61198.72	0.51	.476
Pseudo	Main Effects	8	122204.31	122271.44	-61094.16	209.13	< .001
Pseudo	Interactions	19	122197.68	122357.09	-61079.84	28.64	.003
Real	Intercept Only	2	151926.51	151943.30	-75961.26	NA	NA
Real	Random Intercept	3	150478.20	150503.38	-75236.10	1450.32	< .001
Real	Adjustor Variable	4	150457.14	150490.72	-75224.57	23.06	< .001
Real	Main Effects	8	150354.18	150421.34	-75169.09	110.96	< .001
Real	Interactions	19	150315.00	150474.50	-75138.50	61.18	< .001

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

Table 2

Area under curve model statistics

Word Type	Model	df	AIC	BIC	$\chi^2$	$\Delta \chi^2$	p
Pseudo	Intercept Only	2	13715.38	13727.76	-6855.69	NA	NA
Pseudo	Random Intercept	3	12190.54	12209.10	-6092.27	1526.85	< .001
Pseudo	Adjustor Variable	4	12192.17	12216.92	-6092.08	0.37	.543
Pseudo	Main Effects	8	12152.25	12201.76	-6068.13	47.91	< .001
Pseudo	Interactions	19	12127.85	12245.44	-6044.93	46.40	< .001
Real	Intercept Only	2	16133.38	16145.76	-8064.69	NA	NA
Real	Random Intercept	3	15922.15	15940.72	-7958.08	213.23	< .001
Real	Adjustor Variable	4	15898.69	15923.45	-7945.35	25.46	< .001
Real	Main Effects	8	15812.18	15861.69	-7898.09	94.51	< .001
Real	Interactions	19	15734.48	15852.06	-7848.24	99.71	< .001

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