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10 Prevalence of Generative Artificial Intelligence Sexualized Image Usage by Adolescents in the
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32 **Abstract**

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34 The prevalence of generative artificial intelligence (GenAI) usage related to sexualized images

35 amongst adolescents is a critical emerging research area. In this exploratory study, a nationally

36 representative online survey of 557 English-speaking individuals aged 13 - 17 was conducted.

37 Participants were asked about their consensual and non-consensual usage and interactions with

38 sexualized GenAI images. The survey included questions on images created with nudification

39 software as well as content creation software, and asked participants about their creation,

40 sharing, and viewing of this content, as well as that of their peers. Use of nudification tools was

41 widespread, with 55.3% (n=308) of participants reporting having created and 54.4% (n=303)

42 having received at least one image. Reported victimization levels of participants was substantial,

43 with 36.3% (n=202) of individuals reporting having a non-consensual image created and 33.2%

44 (n=185) of individuals having had at least one image non-consensually shared. Usage was

45 similar across demographic categories, though male participants had higher degrees of regular

46 GenAI sexual image creation and distribution, both consensual and non-consensual.

47 Policymakers need to consider the extensive usage and its normalization and consider

48 educational interventions, and practitioners need to be aware of the high degree and nature of

49 victimization occurring.

50

51 **Introduction**

52

53 Recent advances in generative artificial intelligence (GenAI) have ushered in a paradigm shift in

54 image generation and manipulation. Tools such as Stable Diffusion, DALL-E and Midjourney

55 allow users to create images from text prompts, leveraging large language models (LLMs),

56 through web-based services. This expands artistic possibilities and allows individuals to express

57 their creativity in new and imaginative ways [1,2]. Similarly, locally installed applications on
58 smartphones and computers allow users to create their own visual content and modify existing
59 visual content (for example, placing real individuals into generated or existing photographs
60 seamlessly) using techniques such as inpainting [3]. These techniques have been applied to
61 many legitimate, novel uses, ranging from artificially aging photographs for forensic purposes
62 [4] to enabling smartphone applications that allow individuals to virtually “try on” clothing
63 before buying it [5].

64 As with many new technologies, GenAI can also be used for maladaptive purposes. For child
65 sexual exploitation material (CSEM) offending, there are new affordances available using these
66 tools. Some of these affordances involve the creation of toolsets (e.g., training models
67 specifically with CSEM content), but the key uses for most individuals engaged with GenAI for
68 CSEM offending mirror the legitimate uses of the technology. These include the generation of
69 CSEM content directly using text prompts and common diffusion models [6], the use of
70 nudification tools to visualize what individuals might look like without clothing [7,8], and the
71 alteration of adult sexual exploitation material to include the faces of minors [9].

72 One area of particular concern is the increase in the creation and distribution of self-generated
73 sexualized images by minors. These images are now a substantial portion of the images found
74 in CSEM reports, with the Internet Watch Foundation (IWF) reporting that 44% of CSEM shared
75 online was self-produced [10]. Initial estimates of sexting amongst adolescents in the United
76 States found 2.5% created (or had created) sexualized images of themselves, and 7.1% received
77 these images [11]. More recent work identified higher rates of creation and receipt of these
78 images, with usage increasing over time [12], and a meta-analysis found creation and receipt
79 rates of 14.8% and 27.4%, respectively [13]. Additionally, the creation of these images has

80 shown some normalization within the attitudes of adolescents towards these activities [14], and
81 they have become a routine part of adolescent sexual exploration. The IWF reporting at the time
82 and the more recent statistics on sexting, however, did not differentiate between GenAI and non-
83 GenAI images, due both to methodological (the protocols did not look for GenAI images) and
84 technological (the difficulty in differentiating GenAI from other images) reasons.

85 In addition to the increase in self-generated, sexual images through sexting, the adoption of
86 GenAI technologies by adolescents is expected to continue to grow. In November 2024, Ofcom
87 reported that 54% of British children aged 8-15 had used GenAI in the past year [15].

88 Smartphone applications that utilize augmented reality (using similar technology to nudification
89 apps) for the purposes of trying on virtual clothing are also more accepted by younger users [16].
90 With nudification apps becoming easier to find and use [17], familiarization with and
91 normalization from the usage of similar GenAI applications may increase adoption amongst
92 adolescents.

93 Adolescents are already exposed to sexualized GenAI images in the form of deep fakes. In 2024,
94 14% of British adolescents under 16 had come across a sexualized image or video created with
95 GenAI in the prior year over social media, video sharing platforms, or email [15]. Deep fakes in
96 general have been found to be met with negative sentiments [18], but age-related differences
97 have not been well studied. Teens in particular may exhibit different behaviors in the creation
98 and distribution of sexualized GenAI images, consistent with the normalization of self-generated
99 sexually explicit images [14].

100 In addition to general exposure, the harms of deep fakes on victims are just starting to be
101 explored. Victims of GenAI-based sexual exploitation reported issues consistent with other

102 forms of CSEM victimization, including fears of who may have seen the images when in public
103 (hypervigilance) and general avoidance of social media usage, as well as a sense of
104 powerlessness to prevent it and general dehumanization, resulting in permanent life disruptions
105 [19,20]. Limitations based on current laws further constrain the resources available to victims of
106 malicious deep fakes, particularly those of a sexualized nature [21].

107 The production, viewing, and distribution of pornographic GenAI images of individuals under 18
108 is illegal in the United States under federal law (18 U.S.Code § 1466A), and does not require that
109 a real individual be depicted [22]. In both the United States and most of Europe, there are gaps
110 in the current laws related to GenAI CSEM, including the legality surrounding the possession
111 and training of tools to produce content [23]. Additionally, while there are no safe harbor
112 provisions in United States law for consensually produced adolescent content, in practice
113 prosecutorial discretion is used to allow for education and administrative remedies in lieu of
114 criminal charges. A full review of the legal issues surrounding GenAI CSEM can be found at
115 [24].

116 While there has been an increase in individual reports of GenAI misuse by adolescents [25–27],
117 overall prevalence rates are unknown. This research is the first large-scale exploratory effort to
118 measure the usage of GenAI tools to create sexualized images, including nudification tools and
119 image creation tools (based on text prompts or inpainting of existing images), by adolescents.
120 Both consensual and non-consensual use cases are evaluated, including creation and distribution
121 by both participants and by their peers, using an anonymous, online survey.

122 **Materials and Methods**

123 This study consisted of an anonymous, cross-sectional, Internet-based survey targeting
124 adolescents and their interaction with GenAI sexual imagery.

125 **Participants and Setting**

126 This research obtained data through an online, anonymous survey hosted by Qualtrics. The
127 survey consisted of multiple questions related to the interactions of participants with sexualized
128 images created using GenAI. The population of the survey was English-speaking adolescent
129 teens living in the United States between the ages of 13 and 17, inclusive. A target sample of
130 500 individuals was solicited, based on the ability to identify a medium effect size (.3) for in-
131 group demography variations, between January 11 and January 24, 2025. A pilot distribution of
132 50 responses was solicited prior to the full launch to ensure survey flow, response controls, and
133 timing were performing as expected. The researchers had no direct involvement in the selection
134 of participants, who were drawn by the providers from a group of individuals whose parents pre-
135 identified them as potential candidates as part of a panel service offered by Qualtrics. The
136 sample was designed to be representative and used a non-probability quota-based methodology
137 based on demographic data previously provided by the participants [28]. No direct
138 compensation was provided to participants by the research team, however Qualtrics provides
139 compensation to panel members ranging from gift cards to airline miles based on proprietary
140 algorithms.

141 Due to the age of the participants, a two-stage process of informed consent was utilized. First,
142 the parents of potential participants were provided a link to take a survey by Qualtrics. Those
143 parents who clicked the link were provided details on the survey's nature, including a copy of the
144 questions as well as a detailed consent form outlining the risks and benefits of the study. The

145 information provided to the parents confirmed the results of the study would be anonymous and
146 the anonymized data available to future institutional review board approved studies, and that they
147 would not have access to their child's responses. Parents were requested to allow their children
148 privacy to fill out the questions but encouraged to discuss the questions and the topic with their
149 children afterward. Of the 9098 parents who were presented the form, 967 agreed to their child's
150 participation. The children of parents who provided consent were then presented with a detailed
151 assent form containing age-appropriate language and detailing the same information as the
152 consent form. Of the 967 adolescents presented the form, 936 agreed to continue with the
153 survey. Following their completion of the survey, adolescents were encouraged to voluntarily
154 discuss the issues addressed in the survey with their parents.

155 The surveys were anonymous, and no identifying information, including network information
156 (e.g., IP addresses), was retained. The parents were provided an anonymous link to the consent,
157 and, if they chose to consent, their child was provided a separate, unrelated anonymous link to
158 the survey. The specific demographic information collected was limited to avoid unintentionally
159 allowing ex post facto identification of a specific participant through aggregate demographic
160 details (e.g., no zip codes were collected). The survey was additionally designed to avoid any
161 free-text entry boxes through which participants may inadvertently self-identify. Due to the
162 anonymous nature of the study, no direct debriefings were possible, but links to counseling
163 resources were provided to both the participants and their parents if they wanted to discuss any
164 reported victimization or the topic area in general. Additionally, law enforcement contacts were
165 provided to both participants and parents for voluntary reporting of any related events. Finally,
166 contacts were provided for both the research team and the institutional review board to both
167 parents and participants should they have any detailed questions or concerns about the research.

168 The adolescents were permitted to cease their participation in the survey at any point up until
169 submission but following submission their data was no longer identifiable, and only anonymized
170 data was available to the research team.

171 There were multiple levels of quality assurance built into the process. Qualtrics used browser
172 fingerprinting to avoid duplicate entries, anti-botnet features such as CAPTCHAs, and IP
173 geolocation to ensure respondents were within the United States. They additionally included a
174 timing check, and any responses below half the median time were discarded as “speeders”.
175 Finally, they included straight-line and Christmas-tree checks (any that were consistently
176 answered across three separate matrices) to ensure answer integrity [29]. In addition to integrity
177 checks built-in by Qualtrics, two additional checks were provided by the researchers. First, a
178 question confirming the ages of the participants was asked. Those identifying an age outside of
179 the 13 - 17-year-old range were excluded from continuing with the survey. Second, an attention
180 check was built into the survey to improve response quality. A total of 115 participants failed to
181 complete the survey and integrity checks, and 264 were identified as outside of the target age
182 group, resulting in n=557 total surveys available for analysis.

183 **Questionnaire**

184 Demographic questions on race, gender (participants were asked how they describe themselves),
185 sexual orientation, region and age were collected from all participants. For the purposes of this
186 study sexualized GenAI images were operationally defined as naked, still images or videos,
187 either of their peers (“other individuals between the **ages of 13 and 17 years old**”) or of adults
188 (“any other individuals who are 18 years of age or older”). The primary questions were related
189 to participants’ interactions with relevant content, and specific questions were asked about both

190 consensual and non-consensual creation and distribution of naked GenAI images. Questions
191 were focused on four areas:

192 • Use of nudification tools by participants
193 • Use of nudification tools by others in their peer group
194 • Use of general GenAI image creation/alteration tools by participants
195 • Use of general GenAI image creation/alteration tools by others in their peer group

196 Participants were asked about their own creation of images (of themselves or others), their
197 distribution of images, and their receipt of naked images of both adults and their peers. For each
198 question, a five-point, custom Likert scale was used with the following values:

199 • I have never done this
200 • I have only done this once or twice
201 • I have done this infrequently
202 • I have done this frequently
203 • I have done this on a regular basis

204 The questions were specifically worded based on prior research questions related to CSEM
205 activities to minimize the impact of social desirability bias on responses [30].

206 Participants were provided age-appropriate, plain text operational definitions and directions for
207 each of the sections. For example, nudification tools were defined as follows:

208 The following questions are about the use of nudification tools. These are smartphone
209 apps, websites, or other software that shows what individuals might look like without
210 clothing.

211 AI image creation tools were operationally defined as those that “use text prompts or existing
212 images as inputs... [creating] new images from scratch or using existing images and altering
213 them with AI” to create naked content.

214 Finally, respondents were asked to confirm if they filled out the questions in private, with a
215 parent present, or with another individual present to allow researchers to identify any potential
216 observer influence on responses.

217 **Analysis**

218 The results were collected and stored on the university file share, and all analysis was performed
219 in R (4.5) using R Studio. Wilcoxon rank-sum tests for two category comparisons and chi-
220 squared tests for multi-category comparisons were utilized for the ordinal and categorical
221 comparisons, and Spearman correlation for age comparisons. An additional chi-squared test was
222 performed for age comparisons as well to identify potential non-linear effects. A statistical
223 significance level of .01 was used where p values were reported, with Bonferroni corrections
224 applied.

225 **Ethics**

226 The study design and protocols, including an analysis of the potential risks and benefits, was
227 approved by the George Mason University Institutional Review Board on 19 December 2024.

228 **Results**

229 The research sample consisted of a representative, demographically diverse group of English-
230 speaking, United States teens. The survey was completed by 557 individuals (51.0% male,

231 48.3% female, 0.5% non-binary, 0.2% not specified; 73.2% Caucasian, 15.6% African
 232 American, 2.3% Asian, 1.4% Native American, 4.7% with two or more races; 89.2% identified
 233 as heterosexual, 6.5% as bisexual, and 3.2% as homosexual). With respect to privacy when
 234 taking the survey, 70.6% reported taking it alone, 20% with a parent present, 0.5% with another
 235 individual present, and 9% chose not to answer. Table 1 shows the overall sample
 236 characteristics.

237 **Table 1. Sample Demographics.**

Characteristic	n	Percentage	95% CI
<i>Age</i>			
13	97	17.4%	(14.3 - 20.6)
14	92	16.5%	(13.4 - 19.6)
15	109	19.6%	(16.3 - 22.9)
16	105	18.9%	(15.6 - 22.1)
17	154	27.6%	(23.9 - 31.4)
<i>Race</i>			
American Indian/Native American or Alaska Native	8	1.4%	(0.4 - 2.4)
Asian	13	2.3%	(1.1 - 3.6)
Black or African American	87	15.6%	(12.6 - 18.6)
Multiple Races	26	4.7%	(2.9 - 6.4)
Native Hawaiian or Other Pacific Islander	2	0.4%	(-0.1 - 0.9)
Other	12	2.2%	(0.9 - 3.4)
Prefer not to say	1	0.2%	(-0.2 - 0.5)
White or Caucasian	408	73.2%	(69.6 - 76.9)
<i>Spanish, Hispanic, or Latino Origin</i>			
Yes	94	16.9%	(13.8 - 20)
No	463	83.1%	(80 - 86.2)
<i>Gender</i>			
Female	269	48.3%	(44.1 - 52.4)
Male	284	51.0%	(46.8 - 55.1)
Non-binary / third gender	3	0.5%	(-0.1 - 1.1)
Prefer not to say	1	0.2%	(-0.2 - 0.5)
<i>Sexual Orientation</i>			
Bisexual	36	6.5%	(4.4 - 8.5)

Heterosexual	497	89.2%	(86.7 - 91.8)
Homosexual	18	3.2%	(1.8 - 4.7)
Other	1	0.2%	(-0.2 - 0.5)
Prefer not to say	5	0.9%	(0.1 - 1.7)
<i>Region</i>			
Midwest	125	22.4%	(19 - 25.9)
Northeast	102	18.3%	(15.1 - 21.5)
South	145	26.0%	(22.4 - 29.7)
West	181	32.5%	(28.6 - 36.4)
Other	4	0.7%	(0 - 1.4)

238

239 Overall, usage of GenAI tools to create naked images was widespread. Self-generation of
 240 nudified images (use of a nudification tool to create an image of oneself) was the highest
 241 modality, with 55% of participants having done this at least once, 44% having shared a self-
 242 image created with nudification tools, and with 54% of recipients having received a self-
 243 generated image from a nudification tool. Content creation and sharing appeared to be generally
 244 focused on peer group images and not adult images, with only 35% having created and 33%
 245 having shared naked images of adults. Additionally, the usage of nudification tools was
 246 significantly higher than the usage of GenAI creation tools ($z=-3.66$, $p<.001$) comparing any
 247 nudification image interaction with any traditional GenAI image interaction, indicating a higher
 248 impact of victimization (for the purposes of this research, creation or distribution of images
 249 without their permission) as nudification tools require an actual photo of a real individual as a
 250 basis image. Overall details by activity are available in the online supplemental material (S1
 251 Table).

252 Only male and female gender identities were analyzed for differences as insufficient numbers of
 253 non-binary gender identities were present in the sample to evaluate them effectively. There was
 254 an overall similar usage between genders, but seven categories of usage had higher usage by
 255 males, including the creation of peer images with nudification tools both with ($W= 31967$,

256 p<.001) and without (W=32592, p<.001) their permission; having received re-distributed images
257 of their peers (W=31953, p<.001); using image creation tools to create images of adults
258 (W=32521, p<.001); sharing naked images of themselves (W=31588, p<.001) and of others in
259 their peer group (W=32496, p<.001) created with image creation tools; and having had images of
260 adults (W=32278, p<.001) and other peers (W=30863, p<.001) shared with them by their peer
261 group (S2 Table).

262 There were no statistically significant differences in usage based on age across any of the
263 activities, either through a linear fit based on age or categorically (S3 Table). When comparing
264 prevalence rates across sexual orientation, only sharing of AI-generated images of adults differed
265 significantly across orientations, $X^2(16)=42.47$, p<.001 (S4 Table). A post-hoc analysis of the
266 results with Bonferroni correction identified individuals who identified as “Other” having a
267 significantly higher residual (4.823, p<.001) (S5 Table). No statically significant prevalence
268 differences were identified associated with race (S6 Table).

269 No statistically significant differences were identified based on the presence of a parent or other
270 individual when taking the survey in the responses (S7 Table).

271

272 **Discussion**

273

274 Better education is needed on the healthy and safe usage of GenAI technology, which has been
275 previously believed to disproportionately impact women [20] and was borne out in this research,
276 though both genders were found to be substantially impacted in this study as both users and
277 victims. While only a few questions showed statistically significant distributions by gender, both
278 self and peer creation and distribution were higher for males in terms of regular use and for

279 having experimented (engaged at least one time) with the technologies. The lack of significant
280 gender differences with a large effect size for most of the actions asked about is consistent with
281 Madigan et al.’s work on youth sexting, and may be a continuation of that trend reflecting
282 technological advances [13]. The one differentiator related to sexual orientation was a higher
283 incidence of “Never” within those who identified as “Other”. This may be due to individuals
284 who self-identify as asexual being included in this category, but additional research with a larger
285 dataset and more detailed categorization is needed to verify this hypothesis.

286 Education interventions additionally need to be multimodal and need to occur at a young enough
287 age to address the issue before it occurs [31]. Since there was no statistically significant age
288 difference in either usage or victimization across the age groups (though as with gender, smaller
289 effect sizes may be identified with higher powered studies), the education should occur prior to
290 age 13 to be most impactful. Prior work on rape myth acceptance, particularly that targeting
291 bystanders (in this context, individuals receiving or being shown unsolicited images) and how
292 they should respond, as well as impact education to prevent perpetration are warranted and can
293 potentially inform effective approaches [32].

294 In terms of victimization, the usage of and sharing of content from nudification tools was higher
295 than that of general GenAI creation tools, portending a higher degree of direct victimization as
296 nudification tools generally involve a known/direct child victim. These results represent a lower
297 bound in victimization statistics, as perpetration by adults (use of these tools on minors or
298 sending images to minors) was not incorporated into this study. Secondary victimization through
299 the further distribution of images without consent is likewise a lower bound as participants may
300 not know the extent of onward transmission. Further research, particularly longitudinal research,

301 on the long-term impacts of creating sexualized images (particularly self-images) using these
302 tools is needed.

303 Most of the activities surveyed represent the creation of CSEM, a violation of federal law in the
304 United States. Because many of the actions were consensual and involving peers, policymakers
305 need to consider if these activities, which may be considered part of normal sexual exploration,
306 warrant specific legal exceptions beyond current discretionary prosecution. Specifically, carve-
307 outs for consensual generation and sending between individuals in the same age group warrant
308 consideration. Better controls on the usage of nudification and other GenAI applications to
309 detect CSEM production is also needed. While calling for application makers to provide better
310 controls to limit their use by minors is necessary, many of the applications are gray market tools
311 and age restriction controls are not likely to be fully successful [17] unless coupled with an
312 education strategy.

313 The long-term impacts on CSEM distribution additionally need to be considered in light of these
314 findings, assuming they remain stable over time. GenAI detection tools need to specifically
315 target modified images to ensure victim identification given the significant use of nudification
316 applications. Finally, the prevalence of GenAI nudification and creation tool usage by adults,
317 both legally and illegally (to produce CSEM or non-consensual explicit images) needs to be
318 studied [33].

319 **Limitations**

320

321 This research was an exploratory study conducted on English-speaking teens 13 - 17 years of age
322 within the United States, and additional work would be required for generalizability to other
323 populations. Because of the relatively recent introduction of the latest generation GenAI

324 technologies, these results only represent a point-in-time analysis and future research once the
325 technologies mature, as well as longitudinal research, are needed before drawing any broad
326 conclusions. Although controls were put in place both by Qualtrics and the research team to
327 obtain accurate survey results through attention checks, timing issues, and verification of
328 privacy, there are general issues with Internet surveys that will always be present in this type of
329 research. Specific to this research, the IRB-approved protocol provided parents the ability to
330 review the full questionnaire before allowing their children to participate. This introduces a
331 potential selection bias, where parents who opt out of participation for social, religious, cultural,
332 or other reasons may have adolescents that differ in their GenAI usage from those whose parents
333 opted in. The number of respondents was selected to have sufficient power to perform larger
334 subgroup analyses but was not sufficient to draw conclusions about smaller subgroups (e.g., non-
335 binary teens) or to identify small effect sizes. Participants were asked about their use of GenAI
336 and interactions with sexualized images of others in their peer group and of adults, however no
337 questions were asked of their use of these tools to create images of those younger than 13, so the
338 numbers represent an upper bound of individuals in this group using GenAI to interact with
339 CSEM imagery. Finally, the questions asked about the generation and sharing of naked images -
340 other sexualized imagery with no nudity is possible, and some naked images may not be
341 sexualized, however this language was chosen for ease of understanding and interpretation.

342

343 **Conclusions**

344

345 A significant percentage of adolescents in this national survey were found to engage in the use of
346 GenAI applications to create sexualized images, including nudification applications. This
347 represents a potentially significant source of CSEM creation and distribution, and shows that, to

348 some degree, risky engagement with GenAI is widespread in this population. The normalization
349 of these activities, and the harm to victims for non-consensual creation and/or distribution needs
350 further study for its impact on prevention, treatment and deterrence efforts. While the usage was
351 widespread across demographic categories, most of the usage appears to have been exploratory
352 rather than habitual based on the frequencies reported, providing an opportunity for positive
353 intervention. Because there were no age-related differences in GenAI usage, age-appropriate
354 education-based intervention efforts on the risks associated with AI tools and sharing of images
355 need to start prior to the age of 13. Additional training for law enforcement on handling both
356 offenders and victims, as well as digital forensics specialists on the identification of GenAI
357 images, is needed. Finally, decision makers need to consider the prevalence and nature of these
358 use cases in developing new legislation related to GenAI CSEM.

359

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462 **Supporting Information**

463 **S1 Table. Descriptive Statistics of Individual AI Activities.**

464 **S2 Table. Gender Differences in Sexualized Generative AI Usage.**

465 **S3 Table. Age Differences in Sexualized Generative AI Usage.**

466 **S4 Table. Summary of Sexual Orientation-Related Differences in Sexualized Generative**
467 **AI Usage.**

468 **S5 Table. Post-hoc Analysis of Sexual Orientation.**

469 **S6 Table. Summary of Race-related Differences in Sexualized Generative AI Usage.**

470 **S7 Table. Summary of the Impact of Other Individuals Presence During Survey Taking.**

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