

Addressing the Black Box of AI – A Model and Research Agenda on the Co-Constitution of Aging and Artificial Intelligence

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Abstract

Algorithmic technologies and (large) data infrastructures, often referred to as Artificial Intelligence (AI), have received increasing attention from gerontological research in the last decade. While there is much literature that dissects and explores the development, application, and evaluation of AI relevant for gerontology, this article makes a novel contribution by critically engaging with the theorizing in this growing field of research. We observe that gerontology's engagement with AI is shaped by an interventionist logic that situates AI as a black box for gerontological research. We demonstrate how this black box logic has neglected many aspects of AI as a research topic for gerontology and discuss three classical concepts in gerontology to show how they can be used to open various black boxes of aging and AI in the areas: a) the datafication of aging, b) the political economy of AI and aging, and c) everyday engagements and embodiments of AI in later life. In the final chapter, we propose a model of the co-constitution of aging and AI that makes theoretical propositions to study the relational terrain between aging and AI and hence aims to open the black box of AI in gerontology beyond an interventionist logic.

Keywords: *Gerontechnology, Datafication, Algorithm, Ageism*

Introduction

Artificial intelligence (AI) and algorithmic technologies have gained increasing relevance and attention in society, including its health and care systems (Berridge & Grigorovich, 2022; Lukkien et al., 2023). While it is difficult if not impossible to give a clear definition of AI, the term usually refers to some form of algorithmic automation – for instance, automated decisions, decision support or classifications – based on large quantities of data about behaviors of people.¹ For instance, social media, streaming services or fitness trackers use large amounts of data to make recommendations about what news items to consume, what movies to watch or how much calories still need to be burned in a day. Companies and government organizations increasingly rely on automated decision making, fed by data about a client's or citizen's identity and behaviors, for instance in granting mobile phone contracts, deciding about access to social benefits schemes, or policing traffic offenders. Large Language Models (LLMs) such as ChatGPT use sophisticated autocomplete algorithms, operating on large quantities of previously published language to generate texts that appear meaningful.

It is increasingly clear that AI also plays a role in the lives of older people. For one, the everyday lives of older people are no exception to the prevalence of data-collecting and -processing technologies like smartphones, connected cars or smart speakers. At the same time, much hope is placed on AI based technologies specifically designed for older people, for instance in the form of decision support systems in dementia care, remote risk detection and monitoring systems to detect falls or deviations from “normal” daily routines, or social robots that can hold conversations to mitigate loneliness. Much attention and large-scale financial investments are directed to such AI in gerontechnology (Rubeis, 2020). The overriding hope is that AI in gerontechnology will support health and care professionals (e.g. in clinical decision making, remote monitoring or predictive analysis), and ultimately enable older adults to live and age autonomously (Chen, 2020).

The relevance of AI for the lives of older people has been recognized in gerontology (Chu et al., 2022; Lukkien et al., 2023). So far, however, available studies have focused on evaluating the “impact” of AI technologies on the lives of older people, predominantly in the context of formal and informal care (Loveys et al., 2022). Critical engagement with theory has been largely absent in these studies, which are typically designed as intervention studies – a specific AI technology is implemented temporarily in the lives or care environments of a selected group of older people to measure in how far certain parameters such as depression symptoms, quality of life or agitation changed (Loveys et al., 2022).

In this article, we discuss the need for gerontology to more deeply engage with theorizing in this growing field of AI and ageing research. Our main argument is that gerontology’s focus on intervention studies goes hand in hand with what Peine & Neven (2019) have called an “interventionist logic”. This logic analytically separates the lives of older people from the design and use of technologies so that technology can neatly be conceptualized as an intervention with defined and measurable parameters for its success (or failure). It thus renders invisible, and makes inaccessible to critical theoretical reflection, the dynamics of ageing and technology relations in both the worlds of technological design and the lives of older people.

Hence, the interventionist logic renders AI as a black box (Latour, 2000) in gerontology, making invisible the construction of aging in AI design, its embedding in socio-material infrastructures, and the engagements of older adults with it. To rectify this, we draw on the co-constitution of aging and technology (Peine & Neven, 2021) which studies how ageing and technology come into existence in relation to each other. This notion allows us to conceptually explore the relationship between aging and AI, to highlight the socio-material associations that currently exist between aging and AI, and to outline questions for future gerontological research on ageing and AI. This article, therefore, is a timely intervention that seeks to broaden gerontology’s perspective on AI and open up this topic for gerontological debate and research beyond an interventionist logic.

Black-boxing AI in gerontology

The figure of the black box is often used to highlight forms of opacity in the design and deployment of AI systems (Jarke & Heuer, forthcoming). On a technical level, this opacity stems from algorithmic processes that infer correlations, representations, and categories based on large amounts of data. The ways in which these outputs are created often remain incomprehensible to human beings (Carabantes, 2020), partly because companies refuse transparent data documentation practices. The figure of the black box is also used to describe how AI is mystified (Soraa, 2022) as an entity that detects patterns beyond human intelligence or imagination (Campolo & Crawford, 2020). Such mystification can be seen as a deliberate attempt to distract from the more immediate challenges that arise from the marketization of AI in an age of surveillance capitalism (Zuboff, 2019). Black-boxing also relates to the way in which AI systems are embedded in healthcare and long-term care infrastructures and the lives of its users. These embeddings and the associated changes in decision making processes are widespread, yet they often remain opaque to those interacting within them.

Hence, the black-boxing of AI not only exists due to technological complexity, but it may also be intentionally or unintentionally made to be so by a variety of actors. These actors, we argue, include gerontologists that neglect technological development, implementation, and related social processes as objects of scientific study. We hence argue that a black box can also be understood as an empirical and conceptual problem which can and should be “opened” or “unpacked” (Bucher, 2018) by those who study aging and technology.

In what follows, we aim to open the black box of AI in gerontology by revisiting a range of classical concepts in gerontology and discussing how they can be used to open various black boxes of aging and AI. Opening these black boxes, we argue, holds two potentials for gerontological research: On

the one hand, it offers possibilities for more in-depth engagements with AI as a gerontological topic of research. On the other hand, it also offers new ways of theorizing the relationship between ageing and AI that go beyond an interventionist logic and instead focus on the co-constitution of ageing and technology (Peine and Neven 2019; 2021). We conclude by offering a model of the co-constitution of aging and AI that makes theoretical propositions to study the relational terrain between aging and AI beyond an interventionist gaze.

The datafication of aging

In *Disciplining Old Age*, Katz (1996) employed a Foucauldian analysis of the classification practices that constitute aging bodies as inherently senescent and risky, yet also amenable to discipline and remediation. He provided a portrait of aging bodies as historically configured through webs of knowledge and power. Such classification practices are at the heart of AI and algorithmic technologies (Joyce et al., 2021), which makes the webs of knowledge and power that constitute old age today increasingly technologically mediated, more durable and persistent.

One way of unpacking the black box of AI in the context of aging lies in exploring how AI is not only a technological system, but a web of knowledge that constitutes ageing bodies as measurable and quantifiable. AI systems tend to prioritize information that can be quantified, categorized and classified. Based on these data, AI systems produce output that is taken to be objective, authoritative, and fair. This suggests the collection of data on older populations as universally able to represent an objective, external reality about aging and views the application of technological knowledge as the most effective and economical approach to solving the “problems” linked to population aging (Moreira, 2017). This techno-solutionist thinking, however tends to black box the situated nature of both algorithms and data,

as it makes the role of humans in making and constructing these data, and the “constructed, polyvalent nature of human data” (Joyce et al., 2021) invisible.

Another way of unpacking the black box of AI in gerontology hence lies in theorizing data beyond such a techno-solutionist logic. For example, Beer (2019) refers to the value offered up by data analytics as comprising a “data imaginary,” promising speed, accessibility, insight, prediction and efficiency. When data imaginaries are enacted, for example, through ambient monitoring technologies for aging in place, AI driven analytics are promoted as the most cost-efficient and “smart” solutions to caring for an aging population. However, as Beer (2016, p. 60) remarks, such forms of measurement are not only powerful for what they capture, but also for what they conceal or ignore. This truncated data imaginary risks devaluing or obscuring other kinds of knowledge or experience, rendering factors outside of the machine’s calculations as less relevant to the production of knowledge. Through datafied classification and categorization, the visibility of an older body is limited to the data produced by AI systems, which is then “aggregated and itemized into risk assessments and patterns of behavior” (Ellison et al, 2022). Older adults get placed into categories with consequences for their lives – they can become “fallers”, “at risk”, “frail”, “aggressive”, in need of residential care or “untrustworthy” reporters (Berridge & Grigorovich, 2022), while their subjective experiences of these categories become less relevant forms of knowledge.

The political economy of aging and AI

Political economy perspectives in gerontology (Estes et al., 1982) have articulated the ways in which capitalism and neoliberal profit imperatives drive interest in older adults, ultimately framing them as “profit-making commodities” (Estes, 1993). There is a similar profit-making logic involved in massive investments of policy makers and companies in the development of AI gerontechnologies (Sadowski,

2019). As the black box logic of AI tends to make commodification and profit generation through AI invisible, one way of unpacking the black box lies in engaging more deeply with the commodification of ageing in the political economy of AI, and asking which forms of profit generation become relevant as AI for older adults is developed and implemented.

In the political economy of aging and AI, older adults tend to be viewed as data suppliers who generate profits for AI companies (Birch & Cochrane, 2022). Chu et al. (2022) note that there is hardly enough data about older adults available to train AI models towards the needs of this population. Available data infrastructures often show explicit or implicit age-related bias (Diaz et al., 2018). This points to a major structural problem for the creation of inclusive and fair AI systems (Stypinska, 2021, 2022), but also enables AI companies to collect ‘unique aged-data’ to create new market segments. However, even if the data are derived from, for example, racially diverse or economically marginalized older populations, any divergence gets subsumed within the neoliberal logic of difference (Ludwig, 2016; McNay, 2009) rather than the politics of intersectionality (Ranjan-Rankin, 2018). In this way, efforts to create more representative data sets of older adults commodify “difference” as marginalized groups of older adults become unique and valuable data suppliers. For AI companies, inclusion efforts thus become a new means of generating profits through the ‘differential quality of user data and engagement’ (Birch & Cochrane, 2022, p. 51).

Further, a political economy perspective on AI gerontechnologies enables questioning the role of (paid and unpaid) labor which is necessary to implement AI systems. Even though AI is often perceived as a neutral artefact that “parachutes into” the lives of older adults, specific uses of AI require contextualization and local embedding (Lukkien et al. 2022). Consequently, recent research has stressed the relevance of invisible – and at times unpaid – care giving, that is key to achieving the implementation of AI in diverse contexts of aging (Gallistl & von Laufenberg 2023). This not only includes the work of software designers and programmers, but also work of health care professionals, care staff

in long-term care organizations, or the data work that older adults and family carers need to provide to make AI systems run in practice.

A political economy perspective on aging and AI demands reflection about and analysis of the “ground-truthing, programming and formulating” (Jaton, 2021) that happens in AI companies and transparency about what data a company is collecting with what purpose and how profit is generated through this data (Tucker, 2022).

Everyday embodiments of AI in later life

Cultural gerontology has put forward an embodied understanding of old age and later life (Gilleard & Higgs, 2014; Twigg, 2000). This has sought to move away from a bio-medical gaze on the older body (e.g. Marshall, 2022; Öberg, 1996) and instead focusses on how ageing is embodied as an everyday experience. The embodiments and engagements with AI in the everyday lives of older adults are often overlooked and older adults’ agency is rarely discussed in the context of AI (Neves et al., 2023). However, these engagements are crucial to enhance the design and implementation of technologies into the lives of older adults (van Leersum et al., 2023). One way of unboxing the black box of AI in gerontology lies in theorizing the ways in which datafied embodiment (Lupton et al., 2022) gains relevance in the lives of older adults, and in questioning how older adults perceive, make sense of, or stand in opposition to the implementation of AI and its related data practices (for discussion of refusal, see Berridge & Grigorovich, 2022; 2023; Brewer, 2022).

Understanding older adults’ embodiments and agency in the context of AI includes further unpacking the ways in which the everyday lives of older people are becoming sites of datafication, monitoring, and surveillance (Katz and Marshall, 2018; Dalmer et al., 2022; Ellison et al., 2022). This might also include making visible how the embodiment of aging in everyday life is more-than-human,

and how diverse materialities (including AI systems) are part of the constitution of aging in the lives of older adults. Traditionally, gerontology has theorized aging as a human phenomenon, that happens to or within an aging body. These boundaries of aging are, however, increasingly blurred as datafied forms of embodiment and data doubles become part of the everyday experience of ageing. Another way of unpacking the black box of AI in gerontology lies in exploring how diverse forms of materiality – including sensors, wearables, social or assistive robots or other monitoring devices – become an integral part of the experience of ageing. These entanglements between humans and non-humans can be observed and mapped as a means to understand more about ageing as a more-than-human phenomenon (Gallistl & Wanka, 2023).

The focus on everyday embodiments and engagements of AI also offers a novel perspective on ethics and values that underpin AI. So far, ethical debates on AI in ageing research have largely been informed by a principlist approach, which has narrowed ethical reflections to balancing risks and benefits of a certain AI system (Grigorovich & Kontos, 2020). An everyday perspective on ethics in AI highlights that ethical questions cannot be captured and solved at one point in time but require an ongoing attention to the multiple values that inform the development, implementation, and use of AI (van Hees et al. 2023, Gallistl & Wanka, 2022). Such a valuation approach towards ethics in AI might not seek to define and solve ethical questions, but rather explore the practices of valuation through which manifold forms of value are “produced, diffused, assessed, and institutionalized” (Lamont, 2012: 201) throughout the lifecycle of an AI system. It might ask, for example: What are diverse notions of ‘good’ that inform the development and the implementation of an AI system? How are they re-constituted in such development and implementation?

Conclusion: Co-Constitution of Aging and Artificial Intelligence – A research agenda and model

Considerable public investment in AI, particularly in the care for older adults, means that relevance of AI for older adults will grow, and so will the need to study AI as a gerontological subject matter. We conclude by putting forward four pathways for research on the relational terrain between aging and AI (Figure 1). These pathways build on the overarching conceptual themes we have drawn out from prior gerontological scholarship: classification in knowledge and power, political economy, and embodiment. This model proposes four arenas in which the co-constitution of aging and AI can be studied.

First, gerontology ought to interrogate practices of designing AI and its related logics of commodification and value creation within a political economy of aging. We propose that gerontological research needs to engage with the ways in which AI is imagined, developed, and evaluated for various older target groups, and critically analyze the images of aging that guide this design. This might also include asking which myths are created around AI for older adults with the aim of creating hype and investment about these systems (Hoffman et al., 2022) and exploring the ways in which AI is marketed across various target groups (e.g., healthcare systems, the long-term care industry, older adults, family care partners, etc.).

It also means engaging critically with persisting and emerging forms of exploitation within a political economy of AI and aging. As a field that is “data-rich” (Birren & Bengtson, 1988) gerontology is well positioned to investigate the representation of older adults in data infrastructures that are available to train AI, as well as the consequences of under- or over- representation in these infrastructures.

Examining potential harms is an ethical imperative and will be most impactful when it is responsive to

the realities of those who experience marginalization, including 2SLGBTQI+,² BIPOC,³ disabled, and economically vulnerable older people and care workers (Berridge & Grigorovich, 2022).

Second, we invite gerontologist to explore how AI systems for older adults are practically contextualized, implemented, and locally embedded in care arrangements and everyday life activities; and ask how these care arrangements are changing through the embedding of AI and big data logics. Recent reviews have suggested that there is lack of knowledge on how innovation through AI is practically achieved in context (Lukkien et al. 2023). On the one hand, this is because AI companies aim to offer somewhat standardized and scalable solutions, but it is also because AI is portrayed as an objective and neutral technological actor, which tends to make the human labor involved in the creation and implementation of these systems invisible (Gallistl & von Laufenberg, 2023).

Third, our work has highlighted the need to explore empirically how older adults make sense of, engage, and tinker with AI in their everyday lives. Discourses that portray older adults as incompetent, uninterested or invisible users of technologies (Mannheim et al., 2022) tend to black box the active engagements of older adults with technologies. However, research in the field of Socio-gerontechnology (Peine et al., 2021) has highlighted that older adults – including people with high care needs – routinely and actively engage with technological innovations (Gallistl et al., 2021), underscoring the importance of older adult's agentic engagement and subjectivity in research on AI. Recent work on explainability of AI has proposed to focus on sense-making practices as a crucial element in understanding how people perceive and understand AI (Papagni et al. 2023). There is significant room in the context of aging for work on explainability, as well as expanding participation in AI development and governance (Gilman, 2023).

Fourth, gerontology ought to interrogate the meanings of data in the context of AI and the practices of datafication that go hand in hand with the development and implementation of AI. Many AI

systems rely on massive amounts of data available through the health and care sectors with the built-in assumption that these data are neutral and truthful representations of reality (Hoffman et al., 2022). In contrast, we highlight that what we understand as data, and the value we connect to it in the context of aging, is contingent, ambivalent and ever-changing. We therefore invite gerontologists to engage with data imaginaries (Beer, 2019), to question gerontology's values and expectations around data, and to explore the research culture that forms around these values. This also means exploring the ways in which data are collected, curated and used to build AI for an aging society, as well as questioning what kinds of reality about aging are represented in and created through these data.

Finally, we highlight the need to reflect on how the relationship between aging and AI might be imagined otherwise. The emergence of AI applications poses several ethical questions and challenges, but it also presents an opportunity to ask how sociotechnical arrangements can bring about better futures (Joyce et al. 2021; Ho, 2023). This holds true for imagining a more age-inclusive and age-friendly society. As Onuaha (2022) articulated, technology is created in the service of something, and when we do not have clarity about what aim it is in service of, "it will just default to supporting the dominant model of power that exists at the time, or the means to which you can attain power." We invite gerontology to ask, what is the purpose or benefit of AI applications in gerontechnology beyond the service of capitalist ends? And, what are the liberatory purposes that we may reorient towards instead? For example, can we develop AI gerontechnologies to prevent the transformation of aging subjects into data capital to be exchanged or monopolized by AI companies for profits? Can we disentangle efficiency from care quality in a way that is accountable to older adults? Can AI applications support various ways of aging without or beyond commodification, profit generation or cost reduction?

Drawing on empirical engagements with the four areas of practice we outlined in our model, we hope that gerontologists will explore these questions as part of a gerontological research agenda and

further develop the theoretical tools needed to imagine socio-technical futures of aging that are grounded in the contexts of peoples' lives rather than today's context of AI hype.

Author Notes

¹It is a core argument that the term AI is vague and fluid because it is used in different ways by different parties, often in strategic and deliberately obfuscating ways. For us, it is thus an empirical question what AI is and does in different contexts. We try to be specific, though, about the core technologies and processes that the products and services have in common that populate discourses around AI, hence our reference to data infrastructures and algorithmic decision making and classification. Broussard (2018) provides a succinct and easy to understand introduction to the technological core of many AI technologies. Our description of AI as automation based on large quantities of data has been inspired by the work Emily Bender, who succinctly summarizes it here: <https://medium.com/@emilymenonbender/opening-remarks-on-ai-in-the-workplace-new-crisis-or-longstanding-challenge-eb81d1bee9f> (accessed: 15 January 2024).

²Acronym used to refer to 2S: Two-Spirit people as the first 2SLGBTQI+ communities; L: Lesbian; G: Gay; B: Bisexual; T: Transgender; Q: Queer; I: Intersex; +: is inclusive of people who identify as part of sexual and gender diverse communities, who use additional terminologies.

³Acronym used to refer to Black, Indigenous and People of Color.

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References

- A Conversation with Mimi Onuaha. (2022, May 31). Data Science By Design.
<http://datasciencebydesign.org/blog/conversation-with-mimi>
- Beatty, J. S., Chakraborty, R. S., Gal, R. P., Girard, A. K., MacDevitt, J., & Sengupta, U. (2020). *Algorithmic Culture: How Big Data and Artificial Intelligence are Transforming Everyday Life*. Lexington Books.
- Beer, D. (2016). *Metric Power*. Palgrave MacMillan.
- Beer, D. (2019). *The Data Gaze: Capitalism, Power and Perception*. Sage.
- Berridge, C. & Grigorovich, A. (2022). Algorithmic harms and digital ageism in the use of surveillance technologies in nursing homes. *Frontiers in Sociology*.
<https://doi.org/10.3389/fsoc.2022.957246>.
- Berridge, C. & Grigorovich, A. (March 29, 2023). We need to talk about “digital ageism.” *Points*. Data & Society. <https://points.datasociety.net/we-need-to-talk-about-digital-ageism-21e4c0c7dff3>

Berridge, C., & Wetle, T. F. (2020). Why Older Adults and Their Children Disagree About In-Home Surveillance Technology, Sensors, and Tracking. *The Gerontologist*, 60(5), 926–934.

<https://doi.org/10.1093/geront/gnz068>

Birch, K., & Cochrane, D. T. (2022). Big tech: Four emerging forms of digital rentiership. *Science as Culture*, 31(1), 44-58. <https://doi.org/10.1080/09505431.2021.1932794>

Broussard M (2018) *Artificial Unintelligence: How Computers Misunderstand the World*. MIT Press.

Bucher, T. (2018). *If...then: Algorithmic power and politics*. Oxford University Press.

Brewer, R. (2022). “If Alexa knew the state I was in, it would cry”: Older Adults’ Perspectives of Voice Assistants for Health. In Extended Abstracts of the 2022 CHI Conference on Human Factors in Computing Systems (CHI EA '22). Association for Computing Machinery (1–8).

Campolo, A., & Crawford, K. (2020). Enchanted determinism: Power without responsibility in artificial intelligence. *Engaging Science, Technology, and Society*.

Carabantes, M. (2020). Black-box artificial intelligence: an epistemological and critical analysis. *AI & society*, 35(2), 309-317. <https://doi.org/10.1007/s00146-019-00888-w>

Chen, L. K. (2020). Gerontechnology and artificial intelligence: better care for older people. *Archives of gerontology and geriatrics*, 91, 104252. <https://doi.org/10.1016/j.archger.2020.104252>

Chu, C. H., Nyrup, R., Leslie, K., Shi, J., Bianchi, A., Lyn, A., McNicholl, M., Khan, S., Rahimi, S., & Grenier, A. (2022). Digital ageism: Challenges and opportunities in artificial intelligence for older adults. *The Gerontologist*, 62(7), 947–955. <https://doi.org/10.1093/geront/gnab167>

D'Ignazio, C., & Klein, L. F. (2020). *Data Feminism*. MIT Press.

Dalmer, N., Ellison, K., Katz, S., & Marshall, B. (2022). Ageing, embodiment and datafication: dynamics of power in digital health and care technologies. *International Journal of Ageing and Later Life*, 15(2), 77-101. <https://doi.org/10.3384/ijal.1652-8670.3499>

Ellison, K. L., Martin, W., Pedersen, I., & Marshall, B. L. (2022). Visualizing the datasphere: Representations of old bodies and their data in promotional images of smart sensor technologies for aging at home. *Frontiers in Sociology*, 7, 1008510–1008510. <https://doi.org/10.3389/fsoc.2022.1008510>

Estes, C. L. (1979). *The aging enterprise*. Jossey-Bass.

Estes, C. L., Swan, J. H., & Gerard, L. E. (1982). Dominant and competing paradigms in gerontology: Towards a political economy of ageing. *Ageing & Society*, 2(2), 151-164.
<https://doi.org/10.1017/S0144686X00009405>

Estes, C. L. (1993). The aging enterprise revisited. *The gerontologist*, 33(3), 292-298.
<https://doi.org/10.1093/geront/33.3.292>

Fernández-Ardèvol, M., & Grenier, L. (2022). Exploring data ageism: What good data can ('t) tell us about the digital practices of older people? *New Media & Society*, 14614448221127261.
<https://doi.org/10.1177/14614448221127261>

Fraser, N. (2016). Contradictions of capital and care. *New Left Review*, 100(100), 99–117.
<https://newleftreview.org/issues/ii100/articles/nancy-fraser-contradictions-of-capital-and-care>

Hoffman, S. G., Joyce, K., Alegria, S., Bell, S. E., Cruz, T. M., Noble, S. U., ... & Smith-Doerr, L. (2022). Five Big Ideas About AI. *Contexts*, 21(3), 8–15. <https://doi.org/10.1177/15365042221114975>

Jarke, J., & Heuer, H. (2024). Reassembling the Black Box of Machine Learning: Of Monsters and the Reversibility of Foldings. In Jarke, J., Prietl, B., Egbert, S., Boeva, Y., Heuer, H., Arnold, M. (Eds). *Algorithmic Regimes. Methods, Interactions, Politics*. University Press.

Jarke, J., & Macgilchrist, F. (2021). Dashboard stories: How narratives told by predictive analytics reconfigure roles, risk and sociality in education. *Big Data & Society*, 8(1), 205395172110255. <https://doi.org/10.1177/20539517211025561>

Joyce, K., Smith-Doerr, L., Alegria, S., Bell, S., Cruz, T., Hoffman, S. G., ... & Shestakofsky, B. (2021). Toward a sociology of artificial intelligence: A call for research on inequalities and structural change. *Socius*, 7, 2378023121999581. <https://doi.org/10.1177/23780231219995>

Katz, S. (1996). *Disciplining Old Age: The Formation of Gerontological Knowledge*. University Press of Virginia.

Latour, B. (2000). *Pandora's hope: Essays on the reality of science studies* (2. print). Harvard University Press.

Loveys K et al. (2022) Artificial intelligence for older people receiving long-term care: A systematic review of acceptability and effectiveness studies. *The Lancet Health Longevity* 3:e286-e297. [https://doi.org/10.1016/S2666-7568\(22\)00034-4](https://doi.org/10.1016/S2666-7568(22)00034-4)

Ludwig, G. (2016). Desiring neoliberalism. *Sexuality Research and Social Policy*, 13, 417-427. <https://doi.org/10.1007/s13178-016-0257-6>

Lupton, D., Clark, M., & Southerton, C. (2022). Digitized and datafied embodiment: A more-than-human approach. In *Palgrave Handbook of Critical Posthumanism*. (pp. 1–23). Springer International Publishing.

Mannheim, I., Wouters, E. J., Köttl, H., Van Boekel, L. C., Brankaert, R., & Van Zaaen, Y. (2022). Ageism in the discourse and practice of designing digital technology for older persons: A scoping review. *Gerontologist*, 63(7), 1188–1200. <https://doi.org/10.1093/geront/gnac144>

McNay, L. (2009). Self as enterprise: Dilemmas of control and resistance in Foucault's *The Birth of Biopolitics*. *Theory, Culture & Society*, 26(6), 55–77. <https://doi.org/10.1177/0263276409347697>

Mejias, U. A., & Couldry, N. (2019). Datafication. *Internet Policy Review*, 8(4), 1–10. <https://doi.org/10.14763/2019.4.1428>

Morrison, M. (2022). Making bio-objects mobile: behind the scenes of a translational stem cell banking consortium. *BioSocieties*, 17(1), 145-168. <https://doi.org/10.1057/s41292-020-00207-3>

Moreira, T. (2017). *Science, Technology and the Ageing Society*. Routledge.

Neves, B. B., Petersen, A., Vered, M., Carter, A., & Omori, M. (2023). Artificial intelligence in long-term care: Technological promise, aging anxieties, and sociotechnical ageism. *Journal of Applied Gerontology*, 42(6), 1274-1282. <https://doi.org/10.1177/07334648231157370>

Papagni, G., de Pagter, J., Zafari, S., Filzmoser, M., & Koeszegi, S. T. (2023). Artificial agents' explainability to support trust: considerations on timing and context. *AI&Society*, 38, 947-960. <https://doi.org/10.1007/s00146-022-01462-7>

Rajan-Rankin, S. (2018). Race, embodiment and later life: Re-animating aging bodies of color. *Journal of Aging Studies*, 45, 32-38. <https://doi.org/10.1016/j.jaging.2018.01.005>

Rubeis G (2020) The disruptive power of artificial intelligence. Ethical aspects of gerontechnology in elderly care. *Archives of Gerontology and Geriatrics* 91:104186. <https://doi.org/10.1016/j.archger.2020.104186>

Ruckenstein, M. (2014). Visualized and Interacted Life: Personal Analytics and Engagements with Data Doubles. *Societies*, 4(1), 68–84. <https://doi.org/10.3390/soc4010068>

Sadowski, J. (2019). When data is capital: Datafication, accumulation, and extraction. *Big data & society*, 6(1), 2053951718820549. <https://doi.org/10.1177/2053951718820549>

Søraa, R. (2023). *AI for Diversity*. CRC Press.

Stypinska, J. (2022). AI ageism: A critical roadmap for studying age discrimination and exclusion in digitalized societies. *AI & society*, 1–13. <https://doi.org/10.1007/s00146-022-01553-5>

Stypińska, J., Rosales, A., & Svensson, J. (2023). Silicon Valley ageism—ideologies and practices of expulsion in the technology industry. In *Digital Ageism* (pp. 53-70). Routledge.

Tucker, E. (2022, March 17.). *Artifice and Intelligence*. Tech Policy Press. <https://techpolicy.press/artifice-and-intelligence/>

van Hees S et al. (2023) Valuation in health and ageing innovation practices. *Ageing and Society* 43:2022-2040. <https://doi.org/10.1017/S0144686X21001483>

Wanka, A., & Gallistl, V. (2018). Doing Age in a Digitized World—a Material Praxeology of Aging With Technology. *Frontiers in Sociology*, 3(6). <https://doi.org/10.3389/fsoc.2018.00006>

Wright, J. (2023). *Robots Won't Save Japan: An Ethnography of Eldercare Automation*. Cornell University Press.

Zuboff, Shoshana (2019). The Age of Surveillance Capitalism - the Fight for the Future At the New Frontier of Power. Profile Books.

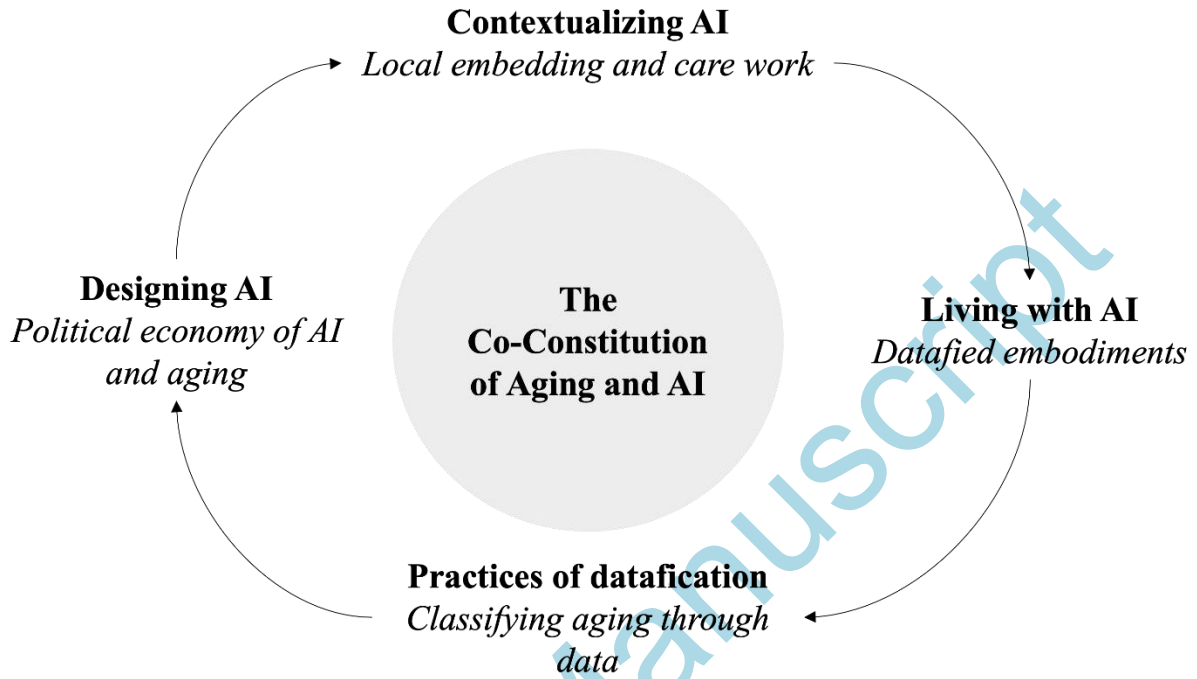
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Figures

Figure 1. Model of the Co-Constitution of Aging and AI (developed further from Peine & Neven 2020)

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Figure 1



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