

# Synthetic Sociality: 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 Synthetic Sociality, 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**; • **Computing methodologies** → **Philosophical/theoretical foundations of artificial intelligence**; *Computer graphics*.

Additional Key Words and Phrases: critical theory, critical AI, artificial intelligence, platform capitalism, data colonialism, technofeudalism, social media content, sociality, commodification, dead labor, chatbot companions, AI art, computer-mediated communication

*Again, by [artificial general intelligence], we mean a highly autonomous system that outperforms humans at most economically valuable work. — OpenAI [74]*

## 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 [8, 76, 93, 102]. 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 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 impacts, 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 [76] or only language modeling [93, 102]. 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” [4, 76], intelligence immanent to language and its recombining [93, 102], or artistic production [27]. 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

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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 [33], 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). Silicon Valley's privately owned and governed models alter our social fabric into a partially generated hollow imitation—a *simulacrum*—which we dub Synthetic Sociality (§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 Tenen [93] points out, language models fit squarely into the tradition of tools meant to automate away parts of the writing process, and one could likely make similar arguments for visual art, tracing the histories of photography and computer graphics technologies. Similarly, Weatherby [102] explores how automation of linguistic and intellectual capacities (e.g., making metaphors, doing math), shapes what we consider human. Zooming out, Pasquinelli [76] and Biondi [8] frame artificial intelligence (AI) as *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 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 [51], 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. [29], 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 [4, 89]. He was inspired by De Prony, who divided the mental labor for computing logarithms into atomized steps to be performed by deskilled factory workers [22]; hearing of this, Babbage suggested that the deskilled labor could be replaced by machines, ones which we now call computers [4, 38, 76, 104]. Computers thus became as machinery—fixed capital to be privately owned, to self-value [4, 89]. Following this line of reasoning, prior critical theory has identified “artificial intelligence” as yet another form of automation and deskilling of intellectual labor [76, 93].

Machinery is a fundamental constitutive element of the capitalist mode of production. Fiduciary duty to the profit motive often requires lowering costs, but labor comes with a lower bound on its price: the lowest wage compatible with the so-called *common humanity*, one necessary for the worker’s survival [89]. At some point, further decreasing production costs necessitates automation [57]. 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 of a machine as a magical object, a force of nature whose 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 [45,

53, 59]. 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 [90]. Interpretations of the datafied economy vary, some starting from a labor perspective [46, 79, 94], others from capitalism’s relationship to colonialism [63, 68], and others yet feudalism [99]. Our analysis should be compatible with any of these perspectives; building upon Lukács [56], 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) [89]. 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 [89]. 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 [63, 79, 90, 99].

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 introducing theoretical tools for discussing the commodification of social ties (§4.1) and apply it to a historical account (§4.2), before returning to generative models (§5) as supported via our literature survey (§5.1, §5.2). Lastly, we offer a normative analysis via our theory and discuss future design opportunities (§6).

##### 4.1 Social Doing and Content

We now introduce a theoretical framework and establish a common language to explain this historical process. We use *social relation* in the sociological sense as the concept referring to any connection between people—we use *social tie* as a synonym. Social ties weave together into the *social fabric*, the grand sum of all social relations. 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 social fabric. 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. Some scholars would assert all social doing is labor power [30, 31, 53, 94]. We intentionally steer clear from this form of economic reductionism, viewing social doing as a broader category. Instead, following Lukács [56] we find commodification a necessary and sufficient condition for discussing social relations. Therefore, social doing is not *necessarily* labor power, rather a resource which sometimes becomes a commodity. We discuss this choice in the context of its alternatives [30, 31, 33, 46], below.

Social doing is inseparable from the economy: not only is labor power a form of social doing, but Fraser [33] teaches us capital transforms all relations, regardless of proximity to capital. A key influence is the commodification of social doing. Thus, 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 [33, 56]; music, film, therapy, call centers, and sex work are just some of the 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 or screen writer, therapist, or sex worker. The use value of these connections is already subsumed by their exchange value. The digital offered new avenues for social doing subsumption. Unlike in prior eras, our private conversations and pictures, our likes and dislikes all undertake a commodity form on privately owned platforms [17]. Increasingly, existing in society is existing online; willingly, reluctantly, or unknowingly, we become commodities to be consumed [9]. We offer a deeper historical account in §4.2.

The fundamental operational logic of digital capitalism, its technological extractive apparatus, is the subsumption of social doing into its exchange form via the manufacture the twin commodities of Content and Data. By capital “D” Data, we mean data that is produced and commodified via digital platforms. Manually timing your heart rate while running measures data; a smart watch company doing the same and commodifying it creates Data. We use “Content” to refer to a subset of Data that was made via social doing: heart rate Data *typically* contains no social doing, but heart rate Data shared on social media does and so becomes Content. Content is 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 [45]; 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 [9], likes and dislikes, i.e., their social doing—are transformed into abstract Content. Alienated, our social doing is an object in the cloud, owned privately.

*Discussion.* Intentionally, we differentiate between our concept of social doing and the concept immaterial labor, i.e., labor in manipulating information or affect [53], or of social reproduction [31, 33] from the Wages for Housework movement [30, 31], i.e., of labor that indirectly contributes to capital by reproducing it. These were extended to the digital sphere to argue social acts on platforms are unpaid labor [46, 79, 94]. First, per Srnicek [90], “[d]ata are not immaterial, as any glance at the energy consumption of data centres will quickly prove,” a resonant statement in face of the ecological impact of generative models. Second, these theories ask us to concede that *every* social act on platforms presents a form of labor. However, labor implies labor relations, material or immaterial, paid or unpaid; Srnicek [90]:

All social interaction becomes free labour for capitalism, and we begin to worry that there is no outside to capitalism. Work becomes inseparable from non-work and precise categories become blunt banalities. [...] [L]abour has a very particular meaning: it is an activity that generates a surplus value within a context of markets for labour and a production process oriented towards exchange. [...] Beyond the intuitive hesitation to think that messaging friends is labour, any idea of socially necessary labour time—the implicit standard against which production processes are set—is lacking. [90]

While our framing aligns closely with social reproduction, we simply steer clear from claiming, as Jarrett [46] does, that a child expressing love to a parent or that sending vacation photos to friends—acts that platforms *do commodify*—count

as labor comparable to housework or factory work. Third, we do not find it analytically useful to view *all* social acts on and off platforms in terms of their ability to reproduce capital, such may they sometimes be. This does not limit our analysis: we still rely on historical accounts of social doing as told by these scholars (§4.2). Moreover, labor power remains a privileged form of social doing as it shapes all social relations, and thus organized labor retains power to reshape the commodification of sociality (§6). Lastly, these lenses can be imposed on top of our analysis by designating all social doing as labor power. This leads to different conclusions and we are excited for future work in this direction.

#### 4.2 Commodification and Fungibility of Social Doing

The account of automation of MacKenzie [57] claims that “preceding organizational changes created the ‘social space,’ as it were, for the machine; and that the limitations of those changes created the necessity for it.” Which organizational changes, here of the social fabric itself, made space for generative models to automate social doing production? We argue that our social fabric as organized through private platform infrastructure implies a commodified and fungible social doing. Once a part of the valorization process, automation becomes tempting [57].

From an economic perspective, commodities are fundamentally fungible. To be a commodity is to be exchangeable. Feminist and colonial scholars have long told that commodity relations grant access to others [31, 50]. Archetypal form of commodity social doing is labor power. Workers must engage with society by selling labor power, meaning (abstract) labor power is fungible. While reproduction enters production indirectly [30, 31, 33], it nonetheless leads to a marketplace fungibility of women (e.g., surrogacy, sex work). Adorno [1] teaches how commodity fetishism makes art fungible, while billboards, video adverts, and brand logos have always been distant from a genuine connection through an artistic medium. While capital shapes all ties [33], most remained peripheral to production itself [17, 18, 90]; recently, platforms emerged with new capture mechanisms in economic, private, and cultural spheres. Couldry and Mejias [18]:

Digital platforms are the technological means that produce a new type of “social” for capital [...] Platforms are a key means whereby the general domain of everyday life, much of it until now outside the formal scope of economic relations, can be caught within the net of marketization. [18]

The extractive apparatus of platforms datafies social doing into Content. For some social doing, like artistry or sexuality, avenues of subsumption change; others enter valorization directly for the first time. As platforms recognized the fungibility of exchange social doing they exchanged friends for “influencers”, laborers objectifying themselves into Content [9]. Algorithmic feeds became ultimate arbiters of what Content one sees, making social doing exchangeable not at the level of people, but at the level of Content. An average person has increasingly many paid parasocial relationships, either directly or by trading in their time to advertisers. Talk therapy, already commodified, is further subsumed via SimplePractice and BetterHelp. Most art online is at the mercy of recommender algorithms as just more Content. Sex work is on the rise and subject to algorithmic feeds of Content. Women are commodified more broadly too, with the most dating app subscriptions paid by men [80]. Instead of getting companionship from, or sharing a hobby or discussing politics with one friend, we get each from a different supplier of abstract Content.

While platforms give Content its *Data character* and provide a mechanism of commodification, they are not the sole progenitor of its *commodity character*. The presence of commodity social doing online is not determined by platform infrastructure alone—it provides a mechanism and a marketplace—but by broader social processes. Some examples follow. Workers have little choice whether their social doing is alienated into Content on automated management platforms. Ideological framing of care and art as “labor of love” reinforces their unpaid extraction, including on platforms [26, 71]. Commodification of reproduction is influenced by patriarchal, legal [31, 50], and macroeconomic conditions [33],

empowering companies to cheaply accumulate reproductive social doing. Generative models themselves further devalue social doing and put producers in precarity, forming an extractive reinforcement loop. Listing all of antecedents is out of scope. Rather, these illustrate how accumulated social doing embodies a broad spectrum of relations beyond capital.

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. Exchange social doing in Content meanwhile 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 [9]. 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.

Our social ties are turning fungible; individual apps, randomized humans with atomic roles; political streamers, virtual care workers, fitness gurus; hobbies and connections we observe—not experience; each a piece of abstract Content that begets more Content, that self-valorizes. Until recently, owners needed humans to alienate social doing into commodities. Today—commodified, fungible, and valorizing—social doing 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 social ties, generative models have in them enclosed the social doing of all Content their makers found.

We identify two different modes in 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.

Fundamentally, social doing enters generative models via its presence in the Content used for training. The models are then deployed to substitute and mediate social doing. A generative model in it includes, of course, dead labor [69] required to churn Content into 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. This may not present novelty in mental labor automation, but it does present a large quantitative change in capacity for industrial scale automated social doing manufacture. So far, algorithmic feeds chose which social ties come to be, but could only manipulate social doing manufacture indirectly via platform incentives. Video games and “beauty” filters mediate social doing, but are limited, either to the game world, or in the magnitude of possible change. We discuss other historical precedents of technologies that substitute for or mediate social doing, such as procedural chatbots, “beauty” filters, internet “memes”,

procedural computer graphics in sections that follow—we find that models significantly transform these technologies and corporate power with it. Whether this quantitative change will turn qualitative remains to be seen.

As we will see, generative models can alter the social fabric at an undemocratic whim of private entities. Owners have unilateral power over automated social doing, forcefully expanding on their existing capacities for social fabric manipulation, such as via algorithmic feeds, video games, procedural chatbots, or “glamour” filters. Generative models can be seen as a natural evolution, manufacturing social doing at an industrial scale, blending it with reality, without the need for human creators. This manufactured simulacrum of a social fabric deserves a name as a provocation meant to foreground and coalesce around a key concern raised by our framework. We dub it *Synthetic Sociality*. Much like the visual artifact themselves [48, 77, 82], it is a surreal social fabric, a blend of the real with the unreal where the two become one.

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 [11, 62], with the US Surgeon General pronouncing a “loneliness epidemic” [97]. 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 [5]. 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” [5]. From a materialist lens, the “epidemic” can be understood via capital’s social-reproductive “crisis tendency”—see Fraser [33].

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 [34, 43], but fully removes the social being on the other side: a pure commodity form 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 generative (non-consensual) sexually explicit imagery [42, 96, 103]. 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 other uses 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 [84], only a year after the announcement of ChatGPT, whereas in 2025, *Character AI* [13] 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 [58], while at the same time relentlessly integrating into Messenger, WhatsApp and Instagram chatbots [72] that imitate everything from an anthropomorphic dog [65] to Snoop Dogg [64]. 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 [12, 16]; 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 [19]. In contrast, chatbots based on generative models, such as Replika, excel at user retention precisely via these mechanisms [52, 88, 91, 107]. 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 motivations for relying on chatbots for social needs. As we alluded to and somewhat unsurprisingly, the primary antecedent to the rise of companion chatbots is loneliness [52, 88, 107], and not, e.g., entertainment [78]. In addition, Laestadius et al. [52] found users turned to Replika due to anxiety, depression, suicidality, and other conditions, whereas a large-scale literature survey [78] discovered social exclusion, distress, and anxiety [2, 21, 35, 73]. Offering tools to guide people through acute hardships—like the COVID-19 pandemic [66, 107]—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. [52]

Even if these issues can be addressed, the profit motive does not necessarily incentivize long term systemic solutions, instead prioritizing financial metrics. 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 [88]. Once a “relationship” is established, a recurring theme is the danger of emotional dependence on the chatbot [52, 88, 98, 107], with multiple studies even reporting a minority of users who described their feelings as addiction [88, 107]. 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...” [10]

Unlike most commodities, users anthropomorphize chatbots and worry about hurting their feelings [52, 88] and report experiencing distress if separated [107]. Once user decides to delete the app, 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. [52]

Sudden changes to the chatbots also leave the users exposed to mental health harms. Luka, Inc., introduced changes to their model, including making it “more mental health focused” and turning certain romantic features into paid subscriptions [52, 88, 107]. According to Laestadius et al. [52], 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” [3]. Retention tactics remain active research, albeit under different names [32, 86, 87].

Concerns over the private control of the chatbot were expressed by multiple studies [10, 52, 88]. Brandtzaeg et al. [10] 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 a synthetic sociality 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 [89].

## 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 [70]. 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 [14], and looking at generative models broadly via public sentiment analysis [108] or large-scale surveys [85] sharply increases that fraction.

Moreover, many deployments are purposefully socially embedded. Media reports on the company Midjourney refer to it as a “social app” [15, 101] as it offers a (mediated) social experience in a Discord server. Midjourney’s founder emphasized its social aspect [101] and noted that theirs was the “biggest active Discord server by far” [82], with nearly 20 million users at time of writing. Midjourney, OpenAI, and xAI offer platforms with interfaces similar to other image sharing platforms (e.g., Flickr, ArtStation), with social media features of a feed, likes and follows. Moreover, OpenAI launched Sora, a “social app” with a video feed of purely generated Content [23, 75]. A big innovation were so-called “characters”, which allow users to effortlessly create DeepFakes of themselves or their friends [75]. After a commodification of the self for 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 [6], Meta Platforms, Inc. offers AI generation features in Facebook, Instagram, and WhatsApp [58, 64, 65, 72], and X.com has

added the ability to generate and share visual content [92, 106]. Downstream from the generation capabilities, Facebook served generative images and videos hundreds of millions of times in 2024 [24], whereas, according to one report, 21% of YouTube Shorts served to a new account were low-effort generated videos [20]. Looking at the motivations behind generative Content creators Harwell [41] and DiResta and Goldstein [24] 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. [41]

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 social ties.

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 [44], or generated objects belonging to a specific category, such as trees, terrains, or city blocks [83]. In both cases, the tools are manually crafted "symbolic" algorithms and typically offer countless 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; see "dearth of the author" of Kreminski [51]. This difference in quantity could become a difference in quality. Instead of offering new paint brushes, generative models automate away the act and art of painting itself: they 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.

Automated social doing can alter one's social doing as it substitutes and supplants capacities of the self with the recycled dead social doing of others—synthetic sociality morphs what we do, not just what we see. For example, Mieczkowski et al. [67] 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 [55]. Existing language models can intentionally tune messages to convey status, trustworthiness, attractiveness, and even adapt tone based on inferred recipient preferences [39]. These systems actively commodify self-representation and outsource parts of the self to machinery, not unlike TikTok and Instagram "beauty" filters [25]. This raises questions on how the altered synthetic sociality will impact individuals and culture broadly. Computer-Mediated Communication (CMC) scholars Hancock et al. [39] call for more research and point towards potentials for "identity shift" [37] and effects on "intimacy, attraction, and relationship maintenance" [95], and, on a larger scale, warn:

Gmail’s overly-positive language suggestions have the potential to shift language norms [...] even when communicators are not using these tools, and produce long-term language change over time. [39]

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 [77]. 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. [49]

Peng et al. [77] warn of the potential for disinformation and misinformation of highly viral synthetic content, and Goldstein et al. [36] demonstrates generative models’ capacity for creating propaganda. Looking at social media usage of generative models, Rosenbaum [81] 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 synthetic sociality.

## 6 ANALYSIS

As shown, generative models promise to displace not just mental labor, but also social roles previously occupied by other social beings. 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.

Throughout, we interweave technical design opportunities for future models. In doing so, our point is not that we can “fix” generative models via better design or that wide-scale deployment of said design is likely under the existing mode of production. As in §4.1 and §5, we call for organized labor as the mechanism for resisting synthetic sociality and addressing counter-incentives (see, e.g., May et al. [61] for an analysis of recent screen actor protests). Instead, in including opportunities, we aim to emphasize that modern generative models embody the existing political economy via implicit design principles [57, 105]; e.g., models without artist attribution further profits and empower their owners [57]. Even under a better mode of production, generative models would still propagate the injustices of our current one. Therefore, we aim to explicitize new goals and identify “contingency” designs [57], and ask how we can “[effect] a change of form in the materials of nature” [57, 60]. We emphasize that these goals cannot be isolated from broader socio-political change, and emphasize their speculative, preliminary, and provocative nature.

### 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 [28]. 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 model that defetishizes the dead social doing in it—one could imagine, e.g., a model that makes the machine transparent and connects us to the artists whose art went into each generation. What are the socioeconomic outcomes of such a model? Technically, how do we make such a system real?

### 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 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 [88] 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 [45], the cultural manifestations of the era of digital capitalism should be evident in Content, and, therefore, in the artifacts of generative models.

In some ways, the models' outputs' exchange character makes them into postmodern artifacts, pastiche generators, tools of machine reproduction happily combining anachronistic elements with “styles”, be it historical eras or mimicry of specific humans [48], all devoid of use social doing, and thus of parody: they generate a Las Vegas of art [100]. 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 [45]. In some sense they elucidate just how deep-seated postmodern thinking has become that Tenen [93] can claim human creativity itself is nothing but “imitating and riffing”, nothing but a pastiche [48]. 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 social in the digital era's cultural dominant, the content regime. Being excluded from the category of social being and thus unable to convey use social doing (cf. Jiang et al. [47]), image generators must rely on visual extremes to convince us of their suitability as substitutes: vivid, hyperreal or surreal images, often featuring humans [48, 77, 82]. Being excluded from expressing affect as a form of social doing, they offer their user an empty opulence of affect. Their multimodality exemplifies the formlessness of Content, and, on flip side, the limitations of Content are limitations of the models. Their virtualness makes them into artifacts which can neither coexist with nature nor with an embodied sociality, meaning they cannot negate or reject what was before, nor create 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 Synthetic Sociality

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 [78]. Similarly, synthetic content influences people in scams or propaganda [24, 36, 41]. 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 a social reality and a synthetic sociality.

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 intelligence or a continued automation thereof. Understanding generative models from a materialist perspective necessitates rejecting this hegemonic debate. Our theory has offered a new historical perspective from the lens of mass-scale automation of exchange social doing.

*Future work.* We have only explored a small sliver of the kinds of social relations that exist, and it remains to be seen how generative models will impact others, such as politics, science, and education . Labor impacts could be explored further, e.g., by placing ethnographic findings of Muldoon et al. [69] within our framework, or via a macroeconomic lens. As with machinery of prior, synthetic sociality will undoubtedly extend to and distort the social relation of labor organizing itself [57], but will do so to capital relations too. Our analysis and the synthetic sociality applies to a narrow geographical context. As in §5, it would be a mistake to assume it applies globally: hypothetically, a stronger state may mean control over the simulacrum is no longer privately held. Future work should explore its compatibility with different technological, cultural, or regulatory contexts. Due to the novelty of generative models, further empirical research is needed. Specifically, we join CMC scholars Hancock et al. [39] in calls for more research on mediative automation. Our framework would have to be expanded to allow drawing conclusions on other avenues of injustice due to generative models, such as ecological impacts. Lastly, future work always includes that of praxis to change the relations of the mode of production that birthed—and continues to reinforce—generative models.

Finally, we must remember that *use* social doing matters in ways that are inescapable, e.g., love means use love, a hollow imitation will never suffice. Generative models' forte will always be in automating the already most commodified forms of social doing. A potential avenue for contestation may lay in knowing that they can never automate use social doing—love, art, and society broadly cannot exist without humans connecting to one another. Nonetheless, it remains to be seen how they and we appear from within our synthetic sociality.

## REFERENCES

- [1] Theodor W. Adorno. 1938. Über den Fetischcharakter in der Musik und die Regression des Hörens. *Zeitschrift für Sozialforschung* 7 (1938), 29–60.
- [2] Fayaz Ali, Qingyu Zhang, Muhammad Zubair Tauni, and Khuram Shahzad. 2024. Social Chatbot: My Friend in My Distress. *International Journal of Human-Computer Interaction* 40, 7 (2024), 1702–1712. <https://doi.org/10.1080/10447318.2022.2150745> arXiv:<https://doi.org/10.1080/10447318.2022.2150745>
- [3] Dani Anguiano. 2025. AI lovers grieve loss of ChatGPT's old model: 'Like saying goodbye to someone I know'. *The Guardian* (Aug. 2025). <https://www.theguardian.com/technology/2025/aug/22/ai-chatgpt-new-model-grief>
- [4] Charles Babbage. 1832. *On the Division of Mental Labour*. Charles Knight. <https://books.google.com/books?id=9EYa0QEACAAJ>
- [5] Milena Batanova, Richard Weissbourd, and Joseph McIntyre. 2025. <https://mcc.gse.harvard.edu/reports/loneliness-in-america-2024>

- [6] Dina Berrada. 2025. Unpacking the magic of our new creative tools. <https://blog.youtube/news-and-events/generative-ai-creation-tools-made-on-youtube-2025/>
- [7] Alexander Bick, Adam Blandin, and David J. Deming. 2024. The Rapid Adoption of Generative AI. 32966 (2024). <https://doi.org/10.3386/w32966> DOI: 10.3386/w32966.
- [8] Zachary Biondi. 2023. The Specter of Automation. *Philosophia* 51, 3 (July 2023), 1093–1110. <https://doi.org/10.1007/s11406-022-00604-x>
- [9] Grant Bollmer and Katherine Guinness. 2024. *The Influencer Factory: A Marxist Theory of Corporate Personhood on YouTube*. Stanford University Press.
- [10] Petter Bae Brandtzaeg, Marita Skjuve, and Asbjørn Følstad. 2022. My AI Friend: How Users of a Social Chatbot Understand Their Human–AI Friendship. *Human Communication Research* 48, 3 (2022), 404–429. <https://doi.org/10.1093/hcr/hqac008>
- [11] John T Cacioppo and Stephanie Cacioppo. 2018. The growing problem of loneliness. *Lancet (London, England)* 391, 10119 (Feb. 2018), 426. [https://doi.org/10.1016/S0140-6736\(18\)30142-9](https://doi.org/10.1016/S0140-6736(18)30142-9)
- [12] Henry Chandonnet. 2025. I used Grok’s AI companions for a week. The foul-mouthed red panda is hilarious – the flirty anime girl is worrying. <https://www.businessinsider.com/grok-bad-rudi-ani-levels-ai-companion-xai-elon-musk-2025-7>
- [13] Character AI. 2025. *Harnessing Data at Scale: Character.AI’s Transition to WarpStream*. <https://blog.character.ai/harnessing-data-at-scale-character-ais-transition-to-warpstream/>
- [14] Aaron Chatterji, Thomas Cunningham, David J. Deming, Zoe Hitzig, Christopher Ong, Carl Yan Shan, and Kevin Wadman. 2025. *How People Use ChatGPT*. Technical Report w34255. National Bureau of Economic Research. <https://doi.org/10.3386/w34255>
- [15] Thomas Claburn. 2022. Holz, founder of AI art service Midjourney, on future images. [https://www.theregister.com/2022/08/01/david\\_holz\\_midjourney/](https://www.theregister.com/2022/08/01/david_holz_midjourney/)
- [16] Ann-Marie Corvin. 2025. <https://cybernews.com/ai-news/project-skippy-turn-employee-data-anime-girlfriend/>
- [17] Nick Couldry and Ulises A. Mejias. 2019. *The Costs Of Connection: How Data Is Colonizing Human Life and Appropriating It for Capitalism*. Bloomsbury Publishing.
- [18] Nick Couldry and Ulises A. Mejias. 2019. Data Colonialism: Rethinking Big Data’s Relation to the Contemporary Subject. *Television & New Media* 20, 4 (2019), 336–349. <https://doi.org/10.1177/1527476418796632> arXiv:<https://doi.org/10.1177/1527476418796632>
- [19] Emmelyn A. J. Croes and Marjolijn L. Antheunis. 2021. Can we be friends with Mitsuku? A longitudinal study on the process of relationship formation between humans and a social chatbot. *Journal of Social and Personal Relationships* 38, 1 (2021), 279–300. <https://doi.org/10.1177/0265407520959463> arXiv:<https://doi.org/10.1177/0265407520959463>
- [20] Liam Curtis. 2025. AI Slop Report: The Global Rise of Low-Quality AI Videos. <https://www.kapwing.com/blog/ai-slop-report-the-global-rise-of-low-quality-ai-videos/>
- [21] Mauro De Gennaro, Eva G Krumhuber, and Gale Lucas. 2020. Effectiveness of an empathic chatbot in combating adverse effects of social exclusion on mood. *Frontiers in psychology* 10 (2020), 495952.
- [22] Gaspard de Prony. 1824. *Notice sur les grandes tables logarithmiques et trigonometriques: adaptées au nouveau système métrique décimal*. Didot.
- [23] Gerrit De Vynck and Drew Harwell. 2025. On Sora, Silicon Valley’s hottest new social network, everything is fake. <https://www.proquest.com/newspapers/on-sora-silicon-valleys-hottest-new-social/docview/3256585135/se-2?accountid=12492>
- [24] Renée DiResta and Josh A. Goldstein. 2024. How spammers and scammers leverage AI-generated images on Facebook for audience growth. <https://doi.org/10.37016/mr-2020-151>
- [25] Miriam Doh, Corinna Canali, and Nuria Oliver. 2025. What TikTok Claims, What Bold Glamour Does: A Filter’s Paradox. In *Proceedings of the 2025 ACM Conference on Fairness, Accountability, and Transparency (FAccT ’25)*. Association for Computing Machinery, New York, NY, USA, 1902–1915. <https://doi.org/10.1145/3715275.3732126>
- [26] Brooke Erin Duffy. 2017. *(Not) Getting Paid to Do What You Love: Gender, Social Media, and Aspirational Work*. Yale University Press, New Haven, CT.
- [27] Ziv Epstein, Aaron Hertzmann, the Investigators of Human Creativity, Memo Akten, Hany Farid, Jessica Fjeld, Morgan R. Frank, Matthew Groh, Laura Herman, Neil Leach, Robert Mahari, Alex “Sandy” Pentland, Olga Russakovsky, Hope Schroeder, and Amy Smith. 2023. Art and the science of generative AI. *Science* 380, 6650 (2023), 1110–1111. <https://doi.org/10.1126/science.adh4451> arXiv:<https://www.science.org/doi/pdf/10.1126/science.adh4451>
- [28] Ziv Epstein, Sydney Levine, David G. Rand, and Iyad Rahwan. 2020. Who Gets Credit for AI-Generated Art? *iScience* 23, 9 (Sept. 2020), 101515. <https://doi.org/10.1016/j.isci.2020.101515>
- [29] Henry Farrell, Alison Gopnik, Cosma Shalizi, and James Evans. 2025. Large AI models are cultural and social technologies. *Science* 387, 6739 (2025), 1153–1156. <https://doi.org/10.1126/science.adt9819> arXiv:<https://www.science.org/doi/pdf/10.1126/science.adt9819>
- [30] Silvia Federici and Power of Women Collective. 1975. *Wages against Housework* (1st ed.). Power of Women Collective; Falling Wall Press, London; Bristol.
- [31] L. Fortunati and J. Fleming. 1995. *The Arcane of Reproduction: Housework, Prostitution, Labor and Capital*. Autonomedia.
- [32] Jesse Fox and Andrew Gambino. 2021. Relationship Development with Humanoid Social Robots: Applying Interpersonal Theories to Human–Robot Interaction. *Cyberpsychology, Behavior, and Social Networking* 24, 5 (May 2021), 294–299. <https://doi.org/10.1089/cyber.2020.0181>
- [33] Nancy Fraser. 2016. Contradictions of Capital and Care. *New Left Review* 100 (2016), 99–117.

- [34] David C. Giles. 2002. Parasocial Interaction: A Review of the Literature and a Model for Future Research. *Media Psychology* 4, 3 (Aug. 2002), 279–305. [https://doi.org/10.1207/S1532785XMEP0403\\_04](https://doi.org/10.1207/S1532785XMEP0403_04)
- [35] Omri Gillath, Ting Ai, Michael S Branicky, Shawn Keshmiri, Robert B Davison, and Ryan Spaulding. 2021. Attachment and trust in artificial intelligence. *Computers in human behavior* 115 (2021), 106607.
- [36] Josh A Goldstein, Jason Chao, Shelby Grossman, Alex Stamos, and Michael Tomz. 2024. How persuasive is AI-generated propaganda? *PNAS Nexus* 3, 2 (Feb. 2024), pgae034. <https://doi.org/10.1093/pnasnexus/pgae034>
- [37] Amy L. Gonzales and Jeffrey T. Hancock. 2008. Identity Shift in Computer-Mediated Environments. *Media Psychology* 11, 2 (2008), 167–185. <https://doi.org/10.1080/15213260802023433> arXiv:<https://doi.org/10.1080/15213260802023433>
- [38] Ivor Grattan-Guinness. 2003. *The computation factory: de Prony's project for making tables in the 1790s*. Oxford University Press. <https://doi.org/10.1093/acprof:oso/9780198508410.003.0005> DOI: 10.1093/acprof:oso/9780198508410.003.0005.
- [39] Jeffrey T Hancock, Mor Naaman, and Karen Levy. 2020. AI-Mediated Communication: Definition, Research Agenda, and Ethical Considerations. *Journal of Computer-Mediated Communication* 25, 1 (March 2020), 89–100. <https://doi.org/10.1093/jcmc/zmz022>
- [40] Kunal Handa, Alex Tamkin, Miles McCain, Saffron Huang, Esin Durmus, Sarah Heck, Jared Mueller, Jerry Hong, Stuart Ritchie, Tim Belonax, Kevin K. Troy, Dario Amodei, Jared Kaplan, Jack Clark, and Deep Ganguli. 2025. Which Economic Tasks are Performed with AI? Evidence from Millions of Claude Conversations. arXiv:2503.04761 [cs.CY] <https://arxiv.org/abs/2503.04761>
- [41] Drew Harwell. 2025. Making cash off 'AI slop': The surreal video business taking over the web. <https://www.proquest.com/blogs-podcasts-websites/making-cash-off-ai-slop-surreal-video-business/docview/3240553260/se-2> Name - TikTok Inc; Copyright - Copyright WP Company LLC d/b/a The Washington Post Aug 18, 2025; Last updated - 2025-08-19.
- [42] Will Hawkins, Brent Mittelstadt, and Chris Russell. 2025. Deepfakes on Demand: The rise of accessible non-consensual deepfake image generators. In *Proceedings of the 2025 ACM Conference on Fairness, Accountability, and Transparency (FAccT '25)*. Association for Computing Machinery, New York, NY, USA, 1602–1614. <https://doi.org/10.1145/3715275.3732107>
- [43] Donald Horton and R. Richard Wohl. 1956. Mass Communication and Para-Social Interaction. *Psychiatry* 19, 3 (1956), 215–229. <https://doi.org/10.1080/00332747.1956.11023049> arXiv:<https://doi.org/10.1080/00332747.1956.11023049> PMID: 13359569.
- [44] Jennifer Jacobs, Joel R. Brandt, Radomír Meunđefinedh, and Mitchel Resnick. 2018. Dynamic Brushes: Extending Manual Drawing Practices with Artist-Centric Programming Tools. In *Extended Abstracts of the 2018 CHI Conference on Human Factors in Computing Systems* (Montreal QC, Canada) (*CHI EA '18*). Association for Computing Machinery, New York, NY, USA, 1–4. <https://doi.org/10.1145/3170427.3186492>
- [45] Fredric Jameson. 1991. *Postmodernism, or, The Cultural Logic of Late Capitalism*. Duke University Press. <http://www.jstor.org/stable/j.ctv12100qm>
- [46] Kylie Jarrett. 2014. The Relevance of “Women’s Work”: Social Reproduction and Immaterial Labor in Digital Media. *Television & New Media* 15, 1 (Jan. 2014), 14–29. <https://doi.org/10.1177/1527476413487607>
- [47] Harry H. Jiang, Lauren Brown, Jessica Cheng, Mehtab Khan, Abhishek Gupta, Deja Workman, Alex Hanna, Johnathan Flowers, and Timmit Gebru. 2023. AI Art and its Impact on Artists. In *Proceedings of the 2023 AAAI/ACM Conference on AI, Ethics, and Society* (Montréal, QC, Canada) (*AIES '23*). Association for Computing Machinery, New York, NY, USA, 363–374. <https://doi.org/10.1145/3600211.3604681>
- [48] Caroline A. Jones, Huma Gupta, and Matthew Ritchie. 2024. Visual Artists, Technological Shock, and Generative AI. (March 2024). <https://doi.org/10.21428/e4baedd9.b4f754fd>
- [49] Andrej Karpathy. 2025. Very impressed with Veo 3 and all the things people are finding on r/aivideo etc. <https://x.com/karpathy/status/1929634696474120576>
- [50] Alexandra Kollontai. 1921. Prostitution and Ways of Fighting It. <https://www.marxists.org/archive/kollonta/1921/prostitution.htm> Online English translation available at Marxists Internet Archive.
- [51] Max Kreminski. 2024. The Dearth of the Author in AI-Supported Writing. In *Proceedings of the Third Workshop on Intelligent and Interactive Writing Assistants* (Honolulu, HI, USA) (*In2Writing '24*). Association for Computing Machinery, New York, NY, USA, 48–50. <https://doi.org/10.1145/3690712.3690725>
- [52] Linnea Laestadius, Andrea Bishop, Michael Gonzalez, Diana Illeňćik, and Celeste Campos-Castillo. 2024. Too human and not human enough: A grounded theory analysis of mental health harms from emotional dependence on the social chatbot Replika. *New Media & Society* 26, 10 (Oct. 2024), 5923–5941. <https://doi.org/10.1177/14614448221142007>
- [53] Maurizio Lazzarato. 1996. Immaterial Labor. In *Radical Thought in Italy: A Potential Politics*, Paolo Virno and Michael Hardt (Eds.). University of Minnesota Press, Minneapolis, 132–146.
- [54] Yier Ling and Alex Imas. 2025. Underreporting of AI use: The role of social desirability bias. 5232910 (May 2025). <https://doi.org/10.2139/ssrn.5232910>
- [55] Jiaxu Lou and Yifan Sun. 2025. Anchoring bias in large language models: an experimental study. 9 (Dec. 2025), 11. <https://doi.org/10.1007/s42001-025-00435-2>
- [56] Georg Lukács. 1971. Reification and the Consciousness of the Proletariat. In *History and Class Consciousness*. MIT Press, Cambridge, MA, 83–222.
- [57] Donald MacKenzie. 1984. Marx and the Machine. *Technology and Culture* 25, 3 (1984), 473–502. <https://doi.org/10.2307/3104202>
- [58] Gili Malinsky. 2025. Mark Zuckerberg says people can fill the need for friends with AI, but “there is no replacement” for human relationships, psychologist says. <https://www.cnn.com/2025/05/09/mark-zuckerberg-says-ai-can-replace-human-relationships-expert-disagrees.html>
- [59] Ernest Mandel. 1975. *Late capitalism* (revised edition, ed.). Verso, London.
- [60] Karl Marx. 1976. *Capital: A Critique of Political Economy, Volume I*. Penguin Books, Harmondsworth, UK.

- [61] Emma May, Britt Paris, and Serita Sargent. 2026. Dis/engaging the ‘common sense’ of AI: Labor strategies from the 2023 SAG-AFTRA around data-driven technologies. *Big Data & Society* 13, 1 (2026), 20539517261421466. <https://doi.org/10.1177/20539517261421466> arXiv:<https://doi.org/10.1177/20539517261421466>
- [62] Miller McPherson, Lynn Smith-Lovin, and Matthew E. Brashears. 2006. Social Isolation in America: Changes in Core Discussion Networks over Two Decades. *American Sociological Review* 71, 3 (2006), 353–375. <https://doi.org/10.1177/000312240607100301> arXiv:<https://doi.org/10.1177/000312240607100301>
- [63] U.A. Mejias and N. Couldry. 2024. *Data Grab: The new Colonialism of Big Tech and how to fight back*. Ebury Publishing. <https://books.google.com/books?id=qAm-EAAAQBAJ>
- [64] Meta Platforms. 2023. *Introducing Social Profiles for Meta’s AIs*. <https://about.fb.com/news/2023/09/social-profiles-for-metas-ai-characters/> Accessed: 2025-12-12.
- [65] Meta Platforms. 2024. *Create Your Own Custom AI With AI Studio*. <https://about.fb.com/news/2024/07/create-your-own-custom-ai-with-ai-studio/> Accessed: 2025-12-12.
- [66] Cade Metz. 2020. Riding Out Quarantine With a Chatbot Friend: ‘I Feel Very Connected’. *The New York Times* (2020). <https://www.nytimes.com/2020/06/16/technology/chatbots-quarantine-coronavirus.html>
- [67] Hannah Mieczkowski, Jeffrey T. Hancock, Mor Naaman, Malte Jung, and Jess Hohenstein. 2021. AI-Mediated Communication: Language Use and Interpersonal Effects in a Referential Communication Task. *Proc. ACM Hum.-Comput. Interact.* 5, CSCW1 (April 2021), 17:1–17:14. <https://doi.org/10.1145/3449091>
- [68] Evgeny Morozov. 2022. Critique of Techno-Feudal Reason. *New Left Review* 133/134 (April 2022), 89–126.
- [69] J. Muldoon, M. Graham, and C. Cant. 2024. *Feeding the Machine: The Hidden Human Labour Powering AI*. Canongate Books. <https://books.google.com/books?id=QQRqEAAAQBAJ>
- [70] Muzeum Sztuki w Łodzi. 2021. Errorism. Agnieszka Kurant. <https://msl.org.pl/en/errorism-agnieszka-kurant>
- [71] Premilla Nadasen. 2021. Rethinking Care Work: (Dis)Affection and the Politics of Caring. *Feminist Formations* 33, 1 (2021), 165–188.
- [72] Katie Notopoulos. 2025. Meta’s dream of AI-generated users isn’t going anywhere. <https://www.businessinsider.com/meta-ai-generated-users-facebook-messenger-instagram-2025-1> Accessed: 2025-12-12.
- [73] Gaby Odekerken-Schröder, Cristina Mele, Tiziana Russo-Spena, Dominik Mahr, and Andrea Ruggiero. 2020. Mitigating loneliness with companion robots in the COVID-19 pandemic and beyond: an integrative framework and research agenda. *Journal of Service Management* 31, 6 (Aug. 2020), 1149–1162. <https://doi.org/10.1108/JOSM-05-2020-0148>
- [74] OpenAI. 2025. Our Structure. <https://web.archive.org/web/20250819200810/https://openai.com/our-structure/> Archived from <https://openai.com/our-structure/>
- [75] OpenAI. 2025. Sora 2 is here | OpenAI. <https://openai.com/index/sora-2/>
- [76] Matteo Pasquinelli. 2023. *The eye of the master: a social history of artificial intelligence*. Verso, London New York.
- [77] Qiyao Peng, Yingdan Lu, Yilang Peng, Sijia Qian, Xinyi Liu, and Cuihua Shen. 2025. Crafting Synthetic Realities: Examining Visual Realism and Misinformation Potential of Photorealistic AI-Generated Images. In *Proceedings of the Extended Abstracts of the CHI Conference on Human Factors in Computing Systems*. 1–12. <https://doi.org/10.1145/3706599.3719834> arXiv:2409.17484 [cs].
- [78] Iryna Pentina, Tianling Xie, Tyler Hancock, and Ainsworth Bailey. 2023. Consumer–machine relationships in the age of artificial intelligence: Systematic literature review and research directions. *Psychology & Marketing* 40, 8 (2023), 1593–1614. <https://doi.org/10.1002/mar.21853>
- [79] Laurel Ptak. 2014. Wages for Facebook. <http://wagesforfacebook.com>. Online manifesto/project, accessed YYYY-MM-DD.
- [80] Lucien Rochat, Elena Orita, Emilien Jeannot, Sophia Achab, and Yasser Khazaal. 2023. Willingness to Pay for a Dating App: Psychological Correlates. *International Journal of Environmental Research and Public Health* 20, 3 (2023). <https://doi.org/10.3390/ijerph20032101>
- [81] J. Rosenbaum. 2025. I need husband: AI beauty standards, fascism and the proliferation of bot driven content. <https://doi.org/10.1007/s00146-025-02491-8>
- [82] Rob Salkowitz. 2022. Midjourney Founder David Holz On The Impact Of AI On Art, Imagination And The Creative Economy. <https://www.forbes.com/sites/robsalkowitz/2022/09/16/midjourney-founder-david-holz-on-the-impact-of-ai-on-art-imagination-and-the-creative-economy/>
- [83] Michael Schwarz and Peter Wonka. 2015. Practical grammar-based procedural modeling of architecture: SIGGRAPH Asia 2015 course notes. In *SIGGRAPH Asia 2015 Courses* (Kobe, Japan) (SA ’15). Association for Computing Machinery, New York, NY, USA, Article 13, 12 pages. <https://doi.org/10.1145/2818143.2818152>
- [84] Sangeeta Singh-Kurtz. 2023. The Women Falling in Love With Their AI Boyfriends. <https://www.thecut.com/article/ai-artificial-intelligence-chatbot-replika-boyfriend.html>
- [85] Marita Skjuve, Petter Bae Brandtzaeg, and Asbjørn Følstad. 2024. Why do people use ChatGPT? Exploring user motivations for generative conversational AI. *First Monday* (Jan. 2024). <https://doi.org/10.5210/fm.v29i1.13541>
- [86] Marita Skjuve, Asbjørn Følstad, and Petter Bae Brandtzaeg. 2023. A Longitudinal Study of Self-Disclosure in Human–Chatbot Relationships. *Interacting with Computers* 35, 1 (2023), 24–39. <https://doi.org/10.1093/iwc/iwad022>
- [87] Marita Skjuve, Asbjørn Følstad, Knut Inge Fostervold, and Petter Bae Brandtzaeg. 2021. My Chatbot Companion - a Study of Human-Chatbot Relationships. *International Journal of Human-Computer Studies* 149 (May 2021), 102601. <https://doi.org/10.1016/j.ijhcs.2021.102601>
- [88] Marita Skjuve, Asbjørn Følstad, Knut Inge Fostervold, and Petter Bae Brandtzaeg. 2022. A longitudinal study of human–chatbot relationships. *International Journal of Human-Computer Studies* 168 (Dec. 2022), 102903. <https://doi.org/10.1016/j.ijhcs.2022.102903>

- [89] Adam Smith. 1776. *An Inquiry into the Nature and Causes of the Wealth of Nations* (none ed.). History of Economic Thought Books, Vol. None. McMaster University Archive for the History of Economic Thought. <https://doi.org/None>
- [90] Nicholas Srnicek. 2016. *Platform Capitalism*. Polity.
- [91] Vivian Ta, Caroline Griffith,Carolynn Boatfield, Xinyu Wang, Maria Civitello, Haley Bader, Esther DeCero, and Alexia Loggarakis. 2020. User Experiences of Social Support From Companion Chatbots in Everyday Contexts: Thematic Analysis. *Journal of Medical Internet Research* 22, 3 (March 2020), e16235. <https://doi.org/10.2196/16235>
- [92] Katherine Tangelakis-Lippert. 2025. Elon Musk says Grok’s latest feature is the new Vine. <https://www.businessinsider.com/elon-musk-grok-new-image-generator-restore-vine-2025-8>
- [93] D.Y. Tenen. 2024. *Literary Theory for Robots: How Computers Learned to Write*. W. W. Norton. <https://books.google.com/books?id=pw3HEAAAQBAJ>
- [94] Tiziana Terranova. 2000. Free Labor: PRODUCING CULTURE FOR THE DIGITAL ECONOMY. *Social Text* 18, 2 (63) (2000), 33–58. [https://doi.org/10.1215/01642472-18-2\\_63-33](https://doi.org/10.1215/01642472-18-2_63-33)
- [95] Stephanie Tom Tong and Joseph B. Walthers. 2011. Just say “no thanks”: Romantic rejection in computer-mediated communication. *Journal of Social and Personal Relationships* 28, 4 (2011), 488–506. <https://doi.org/10.1177/0265407510384895>
- [96] Rebecca Umbach, Nicola Henry, and Gemma Beard. 2025. Prevalence and Impacts of Image-Based Sexual Abuse Victimization: A Multinational Study. In *Proceedings of the 2025 CHI Conference on Human Factors in Computing Systems (CHI ’25)*. Association for Computing Machinery, New York, NY, USA, Article 517, 20 pages. <https://doi.org/10.1145/3706598.3713545>
- [97] US HHS, OSG. 2023. Our Epidemic of Loneliness and Isolation: The U.S. Surgeon General’s Advisory on the Healing Effects of Social Connection and Community. <https://www.hhs.gov/sites/default/files/surgeon-general-social-connection-advisory.pdf>. Accessed: 2025-12-12.
- [98] Aditya Nrusimha Vaidyam, Hannah Wisniewski, John David Halamka, Matcheri S. Kashavan, and John Blake Torous. 2019. Chatbots and Conversational Agents in Mental Health: A Review of the Psychiatric Landscape. *Canadian Journal of Psychiatry. Revue Canadienne De Psychiatrie* 64, 7 (2019), 456–464. <https://doi.org/10.1177/0706743719828977>
- [99] Yanis Varoufakis. 2023. *Technofeudalism: What Killed Capitalism*. Vintage Digital.
- [100] Robert. Venturi, Denise Scott Brown, and Steven Izenour. 1972. *Learning from Las Vegas*. MIT Press, Cambridge, Mass.
- [101] James Vincent. 2022. “An engine for the imagination”: an interview with David Holz, CEO of AI image generator Midjourney. <https://www.theverge.com/2022/8/2/23287173/ai-image-generation-art-midjourney-multiverse-interview-david-holz>
- [102] Leif Weatherby. 2025. *Language Machines: Cultural AI and the End of Remainder Humanism*. University of Minnesota Press, Minneapolis. [https://muse.jhu.edu/pub/23/edited\\_volume/book/129503](https://muse.jhu.edu/pub/23/edited_volume/book/129503)
- [103] Yiluo Wei, Yiming Zhu, Pan Hui, and Gareth Tyson. 2024. Exploring the Use of Abusive Generative AI Models on Civitai. In *Proceedings of the 32nd ACM International Conference on Multimedia (Melbourne VIC, Australia) (MM ’24)*. Association for Computing Machinery, New York, NY, USA, 6949–6958. <https://doi.org/10.1145/3664647.3681052>
- [104] Meredith Whittaker. 2023. Origin Stories: Plantations, Computers, and Industrial Control – logicmag.io. <https://logicmag.io/supa-dupa-skies/origin-stories-plantations-computers-and-industrial-control/>. [Accessed: 31-10-2025].
- [105] Langdon Winner. 1980. Do Artifacts Have Politics? *Daedalus* 109, 1 (1980), 121–136.
- [106] xAI. 2024. <https://x.ai/news/grok-image-generation-release>
- [107] Tianling Xie and Iryna Pentina. 2022. Attachment Theory as a Framework to Understand Relationships with Social Chatbots: A Case Study of Replika. <https://doi.org/10.24251/HICSS.2022.258>
- [108] Marc Zao-Sanders. 2024. How People Are Really Using GenAI. *Harvard Business Review Digital Articles* (March 2024), 1–9. <https://research.ebsco.com/linkprocessor/plink?id=b573fa74-cc50-380e-a17f-402b527c27ba> Publisher: Harvard University.

## A MEDIATIVE AUTOMATION: SURVEYS

Much of the existing literature focuses either on a single platform, such as a single chat interface for an LLM [14, 40], or focus exclusively on uses within the workplace [7, 40]. Moreover, early evidence points to underreporting of certain use-cases of generative models due to social pressures [54].

Zao-Sanders [108] 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. [14] 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. [85] 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. [7] 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. [40] 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.