1	Reliably Measuring Learning-Dependent Distractor Suppression with Eye Tracking				
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Abstract

28 In the field of psychological science, behavioral performance in computer-based 29 cognitive tasks often exhibits poor reliability. The absence of reliable measures of cognitive 30 processes contributes to non-reproducibility in the field and impedes investigation of individual differences. Specifically in visual search paradigms, response time-based measures have 31 32 shown poor test-retest reliability and internal consistency across attention capture and distractor 33 suppression, but one study has demonstrated the potential for oculomotor measures to exhibit 34 superior reliability. Therefore, in this study, we investigated three datasets to compare the 35 reliability of learning-dependent distractor suppression measured via distractor fixations 36 (oculomotor capture) and latency to fixate the target (fixation times). Our findings reveal superior 37 split-half reliability of oculomotor capture compared to that of fixation times regardless of the 38 critical distractor comparison, with the reliability of oculomotor capture in most cases falling 39 within the range that is acceptable for the investigation of individual differences. We additionally 40 find that older adults have superior oculomotor reliability compared with young adults, potentially 41 addressing a significant limitation in the aging literature of high variability in response time 42 measures due to slower responses. Our findings highlight the utility of measuring eye 43 movements in the pursuit of reliable indicators of distractor processing and the need to further 44 test and develop additional measures in other sensory domains to maximize statistical power, 45 reliability, and reproducibility.

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47 **Keywords**: reliability; attention capture; distractor suppression; visual search

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Introduction

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50 The field of psychological science was challenged in the past decade to improve the 51 replicability of behavioral research based on large scale examples of non-reproducibility 52 (Johnson et al., 2017; Open Science Collaboration, 2012, 2015). Nosek and colleagues define 53 reproducibility, robustness, and replicability as "testing the reliability of a prior finding" and 54 propose that maximizing the reliability of research findings will improve research credibility and 55 the translation of knowledge into application (Nosek et al., 2022). The reliability of 56 measurements is particularly important when maximizing the power of significance tests, and 57 measures with poor reliability are not sensitive in detecting individual differences (Zimmerman et 58 al., 1993). Researchers have commonly utilized two types of measurements of reliability: test-59 retest reliability and internal consistency (split-half correlation). These tests have often revealed 60 poor reliability of behavioral measures in the field of psychological science (Dang et al., 2020; 61 Draheim et al., 2019; Paap & Sawi, 2016), calling researchers in the field to identify and develop 62 more reliable measures that can be consistent across multiple experimental paradigms. 63 As the critical need for reliable measures is increasingly recognized, researchers utilizing visual search paradigms have recently highlighted the poor reliability of measures specifically 64 65 using behavioral response times. Ivanov et al. (2023) investigated whether difference scores in 66 manual response times and accuracy were reliable and could be utilized as an individual-level 67 measure. Utilizing both split-half and test-retest reliability measurements, the authors 68 investigated whether attention capture, learned distractor suppression at a high-probability 69 location in the visual search array, and corresponding suppression of targets at the high-

probability location could serve as reliable measures for investigating individual differences (Ivanov et al., 2023). Over the three measures, the authors report poor to moderate split-half reliability over response times and poor reliability over accuracy, in addition to poor test-retest reliability with respect to both response times and accuracy. Furthermore, three studies investigating selection history effects of reward learning in visual search also reported poor test-

retest reliability of behavioral response times (Anderson & Kim, 2019; Freichel et al., 2023;
Garre-Frutos et al., 2024). These studies collectively identified that response time exhibits poor
reliability over experience-driven attention effects. However, in Anderson and Kim (2019), valuedriven oculomotor capture exhibited strong test-retest reliability, suggesting that oculomotor
capture may be more sensitive and reliable in contrast to oculomotor fixation times and even
more so when compared with manual response times (Anderson & Kim, 2019; Weichselbaum et
al., 2018).

82 Therefore, in the current study, we investigated whether oculomotor measures of 83 distractor fixations provide superior reliability compared to response time-based measures 84 (fixation time or time to make an eve movement to the target). We investigated oculomotor 85 measures in three studies containing a total of 8 experiments that utilized a visual search task 86 incorporating attention capture and/or distractor suppression. The selected studies were limited 87 to investigating the reliability of distractor suppression in the context of selection history effects. 88 given pessimistic findings concerning manual response time measures (Ivanov et al., 2023). We 89 aimed to examine the reliability of oculomotor measures in visual search across multiple 90 experimental paradigms incorporating statistical learning of a high-probability distractor location, 91 learned value-associations with the distractor in a context in which these associations lead to 92 reduced distractor interference, and proactive distractor suppression (feature-search) vs. 93 reactive distractor disengagement (singleton-search). Thus, we look to evaluate the reliability of 94 oculomotor measures across numerous critical distractor comparisons. In two cases, data from 95 both older and younger adults was available, permitting an assessment of the reliability of 96 oculomotor measures as a function of age. Based on the findings of Anderson and Kim (2019), 97 we hypothesize that the reliability of oculomotor capture measures will be superior to that of 98 measures involving fixation time, and that these oculomotor measures will also demonstrate 99 high reliability that is superior to the characteristically low reliability associated with manual 100 response time measures as observed in the literature.

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102	Methods
103	Datasets
104	We evaluated three datasets that incorporated oculomotor measures in visual search
105	tasks to investigate the reliability of oculomotor capture by the distractor and fixation times
106	(oculomotor response times) between two critical distractor conditions (Grégoire et al., 2022;
107	Kim et al., 2024; Kim & Anderson, 2022). In Kim and Anderson (2022), the critical distractor
108	comparison was a distractor appearing at a high-probability location vs. a distractor appearing at
109	a low-probability location (statistical learning of a high-probability distractor location). In Grégoire
110	et al. (2022), the critical distractor comparison was previously conditioned distractors (CS+;
111	associated with reward or electric shock) vs. neutral distractors (value- and threat-modulated
112	attentional capture). In this latter study, we separated findings over the three experiments
113	(focusing on the first two in which distractor suppression was observed). In Kim et al. (2024), the
114	critical distractor comparison was attention capture by the distractor on distractor-present trials
115	(first saccade to the distractor) vs. first fixation to a single non-target in distractor-absent trials
116	(attention capture by a physically salient distractor when engaging in feature-search or
117	singleton-search mode); reliability scores were separated by both experiments (feature-search
118	vs. singleton-search) and calculated separately among young and older adult samples to probe
119	potential age differences.
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121	Split-Half Reliability

122 Instead of utilizing an arbitrary odd vs. even split, we estimated internal consistency by 123 utilizing a permuted random split procedure as in Garre-Frutos et al. (2024). In this procedure, 124 all trials were randomly split into two halves with an equal number of observations in each half 125 per condition per run to account for time-dependent effects (e.g., learning or extinction). Trials 126 for each half were then concatenated over all runs. Then, a difference score between the two

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127	critical distractor conditions was computed for each concatenated half for each participant and
128	correlated to get a Pearson's <i>r</i> correlation coefficient. This procedure was repeated 1000 times
129	and the correlation coefficients were averaged to compute the mean split-half correlation.
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131	Non-Parametric Randomization Tests
132	To determine whether estimates of reliability for oculomotor capture and fixation times
133	were significantly different across conditions, we conducted non-parametric randomization tests.
134	Based on the 1000 split-half correlation coefficients calculated for each measure (before
135	averaging), we first computed the mean of the difference scores between the oculomotor
136	capture and fixation time measures as the true sample mean. Then, from the combined 2000
137	coefficient values for both measures, we randomly assigned 1000 values to each measure to
138	create two unique sample groups and computed the difference of these group mean r values
139	(random sample), under the null hypothesis that there was no difference between split-half
140	reliability obtained using each measure and, thus, random assignment of reliability to a
141	dependent measure should tend to produce a similar difference score to the difference score
142	observed between the two measures in the actual data. This randomization procedure was
143	repeated 1000 times and the p-value was manually calculated from the z-score using the
144	observed sample mean.
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146	Data Availability
147	The datasets analyzed in the current study are publicly available in the Open Science
148	Framework repository, https://osf.io/fkj92/.
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150	Results
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151 Kim and Anderson (2022)

152 In Kim and Anderson (2022), visual search required fixating on a target shape singleton 153 in the absence and presence of a salient color singleton distractor. Critically, the location of the 154 color distractor in distractor-present trials was in a high-probability location 45% of the time and 155 equally often in the other low-probability locations (5 low-probability locations). When comparing 156 the oculomotor measures, the split-half correlation for the learning-dependent reduction in 157 oculomotor capture (probability of fixating the distractor on low-probability minus high-probability 158 trials) was r = .802 and for fixation time (latency to fixate the target on low-probability minus 159 high-probability trials) was r = .698. Using non-parametric randomization tests, we found that 160 the reliability of oculomotor capture was significantly superior compared to fixation time, p 161 < .001 (see Figure 1). 162

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166 **times.** Bar graphs depict Pearson's correlation values over attention capture by the distractor

167 (oculomotor capture) and fixation times across multiple datasets. Regardless of critical distractor 168 comparisons (high- vs. low-probability location; reward/threat-related vs. neutral; distractor-169 present vs. distractor-absent), type of visual search attentional template (feature-search vs. 170 singleton-search), and age groups (young adults vs. older adults), the reliability of oculomotor 171 capture was superior to the reliability of fixation times. Furthermore, reliability of older adults 172 was higher than that of young adults. *p < 05. ***p < .001.

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174 Grégoire et al. (2022)

175 All three experiments in Grégoire et al. (2022) incorporated a paradigm that required participants to search for a unique shape singleton (circle among diamonds or diamond among 176 177 circles), requiring participants to engage in singleton-search mode in the presence of color 178 singleton distractors. Data from Experiments 1 and 2 were of particular interest given that 179 reduced processing of valent (reward- and threat-related) distractors relative to neutral 180 distractors was observed in these experiments whereas the opposite was observed in 181 Experiment 3, although reliabilities from all three experiments are reported for completeness. 182 Data from both the training and test phases of each experiment were combined given that 183 mechanisms of attention capture by the distractor were identical in both phases and the only 184 difference in the test phase was the absence of feedback, which provided sufficient data to 185 conduct a split-half analysis. Over all experiments, the critical distractor condition comparison 186 was attention capture by the reward (Experiments 1-3) or threat-related distractor (Experiment 1 187 only) vs. the neutral distractor.

188 When comparing the difference in oculomotor measures between the threat-related vs. 189 neutral distractor in Experiment 1, correlation values over the measure of oculomotor capture 190 was r = .609 and over fixation time was r = .419. Like in Kim and Anderson (2022), we found 191 that the reliability of the learning-dependent reduction in oculomotor capture was significantly 192 superior compared to that observed using fixation time, p < .001. When comparing oculomotor

193 measures between the reward-related vs. neutral distractor, the correlations between the critical 194 distractor conditions over oculomotor capture were r = .692 and r = .658, and over fixation time 195 were r = .351 and r = .372, across Experiments 1 and 2, respectively. Using non-parametric 196 randomization tests, we again found that the reliability of the learning-dependent reduction in 197 oculomotor capture was significantly superior compared to that observed using fixation time 198 across both Experiments, ps < .001 (see Figure 1). Similar results were obtained in the context 199 of oculomotor capture in the third experiment, although overall reliability was somewhat reduced 200 (r = .492 for oculomotor capture and r = .272 for fixation time, p < .001)

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202 Kim et al. (2024)

203 In Experiment 1 of Kim et al. (2024), the task required searching for a specific target 204 shape (circle or diamond, counterbalanced across participants), requiring participants to engage 205 in feature-search mode, which generally promotes the suppression of salient distractors 206 (Gaspelin et al., 2015, 2017; Gaspelin & Luck, 2018). We compared trials in which a salient 207 color singleton distractor was present vs. absent (equally often) and separately for young adults 208 (18-23 years old) and older adults (51-79 years old). Given that we measured attention capture 209 by first fixations to the distractor on distractor-present trials, we summed the first fixations on 210 non-targets in distractor-absent trials and divided the total by the number of non-targets in the 211 visual search array to calculate the probability of fixating at any one non-target (proxy distractor 212 on distractor-absent trials). When comparing oculomotor measures between these distractor 213 conditions, correlations over oculomotor capture (probability of fixating a [proxy] distractor on 214 distractor present vs. absent trials) were r = .656 for young adults and r = .765 for older adults 215 while correlations over fixation times (latency to fixate the target on distractor present vs. absent 216 trials) was r = .547 for young adults and r = .586 for older adults. Both young and older adults 217 demonstrated superior reliability for oculomotor capture compared to fixation times, ps < .001218 (see Figure 1). In addition, older adults demonstrated superior oculomotor capture reliability

compared to young adults, p < .001 (see Figure 1). However, fixation time reliability was not significantly different between age groups, p = .229.

221 In Experiment 2, the task required searching for a unique shape singleton (circle among 222 diamonds or diamond among circles) necessitating participants to engage in singleton-search 223 mode. Under these conditions, attentional capture by the color singleton distractor is robust and 224 difficult to suppress, requiring reactive distractor disengagement to complete the task (Bacon & 225 Egeth, 1994; Geng, 2014; Theeuwes, 1992; Theeuwes et al., 1998). Again, we compared trials 226 in which the distractor was present vs. absent (equally often) and separately for young adults 227 (19-30 years old) and older adults (57-80 years old). When comparing oculomotor measures 228 between these distractor conditions, correlations over oculomotor capture were r = .815 for 229 young adults and r = .890 for older adults while correlations over fixation times were r = .540 for 230 young adults and r = .693 for older adults. As in Experiment 1, both young and older adults 231 demonstrated superior reliability for oculomotor capture compared to fixation times, ps < .001232 (see Figure 1). Furthermore, older adults demonstrated superior oculomotor capture reliability 233 compared to young adults, p = .016, in addition to superior fixation time reliability, p < .001 (see 234 Figure 1).

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Discussion

237 Our findings demonstrate that, as a measure, oculomotor capture produces superior 238 reliability compared to measures computed from fixation time across numerous critical distractor 239 comparisons. Using the probability of fixating the distractor, reliable learning-dependent 240 reductions in distractor processing can be observed (Grégoire et al., 2022; Kim & Anderson, 241 2022), in addition to a measure of attention capture that is reliable for both young and older 242 adults regardless of whether capture is overall suppressed under conditions of feature-search 243 vs. singleton-search. Even when accounting for the increased variance in difference score 244 calculations (Miller & Ulrich, 2013; Paap & Sawi, 2016; Weichselbaum et al., 2018), we

demonstrate that oculomotor measures of attention capture on average exhibit strong reliability (mean across acquired values, r = .735) and are considerably more reliable than response timebased measures (Anderson & Kim, 2019; Freichel et al., 2023; Garre-Frutos et al., 2024; Ivanov et al., 2023).

249 Experimental psychologists have largely undervalued the utility of individual differences. 250 and relationships between mechanisms of attentional control and other cognitive or self-report 251 measures have been relatively unexplored. However, researchers investigating working memory 252 capacity have examined individual differences to identify interactions between neural networks 253 of memory and attention. Prior findings reveal that individuals with low working memory capacity 254 exhibited stronger value-driven attentional capture (Anderson et al., 2011) and also took longer 255 to disengage attention from a task-irrelevant distractor (Fukuda & Vogel, 2011). This relationship 256 between working memory and attention is thought to be mediated by the locus coeruleus-257 noradrenaline system, particularly through modulation of the fronto-parietal attention networks 258 (Unsworth & Robison, 2017). However, individual differences in working memory capacity were 259 unable to predict performance in visual search tasks requiring feature or conjunction search 260 (Kane et al., 2006). The lack of a relationship here is informed by the findings of Ivanov et al. 261 (2023) in which attention capture and learning-dependent distractor suppression were 262 investigated as potentially useful measures of individual differences using manual response 263 times. Unfortunately, both within- and between-session reliability for both measures were poor 264 despite robust group level differences across conditions, suggesting that inconsistent findings 265 relating individual differences in working memory capacity to attention may be due in part to the 266 use of measures with poor reliability (all of the aforementioned studies and many similar studies 267 used attention measures derived from manual response times). Interestingly, when value-driven 268 attentional capture was measured from distractor fixations (Anderson & Yantis, 2012), the 269 reported correlation with working memory capacity was numerically quite a bit stronger than 270 when value-driven attentional capture was measured from manual response times (Anderson et

al., 2011). Our findings suggest a potential path toward more consistent outcomes relating
attention measures to other cognitive processes like working memory, and to the more fruitful
exploration of individual differences in the learning-dependent control of attention more
generally through fixation-based measures of attentional selection. More reliable measures of
attentional control are of particular importance if the goal is to predict the progression of
neurodegenerative diseases and other clinical outcomes, and our findings point to the value of
eye tracking in the pursuit of such measures.

278 The set of experiments in Kim et al. (2024) additionally revealed that older adults exhibit 279 greater reliability compared with young adults. Older adults generally have slower response 280 times compared with young adults and this becomes problematic as overall slower response 281 times have greater variability (Kim et al., 2024; Tse et al., 2010). Although Experiment 2 282 demonstrated that older adults make more first fixations to the distractor compared with young 283 adults, superior reliability cannot be reduced to a product of this greater capture effect given that 284 Experiment 1 showed similar oculomotor suppression by the distractor in both age groups but 285 still greater reliability in older adults. The strong reliability of oculomotor measures in older 286 adults can address a significant issue in the aging literature of low reliability due to increased 287 error variance in measures like response time. Furthermore, the relatively higher reliability in 288 Kim et al. (2024) suggests that the reliability of salience-driven capture may be higher compared 289 with statistically learned distractor suppression (Grégoire et al., 2022; Kim & Anderson, 2022), 290 which is in line with the results of Ivanov et al. (2023).

A natural question posed by the findings of the present study is why oculomotor capture produces a more reliable measure of distractor processing than fixation time in addition to what is typically observed in the literature with respect to manual response time. Although we can only speculate, this superior reliability may be found in the ballistic nature of the measure. Oculomotor capture essentially measures the probability that a task-irrelevant stimulus evokes greater attentional priority than the target at the time of saccade initiation, being directly linked to

297 distractor-target competition in the visual system. Manual response time-based measures add a 298 host of post-selection processes that are tied to target-response mappings and the execution of 299 a manual response (often a keypress), all of which contribute variability that is removed when 300 assessing oculomotor capture. Even in the context of fixation time, the time required to 301 disengage attention from any non-target that is fixated and the efficiency with which the 302 subsequent eye movement is targeted contribute additional variability that occurs after 303 oculomotor capture is assessed, during which there is additional opportunity for task-unrelated 304 processes (e.g., mind wandering) to randomly slow responses. If the goal is to measure 305 distractor processing, the probability of initially fixating the distractor (oculomotor capture) may 306 be the purest and most direct means of assessing it.

307 Our findings across multiple experiments suggests that the superior reliability of 308 oculomotor capture relative to even response time-based measures derived from eye tracking 309 may reflect a more general property of the measurements that would further generalize to other 310 tasks and experimental situations. However, determining whether this is the case requires 311 further investigation, in addition to the extent to which specific mechanisms of distractor 312 processing (e.g., learning effects that promote capture vs. suppression, salience-driven vs. 313 learning-dependent priority) are differently reliable. Similarly, it would also be important to 314 investigate whether the observed high reliability of oculomotor capture as a measure extends to 315 other mechanisms of distractor processing (e.g., contingent attention capture, emotion-316 modulated distraction).

The present study suggests a potential avenue forward for the field of psychological science to maximize reproducibility by utilizing oculomotor measures that exhibit high reliability. However, the biggest limitation in acquiring such measures is the accessibility of eye tracking technology. All of the datasets analyzed utilized an EyeLink 1000 plus eye tracker (SR Research) that is far less accessible than what is required to conduct research using manual response time measures, both with respect to financial cost and training. The development of

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323	more reliable measures of visual information processing involving manual response time that
324	can more closely approximate what we were able to achieve with oculomotor measures is
325	therefore an important target for future research. At least for the time being, until more reliable
326	response time-based measures are developed, we recommend that researchers consider
327	investing in oculomotor measures particularly when individual differences in distractor
328	processing are of scientific interest. Oculomotor measures are naturally bound to experiments
329	involving the processing of visual information, and it is also important to identify reliable
330	measures of information processing in other sensory modalities in an effort to maximize
331	statistical power and reproducibility.
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