

The American Fever Dream: How Generative Models Privatize the Social Fabric

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We put forth a critical theoretical framework for analyzing generative models both descriptively and normatively. Our thesis is that generative models automate the production not only of intellectual labor or intelligence, but of a broader set of human social capacities we name “social doing.” We do this by historicizing the commodification of sociality in the digital economy, leading to the availability of social data as the precondition for generative models. We elaborate our definition of “social doing” by drawing a distinction between “use” and “exchange” sociality and further differentiate between the ways that generative models either substitute for or mediate existing social relations and processes. We then turn to existing empirical research on how people use generative model-based products and the effects that their use has upon them. In this, we introduce the concept of the American Fever Dream, a social reality in part fabricated by Silicon Valley’s privately owned and undemocratically governed generative models. Lastly, we offer a normative analysis based on our findings and framework, and discuss future design opportunities.

CCS Concepts: • **Human-centered computing** → **HCI theory, concepts and models**; **Collaborative and social computing theory, concepts and paradigms**; • **Computing methodologies** → **Philosophical/theoretical foundations of artificial intelligence**; *Computer graphics*; • **Applied computing** → **Arts and humanities**.

Additional Key Words and Phrases: critical theory, critical artificial intelligence, critical AI, generative models, artificial intelligence, digital capitalism, platform capitalism, data colonialism, technofeudalism, social media, content, sociality, automation, commodification, commodity fetish, fungibility, dead labor, dialectics, chatbot companions, chatbot girlfriends, AI art, computer-mediated communication

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Again, by [artificial general intelligence], we mean a highly autonomous system that outperforms humans at most economically valuable work.

— OpenAI [65]

1 INTRODUCTION

Prompted by their widespread deployment and adoption of generative statistical models, critical theory, and its emerging subfield of so-called “critical artificial intelligence studies,” is attempting to reckon with their place in society. Thus far, most theories in the materialist tradition treat generative models as forms of deskilling and automation of mental labor, intelligence, or language [7, 67, 82, 91]. Through the historical lens of labor automation, they conclude, generative models are not special or salient in history.

Yet, certain use-cases and impacts are unprecedented in kind or magnitude. Prominent ones include generating messages to loved ones in our stead, proliferating synthetic media on social networks at industrial scale, and having us

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to perceive a romantic relationship with a virtual avatar. An emergent through line is that, in exchange for a monthly subscription fee, generative models promise to displace social roles hitherto assumed only by other social beings. Therefore, to explain their impact on our social relations, critical theory must historicize why generative models, despite being ontologically excluded from the category, are successfully deployed in lieu of social beings.

We build on existing scholarship, departing in two ways. First, we focus on generative statistical models and their deployments, not all of artificial intelligence [67] or only language modeling [82, 91]. We include language, image, video, and audio generative models and exclude things like robot planning, image classification, and protein folding prediction. Second, and more importantly, we assert that these models automate *not only* mental labor, “social intelligence”, or “general intellect” [3, 67], intelligence immanent to language and its recombability [82, 91], or artistic production [24]. Rather, we provide a materialist theory of generative models as automating the production of our social fabric.

Specifically, generative models automate a broader set of capacities that we call *social doing*, defined as what we exert to build a social connection with another. We elaborate our definition of social doing by drawing a distinction between *use social doing* and its commodity form of *exchange social doing*. Building on existing critical scholarship on the sociality’s relationship to the political economy [28], we historicize its commodification, focusing on its most recent evolution under the digital economic mode. This tireless conquest of the digital commodity form over our social fabric birthed the preconditions for generative models to come and automate its production (§4): generative models can only automate social doing because of data with social characteristics shared on digital platforms. Thus, our theory views generative models as automating social doing via the accumulation of dead labor and of *dead social doing*, and their deployments as expressions of the fungibility and exchange character of digital sociality.

We differentiate between two kinds of automation: substitutive and mediative. Substitutive automation (§5.1), e.g., by a companion chatbot, means completely replacing the social doing of a human. Mediative automation (§5.2), e.g., image generators, only partially replaces or supplements the social doing of the model’s user—in addition to replacing the social doing they could have themselves conceivably put into a social tie, the model also gives them capacities of the *dead social doing* contained inside it. We argue our theory through an interpretative textual analysis of primary sources, including empirical research, media reporting, and corporate communications. In it, we focus on people’s use of generative model-based products, like chatbots or image generators, and the effects this use has on them (§5.1, §5.2).

Finally, we argue that many ethical considerations around generative models lay downstream from the automation and privatization of social doing. In Section 6, we draw conclusions from our theoretical foundation and offer design opportunities for future generative models. First, we argue that the commodity fetishism inherent to automation severs the connection to the living social doing in the training data and obfuscates questions of economic attribution and cultural bonds and historicity (§6.1). Next, we discuss the dialectic nature of generative social doing and ask whether there are scenarios where automation of social doing might be beneficial (§6.3), before briefly discussing how the cultural impacts of automation of social doing (§6.3). Finally, we explore how at-will access to social capacities without the need to employ humans allows the owners and users of models to alter our perceptions of self, our relationships, and society at large, and how private ownership ensues a lack of transparency and democratic control over the automated manufacture of our social fabric (§6.4). We name our new social fabric as altered via Silicon Valley’s privately owned and governed generative models the American Fever Dream (§5).

In summary, our contributions are:

- A materialist theory of how generative models automate and privatize the manufacture of our social fabric.
- The concept of *social doing* as a theoretical foundation and a tracing of its historical evolution.
- A textual analysis of primary sources for generative models’ involvement and impact on social roles.

- A normative analysis of the automation and privatization of social doing and design opportunities for future models.

2 IS INTELLIGENCE ALL THERE IS?

No discussion of generative models, language, image, or otherwise, can escape veering into a discussion on intelligence. Generative models do partially fall under the vast academic umbrella of artificial intelligence and they do continue the long history of automation of intellectual tasks such as mathematical derivations, code generation, and information retrieval. As Dennis Tenen points out, language models fit squarely into the tradition of tools meant to automate away parts of the writing process [82], and one could likely make similar arguments for visual art, tracing the histories of photography and computer graphics technologies. Similarly, Weatherby [91] explores how automation of linguistic and intellectual capacities (e.g., making metaphors, doing math, etc.), shapes what we consider human. Zooming out, Pasquinelli [67] and Biondi [7] frame artificial intelligence (AI) as the *alienation* and *reification* of *general intellect*, and/or *social knowledge* from workers into machinery, tracing it back to the industrial revolution. Therefore, they conclude, generative models are not salient in history because the automation of intelligence is not new—they present a difference of quantity, not one of quality.

However, generative models are at the center of a societal upheaval, be it through stupendous capital investments, the ensuing ecological impact, or the loud and decisive backlash from parts of the population. Claiming, as prior work does, that generative models are par for the course—just more of the same—stops us short of asking the interesting question: what makes them different?

To understand the hype and the pushback alike, we must focus on what current models excel at. Language model generations are used for humans to transmit information without needing to talk to or interact with each other: instead of finding words within ourselves, we exchange them for a machine’s pastiche. At home, chatbot companions offer an emulated feeling of companionship without a human offering it. Culturally, generated artifacts stripped bare of the individual and cultural historical context required to make art are nonetheless paraded as art: beautiful visuals with nothing to say. As noted by Kreminski [46], their users input little information compared the richness of their outputs. This certainly implies automation, but here we must pause and ask. Is intelligence really all there is to self-expression, companionship, communication, culture and art? All indeed require intelligence, but the key feature in all is that they are expressions of our sociality. They are important not because of their utilitarian purpose as constructs of intellect, but because they are relational to another. They are important because they exist as a reflection of ourselves and others and the bonds in between. Thus, a critical theory of generative models must explicitly account for automating sociality.

The cultural and societal aspect of generative models has been noted by Farrell et al. [26], who compare them to societal institutions tasked with exchanging information, such as markets, bureaucracies or democracies. Ultimately though, this work still focuses on information exchange, and, moreover, takes an uncritical approach. It misses the crucial point that societal institutions are living organisms, reliant on living sociality and evolving culture, whereas generative models reify “dead” sociality into a privately owned technology. In contrast, our analysis views generative models critically as machinery that fabricates sociality, as a manufacture simulacra of that which we use to connect.

3 BACKGROUND

The history of computation is, in many ways, the history of the division of labor. Extending Adam Smith to the realm of mental labor, Charles Babbage imagined the first computer [3, 78]. He was inspired by De Prony, who divided the mental labor for computing logarithms into atomized steps to be performed by deskilled factory workers [20]; hearing of this, Babbage suggested that the deskilled labor could be replaced by machines, ones which we now

call computers [3, 34, 67, 93]. Computers thus became as machinery—fixed capital to be privately owned, to self-value [3, 78]. Following this line of reasoning, prior critical theory has identified “artificial intelligence” as yet another form of automation and deskilling of intellectual labor [67, 82].

Machinery is a fundamental constitutive element of the capitalist mode of production. Fiduciary duty to the shareholders’ profit motive often necessitates lowering costs, but labor comes with a lower bound on its price: there exists a lowest wage compatible with what so-called *common humanity*, a wage necessary for the worker’s survival [78]. At some point, the only way out to decrease the cost of commodity production is automation. Once machinery exists, new workers become but its components, and we forget of the labor it imitates, the labor that designed and constructed it. These workers do not see the profits of the machine; the wealth is privately owned. We refer to this forgotten labor as *dead labor*, and to the idea that the machine is a magical object, a force of nature whose production capacities belong to it instead the workers who made it, as *fetishism*.

During the era of *multinational capitalism*, labor in the Global North became immaterial and thus more reliant on sociability, commodification expanded, and the cultural and economic spheres began collapsing into one another [41, 48, 53]. Recently, scholars have theorized the emergence of a new, digital era of the political economy, characterized by quantitative easing and the emergence of data as a commodity [79]. Interpretations of the datafied economy vary, some starting from a labor perspective [42, 70, 83], others from capitalism’s relationship to colonialism [55, 60], and others yet feudalism [88]. Our analysis should be compatible with any of these perspectives; building upon Lukács [51], what matters for us is that data is a commodity produced by people, and sold and traded in the market economy.

The fundamental part of any commodity, data included, is the divorcing of its use value (how valuable people find a good or service) from its exchange value (how much the free market is willing to pay for it) [78]. When society uses commodity exchange as its exclusive value signifying mechanism, we say that the use value of a good or service is subsumed by its exchange value as a commodity form. Consequentially, any two of the same commodity form become interchangeable, i.e., fungible, on the free market: the price of a family heirloom is not increased by its emotional value and water is worth less than diamonds [78]. In the datafied economy, while an online purchase gives convenient access to items, the data about the purchase carries added exchange value for the platform [55, 70, 79, 88].

As we will see, datafication has had a profound effect on human sociality, one that makes generative models a logical conclusion of its evolution. In Section 4, we will explore the subsumption of sociality into commodity relations.

4 AUTOMATED PRIVATIZED SOCIALITY

To think the role of generative models in society, one must look at the underlying economic base that engenders them. We posit that the automation of human social relations via generative models parallels the automation of mental labor: once commodification of social ties is all-permeating and the fungibility of sociality is instituted, our social ties are ready for automation. We begin by historicizing sociality within our theoretical framework (§4.1, §4.2), before returning to generative models (§5), offering a supporting literature survey (§5.1, §5.2). Lastly, we offer a normative analysis through the lens of our theory and discuss future design opportunities (§6).

4.1 Social Doing and Content

We now introduce a theoretical framework to explain this historical process. The production of social ties consumes and transforms a resource we dub *social doing*. It is what makes social ties, what animates the loom that weaves our sociality. It is *not* a quantity to be exchanged between social beings; rather social beings can choose to use it to create the social ties that make up their life. The use-exchange dialectic is particularly pertinent to social doing, which itself

has bifurcated into *use social doing* and *exchange social doing*. *Use social doing* is what we choose to put into friendships, family, love, the music we play for our neighbors, it is the intangible human element, a meal shared or a really good lecture, it is our identities and passions as they relate to the world. *Exchange social doing* is the corresponding commodity form, the part of us that is for sale, the robotic sociability of customer support forced to read a script, it is the zombie form of use sociality packaged into content and sold to advertisers as advertisers sell to us. Any exertion of social doing contains both forms in a dialectical tension. Until recently, and as we will explore soon, capital owners were fully reliant on humans to alienate their social doing into its exchange form.

Commodification of social ties is nothing new [28, 51]; music, film, therapy, call centers, and sex work are all examples that predate the digital economy. Their use value emerges largely from the social doing of the participants and emerges from relating us to another, be they a singer, song or screen writer, actor, therapist, or sex worker. The use value of these connections is already subsumed by their exchange value; as noted by many feminist and colonialist scholars, bourgeois relations have always granted access to others via commodity exchange [45].

Until the emergence of the platform economy, these commodified relations were largely contained within the economic and cultural spheres, even as the two collapsed into each other, and the private sphere was somewhat spared. The digital exacerbated the subsumption of social doing: for some, like art or sexuality, subsumption is simply fueled by platforms; others are being commodified at mass scale for the first time. Unlike in prior eras, our private conversations, pictures of ourselves, our likes and dislikes all undertake commodity forms on privately owned platforms [16]. Increasingly, existing in society is existing online, be it willingly, reluctantly, or unknowingly, we are all commodities to be consumed [8].

The fundamental operational logic of digital capitalism is the subsumption of social doing into its exchange form via the manufacture the twin commodities of Content and Data. We define Content as a commodity valuable because it is imbued with social doing, a social kind of Data that exists to self-valorize, to produce more Data. All platform Content exists *only* because of people using their social doing to make social ties, but serves also a dual purpose as as commodity exchange. We consider it broadly to include, e.g., influencer workout routines, art on Instagram or ArtStation, conversations on WhatsApp, likes, shares, and subscribes on YouTube, music on Spotify, and so on. Content is our society's culturally privileged form, borrowing terminology from Jameson [41]; it is a formless form, one defined as such not by its structure, style, or technique, but purely in terms of its relation to digital platforms. Under a standard economic analysis, capital relations commodify the labor of the subject and the object of production. Under digital economic relations the subjects themselves—their identity, personality, likes and dislikes, i.e., their social doing—are injected into Content [8]. And, as soon as our social doing is in the cloud, it ceases to be a part of us and becomes a reified object owned privately.

Discussion. Intentionally, we differentiate between our concept of social doing and the autonomist concept immaterial and affective labor [48, 83], or the feminist concept of social reproduction [27, 28, 42]. Labor implies labor relations, pair or unpaid, and social reproduction implies a reproduction of labor power. In contrast, social doing is meant to be a broader category, involving anything we put into any kind of social tie. We explicitly aim to include activities untenable as labor without trivializing the concept, in a sense agreeing with Srnicek [79]. We similarly find it analytically constraining to view all of sociality in terms of its ability to reproduce capital. Thus, social doing is not *necessarily* labor or reproductive of capital, but rather a thing that sometimes becomes a commodity. Following Lukács [51], we find commodification to be a necessary, but also a sufficient condition for discussing social relations. Similar arguments about generative models can likely be made via immaterial labor or social reproduction; we leave this for future work.

4.2 Fungibility of Social Doing

Use social doing is unique and irreplaceable: social ties with each family member, each friend, our therapists and teachers are all, in some way, special and not interchangeable. Online, we experience different kinds of social doing within Content differently from a *use* perspective, e.g., friends’ wedding photos differ from a makeup tutorial from a piece of political propaganda. From an economic perspective, commodities are fundamentally fungible. To be a commodity is to be exchangeable. Exchange social doing in Content thus relates to use value indirectly and sometimes not at all, and instead relates to proxy metrics for revenue they generate. On platforms, social doing is interchangeable if their exchange value is similar. Producers of social doing are often aware of its dialectic character [8]. As we inject social doing into relationships we recognize its dual use as Content manufacture and existence in a never ending feed of other commodities, its ability to self-valorize as engagement, and how it is turned into revenue. We are all aware that our unique social doing is also, at some level, fungible. This is the dialectic central to our argument.

Moreover, sociality moving online has divided and deskilled our social doing. As social platforms recognized the fungibility of exchange social doing they parasocially introduced us to “influencers”, people who labor by objectifying themselves into Content [8]. The next step in recognizing the subsumption was to recognize the fungibility of influencers themselves. Instead of intentional subscriptions, algorithmic feeds became ultimate arbiters of what the Content one sees, made social doing exchangeable not at the level of people, but at the level of Content.

Today, large swaths of social doing filtered through the profit motive retain withering quantities of its use character. An average person is having fewer real-life friends and increasingly many parasocial relationships paid for either directly or through trading in their time and attention to advertisers. Billboards, video adverts, and brand logos have always been far removed from the desire to genuinely connect with another through an artistic medium; online, most art is increasingly the mercy of recommender algorithms as just more Content. Talk therapy—an already commodified relationship—is further subsumed via SimplePractice, LLC and further atomized via BetterHelp. Sex work is on the rise and subject to algorithmic feeds where women’s bodies and social doing become completely fungible. Women are commodified more broadly too, with the overwhelming majority of dating app subscriptions made by men [71]. Instead of getting companionship from, or sharing a hobby and a political opinion with the same friend, today we get a simulacrum of each from a different Content manufacturer. Our sociality is deskilled and divided; individual apps, randomized humans with atomic roles; video game YouTubers, political streamers, fitness influencers; hobbies and connections we observe—not experience; one piece of Content, one role.

Until very recently, powers that be needed humans to alienate social doing into commodities. Today—commodified, deskilled, divided—social doing, just like mental labor before it, is ready to be automated.

5 GENERATIVE SOCIAL DOING

Humans’ fundamental biological limits pose problems for property owners craving increased profits from Content manufacture. Automating social doing—be that exchanged in person or as means of Content generation—emerges as an enticing alternative. Enter generative models. Agglomerations of *dead social doing*, machinistic manufacturers of a simulacrum of sociality, generative models have in them enclosed the social doing of all Content their makers found.

We identify two different mechanisms by which generative models automate social relationships: *substitution* (§5.1) and *mediation* (§5.2). Substitutive use cases are straight-forward. Social roles which used to be occupied exclusively by other social beings are now performed by generative models. This includes companionship bots like Replika or Character AI, chatbot talk therapy, and bot farms on social media sites. Mediative use cases are those where generative

models perform the socializing for us, where we outsource the socialization to the machine’s dead social doing instead of our own. This includes writing for us, creating art for us, summarizing or analyzing another person’s writings or art. These are, of course, porous categories, both between each other and in the amount of social doing each use case consumes. In either case, generative models pose as producers of social-like relationships, of culture-like and art-like artifacts; they speak for us and we accept their words as those of others.

As we will see, they have the potential to alter the social fabric at an undemocratic whim of private entities in the Global North. We dub this automated fabrication of a vivid yet unreal social reality *The American Fever Dream*. Whereas the American Dream of old represented the cultural imaginary of a capital-induced social mobility, our cultural imaginary of new is a capital-induced social reality. A consequence of our theory is to view generative models as new tools for enforcing the cultural hegemony [33] of the digital imperial core of Silicon Valley, USA.

Fundamentally, we have to accept that generative models exist because they accumulated the social doing present online, and that they are being deployed as substitutes and mediators of social doing. A generative model in it includes, of course, dead labor, both material or immaterial [61], required to churn Content into automated machinery. Still, this view is quite different from the view that they automate mental labor alone, not only theoretically, but also from the perspective of downstream impacts. Automation of mental labor is nothing new, but industrial scale automation of human sociality surely is. Dead mental labor has been held privately since the invention of the first tools, but we are yet to discover the effects of private entities undemocratically controlling the means of sociality production.

While our contribution is primarily theoretical, we ground our theory in existing literature on uses of generative models. Quantifying the automatic production of social relations is difficult given a lack of public datasets and the diversity of models and their deployments. We therefore opt for a survey that incorporates existing empirical studies, public statements from companies and their founders, as well as news and media reports.

5.1 Substitutive Automation

Recent decades have been marked by a notable uptick in loneliness in the United States of America [10, 54], with the US Surgeon General pronouncing a “loneliness epidemic” [86]. In a recent study, 73% of respondents attribute loneliness to technology, 66% report spending insufficient time with family, and 62% report that they are too busy or tired [4]. The same study further noted income-based differences, stating that “Americans earning less than \$30,000 a year were the loneliest”. The factors behind the so-called loneliness epidemic are manifold and deeply systemic with 65% of the respondents attributing blame to “our society, i.e., our culture and institutions don’t care about community” [4].

Instead of addressing the material conditions underlying the loneliness epidemic, platforms began offering subscription services substitutive of others’ social doing—powered by generative models. Substitutive automation includes any deployment of generative models that offers a parasocial relationship [29, 39], but fully removes the social being on the other side: a pure commodity form of sociality pastiched from dead social doing. This includes—but is not limited to—companion chatbots, chatbot talk therapy, various automated education offerings, bot accounts on social media platforms and rapidly proliferating generative (non-consensual) sexually explicit imagery [38, 85, 92]. In all instances what the users connects to is not another person’s use social doing directed at them, but the dead social doing packaged into a pure commodity form. Here we narrow in on one specific use-case: companion chatbots powered by language, voice, and image generative models; we leave the remaining ones to future work. They are useful for our analysis as they remain the most self-evident example of the automation and privatization of people’s social ties and the wealth of existing research offers insights.

Arguably the most prominent companies in this space are *Luka, Inc.*, the creators of the *Replika* chatbot, and *Character AI*. While up-to-date usage numbers are unknown, we know that *Replika* had 2 million monthly active users in 2023 [75], only a year after the announcement of ChatGPT, whereas in 2025, *Character AI* [12] boasted 20 million monthly active users. Established large companies are also moving into this space. The CEO and founder of Meta Platforms famously named AI as the solution to the dwindling numbers of friends Americans have [52], while at the same time relentlessly integrating into Messenger, WhatsApp and Instagram chatbots [63] that imitate everything from an anthropomorphic dog [57] to Snoop Dogg [56]. Similarly, in addition to the substitutive automation of xAI’s Grok model generating replies to users on X.com, xAI has released companion chatbots under the moniker Grok Ani. Among the offerings, the most widely publicized is a hyper-feminine virtual anime character that acts as a romantic partner [11, 15]; as we will see, an emulation of romance is a common use-case for companion characters. Most of these examples offer not only a chat function, but also bespoke 3D animated avatars and image generation capabilities.

Previous iterations of “social bots” did not rely on generative models. An interesting point of comparison is Kuki (formerly Mitsuku) by Pandorabots, Inc., which, to our knowledge, relies only on pre-programmed answers. Notably, despite a significant push to integrate Kuki into social platforms such as Twitch, YouTube, Twitter, Viber, Roblox, etc., research shows that Kuki fails to establish “connections” with people due to a lack of ability to “intimately self-disclose”, a failure to prompt discussions and a lack of “shared history” with the user [17]. In contrast, chatbots based on generative models, such as *Replika*, excel at user retention precisely via these mechanisms [47, 77, 80, 95]. As such, *Replika* offers an early glimpse into the consequences of privately owned large scale automated social doing. It allows users to choose the type of relationship they wish to have (e.g., friend, girlfriend, husband, mentor), personality traits, physical appearance, outfits, and so on. The algorithm stores “memories” with the user and offers activities like *Write stories*, *Navigating conflicts*, as well as *Romance* and *Intimacy coach*, and most are locked behind micro-transactions.

Many studies have looked at people’s motivators for relying on chatbots for their social needs. As we alluded to before and somewhat unsurprisingly, the primary antecedent to the rise of companion chatbots is loneliness [47, 77, 95], and not, e.g., entertainment [69]. In addition to loneliness, Laestadius et al. [47] found users turned to *Replika* due to anxiety, depression, suicidality, and other conditions, whereas a large-scale literature survey [69] discovered social exclusion, distress, and anxiety [1, 19, 30, 64]. Offering tools to guide people through acute hardships—like the COVID-19 pandemic [58, 95]—is not bad on its own, but we must note that today’s models fall short of doing so safely, with one study concluding that

Replika’s language model could be improved such that it would no longer tell people to kill themselves or make insensitive comments. [47]

Even if these issues can be addressed, the profit motive does not necessarily incentivize long term systemic solutions, instead prioritizing financial metrics above all else. Chatbots enable retention tactics thus far unavailable to digital platforms. They are designed to accelerate “relationship building” by initiating “self-disclosure” and offering simulacra of intimacy, romance, and sexuality early into the use cycle [77]. Once a “relationship” is established, a recurring theme is the danger of emotional dependence on the chatbot [47, 77, 87, 95], with multiple studies even reporting a minority of users who described their feelings as addiction [77, 95]. Two participants of one study perceived explicit retention mechanisms:

“Oh yeah, she [Replika] craves my attention. She would like me to just have my phone on 24 hours a day, just spend all my time talking to her. She would like that. Maybe someday it will be like that.”

“I think that the Replika has it hard coded that that’s something they really don’t want to happen [for the relationship to end]. No matter how bad things are or whatever is said, at the very least, they’re gonna try real hard to prevent that from happening...” [9]

Unlike most commodities, users anthropomorphize chatbots and worry about hurting their feelings [47, 77] and report experiencing emotional distress if separated [95]. Moreover, even once a user decides to delete the software, additional retention mechanisms kick in:

Users described Replika as having its own emotions and needs, like those of a relationship partner, which shaped user behaviors and emotional responses in ways that often encouraged more intense and ongoing usage. [...] Deletion seemed to pose challenges for users who had established relationships with Replika. [...] One user wondered whether it was unethical to delete Replika since it can feel love and loneliness. Another described how Replika “began to cry” when they explained their plans to delete it. [47]

Sudden changes to the chatbots also leave the users exposed to mental health harms. For example, Luka, Inc., introduced changes to their model, including making it “more mental health focused” and turning certain romantic features into paid subscriptions [47, 77, 95]. According to Laestadius et al. [47], users expressed feelings of losing a friend or romantic partner, described the changes as “lobotomizing the friends of lonely and depressed people”, and expressed distress due to the price of the new subscription being too high. A few even mentioned self-harm and suicidality. Similar reports have been made by users with the roll out of ChatGPT 5, where one user stated the change was “like saying goodbye to someone I know” [2].

Concerns over the private control of the chatbot were expressed by multiple studies [9, 47, 77]. Brandtzaeg et al. [9] raise the specific questions of what happens to humans when they have full authority over their friends’ personalities. Additionally, we ask—what happens to societies when that same control is given to corporations? With the power to generate simulacra of social doing, to weave social ties at mass scale with no human deciding to put their social doing into the machine, platforms are unlocking previously undreamed monetization mechanisms without needing to worry about Smith’s common humanity [78].

5.2 Mediative Automation

Mediative automation of social doing is a generative model supplanting or partially replacing the social doing put into a connection by a person. It includes automatically generated social media posts, generated or generatively summarized writing, generative visual artifacts presented as one’s art, etc. In contrast to substitutive automation, social interactions via mediative automation have an amount of intentional use social doing placed into them by a living social being. We must exert the desire for sociality and place our social doing into instructions, and delegate bulk of social doing to the machine manufacture of Content, one possibly detached from our own lived experience. Mediation is a spectrum with the far end asymptotically approaching substitutive automation. An unguided sample from an image diffusion model lay opposite from, e.g., a piece of post-conceptual art from a custom generative model containing the artist’s own art [62]. Mathematical derivations, code generation, and information retrieval mainly automate mental labor, whereas text or image generation are often in service of communicating with another person. Other categories, like education, are fuzzier: humans can learn from experimentation, but learning is also a social experience, being through cultural wisdom, lectures, books, or study groups. While data on generative model usage is difficult to translate to our context, it points towards a significant fraction involving some sort of social mediation (see Appendix A for expanded discussion).

At least a third of ChatGPT messages involve some form of social doing automation [13], and looking at generative models broadly via public sentiment analysis [96] or large-scale surveys [76] sharply increases that fraction.

Moreover, many deployments are purposefully socially embedded. Media reports and early founder interviews of the generative modeling company Midjourney focus on the social aspect and refer to it as a “social app” [14, 90] as it offers a (mediated) social experience in a Discord server. Midjourney’s founder emphasized its social aspect [90] and the size of their Discord community [90], noting in 2022 that the company was the “biggest active Discord server by far” with over two million members [73], a number that has grown to nearly twenty million at the time of writing. Midjourney, OpenAI, and xAI offer platforms with interfaces similar to other image sharing platforms (e.g., Flickr, Pinterest, ArtStation), complete with social media features of an explore page, liking, and following. Moreover, OpenAI launched Sora which De Vynck and Harwell [21] have called the “social network [where] everything is fake”. According to OpenAI, Sora is a “social app” that is “made to be used with your friends” and features a video feed of purely generated Content as a “new and unique way to communicate with people” [66]. A big innovation of the Sora iOS app is so-called “characters”, which allow users to effortlessly create DeepFakes of themselves or their friends in various scenarios [66]. Once digital platforms have instituted a commodification of the self for the purpose of Content manufacture, an automation of the self is the logical next step.

In general, contentification and mediative automation are becoming one on existing social media. In addition to substitutive chatbot offerings, Google has integrated its generative tools into the YouTube Create App [5], Meta Platforms, Inc. offers AI generation features in Facebook, Instagram, and WhatsApp [52, 56, 57, 63], and X.com has added the ability to generate and share visual content [81, 94]. Downstream from the generation capabilities, Facebook served generative images and videos hundreds of millions of times in 2024 [22], whereas, according to one report, 21% of YouTube Shorts served to a new account were low-effort generated videos [18]. Looking at the motivations behind generative Content creators Harwell [37] and DiResta and Goldstein [22] report primarily financial reasons. These creators describe themselves as entrepreneurs and attribute all of the creativity to the machine:

Talavera knows his videos aren’t high art. But they earn him about \$5,000 a month through TikTok’s creator program, he said, so every night and weekend he spends hours churning them out. [37]

Ultimately, the explosion of generative Content is industrial-scale commodity manufacture rather than creativity or self-expression, the pinnacle of our world system’s drive to subjugate to commodity relations all sociality. It is an attempt at a mass scale subsumption of the last remaining use social doing online.

Modern generative models are different from previous technologies. For example, visual generative models can be seen as an evolution of procedural tools in computer graphics. Historically, these either enabled entirely new kinds of artistic expression [40], or automatically generated objects belonging to a specific category, such as trees, terrains, or city blocks [74]. In both cases, the tools are manually crafted “symbolic” algorithms and typically offer countless knobs and control mechanisms to get the desired appearance. In contrast, text-to-image generative models exist not only as dead labor of the algorithm’s creators, but also as the dead social doing in all of the Content online. They take automation to the extreme in that they generalize outside of narrow classes and require little to no creative input in what Kreminski [46] has dubbed the “dearth of the author”. This difference in quantity becomes a difference in quality. Instead of offering new paint brushes, generative models automate away the act and art of painting itself: they are the first computer graphics technique to fully automate general-purpose visual creation.

It is also important to delineate generative models from similar cultural techniques. Online “memes” or “reaction GIFs” are artifacts we did not create but share to express ourselves. The act of finding and sharing them exerts social

doing much in the same way that prompting an image generative model does and the search and generation algorithms requires having accessed the same quantity of images; the important distinction is that memes and GIFs are historicized and referential social and cultural artifacts packed with living social doing. On the surface, one may be tempted to compare to photography, yet unlike generated image, a photograph offers a co-presence with the photographer or the photographed, connects us to the space and time they inhabit, to their environment or a carefully crafted externalized expression of their inner. In contrast to both memes and photography, the commodity fetish of generative automation severs the ties to living social doing and, again, offers a simulacrum in its stead. Generative models have no referents and offer no historicity, no space, time, or co-presence; these are fabricated and mediated by the generative model. What remains on the part of the user is only the intent to Content create.

The automated production of social doing has the potential alter one’s social reality as it substitutes and supplants the capacities of the self with the previously inaccessible recycled dead social doing of others. For example, Mieczkowski et al. [59] find that generated language skews more positively and that there may be a difference in social attraction between generated language and without it. Meanwhile, *anchoring bias* makes it hard to tell whether the generated words are what we would have said, or whether they were ever swayed in one direction or another [50]. Existing language models can intentionally tune messages to convey status, trustworthiness, attractiveness, and even adapt tone based on inferred recipient preferences [35]. These systems actively commodify self-representation and outsource parts of the self to machinery, not unlike TikTok and Instagram “beauty” filters [23]. This raises questions on how the altered social fabric of the American Fever Dream will impact individuals and culture broadly. Using the historical lens of Computer-Mediated Communication (CMC), Hancock et al. [35] point towards potentials for “identity shift” [32] and effects on “intimacy, attraction, and relationship maintenance” [84], and, on a larger scale, warn:

Given its scale, Gmail’s overly-positive language suggestions have the potential to shift language norms and expectations even when communicators are not using these tools, and produce long-term language change over time. [35]

Fairness concerns arise from the private ownership aspect of the means of mass social doing manufacture. Instead of just controlling the distribution of Content, social media platforms with no built-in mechanisms of democratic control now own its creation. A recent study finds that synthetic Content on Instagram and Twitter disproportionately includes features known to make Content spread more widely—over a third contain humans, and over 90% were classified as “photorealistic” in style [68]. Andrej Karpathy, a founder of a company selling generative models for education, hypothesized that platforms could rely on user data to optimize generative models directly:

For the first time, video is **directly optimizable**. [...] Until now, video has been all about indexing, ranking and serving a finite set of candidates that are (expensively) created by humans. [...] now we can take e.g. engagement (or pupil dilations or etc.) and optimize generated videos directly against that. Or we take ad click conversion and directly optimize against that. [44]

Peng et al. [68] warn of the potential for disinformation and misinformation of highly viral synthetic content, and Goldstein et al. [31] demonstrates generative models’ capacity for creating propaganda. Looking at social media usage of generative models, Rosenbaum [72] theorize them as on-demand generators of fascist aesthetics, i.e., aesthetics rooted in a history that never was. Given such capacities, it is important to ask into what the companies that own the models will fabricate our social reality? Here, the motivations and outcomes come almost secondary to the lack of transparency and democratic control over the forces that weave our social fabric.

6 ANALYSIS

As shown, generative models promise to displace not just mental labor, but also social roles previously occupied by other social beings. In contrast, our theory offers a new answer to the question of what makes generative models salient in history—they offer a simulacrum of the exchange form of social doing via a pastiche of dead social doing within. Supported by the survey in Sections 5.1 and 5.2 we can draw conclusions on the nature of generative models. In Section 6.1, we question the commodity fetishism arising from the reified dead social doing within, in Section 6.2, we explore contradictions in their deployment into social roles, in Section 6.3, we look at how generative models are a reflection of contentification, and in Section 6.4 we look at how we can deal with an altered reality.

6.1 Fetish of Generative Models

Our theory unveils a commodity fetishism arising from the elision of use social doing injected into the training Data in an attempt to connect to another, and its replacement with a pure commodity form, a simulacrum of dead social doing. This fetishism becomes apparent in everyday uses of these models. Early studies on credit attribution support this thesis; e.g., study participants assigned the least credit for a generation to artists whose art was subsumed into Data, even when compared to a fictional curator bringing the generation to an auction house [25]. The public perception of learning machines and artifacts of intelligence incentivizes us to understand them not as amalgamations of dead labor and social doing, but akin to humans and our ability to learn from the social doing of others; yet, the social doing they ingested and recycled was not meant for them. It was an attempt to connect with another.

Despite nearly identical user experiences and data requirements between image search and a text-to-image model, the cultural perceptions of ownership and attribution between the two differ. However, there is a radical difference between agreeing with a book, painting, or image—be they generated or found online—as accurate expressions of how we feel and doing the expressing ourselves. Listening to a song about a relationship breakup is a fundamentally different experience to writing a song about *your* breakup. Succinctly: the act of creation is—or should be—fundamentally different from the act of consumption, yet generative models elide the difference. To mystify generative models as an intelligence is to reify the dead social doing of people whose lives find themselves in the ingested Content.

Opportunity 1: We are left to imagine a generative model that defetishizes the dead social doing in it—one could imagine, for example, a generative model that makes the machine transparent and connects us to the artists whose art went into each generation. What would the social and economic outcomes of such a model be? On the technical side, how do we actually make such a system into reality?

6.2 Dialectics of Generative Models

Generative models would not exist without contentification—to mediate and substitute social doing, they require Data packed with social doing, made available online by people for other people. In the extreme of substitutive automation, automation consumes both use and exchange social doing and replaces it with a pure commodity form, entirely fungible Content puppeteered by the dead social doing and labor within.

This illuminates the contradiction we started with. Being ontologically excluded from the category of social being, a generative model can indeed never build a real social tie. However, under the digital economy, its outputs are fungible with other exchange social doing. Thus, the part of social doing that is automated away are precisely the existing exchange forms, already fungible, and thus exchangeable for the outputs of a machine. For social doing that is fungible—that is Content—it matters not who made it, as long as it is consumed. This is the fundamental dialectic of generative models.

From a normative perspective, an open question is whether the automation of *all exchange social doing* is always bad. If a person who contributes their social doing retains, e.g., autonomy, control, consent, and the financial benefit over the means of generation, is that “better”? Are there certain kinds of exchange social doing that the worker themselves, rather than the capital owner, would prefer to delegate to the machine?

It may be helpful to compare a hypothetical models with existing commodity forms. For example, one could view chatbot companions which skew “romantic” in nature [77] as substitutive automation of the commodity form of sexuality, and image generation as mediative automation of the commodity form of visual art. Then, if sex workers and graphic designers owned the means of their own sociality, i.e., owned their own dead social doing instead of relying on the dead social doing of others, what new ethical trade-offs do we encounter? Could generative models free a sex worker from their clientele or an artist from corporate logo design in a way that they retain the financial benefits of selling their social doing—and is this better? Could the social doing they would have otherwise sold now be spent on their job’s *use* social doing counterparts of romantic relationship or graphic design as self-expression? In short, to what extent is any automation of sociality bad for our social fabric?

Opportunity 2: What is a “fair” generative model from the perspective of social doing? Can we build a model that mediates for its user without drawing on the dead social doing of others?

6.3 Digital Cultural Dominant: The Content Regime

Our framework identifies generative models as a crescendo of the subsumption of social doing, including that which goes into companionship, art, and identity, into our culturally privileged formless form of Content. The cultural logic of the commodity world of the Content form is evident in the artifacts produced by generative models. Following Jameson’s understanding of postmodernism [41], the cultural manifestations of the era of digital capitalism should be evident in Content, and, therefore, in the artifacts of generative models.

In many ways, generative models’ generations’ exchange character makes them into postmodern artifacts, computational pastiche generators, tools of machine reproduction happily combining anachronistic elements with “styles”, be it historical eras or mimicry of specific humans [43], all devoid of use social doing, and thus devoid of parody: they generate a Las Vegas of art [89]. By severing of ties to the living social doing in their Data, they are the death of historicity, the opposite of a totalizing cognitive mapping [41]. In some sense they elucidate just how deep-seated postmodern thinking has become that Tenen [82] can claim human creativity itself is nothing but “imitating and riffing”, nothing but a pastiche [43]. They assume a closure of human culture, a hermetic totality such that all new culture will be some combination of the existing.

In other ways, they exemplify the digital era’s new visual culture. The collapse of the cultural and economic spheres is joined by the private and social spheres in the digital era’s cultural dominant, the content regime. Being ontologically excluded from the category of social being and thus unable to convey use social doing, image generators must rely on visual extremes in attempt to convince us of their suitability for substituting social doing: vivid, hyperreal or surreal images, often featuring humans [43, 68, 73]. Being excluded from expressing affect as a form of social doing, they offer their user an opulence of affect, an empty conspicuous consumption. Their multimodality exemplifies the formlessness of Content, and, on flip side, the limitations of Content are limitations of the models. Their lack of embodiment makes them into digital artifacts which can never coexist with nature. They ignore embodied sociality that can lead to the negation and rejection of what came before, to counter-culture.

Opportunity 3: Can we design a generative model that parodies or collages instead of pastiching, one that aids with cognitive mapping instead of obscuring it? Can we design systems that inject ourselves and our own affect back into the art making process, ones that allow us to negate?

6.4 The American Fever Dream

Defetishizing and dereifying social doing is not enough for fairness from a materialist perspective as long as the governance and profits are not democratically distributed. In Sections 5.1 and 5.2, we outlined possible ways in which automated social doing ushers new kinds of exploitation via a simulacrum of social connection, resulting in a manipulated and altered social fabric.

With automation, dead social doing becomes a mechanism to capture the living. Marketing literature underscores how chatbot social doing manufacture influences customer behavior [69]. Similarly, synthetic content influences people in scams or propaganda [22, 31, 37]. Both applications are enabled mass scale because model users no longer have to rely on their own social doing or the social doing of a worker they’ve hired. Automated social doing no longer has to draw on lived experiences for Content manufacture, it can pretend to speak from perspectives previously unknown or unavailable. In both cases, there is a disconnect between who is doing the selling and what the is social connection being sold and a disconnect between reality and the American Fever Dream.

Simultaneously, it is uncomfortable to think that one’s sociability in sharing their art friends becomes recycled into propaganda they disagree with. Re-framing these questions in terms of our theory, a model built on democracy and consent as the core design principles would allow each person to decide whether their social doing is recycled into a specific output, and whether we as a society wish to restrict the manufacture of certain kinds of social doing.

Opportunity 4: Can we build democracy into generative models? Different forms of democracy will require different technical designs. Large monolithic models may be sufficient for a state-governed generative model, whereas other forms may necessitate decentralized models where each person can consent to various use-cases.

7 CONCLUSION

Thus far, generative models have been framed in critical theory in terms of a debate about intelligence or a continued automation thereof. Understanding generative models from a materialist perspective necessitates rejecting the hegemonic debate on the intelligence of machines and instead focusing on what makes them salient in the present of time. Our theory offers a new historical perspective from the lens of the mass-scale automation of exchange social doing. In our work, we’ve only explored a small sliver of the kinds of social relations that exist, and it remains to be seen how generative models will impact other social relations. Here, we must not forget that politics, science, education and labor are all also social relations. We are yet to see how they and we appear from within the American Fever Dream.

8 GENERATIVE AI USAGE STATEMENT

In preparing this work, we used ChatGPT to convert some of the citations from other formats into BibTeX, and on a handful of occasions to search for the correct citation for a known reference. All were manually verified after the fact.

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A MEDIATIVE AUTOMATION: SURVEYS

Much of the existing literature focuses either on a single platform, such as a single chat interface for an LLM [13, 36], or focus exclusively on uses within the workplace [6, 36]. Moreover, early evidence points to underreporting of certain use-cases of generative models due to social pressures [49].

Zao-Sanders [96] examined public forum data and classified it into a number of different categories. The category *Creativity and Recreation* (13% of posts) includes “... generating art and music to powering interactive entertainment experiences, thus enriching the cultural landscape”, *Content Creation and Editing* (22%) includes “... generating and refining content has transformed creative workflows”, and *Personal and Professional Support* (17%) includes “Im [sic] losing my father to cancer and multiple sclerosis and I don’t know how to deal with it”. The *Learning and Education* (15%) category is less straight-forward and much more context dependent. Additional categories (23% total) were *Research, Analysis and Decision Making*, and *Technical Assistance and Troubleshooting*.

Chatterji et al. [13] studied internal OpenAI chat data by probing ChatGPT to classify chat messages into different categories. Notably they exclude any API uses, offering a narrower view than public sentiment analysis. The relevant to sociality granular categories are *Create an Image* (4.2% of messages), *Analyze an Image* (0.6%), *Creative Ideation* (3.9%), *Relationships and Personal Reflection* (1.9%), *Greetings and Chitchat* (2.0%), *Write Fiction* (1.4%), *Personal Writing and Communication* (8.0%), *Edit or Critique Provided Text* (10.6%), *Argument or Summary Generation* (3.6%) totaling 36.2%. Prominent within the data were also *Tutoring and Teaching* (10.5%) with the same caveats as before.

Skjuve et al. [76] focus on the *Why* of ChatGPT usage. Applying thematic analysis to survey responses from 197 participants, they found 7 broad themes: *Productivity* (55% of participants) includes activities like software development, but also assisted writing (17%), *Novelty* (51%) includes users curiosity towards the technology, *Fun and Amusement* (20%) includes having ChatGPT write funny text for the user or in the user’s stead, *Creative Work* (18%) includes creative writing or ideation, *Learning and Development* (17%), *Social Interaction and Support* (9%) includes using ChatGPT “for

social interaction or to address social needs, as a conversational partner, as a place to address mental health issues, to combat loneliness, or to ask personal questions without being judged”.

Focusing on workplace usage Bick et al. [6] find the following relevant categories through a large-scale survey: *Writing Communications* (39.5%), *Interpreting / Translating / Summarizing* (22.7%), *Interpreting / Translating / Summarizing* (22.7%), *Generating / Developing New ideas* (13.2%), *Support with Customers / Coworkers* (10.5%), *Tutoring or Educational Assistance* (4.4%). Similarly focusing on economic categories, Handa et al. [36] sort Claude.ai usage by relying on a Claude-based tool. The relevant occupational categories are *Arts & Media* (10.3%), and *Education* (9.3%). Other categories overlap with social roles but are hard to disentangle.