

POLICY FORUM

Large AI models are cultural and social technologies

Implications draw on the history of transformative information systems from the past

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Debates about artificial intelligence (AI) tend to revolve around whether large models are intelligent, autonomous agents. Some AI researchers and commentators speculate that we are on the cusp of creating agents with artificial general intelligence (AGI), a prospect anticipated with both elation and anxiety. There have also been extensive conversations about cultural and social consequences of large models, orbiting around two foci: immediate effects of these systems as they are currently used, and hypothetical futures when these systems turn into AGI agents - perhaps even superintelligent AGI agents. But this discourse about large models as intelligent agents is fundamentally misconceived. Combining ideas from social and behavioral sciences with computer science can help us understand AI systems more accurately. Large Models should not be viewed primarily as intelligent agents, but as a new kind of cultural and social technology, allowing humans to take advantage of information other humans have accumulated.

The new technology of large models combines important features of earlier technologies. Like pictures, writing, print, video, Internet search, and other such technologies, large models allow people to access information that other people have created. Large Models - currently language, vision, and multi-modal - depend on the fact that the Internet has made the products of these earlier technologies readily available in machine-readable form. But like economic markets, state bureaucracies, and other social technologies, these systems not only make information widely available, they allow it to be reorganized, transformed, and restructured in distinctive ways. Adopting Herbert Simon's terminology (1), large models are a new variant of the "artificial systems of human society" that process information to enable large-scale coordination.

Our central point here is not just that these technological innovations, like all other innovations, will have cultural and social consequences. Rather we argue that Large Models are themselves best understood as a particular

type of cultural and social technology. They are analogous to such past technologies as writing, print, markets, bureaucracies, and representative democracies. Then we can ask the separate question about what the effects of these systems will be. New technologies that aren't themselves cultural or social, such as steam and electricity, can have cultural effects. Genuinely new cultural technologies, Wikipedia for example, may have limited effects. However, many past cultural and social technologies also had profound, transformative effects on societies, for good and ill, and this is likely to be true for Large Models.

These effects are markedly different from the consequences of other important general technologies such as steam or electricity. They are also different from what we might expect from hypothetical AGI. Reflecting on past cultural and social technologies and their impact will help us understand the perils and promise of AI models better than worrying about superintelligent agents.

SOCIAL & CULTURAL INSTITUTIONS

For as long as there have been humans, we have depended on culture. Beginning with language itself, human beings have had distinctive capacities to learn from the experiences of other humans and these capacities are arguably the secret of human evolutionary success. Major technological changes in these capacities have led to dramatic social transformations. Spoken language was succeeded by pictures, then by writing, print, film, and video. As more and more information became available across wider gulfs of space and time, new ways of accessing and organizing that information also developed, from libraries to newspapers to Internet search. These developments have had profound effects on human thought and society, for better or worse. 18th century advances in print technology, for example, which allowed new ideas to quickly spread, played an important role in the Enlightenment and the French Revolution. A landmark transformation occurred around 2000 when nearly all the information from text, pictures, and moving images was converted into digital formats - it could be instantly transmitted and infinitely reproduced.

As long as there have been humans, we have also relied on social institutions to coordinate individual information-gathering and decision-making. These institutions can themselves

be thought of as a kind of technology (1). In the modern era, markets, democracies, and bureaucracies have been particularly important. The economist Friedrich Hayek argued that the market's price mechanism generates dynamic summaries of enormously complex and otherwise unfathomable economic relations (2). Producers and buyers do not need to understand the complexities of production: all they need to know is the price, which compresses vast swathes of detail into a simplified but usable representation. Election mechanisms in democratic regimes focus distributed opinion toward collective legal and leadership decisions in a related way. The anthropologist James Scott argued (3) that all states, democratic or otherwise, have managed complex societies by creating bureaucratic systems that categorize and systematize information. Markets, democracies, and bureaucracies have relied on mechanisms that generate lossy (i.e., incomplete, selective, and uninvertible) but useful representations well before the computer. Those representations both depend on and go beyond the knowledge and decisions of individual people. A price, an election result, or a measure like gross domestic product (GDP) summarizes large amounts of individual knowledge, values, preferences and actions. At the same time, these social technologies can also themselves shape individual knowledge and decision-making.

The abstract mechanisms of a market, state, or bureaucracy, like cultural media, can influence individual lives in crucial ways, sometimes for the worse. Central banks, for example, reduced the complexities of the financial economy down to a few key variables. This provided apparent financial stability but at the cost of allowing instabilities to build up in the housing market, which central banks paid little attention to, precipitating the 2008 global financial crisis (4). Similarly, markets may not represent "externalities" like harmful carbon emissions. Integrating such information into prices through, e.g., a carbon tax can help but requires state action.

Humans rely extensively on these cultural and social technologies. These technologies are only possible, however, because humans have distinct capacities characteristic of intelligent agents. Humans, and other animals, can perceive and act on a changing external world, build new models of that world, revise those models as they accumulate more evidence,

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1 and then design novel goals. Individual humans
2 can create novel beliefs and values and convey
3 those beliefs and values through language,
4 print, etc., to others. Cultural and social tech-
5 nologies transmit and organize those beliefs
6 and values in powerful ways, but without those
7 individual capacities, the cultural and social
8 technologies would have no purchase. With-
9 out innovation, there would be no point to im-
10 itation (5).

11 Some AI systems, in robotics for example,
12 do attempt to instantiate similar truth-finding
13 abilities. There is no reason, in principle, why
14 an artificial system could not do so at some
15 point in the future. Human brains do, after all.
16 But at the moment all such systems are very far
17 from these human capacities. We can debate
18 how much to worry now about these potential
19 future AI systems, or how we might handle
20 them if and when they emerge. But this is very
21 different from the question of the effects of
22 Large Models at present and in the immediate
23 future

24 LARGE MODELS

25 Large Models, unlike more agentive systems,
26 have made remarkable and surprising progress
27 in the past few years, making them the focus of
28 the current conversation about “AI” in general.
29 This progress has led to claims that “scaling”,
30 simply taking the current designs and increas-
31 ing the amount of data and computing power
32 they use, will lead to AGI agents in the near fu-
33 ture. But Large Models are fundamentally dif-
34 ferent from intelligent agents and “scaling”
35 won’t change this. For example, “hallucina-
36 tions” are an endemic problem in these sys-
37 tems because they have no conception of truth
38 and falsity (although there are practical steps
39 toward mitigation). They simply sample and
40 generate text and images.

41 Rather than being intelligent agents, Large
42 Models combine the features of cultural and
43 social technologies in a new way. They gener-
44 ate summaries of unmanageably large and
45 complex bodies of human-generated infor-
46 mation. But these systems do not merely sum-
47 marize this information, like library catalogs, In-
48 ternet search, and Wikipedia. They also can
49 reorganize and reconstruct representations or
50 “simulations” (1) of this information at scale
51 and in novel ways, like markets, states and bu-
52 reaucracies. Just as market prices are lossy rep-
53 resentations of the underlying allocations and
54 uses of resources, and government statistics
55 and bureaucratic categories imperfectly repre-
56 sent the characteristics of underlying popula-
57 tions, so too Large Models are ‘lossy JPEGs’ (6)
58 of the data corpora on which they have been
59 trained.

Because it is hard for humans to think

clearly about large-scale cultural and social
technologies, we have tended to think of them
in terms of agents. Stories are a particularly
powerful way to pass on information, and from
fireside tales to novels to video games, they
have done this by creating illustrative fictional
agents, even though listeners know that those
agents aren’t real. Chatbots are the successor
to Hercules, Anansi, and Peter Rabbit. Similarly,
it is easy to treat markets and states as if they
were agents, and agencies or companies can
even have a kind of legal personhood.

But behind their agent-like interfaces and
anthropomorphic pretensions, Large Language
Models (LLM) and Large Multi-modal Models
are statistical models that take enormous cor-
pora of text produced by humans, break them
down into particular words, and estimate the
probability distribution of long word se-
quences. This is an imperfect representation of
language but contains a surprisingly large
amount of information about the patterns it
summarizes. It allows the LLM to predict which
words come next in a sequence, and so gener-
ate human-like text. Large Multi-modal Models
do the same with audio, image, and video data.
Large Models not only abstract a very large
body of human culture; they also allow a wide
variety of new operations to be carried out on
it. LLMs can be prompted to carry out complex
transformations of the data on which they are
trained. Simple arguments can be expressed in
flowery metaphors, while ornate prose can be
condensed into plain language. Similar tech-
niques enable related models to generate
novel pictures, songs, and video in response to
prompts. A body of cultural information that
was previously too complex, large and inchoate
for large-scale operations has been rendered
tractable.

In practice, the most recent versions of
these systems depend not only on massive
caches of text and images generated and cu-
rated by humans but also on human judgment
and knowledge in other forms. In particular,
the systems rely on reinforcement learning
from human feedback (RLHF) or its variants—
tens of thousands of human employees pro-
vide ratings of model outputs. They also de-
pend on prompt engineering—humans must
use both their background knowledge and in-
genuity to extract useful information from the
models. Even the newest ‘chain of thought’
models regularly begin from dialogue with their
human users.

The relatively simple though powerful algo-
rithms that allow large models to extract statis-
tical patterns from text are not really the key to
the models’ success. Instead modern AI rests
atop libraries, the Web, tens of thousands of
human coders, and a growing international

world of active users. Someone asking a bot for
help writing a cover letter for a job application
is really engaging in a technically mediated re-
lationship with thousands of earlier job appli-
cants and millions of other letter writers, RLHF
workers, etc.

CHALLENGES & OPPORTUNITIES

The AI debate should focus on the challenges
and opportunities that these new cultural and
social technologies generate. We now have a
technology that does for written and pictured
culture, what large-scale markets do for the
economy, what large-scale bureaucracy does
for society, and perhaps even comparable to
what print once did for language. What hap-
pens next? Like past economic, organizational,
and informational “general purpose technolo-
gies”, these systems will have implications for
productivity (7), complementing human work
but also automating tasks that only humans
could previously perform, and for distribution,
affecting who gets what (8).

Yet they will also have wider and more pro-
found cultural consequences. We don’t yet
know if these consequences will be as great as
those of earlier technologies like print, markets,
or bureaucracies, but thinking of them as cul-
tural technologies increases rather than de-
creases their potential impact. These earlier
technologies were central to the extensive so-
cial transformations of the 18th and 19th cen-
turies, both as causes and effects. All these
technologies, like Large Models, supported the
abstraction of information so that new kinds of
operations could be carried out at scale. All pro-
voked justified concerns about the spread of
misinformation and bias, cultural homogeniza-
tion or fragmentation, and shifts in the distribu-
tion of power and resources. The emergence of
new communications media, including both
print and television, was accompanied by rea-
sonable worries that the new media would
spread misinformation and strengthen malign
cultural forces. Similarly, the categorization
schemes that bureaucracies and markets de-
ploy often embed oppressive assumptions.

At the same time, these technologies gener-
ated new possibilities for recombining infor-
mation and coordinating actions among mil-
lions of people at a planetary scale. Emerging
debates over the social, economic and political
consequences of LLMs continue deep-rooted
historical worries and hopes about new cultural
and social technologies. Orienting these de-
bates requires both recognizing the commonal-
ities between new arguments and old ones and
carefully mapping the particulars of the new
and evolving technologies.

Such mapping is among the central tasks of
the social sciences, which emerged from the

1 social, economic, and political upheavals of the
2 Industrial Revolution and its aftermath. Social
3 scientists' investigation of the consequences of
4 these past technologies can help us think about
5 less obvious social implications of AI, both neg-
6 ative and positive, and to consider ways that AI
7 systems could be redesigned to increase the
8 positive impacts and reduce the negative. As
9 media, markets, and bureaucratic technologies
10 expanded in the nineteenth and 20th centu-
11 ries, they generated economic losers and win-
12 ners, displacing whole categories of workers,
13 from clerks and typists to human "computers".
14 So too, there are obvious worries that large
15 models and related technologies may displace
16 "knowledge workers".

17 There are also less obvious questions. Will
18 Large Models homogenize or fragment culture
19 and society? Thinking about this in historical
20 context can be particularly illuminating. Cur-
21 rent concerns resemble nineteenth and twen-
22 tieth-century disagreements over markets and
23 bureaucracies. Max Weber worried (9) about
24 the deadening homogenizing consequences of
25 economic and bureaucratic "rationalization,"
26 while John Stuart Mill (10) thought, on the con-
27 trary, that market exchanges would expose
28 participants to different forms of life and sof-
29 ten impulses to conflict ("doux commerce").

30 Large Models are designed to work well—
31 to faithfully reproduce the actual probabilities
32 of sequences of text, images, and video—on
33 average. They therefore have an intrinsic ten-
34 dency to be most accurate in situations most
35 commonly found in their training data and
36 least accurate in situations which were rare in
37 data or entirely novel. This might lead Large
38 Models to worsen the kind of homogenization
39 that haunted Weber.

40 On the other hand, Large Models may al-
41 low us to design new ways to harvest the diver-
42 sity of the cultural perspectives they summa-
43 rize. Combining and balancing these
44 perspectives may provide more sophisticated
45 means of solving complex problems (11). One
46 way to do this may be to build "society-like"
47 ecologies in which different perspectives, en-
48 coded in different Large Models, debate each
49 other and potentially cross-fertilize to create
50 hybrid perspectives (12) or to identify gaps in
51 the space of human expertise (13) that might
52 usefully be bridged. Large Models are surpris-
53 ingly effective at abstracting subtle and non-
54 obvious patterns in texts and images. This sug-
55 gests that such technologies could be used to
56 find novel patterns in text and images that
57 crisscross the space of human knowledge and
58 culture, including patterns invisible to any par-
59 ticular human. We may require new systems
that diversify Large Model reflections and per-
sonas, and produce the same distribution and

diversity as do human societies.

Diversifying systems like this might be par-
ticularly important for scientific progress. For-
mal science itself depended on the emergence
of the new cultural technologies of the 17th
and 18th centuries, from coffee houses and
rapid mail to journals and peer review. AI tech-
nologies have the potential to accelerate sci-
ence further, but this will depend on imagina-
tive ways of using and rethinking these
technologies. By wiring together so many per-
spectives across text, audio, and images, Large
Models may allow us to discover unprece-
dented connections between them for the
benefit of science and society. These technol-
ogies have most commonly been trained to re-
gurgitate routine information as helpful assis-
tants. A more fundamental set of possibilities
might open up if we deployed them as maps to
explore formerly uncharted territory.

There are also less obvious and more inter-
esting ways that new cultural and social tech-
nologies influence economic relationships. The
development of cultural technologies leads to
a fundamental economic tension between the
people who produce information and the sys-
tems that distribute it. Neither group can exist
without the other: a writer needs publishers as
much as the publisher need writers. But their
economic incentives push in opposite direc-
tions. The distributors will profit if they can ac-
cess the producer's information cheaply, while
the producers will profit if they can get their in-
formation distributed cheaply. This tension has
always been a feature of new cultural technol-
ogies. The ease and efficiency of distributing in-
formation in digital form has already made this
problem especially acute, as evidenced by the
crisis in everything from local newspapers to
academic journals. But the very speed, effi-
ciency and scope of Large Models, processing
all the available information at once, combined
with the centralized ownership of those mod-
els, makes these problems loom especially
large. Concentrated power may make it easier
for those who own the systems to skim the
benefits of efficiency at the expense of others.

There are crucial technical questions: to
what extent can the systematic imperfections
of Large Models be remedied, and when are
they better or worse than the imperfections of
systems based around human knowledge
workers? Those should not overshadow the
crucial political questions: which actors are ca-
pable of mobilizing around their interests, and
how might they shape the resulting mix of tech-
nology and organizational capacities? Very of-
ten, commentators within the technology sec-
tor reduce these questions into a simple battle
between machines and humans. Either the
forces of 'progress' will prevail against

retrograde Luddite tendencies, or on the other
hand, human beings will successfully resist the
inhuman encroachment of artificial technol-
ogy. Not only does this fail to appreciate the
complexities of past distributional struggles,
struggles that long predate the computer, but
it ignores the many different possible paths
that future progress might take, each with its
own mix of technological possibilities and
choices (8).

In the case of earlier social and cultural
technologies, a range of further institutions, in-
cluding normative and regulatory institutions,
emerged to temper their effects. These ranged
from editors, peer review, and libel laws for
print, to election law, deposit insurance and the
Securities and Exchange Commission for mar-
kets, democracies, and bureaucracies. These
institutions had varied effectiveness and re-
quired continual revision. These countervailing
forces did not emerge on their own, however,
but resulted from concerted and sustained ef-
forts by actors both within and outside the
technologies themselves.

LOOKING FORWARD

The narrative of AGI, of large models as super-
intelligent agents, has been promoted both
within the tech community and outside it, both
by AI optimist "boomers" and more concerned
"doomers". This narrative gets the nature of
these models and their relation to past techno-
logical changes wrong. But more importantly, it
actively distracts from the real problems and
opportunities that these technologies pose,
and the lessons history can teach us about how
to ensure that the benefits outweigh the costs.

Of course, as we note above, there may be
hypothetical future AI systems that are more
like intelligent agents and we might debate
how we should deal with these hypothetical
systems, but LLM's are not such systems, any
more than were library card catalogs or the
Web. Like catalogs and the Web, Large Models
are part of a long history of cultural and social
technologies.

The social sciences have explored this his-
tory in detail, generating a unique understand-
ing of past technological upheavals. Bringing
computer science and engineering into close
cooperation with the social sciences will help us
understand this history and apply these les-
sons. Will large models lead to greater cultural
homogeneity or greater fragmentation? Will
they reinforce or undermine the social institu-
tions of human discovery? As they reshape the
political economy, who will win and lose?
These and other urgent questions do not come
into focus in debates that treat Large Models as
analogs for human agents.

Changing the terms of debate would lead to

1 better research. It would be far easier for social
2 scientists and computer scientists to cooperate
3 and combine their respective strengths if both
4 understood that LMs are no more - but also no
5 less - than a new kind of cultural and social
6 technology. Computer scientists could bring to-
7 gether their deep understanding of how these
8 systems work with social scientists' compre-
9 hension of how other such large-scale systems
10 have reshaped society, politics, and the econ-
11 omy in previous eras, elaborating existing re-
12 search agendas and discovering new ones. This
13 would help remedy past confusions in which
14 computer scientists have adopted overly sim-
15 plified notions of complex social phenomena
16 (14) while social scientists have failed to under-
17 stand the complex functioning of these new
18 technologies.

19 It would move policy discussions over AI
20 decisively away from simplistic battles be-
21 tween the existential fear of a machine take-
22 over and the promise of a near-future paradise
23 in which everyone will have a perfectly reliable
24 and competent artificial assistant. The actual
25 policy consequences of LMs will surely be dif-
26 ferent. Like markets and bureaucracies, they
27 will make some kinds of knowledge more visi-
28 ble and tractable than they were in the past,
29 encouraging policymakers to focus on the new
30 things that they can measure and see at the ex-
31 pense of those less visible and more confusing.
32 As a result, reflecting past cases of markets and
33 media, power and influence will shift towards
34 those who can fully deploy these technologies
35 and away from those who cannot. AI weakens
36 the position of those upon whom it is used and
37 who provide its data, strengthening AI experts
38 and policymakers (14).

39 Finally, thinking in this way might reshape
40 AI practice. Engineers and computer scientists
41 are already aware of the problem of Large
42 Model bias, and are thinking about their rela-
43 tionship to ethics and justice. They should go
44 further. How will these systems affect who gets
45 what? What will their practical consequences
46 be for societal polarization and integration?
47 Can they be developed to enhance human cre-
48 ativity rather than to dull it? Finding practical
49 answers to such questions will require an un-
50 derstanding of social science as well as engi-
51 neering. Shifting the debate about AI away
52 from agents and toward cultural and social
53 technologies is a crucial first step toward build-
54 ing that cross-disciplinary understanding (15).

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Acknowledgments

All authors contributed equally

10.1126/science.adt9819