

1 Benefits and risks associated with children's and adolescents' interactions with electronic
2 screens: An umbrella review

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Abstract

38 Children's engagement in screen time is a complex issue. While some forms of screen time
39 have consistently been associated with harm, others have been associated with gains, making
40 it difficult to weigh the risks and benefits of use. In this umbrella review, we systematically
41 collate and examined meta-analyses examining the effects of screen use on children and
42 youth. We converted results onto a common metric (Pearson's r) to make comparisons
43 simple, and where possible we reanalysed study-level data to standardise the approach across
44 meta-analyses. We identified 218 meta-analyses, and extracted 274 unique combinations of
45 exposures and outcomes. If duplicate effect sizes from the same combination of exposure and
46 outcome were drawn from different age groups or study designs we retained them, otherwise
47 we chose the effect size with the largest total sample size. We removed effect sizes that could
48 not be harmonised, resulting in 255 effects from 103 reviews. These effects represent the
49 findings of 2,496 primary studies comprised of 2,026,054 participants. When focusing on the
50 meta-analyses with the most statistically robust evidence, we found that general screen use
51 (when content was not indicated), was associated with potential harm on learning, literacy,
52 body composition, and depression. Likewise, social media was consistently associated with
53 risks to health, with no identified benefits. However, we also found that these harms could
54 often be mitigated by certain kinds of content (e.g., educational), or by modifying the context
55 (e.g., co-viewing with a parent). In summary, our findings point to the need for careful and
56 nuanced guidelines that support parents to make the best decisions for their children.

57

Keywords: screen time; youth; health; education

58

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60 screens: An umbrella review

61 **Summary**

62 Children's engagement in screen time is a complex issue. Parents, policymakers, and
63 educators need to weigh the risks that sedentary use of screens present alongside the
64 potential benefits for learning and social connectedness. The lack of comprehensive evidence
65 hampers efforts to make an informed decision. As a Lancet editorial¹ suggested, "Our
66 understanding of the benefits, harms, and risks of our rapidly changing digital landscape is
67 sorely lacking." In this study, we systematically harmonise data from existing meta-analyses
68 of screen time on a range of outcomes, including health, education, and psychology, and
69 identify the most statistically robust relationships. We show that some forms of screen
70 time—such as social media—show consistent evidence of harm for children, with no clear
71 evidence of a benefit. Other relationships are more complex. Video games, for example, are
72 associated with poorer body composition and learning outcomes. However, video games for a
73 specific educational purpose (such as numeracy) are associated with improvements in that
74 subject area. Caregivers must therefore weigh the health risk against the educational benefit.
75 The findings of this study provide parents and other caregivers with the information to make
76 these informed decisions.

77 **Background**

78 In the 16th century, hysteria reigned around a new technology that threatened to be
79 “confusing and harmful” to the mind. The cause of such concern? The widespread
80 availability of books brought about by the invention of the printing press.² In the early 19th
81 century, concerns about schooling “exhausting the children’s brains” followed, with the
82 medical community accepting that excessive study could be a cause of madness.³ By the
83 20th century, the invention of the radio was accompanied by assertions that it would distract
84 children from their reading (which by this point was no longer considered confusing and
85 harmful) leading to impaired learning.⁴

86 Today, the same arguments that were once leveled against reading, schooling, and
87 radio are being made about screen use (e.g., television, mobile phones, and computers).⁵
88 Excessive screen time use is the number one concern parents in Western countries have
89 about their children’s health and behaviour, ahead of nutrition, bullying, and physical
90 inactivity.⁶ Yet, the evidence to support parents’ concerns is inadequate. A Lancet editorial¹
91 suggested that, “Our understanding of the benefits, harms, and risks of our rapidly changing
92 digital landscape is sorely lacking.”

93 While some forms of screen use (e.g., television viewing) may be detrimental to health
94 and wellbeing,^{7,8} evidence for other forms of screen exposure (e.g., video games or online
95 communication, such as Zoom™) remains less certain and, in some cases, may even be
96 beneficial.^{9,10} Thus, according to a Nature Human Behaviour editorial, research to determine
97 the effect of screen exposure on youth is “a defining question of our age”.¹¹ With concerns
98 over the impact of screen use including education, health, social development, and
99 psychological well-being, an overview that identifies potential benefits and risks is needed.

100 Citing the negative effects of screens on health (e.g., increased risk of obesity) and
101 health-related behaviours (e.g., sleep), guidelines from the World Health Organisation¹² and
102 numerous government agencies^{13,14} and statements by expert groups¹⁵ have recommended

that young people's time spent using electronic media devices for entertainment purposes should be limited. For example, the Australian Government guidelines regarding sedentary behaviour recommend that young children (under the age of two) should not spend any time watching screens. They also recommend that children aged 2-5 years should spend no more than one hour engaged in recreational sedentary screen use per day, while children aged 5-12 and adolescents should spend no more than two hours. However, recent evidence suggests that longer exposures may not have adverse effects on children's behaviour or mental health—and might, in fact, benefit their well-being—as long as exposure does not reach extreme levels (e.g., 7 hours per day)¹⁶. Some research also indicates that content (e.g., video games vs television programs) plays an important role in determining the potential benefit or harm of youths' exposure to screen-based media.¹⁷ Indeed, educational screen time is positively related to educational outcomes.¹⁸ This evidence has led some researchers to argue that a more nuanced approach to screen time guidelines is required.¹⁹

In 2016, the American Academy of Pediatrics used a narrative review to examine the benefits and risks of children and adolescents' electronic media²⁰ as a basis for updating their guidelines about screen use.¹⁵ Since then, a large number of systematic reviews and meta-analyses have provided evidence about the potential benefits and risks of screen use.

While there have been other overviews of reviews on screen time, these have tended to focus on a single domain (e.g., health²¹), focus on a particular exposure (e.g., social media^{22,23}) or provide only a narrative summary of the literature.²⁴ Focusing on a single domain or exposure makes it difficult to understand what trade-offs are involved in any guidelines around screen use. For example, prohibiting screen use might reduce exposure to advertising but may also thwart learning opportunities from interactive educational tools. Reviews on either of these exposures or outcomes would likely miss being able to quantify these trade-offs. Overviews are one method of evidence synthesis that helps address these trade-offs, by providing 'user-friendly' summaries of a field of research.²⁵ These overviews provide a reference point for the field and allow for easier comparison of risks and benefits

130 for the same behaviour.

131 In order to synthesise the evidence and support further evidence-based guideline
132 development and refinement, we reviewed published meta-analyses examining the effects of
133 screen use on children and youth. This review synthesises evidence on any outcome of
134 electronic media exposure. We deliberately do not pre-specify outcomes, in order to get a
135 comprehensive list of areas where there is meta-analytical evidence. Adopting this broad
136 approach allowed us to provide a holistic perspective on the influence of screens on children's
137 lives. By synthesising across life domains (e.g., school and home), this review provides
138 evidence to inform guidelines and advice for parents, teachers, pediatricians and other
139 professionals in order to maximise human functioning.

140

Methods

141 We prospectively registered our methods on the International Prospective Register of
142 Systematic Reviews (PROSPERO; CRD42017076051). We followed the Preferred Reporting
143 Items for Systematic Reviews and Meta-Analyses (PRISMA) guidelines.²⁶

144 **Eligibility criteria.** *Population:* To be eligible for inclusion, meta-analyses needed
145 to include meta-analytic effect sizes for children or adolescents (age 0-18 years). We included
146 meta-analyses containing studies that combined data from adults and youth if meta-analytic
147 effect size estimates specific to participants aged 18 years or less could be extracted (i.e., the
148 highest mean age for any individual study included in the meta-analysis was < 18 years). A
149 meta-analysis was still included if the age range exceed 18 years, provided that the mean age
150 was less than 18. We excluded meta-analyses that only contained evidence gathered from
151 adults (age >18 years).

152 *Exposure:* We included meta-analyses examining all types of electronic screens
153 including (but not necessarily limited to) television, gaming consoles, computers, tablets,
154 and mobile phones. We also included analyses of all types of content on these devices,
155 including (but not necessarily limited to) recreational content (e.g., television programs,
156 movies, games), homework, and communication (e.g., video chat). In this review we focused
157 on electronic media exposure that would be considered typical for children and youth. That
158 is, exposure that may occur in the home setting, or during schooling. Consistent with this
159 approach, we excluded technology-based treatments for clinical conditions. However, we
160 included studies examining the effect of screen exposure on non-clinical outcomes (e.g.,
161 learning) for children and youth with a clinical condition. For example, a meta-analysis of
162 the effect of television watching on learning among adolescents diagnosed with depression
163 would be included. However, a meta-analysis of interventions designed to *treat* clinical
164 depression delivered by a mobile phone app would be excluded. *Outcomes:* We included all
165 reported outcomes on benefits and risks.

166 *Publications:* We included meta-analyses (or meta-regressions) of quantitative evidence.

167 To be included, meta-analyses needed to analyse data from studies identified in a systematic
168 review. For our purposes, a systematic review was one in which the authors attempted to
169 acquire all the research evidence that pertained to their research question(s). We excluded
170 meta-analyses that did not attempt to summarise all the available evidence (e.g., a
171 meta-analysis of all studies from one laboratory). We included meta-analyses regardless of
172 the study designs included in the review (e.g., laboratory-based experimental studies,
173 randomised controlled trials, non-randomised controlled trials, longitudinal, cross-sectional,
174 case studies), as long as the studies in the review collected quantitative evidence. We
175 excluded systematic reviews of qualitative evidence. We did not formulate
176 inclusion/exclusion criteria related to the risk of bias of the review. We did, however, employ
177 a risk of bias tool to help interpret the results. We included full-text, peer-reviewed
178 meta-analyses published or ‘in-press’ in English. We excluded conference abstracts and
179 meta-analyses that were unpublished.

180 **Information sources.** We searched records contained in the following databases:

181 Pubmed, MEDLINE, CINAHL, PsycINFO, SPORTDiscus, Education Source, Embase,
182 Cochrane Library, Scopus, Web of Science, ProQuest Social Science Premium Collection, and
183 ERIC. We conducted an initial search on August 17, 2018 and refreshed the search on
184 September 27, 2022. We searched reference lists of included papers in order to identify
185 additional eligible meta-analyses. We also searched PROSPERO to identify relevant
186 protocols and contacted authors to determine if these reviews have been completed and
187 published.

188 **Search strategy.** The search strategy associated with each of the 12 databases can
189 be found in Supplementary File 1. We hand searched reference lists from any relevant
190 umbrella reviews to identify systematic meta-analyses that our search may have missed.

191 **Selection process.** Using Covidence software (Veritas Health Innovation,
192 Melbourne, Australia), two researchers independently screened all titles and abstracts. Two

193 researchers then independently reviewed full-text articles. We resolved disagreements at each
194 stage of the process by consensus, with a third researcher employed, when needed.

195 **Data items.** From each included meta-analysis, two researchers independently
196 extracted data into a custom-designed database. We extracted the following items: First
197 author, year of publication, study design restrictions (e.g., cross-sectional, observational,
198 experimental), region restrictions (e.g., specific countries), earliest and latest study
199 publication dates, sample age (mean), lowest and highest mean age reported, outcomes
200 reported, and exposures reported.

201 **Study risk of bias assessment.** For each meta-analysis, two researchers
202 independently completed the National Health, Lung and Blood Institute's Quality
203 Assessment of Systematic Reviews and Meta-Analyses tool²⁷ (see Table 1). We resolved
204 disagreements by consensus, with a third researcher employed when needed. We did not
205 assess risk of bias in the individual studies that were included in each meta-analysis.

206 **Effect measures.** Two researchers independently extracted all quantitative
207 meta-analytic effect sizes, including moderation results. We excluded effect sizes which were
208 reported as relative risk ratios or odds ratios, as meta-analyses did not contain sufficient
209 information to meaningfully convert to a correlation. We also excluded effect size estimates
210 when the authors did not provide a sample size. Where possible, we also extracted effect
211 sizes from the primary studies included in each meta-analysis.

212 To facilitate comparisons, we converted effect sizes to Pearson's r using established
213 formulae.^{28,29} Effect sizes on the original metric are provided in Supplementary File 2.
214 Throughout the results section we interpret the size of the effects using Funder and Ozer's
215 guidelines:³⁰ very small ($0.05 < r \leq 0.1$), small ($0.1 < r \leq 0.2$), medium ($0.2 < r \leq 0.2$),
216 large ($0.3 < r \leq 0.4$), and very large ($r \geq 0.4$). These are similar to other interpretations
217 based on empirical data.³¹

218 **Synthesis methods.** After extracting data, we examined the combinations of

exposure and outcomes and removed any effects that appeared multiple times (i.e., in multiple meta-analyses, or with multiple sub-groups in the same meta-analysis), keeping the effect with the largest total sample size. In instances where effect sizes from the same combination of exposure and outcome were drawn from different age-groups (e.g., children vs adolescents), or were drawn using different study designs (e.g., cross-sectional vs longitudinal) we retained both estimates in our dataset.

We descriptively present the remaining meta-analytic effect sizes. To remove the differences in approach to meta-analyses across the reviews, we reran the effect size estimate using a random effects meta-analysis via the metafor package³² in R³³ (version 4.2.2) when the meta-analysis's authors provided primary study data associated with these effects. When required, we imputed missing sample sizes using mean imputation from the other studies within that review. From our reanalysis we also extracted I^2 values. To test for publication bias, we conducted Egger's test³⁴ when the number of studies within the review was ten or more,³⁵ and conducted a test of excess significance.³⁶ We contacted authors who did not provide primary study data in their published article. Where authors did not provide data in a format that could be re-analysed, we used the published results of their original meta-analysis.

Evidence assessment criteria. *Statistical Credibility.* We employed a statistical classification approach to grade the credibility of the effect sizes in the literature. To be considered 'credible' an effect needed to be derived from a combined sample of $>1,000$ participants³⁷ and have non-significant tests of publication bias (i.e., Egger's test and excess significance test). We performed these analyses, and therefore the review needed to provide usable study-level data in order to be included.

Consistency of Effect within the Population. We also examined the consistency of the effect size using the I^2 measure. We considered $I^2 < 50\%$ to indicate effects that were relatively consistent across the population of interest. I^2 values of $> 50\%$ were taken to indicate an effect was potentially heterogeneous within the population.

246 *Direction of Effect.* Finally, we examined the extent to which significance testing
247 suggested screen exposure was associated with benefit, harm, or no effect on outcomes. We
248 used thresholds of $P < .05$ for weak evidence and $P < 10^{-3}$ for strong evidence. An effect
249 with statistical credibility but with $P > .05$ was taken to indicate no association of interest.

250 **Deviations from protocol.** We initially planned to describe the findings of
251 systematic reviews that did not include meta-analyses in a narrative summary alongside the
252 main findings from meta-analytic studies. However, we determined that combining results
253 from the meta-analyses allowed readers to compare relative strength of associations more
254 easily. Readers interested in the relevant systematic reviews (i.e., without meta-analysis) can
255 consult the list of references in Supplementary File 4.

256 We altered our evidence assessment plan when we identified that, as written, it could
257 not classify precise evidence of null effects (i.e., from large reviews with low heterogeneity
258 and low risk of publication bias) as ‘credible’ because a highly-significant P -value was a
259 criteria. This would have significantly harmed knowledge gained from our review as it would
260 have restricted our ability to show where the empirical evidence strongly indicated that there
261 was no association between screen time and a given outcome.

262 Results

263 **Search Results.** The searches yielded 50,649 results, of which 28,675 were
264 duplicates. After screening titles and abstracts, we assessed 2,557 full-texts for inclusion. Of
265 those, 218 met the inclusion criteria and we extracted the data from all of these
266 meta-analyses. Figure 1 presents the full results of the selection process.

267 The most frequently reported exposures were physically active video games ($n = 31$),
268 general screen use ($n = 27$), general TV programs and movies ($n = 20$), and screen-based
269 interventions to promote health ($n = 14$). Supplementary File 5 provides a list of all
270 exposures identified. The most frequently reported outcomes were body composition ($n =$

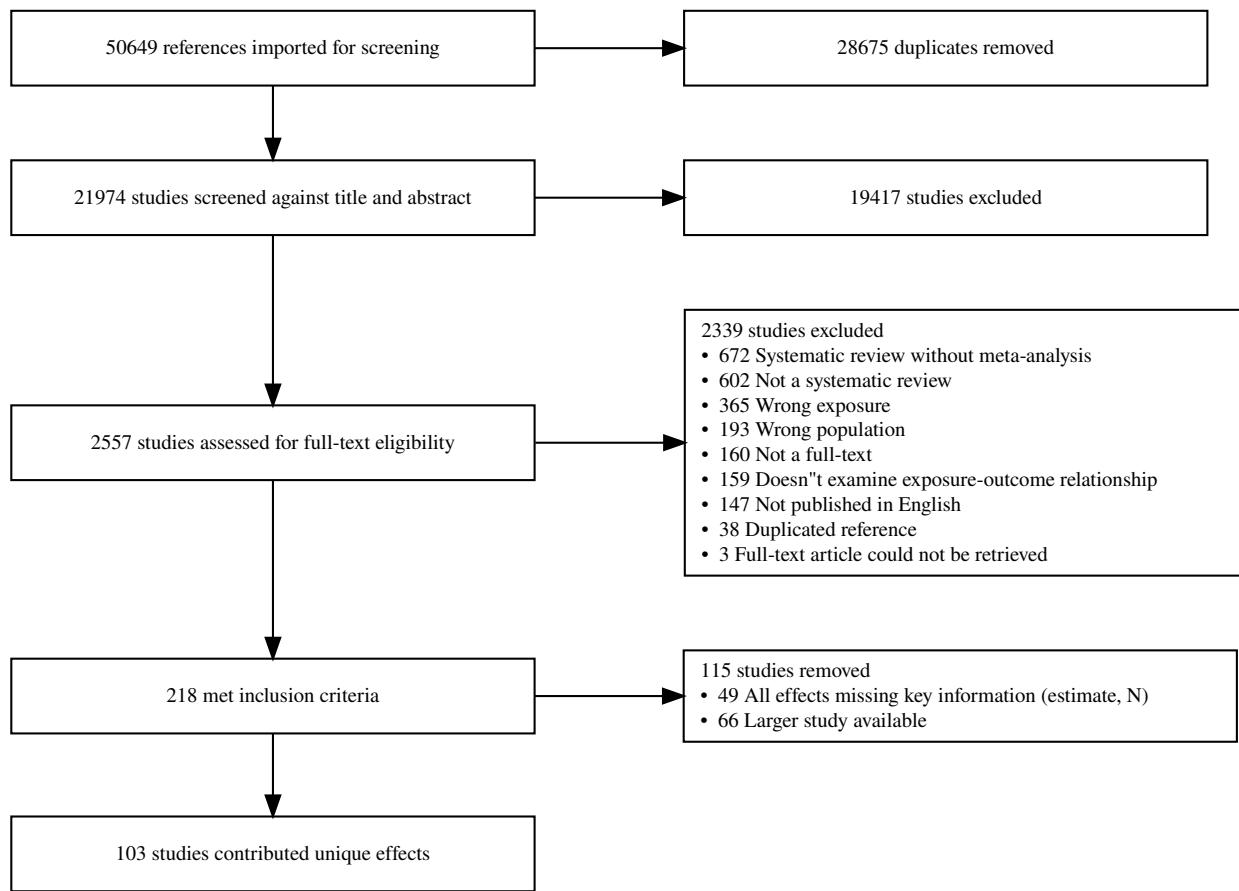


Figure 1. PRISMA Diagram

271 30), general learning ($n = 25$), depression ($n = 13$), and general literacy ($n = 12$). Of the 274
 272 unique exposure/outcome combinations, 242 occurred in only one review, with 23 appearing
 273 twice, and 9 appearing three or more times. Full characteristics of the included studies are
 274 provided in Table 1. After removing reviews with duplicate exposure/outcome combinations,
 275 our process yielded 255 unique effect/outcome combinations (retaining multiple effects for
 276 different age groups or study designs) contributed from 103 reviews. These effects represent
 277 the findings of 2,496 primary studies comprised of 2,026,054 participants. The characteristics
 278 of the included effects are available in Supplementary File 9.

TABLE 1

280 The quality of the included meta-analyses was mixed (see Table 1). Most assessed

281 heterogeneity (n low risk = 94/103, 91% of meta-analyses), reported the characteristics of
282 the included studies (n low risk = 87/103, 84%), and used a comprehensive and systematic
283 search strategy (n low risk = 72/103, 70%). Most reviews did not clearly report if their
284 eligibility criteria were predefined (n unclear = 72/103, 70%). Many papers also did not
285 complete dual independent screening of abstracts and full text (n high risk = 20/103, 19%)
286 or did not clearly report the method of screening (n unclear = 38/103, 37%). A similar trend
287 was observed for dual independent quality assessment (n high risk = 53/103, 51%; n high risk
288 = 19/103, 18%). Overall, only 7 meta-analyses were graded as low risk of bias on all criteria.

289 **Education Outcomes.** There were 89 unique effects associated with education
290 outcomes, including general learning outcomes, literacy, numeracy, and science. We removed
291 28 effects that did not provide individual study-level data, 19 effects with samples < 1,000,
292 and 19 effects with a significant Egger's test or insufficient studies to conduct the test.
293 Effects not meeting one or more of these standards are presented in Supplementary File 6.
294 The remaining 23 effects met our criteria for statistical credibility and are described in
295 Figure 2. These 23 effects came from 18 meta-analytic reviews analysing data from 338
296 empirical studies with 262,537 individual participants.

297 Among the statistically credible effects, general screen use, television viewing, and
298 video games were all negatively associated with learning. E-books that included narration,
299 as well as touch screen education interventions, and augmented reality education
300 interventions were positively associated with learning. General screen use was negatively
301 associated with literacy outcomes. However, if the screen use involved co-viewing (e.g.,
302 watching with a parent), or the content of television programs was educational, the
303 association with literacy was positive and significant at the 95% confidence level (weak
304 evidence). Numeracy outcomes were positively associated with screen-based mathematics
305 interventions and video games that contained numeracy content.

306 As shown in Figure 2, most of the credible results (14 of 23 effects) showed statistically
307 significant associations, with 99.9% confidence intervals not encompassing zero (strong

Associations Between Exposures and Education Outcomes

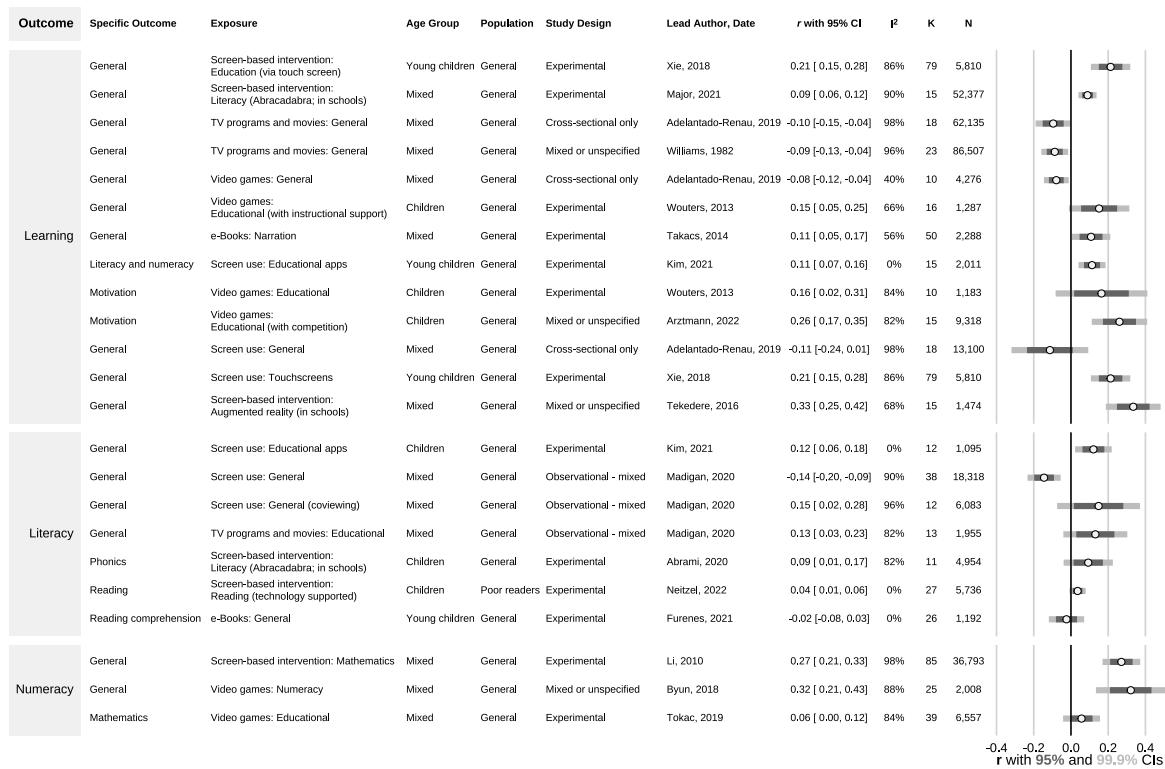


Figure 2. Education outcomes

308 evidence). The remaining six associations were significant at the 95% confidence level (weak
 309 evidence). All credible effects related to education outcomes were small-to-moderate.
 310 Screen-based interventions designed to influence an outcome (e.g., a computer based
 311 program designed to enhance learning³⁸) tended to have larger effect sizes than exposures
 312 that were not specifically intended to influence any of the measured outcomes (e.g., the
 313 association between television viewing and learning³⁹). The largest effect size observed was
 314 for augmented reality-based education interventions on general learning
 315 ($r = 0.33, k = 15, N = 1,474$). Most effects showed high levels of heterogeneity (18 of 23
 316 with $I^2 > 50\%$).

317 **Health and Health-related Behaviours.** We identified 165 unique
 318 outcome-exposure combinations associated with health or health-related behaviour outcomes.
 319 We removed 41 effects that did not provide individual study-level data, 50 effects with

samples < 1,000, and 53 effects with a significant Egger's test or insufficient studies to conduct the test. No remaining studies showed evidence of excessive significance. Effects not meeting one or more of these standards are presented in Supplementary File 7. The remaining 21 meta-analytic associations met our criteria for credible evidence and are described below (see also Figure 3). These 21 effects came from 15 meta-analytic reviews analysing data from 344 empirical studies with 859,562 individual participants.

Associations Between Exposures and Health-related Outcomes

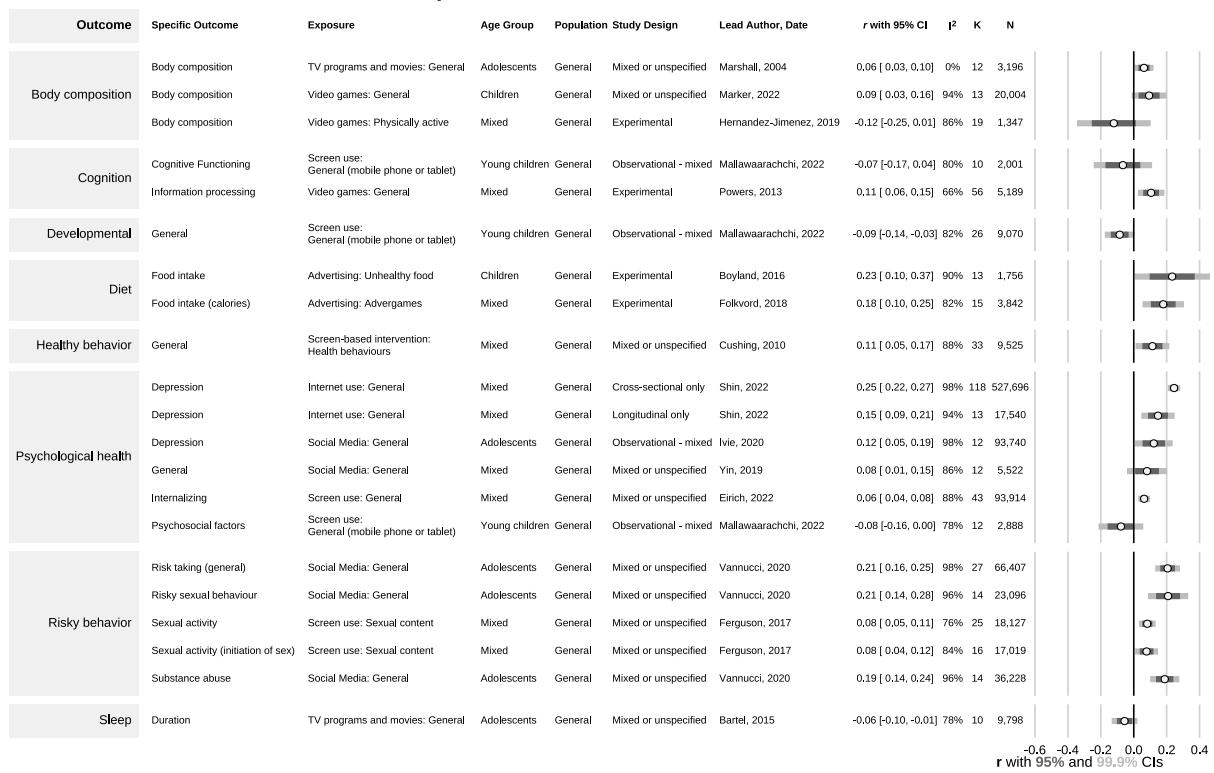


Figure 3. Health and health-related behaviour outcomes

Digital advertising of unhealthy foods—both traditional advertising and video games developed by a brand for promotion—were associated with higher unhealthy food intake. Social media use and sexual content were positively associated with risky behaviors (e.g., sexual activity, risk taking, and substance abuse). General screen use was positively associated with depression, with stronger associations observed for adolescents than other groups. Television viewing was negatively correlated with sleep duration, but with stronger

332 evidence only observed for younger children. All forms of screen use (general, television, and
333 video games) were associated with body composition (e.g., higher BMI). Screen-based
334 interventions which target health behaviours appeared mostly effective.

335 Across the health outcomes, most (14 of 21) effects were statistically significant at the
336 99.9% confidence interval level, with the remaining four significant at 95% confidence.

337 However, most of the credible effects exhibited high levels of heterogeneity, with all but two
338 having $I^2 > 75\%$. Additionally, most effects were small, with the association between screen
339 use and sleep duration the largest at $r = -0.37$ ($k = 10, N = 56,720$). Most of the effect
340 sizes (17/21) had an absolute value of $r < 0.2$.

341 **Discussion**

342 The primary goal of this review was to provide a holistic perspective on the influence
343 of screens on children's lives across a broad range of outcomes. We found that when
344 meta-analyses examined general screen use, and did not specify the content, context or
345 device, there was strong evidence showing potentially harmful associations with general
346 learning, literacy, body composition, and depression. However, when meta-analyses included
347 a more nuanced examination of exposures, a more complex picture appeared.

348 As an example, consider children watching television programs—an often cited form of
349 screen time harm. We found statistically robust evidence for a small association with poorer
350 academic performance and literacy skills for general television watching³⁹. However, we also
351 found evidence that if the content of the program was educational, or the child was watching
352 the program with a parent (i.e., co-viewing), this exposure was instead associated with
353 better literacy.⁴⁰ Thus, parents may play an important role in selecting content that is likely
354 to benefit their children or, perhaps, interact with their children in ways that may foster
355 literacy (e.g., asking their children questions about the program). Similar nuanced findings
356 were observed for video games. The credible evidence we identified showed that video game

357 playing was associated with poorer body composition and learning.^{39,41} However, when the
358 video game were designed specifically to teach numeracy, playing these games showed
359 learning benefits.⁴² One might expect that video games designed to be physically active
360 could confer health benefits, but none of the meta-analyses examining this hypothesis met
361 our thresholds for statistical credibility (see Supplementary Files 6 & 7) therefore this
362 hypothesis could not be addressed.

363 Social media was one type of exposure that showed consistent associations with poor
364 health, with no indication of potential benefit. Social media showed strong evidence of
365 harmful associations with risk taking in general, as well as unsafe sex and substance abuse.⁴³
366 These results align with meta-analytic evidence from adults indicating that social media use
367 is also associated with increased risk of depression.^{44,45} Recent evidence from social media
368 companies themselves suggest there may also be negative effects of social media on the
369 mental health of young people, especially teenage girls.⁴⁶

370 One category of exposure appeared to be consistently associated with benefits:
371 screen-based interventions designed to promote learning or health behaviours. This finding
372 indicates that interventions can be effectively delivered using electronic media platforms, but
373 does not necessarily indicate that screens are more effective than other methods (e.g.,
374 face-to-face, printed material). Rather, it reinforces that the content of the screen time may
375 be the most important aspect. The way that a young person interacts with digital screens
376 may also be important. We found evidence that touch screens had strong evidence for
377 benefits on learning,³⁸ as did augmented reality.⁴⁷

378 Largely owing to a small number of studies or missing individual study data, there
379 were few age-based conclusions that could be drawn from reviews which met our criteria for
380 statistical certainty. If we expand to include those reviews which did not meet this threshold,
381 there remained no clear pattern although there were some age-specific differences in
382 associations (data available in Supplementary Materials). For example, advertising of

383 unhealthy food was associated with unhealthy food choice for young children, but was not
384 statistically significant for other age groups.⁴⁸ Conversely, TV programs and movies were
385 more strongly associated with lower physical activity for adolescents than for younger age
386 groups.⁴⁹ Given the differences in development across childhood and adolescence and the
387 different ways children of various ages use screens, further examination of age-based
388 differences is needed. However, in the absence of this work, our study has shown how
389 children are affected by screens in general.

390 Among studies that met our criteria for statistical certainty heterogeneity was high,
391 with almost all effects having $I^2 > 50\%$. Much of this heterogeneity is likely explained by
392 differences in measures across pooled studies, or in some cases, the generic nature of some of
393 the exposures. For example, “TV programs and movies” covers a substantial range of
394 content, which may explain the heterogeneous association with education outcomes.

395 Implications for Policy and Practice

396 Broadly, our findings align with the recommendations of others who suggest that
397 current guidelines may be too simplistic, mischaracterise the strength of the evidence, or do
398 not acknowledge the important nuances of the issue.^{50–52} Our findings suggest that screen
399 use is a complex issue, with associations based not just on duration and device type, but also
400 on the content and the environment in which the exposure occurs. Many current guidelines
401 simplify this complex relationship as something that should be minimised in all
402 instances.^{12,13} We suggest that future guidelines need to embrace the complexity of the issue,
403 to give parents and clinicians specific information to weigh the pros and cons of interactions
404 with screens.

405 Implications for Future Research

406 Screen use research is extensive, varied, and rapidly growing. Reviews tended to be
407 general (e.g., all screen time) and even when more targeted (e.g., social media) nuances
408 related to specific content (e.g., Instagram vs Facebook) have not been meta-analysed or
409 have not produced credible evidence. Fewer than 20% of the effects identified met our
410 criteria for statistical credibility. Most studies which did not meet our criteria failed to
411 provide study-level data (or did not provide sufficient data, such as including effect estimates
412 but not sample sizes). Newer reviews were more likely to provide this information than older
413 reviews, but it highlights the importance of data and code sharing as recommended in the
414 PRISMA guidelines.²⁶ When study level data was available, many effects were removed
415 because the pooled sample size was small, or because there were fewer than ten studies on
416 which to perform an Egger's test. It seems that much of the current screen time research is
417 small in scale, and there is a need for larger, high-quality studies.

418 Our results highlight the need for the field to more carefully consider if the term 'screen
419 time' remains appropriate for providing advice to parents. Instead, our results suggest that
420 more nuanced and detailed descriptions of the behaviours to be modified may be required.
421 Rather than suggesting parents limit 'screen time', for example, it may be better to suggest
422 that parents promote interactive educational experiences but limit exposure to advertising.

423 Screen time research has a well-established measurement problem, which impacts the
424 individual studies of this umbrella review. The vast majority of screen time research relies on
425 self-reported data, which not only lacks the nuance required for understanding the effects of
426 screen time, but may also be inaccurate. In one systematic review on screen time and sleep,⁷
427 66 of the 67 included studies used self-reported data for *both* the exposure and outcome
428 variable. It has been established that self-reported screen time data has questionable validity.
429 In a meta-analysis of 47 studies comparing self-reported media use with logged measures,
430 Parry et al⁵³ found that the measures were only moderately correlated ($r = 0.38$), with

431 self-reported problematic usage fairing worse ($r = 0.25$). Indeed, of 622 studies which
432 measured the screen time of 0—6 year-olds, only 69 provided any sort of psychometric
433 properties for their measure, with only 19 studies reporting validity.⁵⁴ While some
434 researchers have started using newer methods of capturing screen behaviours—such as
435 wearable cameras⁵⁵ or device-based loggers⁵⁶—these are still not widely adopted. It may be
436 that the field of screen time research cannot be sufficiently advanced until accurate,
437 validated, and nuanced measures are more widely available and adopted.

438 **Strengths and Limitations**

439 Our primary goal for this umbrella review was to provide a high-level synthesis of
440 screen time research, by examining a range of exposures and the associations with a broad
441 scope of outcomes. Our results represent the findings from 2,496 primary studies comprised
442 of 2,026,054 participants. To ensure findings could be compared on a common metric, we
443 extracted and reanalysed individual study data where possible.

444 Our high-level approach limits the feasibility of examining fine-grained details of the
445 individual studies. For example, we did not examine moderators beyond age, nor did we rate
446 the risk of bias for the individual studies. Thus, our assessment of evidence quality was
447 restricted to statistical credibility, rather than a more complete assessment of quality (e.g.,
448 GRADE⁵⁷). As such, we made decisions regarding the credibility of evidence, where others
449 may have used different thresholds or metrics. For this reason, we provide the complete
450 results in the supplementary material, along with the dataset for others to consider
451 alternative criteria.

452 Our high-level approach also means that we could not engage with the specific
453 mechanisms behind each association, and as such, we cannot make strong claims on the
454 directions of causality. These likely depend on the specific exposure and outcome. It is
455 tempting for people to draw inferences that the associations are due to screen time causing

456 these outcomes, but we cannot rule out reverse causality, a third variable, or some
457 combination of influences. Many of the individual reviews go into more detail about the
458 strength of the evidence for causal associations, but those judgements were difficult to
459 synthesise across more than 200 reviews. Readers who wish to more deeply understand one
460 specific relationship are directed to the cited review for that effect, where the authors could
461 engage more deeply with the mechanisms. We converted all effect sizes to a common metric
462 (Pearson's r) to allow for comparisons of magnitude, but acknowledge that this assumes a
463 linear relationship. Some previous research suggests that associations are typically linear.¹⁸
464 However, others have identified instances where non-linear relationships exist, especially for
465 very high levels of screen time.^{17,58,59} Additionally, our conversion may not always adequately
466 account for differences in study design or measures of exposures and outcomes. Care is
467 needed, therefore, when interpreting the effect sizes. In addition, reviews provide only
468 historical evidence which may not keep up with the changing ways children can engage with
469 screens. While our synthesis of the existing evidence provides information about how screens
470 might have influenced children in the past, it is difficult to know if these findings will
471 translate to new forms of technology in the future.

472 Conclusions

473 Screen time is a topic of significant interest, as shown by the wide variety of academic
474 domains involved, parents' concerns, and the growing pervasiveness into society. Our
475 findings showed that the influence of screen time can be both positive (e.g., educational
476 video games were associated with improved literacy) and negative (e.g., general screen use
477 was associated with poorer body composition). The interplay of these findings show that
478 parents, teachers, and other caregivers need to carefully weigh the pros and cons of each
479 specific activity for potential harms and benefits. However, our findings also suggest that in
480 order to aid caregivers to make this judgement, researchers need to conduct more careful and
481 nuanced measurement and analysis of screen time, with less emphasis on measures that

- ⁴⁸² aggregate screen time and instead focus on the content, context, and environment in which
⁴⁸³ the exposure occurs.

484

References

- 485 1. The Lancet. Social media, screen time, and young people's mental health. *The
486 Lancet* **393**, 611 (2019).
- 487 2. Blair, A. Reading Strategies for Coping With Information Overload ca.1550-1700.
488 *Journal of the History of Ideas* **64**, 11–28 (2003).
- 489 3. Bell, A. N. *The sanitarian.* vol. 11 (AN Bell, 1883).
- 490 4. Dill, K. E. *The Oxford handbook of media psychology.* (Oxford University Press,
491 2013).
- 492 5. Wartella, E. A. & Jennings, N. Children and computers: New technology. Old
493 concerns. *The future of children* 31–43 (2000).
- 494 6. Rhodes, A. *Top ten child health problems: What the public thinks.* (2015).
- 495 7. Hale, L. & Guan, S. Screen time and sleep among school-aged children and
496 adolescents: A systematic literature review. *Sleep Medicine Reviews* **21**, 50–58 (2015).
- 497 8. Sweetser, P., Johnson, D., Ozdowska, A. & Wyeth, P. Active versus passive screen
498 time for young children. *Australasian Journal of Early Childhood* **37**, 94–98 (2012).
- 499 9. Li, X. & Atkins, M. S. Early childhood computer experience and cognitive and
500 motor development. *Pediatrics* **113**, 1715–1722 (2004).
- 501 10. Warburton, W. & Highfield, K. Children and technology in a smart device world.
502 in *Children, Families and Communities* 195–221 (Oxford University Press, 2017).
- 503 11. Nature Human Behaviour. Screen time: How much is too much? *Nature* **565**,
504 265–266 (2019).
- 505 12. World Health Organization. *Guidelines on physical activity, sedentary behaviour
506 and sleep for children under 5 years of age.* 33 p. (World Health Organization, 2019).

507 13. Australian Government. *Physical activity and exercise guidelines for all*

508 *Australians*. (2021).

509 14. Canadian Society for Exercise Physiology. *Canadian 24-Hour Movement*

510 *Guidelines for Children and Youth: An Integration of Physical Activity, Sedentary Behaviour,*
511 *and Sleep*. (2016).

512 15. Council On Communication and Media. Media Use in School-Aged Children and

513 Adolescents. *Pediatrics* **138**, e20162592 (2016).

514 16. Ferguson, C. J. Everything in Moderation: Moderate Use of Screens Unassociated

515 with Child Behavior Problems. *Psychiatric Quarterly* **88**, 797–805 (2017).

516 17. Przybylski, A. K. & Weinstein, N. A Large-Scale Test of the Goldilocks Hypothesis:

517 Quantifying the Relations Between Digital-Screen Use and the Mental Well-Being of
518 Adolescents. *Psychological Science* **28**, 204–215 (2017).

519 18. Sanders, T., Parker, P. D., del Pozo-Cruz, B., Noetel, M. & Lonsdale, C. Type of

520 screen time moderates effects on outcomes in 4013 children: Evidence from the Longitudinal

521 Study of Australian Children. *International Journal of Behavioral Nutrition and Physical*

522 *Activity* **16**, 117 (2019).

523 19. Kaye, L. K., Orben, A., Ellis, D. A., Hunter, S. C. & Houghton, S. The Conceptual

524 and Methodological Mayhem of ‘Screen Time’. *International Journal of Environmental*

525 *Research and Public Health* **17**, 3661 (2020).

526 20. Chassiakos, Y. L. R. *et al.* Children and Adolescents and Digital Media. *Pediatrics*

527 **138**, e20162593 (2016).

528 21. Stiglic, N. & Viner, R. M. Effects of screentime on the health and well-being of

529 children and adolescents: A systematic review of reviews. *BMJ Open* **9**, e023191 (2019).

530 22. Valkenburg, P. M., Meier, A. & Beyens, I. Social media use and its impact on

531 adolescent mental health: An umbrella review of the evidence. *Current Opinion in*

- 532 *Psychology* **44**, 58–68 (2022).
- 533 23. Arias-de la Torre, J. *et al.* Relationship Between Depression and the Use of Mobile
534 Technologies and Social Media Among Adolescents: Umbrella Review. *Journal of Medical*
535 *Internet Research* **22**, e16388 (2020).
- 536 24. Orben, A. Teenagers, screens and social media: A narrative review of reviews and
537 key studies. *Social Psychiatry and Psychiatric Epidemiology* **55**, 407–414 (2020).
- 538 25. Pollock, M., Fernandes, R., Becker, L., Pieper, D. & Hartling, L. Chapter V:
539 Overviews of Reviews. in *Cochrane Handbook for Systematic Reviews of Interventions* (eds.
540 Higgins, J. P. *et al.*) (Cochrane, 2022).
- 541 26. Page, M. J. *et al.* *The PRISMA 2020 statement: An updated guideline for*
542 *reporting systematic reviews*. (2020) doi:10.31222/osf.io/v7gm2.
- 543 27. National Health, Lung, and Blood Institute. *Quality Assessment of Systematic*
544 *Reviews and Meta-Analyses*. (2014).
- 545 28. Bowman, N. A. Effect Sizes and Statistical Methods for Meta-Analysis in Higher
546 Education. *Research in Higher Education* **53**, 375–382 (2012).
- 547 29. Jacobs, P. & Viechtbauer, W. Estimation of the biserial correlation and its
548 sampling variance for use in meta-analysis: Biserial Correlation. *Research Synthesis Methods*
549 **8**, 161–180 (2017).
- 550 30. Funder, D. C. & Ozer, D. J. Evaluating Effect Size in Psychological Research:
551 Sense and Nonsense. *Advances in Methods and Practices in Psychological Science* **2**, 156–168
552 (2019).
- 553 31. Gignac, G. E. & Szodorai, E. T. Effect size guidelines for individual differences
554 researchers. *Personality and Individual Differences* **102**, 74–78 (2016).
- 555 32. Viechtbauer, W. *Metafor: Meta-analysis package for r*. (2022).

556 33. R Core Team. *R: A language and environment for statistical computing.* (R

557 Foundation for Statistical Computing, 2022).

558 34. Egger, M., Smith, G. D., Schneider, M. & Minder, C. Bias in meta-analysis

559 detected by a simple, graphical test. *BMJ* **315**, 629–634 (1997).

560 35. Page, M. J., Higgins, J. P. & Sterne, J. A. Chapter 13: Assessing risk of bias due

561 to missing results in a synthesis. in *Cochrane Handbook for Systematic Reviews of*

562 *Interventions* (eds. Higgins, J. P. et al.) (Cochrane, 2021).

563 36. Ioannidis, J. P. & Trikalinos, T. A. An exploratory test for an excess of significant

564 findings. *Clinical Trials* **4**, 245–253 (2007).

565 37. Papadimitriou, N. *et al.* An umbrella review of the evidence associating diet and

566 cancer risk at 11 anatomical sites. *Nature Communications* **12**, 4579 (2021).

567 38. Xie, H. *et al.* Can Touchscreen Devices be Used to Facilitate Young Children's

568 Learning? A Meta-Analysis of Touchscreen Learning Effect. *Frontiers in Psychology* **9**, 2580

569 (2018).

570 39. Adelantado-Renau, M. *et al.* Association Between Screen Media Use and Academic

571 Performance Among Children and Adolescents: A Systematic Review and Meta-analysis.

572 *JAMA Pediatrics* **173**, 1058 (2019).

573 40. Madigan, S., McArthur, B. A., Anhorn, C., Eirich, R. & Christakis, D. A.

574 Associations Between Screen Use and Child Language Skills: A Systematic Review and

575 Meta-analysis. *JAMA Pediatrics* **174**, 665 (2020).

576 41. Poorolajal, J., Sahraei, F., Mohammadi, Y., Doosti-Irani, A. & Moradi, L.

577 Behavioral factors influencing childhood obesity: A systematic review and meta-analysis.

578 *Obesity Research & Clinical Practice* **14**, 109–118 (2020).

579 42. Byun, J. & Joung, E. Digital game-based learning for K-12 mathematics education:

580 A meta-analysis. *School Science and Mathematics* **118**, 113–126 (2018).

- 581 43. Vannucci, A., Simpson, E. G., Gagnon, S. & Ohannessian, C. M. Social media use
582 and risky behaviors in adolescents: A meta-analysis. *Journal of Adolescence* **79**, 258–274
583 (2020).
- 584 44. Yoon, S., Kleinman, M., Mertz, J. & Brannick, M. Is social network site usage
585 related to depression? A meta-analysis of FacebookDepression relations. *Journal of Affective
586 Disorders* **248**, 65–72 (2019).
- 587 45. Vahedi, Z. & Zannella, L. The association between self-reported depressive
588 symptoms and the use of social networking sites (SNS): A meta-analysis. *Current Psychology*
589 **40**, 2174–2189 (2021).
- 590 46. Seetharaman, G. W., Jeff Horwitz and Deepa. Facebook Knows Instagram Is Toxic
591 for Teen Girls, Company Documents Show. *Wall Street Journal* (2021).
- 592 47. Tekedere, H. & Göke, H. Examining the Effectiveness of Augmented Reality
593 Applications in Education: A Meta-Analysis. *International Journal of Environmental and
594 Science Education* **11**, 9469–9481 (2016).
- 595 48. Sadeghirad, B., Duhaney, T., Motaghpisheh, S., Campbell, N. R. C. & Johnston,
596 B. C. Influence of unhealthy food and beverage marketing on children's dietary intake and
597 preference: A systematic review and meta-analysis of randomized trials. *Obesity Reviews* **17**,
598 945–959 (2016).
- 599 49. Marshall, S. J., Biddle, S. J. H., Gorely, T., Cameron, N. & Murdey, I.
600 Relationships between media use, body fatness and physical activity in children and youth:
601 A meta-analysis. *International Journal of Obesity* **28**, 1238–1246 (2004).
- 602 50. Elson, M. *et al.* Do policy statements on media effects faithfully represent the
603 science? *Advances in Methods and Practices in Psychological Science* **2**, 12–25 (2019).
- 604 51. Ashton, J. J. & Beattie, R. M. Screen time in children and adolescents: Is there
605 evidence to guide parents and policy? *The Lancet Child & Adolescent Health* **3**, 292–294

- 606 (2019).
- 607 52. Royal College of Paediatrics and Child Health. *The health impacts of screen time:*
- 608 *A guide for clinicians and parents.* (2019).
- 609 53. Parry, D. A. *et al.* A systematic review and meta-analysis of discrepancies between
- 610 logged and self-reported digital media use. *Nature Human Behaviour* **5**, 1535–1547 (2021).
- 611 54. Byrne, R., Terranova, C. O. & Trost, S. G. Measurement of screen time among
- 612 young children aged 0 years: A systematic review. *Obesity Reviews* **22**, (2021).
- 613 55. Smith, C., Galland, B. C., de Bruin, W. E. & Taylor, R. W. Feasibility of
- 614 automated cameras to measure screen use in adolescents. *American journal of preventive*
- 615 *medicine* **57**, 417–424 (2019).
- 616 56. Ryding, F. C. & Kuss, D. J. Passive objective measures in the assessment of
- 617 problematic smartphone use: A systematic review. *Addictive Behaviors Reports* **11**, 100257
- 618 (2020).
- 619 57. Guyatt, G. *et al.* GRADE guidelines: 1. IntroductionGRADE evidence profiles
- 620 and summary of findings tables. *Journal of Clinical Epidemiology* **64**, 383–394 (2011).
- 621 58. Twenge, J. M. More Time on Technology, Less Happiness? Associations Between
- 622 Digital-Media Use and Psychological Well-Being. *Current Directions in Psychological*
- 623 *Science* **28**, 372–379 (2019).
- 624 59. Kelly, Y., Zilanawala, A., Booker, C. & Sacker, A. Social Media Use and
- 625 Adolescent Mental Health: Findings From the UK Millennium Cohort Study.
- 626 *EClinicalMedicine* **6**, 59–68 (2018).