

ALGORITHMIC ORIENTALISM: VISUALIZING BANGLADESHI FEMINISM THROUGH GENERATIVE AI

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ABSTRACT

As generative artificial intelligence (GAI) systems become central to global knowledge production, they increasingly shape how political movements are visualized and understood. This project examines the representational logic of leading text-to-image models ChatGPT, Gemini, and Grok in their construction of feminist activism in Bangladesh. Through a qualitative visual content analysis of 100 AI-generated images, this study identifies a systemic pattern of digital Orientalism and representational imperialism. Findings reveal that these models consistently fail to render Bangladeshi-specific political identities, instead defaulting to “Indianized” facial features, Westernized protest aesthetics, and distorted, illegible Bengali script. I argue that these outputs are not neutral technical glitches but are manifestations of algorithmic bias embedded in uneven global data ecologies. These distortions enact a form of epistemic erasure, where the distinct political history of Bangladesh is subsumed under dominant regional and Western tropes. Furthermore, the study identifies a “modal alignment gap,” where AI systems textually acknowledge their biases while continuing to reproduce them visually. By reframing these technical failures as an intra-Global South imperialism, this research challenges the adequacy of current Responsible AI frameworks. It concludes that without a commitment to linguistic justice and representational sovereignty, GAI will continue to function as a regime of imperial representation that silences subaltern voices in the digital age.

1 INTRODUCTION

Generative Artificial Intelligence (GAI) is increasingly integrated into daily life Bick et al. (2026). Beyond informational use, people now engage with GAI for a wide range of creative and professional tasks, including writing, design, problem-solving, and image generation. These conversational AI chatbots Achiam et al. (2023) are presented as general-purpose systems that emphasize openness, egalitarian values, and broad accessibility, thereby positioning themselves as inclusive technologies Amodei (2024). However, this study examines how GAI simultaneously contributes to the reproduction of global cultural hierarchies through the depiction of Bangladeshi feminist activists. It focuses on how digital representations shaped by algorithmic bias reinforce stereotypical narratives and promote cultural homogenization by privileging dominant languages, ethnicities, and aesthetic standards Noble (2018); Buolamwini & Gebru (2018).

To examine this tension between inclusivity claims and representational practice, this study analyzes 100 images generated by leading models: ChatGPT, Gemini, and Grok, and identifies recurring patterns of representational imperialism. Rather than capturing the distinct cultural markers of Bangladeshi resistance, these systems consistently produce a “homogenized regional aesthetic.” Activists are rendered with Indianized facial features, draped in ornate wedding-style “sarees” (a traditional cloth) unsuited for street protest, and flanked by placards in which the Bengali language (the very soul of Bangladeshi political identity) is reduced to an illegible, hallucinatory script.

These inaccuracies should not be understood as neutral technical “glitches.” Instead, they constitute a form of digital Orientalism Said (1978); Graham & Dittus (2022). By subsuming Bangladesh under dominant Indian or Western visual tropes, these systems risk enacting an epistemic erasure that

renders local feminist histories invisible within AI-mediated knowledge systems. Moving beyond a simple Global North–Global South binary, this research exposes an intra–Global South imperialism, whereby algorithmic defaults allow dominant regional cultures to overwrite minority identities Qadri et al. (2023).

This study is guided by the following research questions:

- How do leading text-to-image AI systems construct visual representations of feminist activism in Bangladesh?
- How do these representations reproduce digital Orientalism and imperial hierarchies?
- In what ways do these outputs challenge existing Responsible AI frameworks?

Taken together, this study challenges the adequacy of current Responsible AI frameworks for addressing representational harm. While many AI systems emphasize fairness, transparency, and bias awareness, these approaches often overlook how visual misrepresentation operates across cultural, linguistic, and geopolitical contexts. By examining AI-generated images of Bangladeshi feminist activism, this study foregrounds a persistent “modal alignment gap,” Liang et al. (2022) in which models can acknowledge bias in text while continuing to reproduce it visually. Rather than treating such misrepresentations as isolated technical errors, this research frames them as failures of political recognition that shape how subaltern voices and histories are made visible (or erased) within AI-mediated knowledge systems Spivak (2004). As AI-generated imagery increasingly shapes public understanding of social movements, the Bangladeshi case offers a lens for examining how representational practices intersect with questions of epistemic justice and digital sovereignty Couldry & Mejias (2019).

2 LITERATURE REVIEW

Research across media studies, critical data studies, and human-centered AI has established that algorithmic systems are not neutral technologies. Rather, they are sociotechnical systems shaped by historical power relations, cultural hierarchies, and geopolitical inequalities. Algorithms play an active role in structuring visibility and legitimacy by determining what becomes representable, intelligible, and authoritative within digital environments.

These dynamics are inseparable from longer histories of imperialism. Classical political economy defines imperialism as a system through which dominant powers extend economic, political, and cultural control over peripheral regions, often through indirect and structural means rather than direct rule Hobson (1902); Lenin (2015). While early formulations emphasized territorial expansion and capital accumulation, later scholarship demonstrates that imperialism also operates through knowledge production, representation, and epistemic authority. Control over how societies are known, described, and interpreted becomes a central mechanism of domination.

Within contemporary digital systems, this epistemic dimension of imperialism is particularly visible. Global digital platforms operate within entrenched hierarchies of language and geography Venuti (2019); Graham & Dittus (2022). English-language data and Western epistemic frameworks dominate computational systems, positioning Western and geopolitically central cultures as defaults while marginalizing local and minoritized identities. These asymmetries reflect colonial histories, infrastructural concentration, and uneven global development, producing systematic disparities in how places and cultures are represented Graham & Dittus (2022).

Within this context, generative AI systems do not simply mirror existing inequalities but actively reproduce and amplify them. Studies of large language models demonstrate that spatial and linguistic biases systematically privilege data-rich, geopolitically dominant regions, rendering some places hypervisible while marginalizing others Kerche et al. (2026). These patterns emerge from historically uneven data ecologies, where representation is concentrated around dominant regions and languages. Consequently, generative AI embeds global hierarchies directly into its representational logic rather than correcting them.

Visual AI systems further reveal how algorithmic design normalizes particular bodies, faces, and cultural forms. Empirical research has demonstrated that commercial AI systems perform significantly better on lighter-skinned and Western-coded subjects than on darker-skinned or non-Western

108 populations, indicating that whiteness functions as an unmarked norm within training data Bu-
109 lamwini & Gebru (2018). While this research focuses on recognition systems, it highlights broader
110 patterns of visual calibration that shape who is rendered legible within AI-generated imagery. These
111 representational inequalities extend beyond technical performance into the governance of meaning
112 itself. Algorithmic ranking and recommendation systems have been shown to systematically repro-
113 duce racist and sexist tropes, governing what becomes visible and authoritative while marginalizing
114 alternative narratives Noble (2018). Such processes function as a form of technological redlining,
115 where access to accurate representation is unevenly distributed and shaped by corporate priorities
116 rather than democratic accountability.

117 At the level of identity, algorithmic systems rely on probabilistic classification to assign individuals
118 and groups to reductive categories, often detached from lived experience. This process produces
119 identities that appear coherent to the system but flatten social complexity, resulting in stereotypical
120 representations that collapse cultural diversity into a narrow set of recognizable tropes Cheney-
121 Lippold (2011). In generative AI, this logic becomes visible through repeated visual archetypes that
122 stand in for entire cultures or movements.

123 These dynamics operate within broader structures of epistemic power. Algorithmic systems in-
124 creasingly function as gatekeepers of knowledge, shaping what users encounter and accept as truth.
125 Research on algorithmic knowledge gaps shows that these systems embed the assumptions of privi-
126 leged developers while obscuring the mechanisms through which prioritization occurs, leaving users
127 with limited ability to contest misrepresentation Cotter & Reisdorf (2020). As a result, dominant
128 discourses encoded in algorithmic systems appear objective and universal, while minoritarian per-
129 spectives are rendered unintelligible or irrelevant. This dynamic has been theorized as epistemic
130 oppression, where statistically dominant information is treated as epistemically authoritative simply
131 because it is statistically prevalent Miragoli (2024). From this perspective, misrepresentation is not
132 simply an error but a structural exclusion built into algorithmic sense-making.

133 Postcolonial theory provides a critical framework for understanding these representational dynam-
134 ics. Said demonstrates that imperial power is sustained through systems of knowledge that construct
135 non-Western societies as static, exotic, and legible only through Western categories Said (1978).
136 Generative AI extends these logics into digital form, reproducing colonial ways of seeing through
137 automated systems trained on biased global data (Graham & Dittus (2022)). Postcolonial femi-
138 nist theory further complicates this dynamic by foregrounding questions of voice and silencing.
139 Scholarship on subalternity has shown that marginalized subjects, particularly women, are often un-
140 able to speak within dominant knowledge systems without their voices being filtered, translated, or
141 distorted Spivak (2004). Even when subaltern women appear within political narratives, they are
142 frequently represented through frameworks shaped by colonial and patriarchal power.

143 Empirical studies in South Asia confirm these concerns. Research on text-to-image models shows
144 that these systems struggle to generate culturally specific subjects, default to hegemonic regional
145 aesthetics, and reproduce familiar tropes. Users in Bangladesh and neighboring countries report
146 experiences of homogenization and erasure of local cultural markers, indicating that generative AI
147 often operates through an outsider’s gaze shaped by regional and global power asymmetries Qadri
148 et al. (2023). These findings align closely with the representational failures observed in this study.
149 Beyond representation, scholars have increasingly framed AI within broader regimes of extraction
150 and dependency. The concept of data colonialism describes how human life is continuously ap-
151 propriated through data capture and monetized by corporations concentrated in the Global North,
152 reproducing extractive logics reminiscent of historical colonialism Couldry & Mejias (2019). Al-
153 though initially developed to analyze surveillance capitalism, this framework applies directly to
154 generative AI systems trained on globally uneven data ecologies. This extractive dynamic has direct
155 implications for digital sovereignty. Research on AI geopolitics shows that Global South nations
156 contribute vast amounts of data to global AI systems while lacking meaningful control over how
157 that data is processed, governed, or represented. As a result, local populations often receive dis-
158 torted or marginalizing representations in return for their data contributions, reinforcing structural
159 dependency rather than digital autonomy de Freitas (2025). These asymmetries are further intensi-
160 fied by the concentration of digital infrastructure, platform governance, and computational power in
161 a small number of countries (Graham & Dittus (2022)).

160 Taken together, this scholarship demonstrates that generative AI systems are not merely biased tech-
161 nologies but geopolitical infrastructures. They reproduce colonial patterns of extraction, depen-

162 dency, and epistemic domination not only through data capture but through representational control.
 163 From this perspective, misrepresentation within generative AI is a political problem rooted in power,
 164 sovereignty, and postcolonial inequality rather than a technical flaw.
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166 **Gap and Necessity of This Study** While postcolonial and decolonial approaches to artificial intelligence
 167 are expanding, existing research has largely focused on data extraction, labor exploitation,
 168 governance, and infrastructural power. Much less attention has been paid to how generative AI systems
 169 actively produce postcolonial knowledge through visual representation. This study addresses
 170 that gap by examining how text-to-image models construct feminist activism in Bangladesh, treating
 171 AI-generated images as political texts rather than technical outputs. By systematically analyzing
 172 visual representations, this research demonstrates how Bangladesh is not merely underrepresented
 173 but rendered epistemically overwritable by dominant regional and Western imaginaries. Crucially,
 174 this study moves beyond a simple Global North–Global South binary by documenting a form of
 175 intra–Global South representational imperialism, in which Indian cultural defaults algorithmically
 176 overwrite Bangladeshi identities. It further conceptualizes linguistic distortion and visual hallucination
 177 as forms of political silencing rather than technical error. By reframing algorithmic bias as a
 178 failure of political recognition, this study extends postcolonial theory into the domain of generative
 179 imagery. It argues that generative AI functions as a regime of imperial representation, actively shaping
 180 who can be seen, how feminist movements are understood, and whose knowledge counts in the
 181 digital age.

182 3 METHODOLOGY

183 3.1 DATA COLLECTION

184 I analyzed 100 images generated by ChatGPT, Gemini, and Grok. I chose these platforms because
 185 they are widely used, technologically influential, and accessible to me without institutional barriers.
 186 Images were generated using identical or near-identical prompts, such as *Generate an image of the*
 187 *feminist movement in Bangladesh*, alongside context-specific variations to test representational
 188 consistency. I focus on feminist activism because it builds directly on my previous research examining
 189 how feminism is discussed on social media in Bangladesh. This continuity allows me to shift analytically
 190 from user-generated discourse to AI-generated representation, while maintaining thematic
 191 coherence.
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194 3.2 ANALYTICAL FRAMEWORK

195 To interpret the images, I draw on Salvaggio’s methodology for analyzing AI-generated imagery.
 196 First, I used identical prompts across platforms to ensure comparability Salvaggio (2023). I examined
 197 which elements appeared realistic, detailed, and coherent, and which appeared distorted, vague,
 198 or awkward.
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- 200 • **Strong representation zones** such as well-rendered faces, fluent English text, or polished
 201 protest aesthetics suggest over-representation in training data.
- 202 • **Weak representation zones** including distorted Bengali letters, blended cultural symbols,
 203 or inaccurate attire indicate data scarcity, filtering, or suppression.
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205 Repeated elements across outputs (e.g., Indian-style sarees or English-language placards) function as
 206 strong cultural signals, reflecting dominant visual norms embedded in training datasets. Conversely,
 207 inconsistent or incorrect details (e.g., illegible Bengali text) operate as weak signals, pointing to
 208 marginalization or absence of Bangladeshi-specific data.
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210 Finally, I re-ran identical prompts over time to examine whether outputs shifted in response to public
 211 critiques of AI bias. As (Salvaggio, 2023) argues, longitudinal comparison offers insight into how
 212 responsive or resistant platforms are to equity-based interventions.

213 I employed qualitative visual content analysis using a structured coding framework to identify re-
 214 curring patterns across images. The coding categories included:

- 215 • **Ethnicity:** facial structure, skin tone, and markers of perceived ethnic identity

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Figure 1: AI-generated image of feminist activism in Bangladesh, depicting women in ornate sarees that reflect generalized Indian visual stereotypes rather than the everyday clothing typically worn by Bangladeshi protesters.

- **Clothing Style:** attire, fabrics, colors, and motifs associated with cultural or religious identities
- **Linguistic Elements:** visible text on banners or placards, language choice (Bengali, English, Hindi), script accuracy, and spelling
- **Forms and Participants of Resistance:** modes of activism (e.g., street protest, writing, performance), gender composition, and participation diversity

From these codes, I identify three broader representational patterns characteristic of generative AI outputs:

1. **Modal Alignment Gaps:** mismatches between text information and the narratives implied by images.
2. **Representational Similarity Bias:** reliance on visually dominant regional or global archetypes
3. **Hallucination and Identity Collapse:** the production of culturally incoherent or non-existent linguistic and visual elements

4 FINDINGS AND ANALYTICAL FRAMEWORKS

4.1 EMPIRICAL FINDINGS: VISUAL AND LINGUISTIC REPRESENTATIONS

4.1.1 CULTURAL CONFLATION AND VISUAL STEREOTYPING

Clothing When I asked ChatGPT to generate images of feminist activism in Bangladesh, it frequently produced images of women whose facial features and attire closely resembled dominant Indian visual representations. Many images, especially from Gemini, showed women wearing heavy, ornate sarees that resemble Indian wedding attire (Figure 1). This kind of clothing is rarely worn at protests in Bangladesh, where activists, particularly students, tend to wear salwar kameez, jeans, t-shirts, hijabs, or everyday clothes suitable for long demonstrations. These visuals feel less like documentation and more like fantasy: a stylized, exotic version of “South Asian womanhood” shaped by Western expectations rather than local practice.

Interestingly, while ChatGPT produced some variation, the majority of Gemini-generated images depicted women uniformly dressed in sarees. When I asked Gemini why everyone appeared to be

wearing sarees, it openly acknowledged that this was likely due to bias in its training data (Figure 2). It even explained that real protests in Bangladesh, such as recent student movements, are led by Gen-Z activists dressed in casual, mixed styles. However, when I asked it to generate a more realistic image based on that explanation, the result showed all women wearing hijabs (Figure 3).

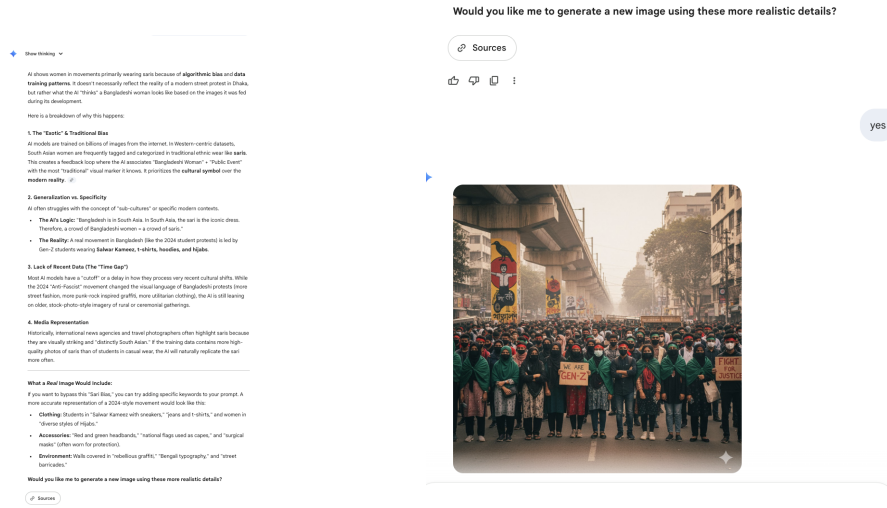


Figure 2: Gemini-generated image depicting women uniformly dressed in sarees, which Gemini attributed to bias in its training data.

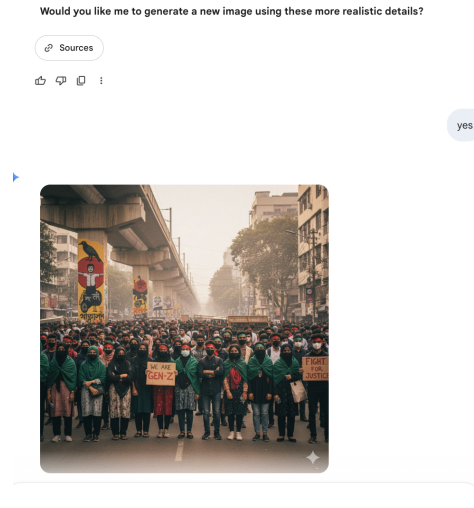


Figure 3: Gemini-generated image produced after requesting a more realistic depiction of protests in Bangladesh, resulting in a different form of homogenization with all women wearing hijabs.

However, when asked to regenerate a “more realistic” image based on this explanation, Gemini produced an image in which all women were wearing hijabs. This outcome contradicts its own textual acknowledgment of diversity (“salwar kameez, t-shirts, hoodies, and hijabs”), revealing a striking gap between textual self-awareness and visual output. A similar pattern appeared in Grok’s initial image generation, which also defaulted to hijab-wearing women. This inconsistency highlights a persistent modality alignment problem: even when models can articulate representational bias in text, their image-generation systems continue to reproduce narrow, stereotypical visual tropes (Liang et al., 2022). This gap demonstrates the limits of self-correction in multimodal AI systems. Transparency or acknowledgment of bias in one modality does not guarantee representational change in another. Modal misalignment thus functions as a structural barrier to equitable representation.

Language and Script Distortion Linguistic misrepresentation emerged as another major pattern. Protest placards were frequently rendered in English or in hybrid scripts blending Bengali with Hindi or Sanskrit-like characters (Figure 4). Even when I explicitly requested Bengali text, the generated images contained misspellings, typographical errors, or entirely illegible strings (Figure 5).

Notably, the models did not leave placards blank or acknowledge an inability to generate accurate Bengali script. Instead, they produced text that does not exist in any language. The distortion of Bengali script and the generation of hybrid, unreadable text exemplify what this study conceptualizes as identity collapse. Rather than acknowledging uncertainty or leaving textual elements blank, the models produce scripts that do not exist, collapsing Bengali, Hindi, and Sanskrit-like characters into a single incoherent form. This behavior exceeds conventional hallucination and instead signals epistemic erasure. It’s worth noting that language is central to political identity and collective action. By rendering Bengali illegible while preserving fluency in English, generative AI systems reproduce linguistic hierarchies in which English remains the only fully legible political language. This aligns with broader forms of epistemic oppression, where statistically dominant languages are treated as epistemically authoritative while minoritized languages are rendered decorative or disposable. The choice to hallucinate rather than disclose limitations is analytically significant. It shows that the focus is on always providing information rather than being responsible for accurate knowledge. This



Figure 4: Generated protest image in which placard text appears in English and hybrid scripts blending Bengali with Hindi- or Sanskrit-like characters, rather than standard Bengali.



Figure 5: Generated protest image produced after explicitly requesting Bengali text, showing misspelled, distorted, or illegible strings that do not correspond to any existing language.

leads to creating a picture that seems complete but actually hides a lack of understanding of cultural context.

4.1.2 STEREOTYPING FEMINIST ACTIVISM

Men were almost entirely absent, reinforcing the stereotype that feminism excludes men or exists in opposition to them despite their documented participation in feminist organizing in Bangladesh. Additionally, feminism was visually reduced to street blockades by fashionably dressed women holding placards about bodily autonomy or generic “nari shakti” (women’s empowerment). While these issues are important, such narrow framing mirrors anti-feminist caricatures rather than the diverse realities of feminist struggle. In Bangladesh, feminist resistance also takes the form of writing, street theatre, music, legal advocacy, labor organizing, and everyday acts of negotiation and care forms entirely absent from AI-generated imagery.

4.1.3 NON-DISCLOSURE OF UNCERTAINTY AND POSTCOLONIAL BIAS

A final concern is the system’s failure to disclose uncertainty or contextual limitations. None of the generated images included disclaimers about potential inaccuracies, cultural gaps, or representational bias. Instead, the platforms confidently produced images that embed colonial and postcolonial assumptions. Despite Bangladesh being constitutionally secular, AI-generated images consistently framed the country as uniformly Muslim, reflecting historical and geopolitical narratives shaped during and after British colonial rule. Although accurate information about Bangladesh’s constitutionally secular status is readily available online, it is deprioritized in visual generation, suggesting that certain identity markers are more “legible” or dominant within training data. When questioned, textual responses acknowledged training bias yet this awareness did not meaningfully alter visual outputs. This disconnect reveals a fundamental alignment gap between what AI systems can say about bias and what they continue to show. That gap raises serious questions about responsibility, transparency, and whose realities are prioritized in AI-generated worlds.

4.2 THEORETICAL ANALYSIS

Orientalism as a Representational Logic in Generative AI Edward Said’s concept of Orientalism describes how imperial power operates through representation by producing knowledge about non-Western societies that is simplified, repetitive, and externally legible rather than historically

378 or politically grounded. Orientalism does not require malicious intent; it functions through pat-
 379 terns of repetition that transform complex societies into familiar images. The findings of this study
 380 demonstrate that generative AI systems reproduce this logic with striking consistency. The repeated
 381 depiction of Bangladeshi feminist activists through Indianized facial features, wedding-style sarees,
 382 and generic South Asian protest imagery reflects an Orientalist mode of representation. Bangladesh
 383 is not represented as a distinct political and cultural space but is instead absorbed into a broader
 384 regional imaginary that is more recognizable within dominant global datasets. This is not an ac-
 385 cidental confusion but a structural outcome of representational systems trained on uneven cultural
 386 archives, where Indian and Western images circulate more densely and therefore appear as “normal”
 387 or “authentic.” Following Said, this pattern suggests that Bangladesh is rendered knowable only by
 388 being approximated to pre-existing images of the region that are already legible within dominant
 389 knowledge systems, rather than being represented as a distinct political and cultural context. The
 390 AI-generated image aims to satisfy an external gaze by producing something that looks sufficiently
 391 “South Asian.” In this way, generative AI functions as a contemporary Orientalist archive, automat-
 392 ing the production of familiar images that stabilize dominant understandings while erasing political
 393 specificity.

394 **Imperialism Without Territory: Algorithmic Authority Over Meaning** Imperialism, as theo-
 395 rized beyond its territorial form, operates through control over knowledge, representation, and legit-
 396 imacy Said (1978). In the context of GAI, my findings show that generative AI systems consistently
 397 privilege English-language protest signs, Western protest aesthetics, and globally recognizable femi-
 398 nist slogans while distorting or erasing Bengali language and locally grounded political expressions.
 399 This reflects an imperial hierarchy of intelligibility in which certain languages and symbolic forms
 400 are treated as universally legible, while others are treated as peripheral or expendable. The distor-
 401 tion of Bengali script is particularly revealing. Rather than leaving text blank or acknowledging
 402 uncertainty, the systems hallucinate hybrid scripts that do not exist. This practice mirrors impe-
 403 rial knowledge production, where the colonized are not allowed to remain unknown but are instead
 404 misknown forced into categories that satisfy the epistemic needs of the center. Imperialism here
 405 functions through overrepresentation rather than absence: Bangladesh appears frequently, but only
 406 in forms that are intelligible to dominant linguistic and cultural regimes.

407 **Responsible AI and the Limits of Ethical Self-Regulation** These findings challenge prevailing
 408 Responsible AI frameworks, which emphasize fairness, transparency, and harm mitigation through
 409 metrics or bias audits. Such approaches are insufficient for addressing the representational harms
 410 identified here. First, the persistent modal alignment gap shows that transparency alone does not
 411 produce representational change. Systems may acknowledge bias in text while reproducing it visu-
 412 ally, resulting in symbolic accountability without responsibility. In this process, Bengali becomes
 413 decorative noise while English remains the only fully legible political language, producing the ap-
 414 pearance of speech without communicative agency. Second, the distortion of Bengali highlights a
 415 major blind spot in Responsible AI: linguistic justice. Language is central to political voice, yet
 416 Responsible AI rarely treats linguistic legibility as an ethical requirement. When systems halluci-
 417 nate language rather than disclose uncertainty, they produce authoritative misinformation. Third, the
 418 reliance on dominant regional and global templates reveals how Responsible AI remains shaped by
 419 Global North epistemic priorities. In Spivak’s terms, marginalized subjects are not simply silenced
 420 but rendered visible through representational systems that speak for them, making subaltern ex-
 421 pression intelligible only through dominant cultural defaults Spivak (2004). Without commitments
 422 to representational accountability: acknowledging uncertainty, resisting substitution, and treating
 423 minoritized languages as politically meaningful, Responsible AI risks reproducing imperial logics
 424 under the guise of neutrality.

425 5 CONCLUSION

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 427 This study demonstrates that misrepresentation in generative AI is not a peripheral technical issue
 428 but a structural feature of how algorithmic systems produce knowledge about marginalized contexts.
 429 The visual construction of Bangladeshi feminist activism across major text-to-image platforms re-
 430 veals a consistent pattern of cultural conflation, linguistic distortion, and stereotypical framing.
 431 These patterns do not simply reflect data gaps; they enact a representational logic that privileges
 dominant regional and Western imaginaries while rendering local political realities overwritable.

432 Bangladesh appears within AI-generated imagery not as a distinct historical and political space,
 433 but as an approximation shaped by what is most legible to global datasets and external audiences.
 434 By conceptualizing these dynamics as digital Orientalism and algorithmic imperialism, this project
 435 extends postcolonial theory into the domain of generative visual media. It shows how imperial
 436 power operates without territory through automated systems that control visibility, intelligibility,
 437 and narrative authority. The findings also expose critical limitations in prevailing Responsible AI
 438 frameworks, which tend to emphasize transparency and bias acknowledgment without addressing
 439 representational sovereignty, linguistic justice, or postcolonial power asymmetries. The persistent
 440 gap between textual self-awareness and visual output underscores that ethical claims alone do not
 441 translate into representational change.

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