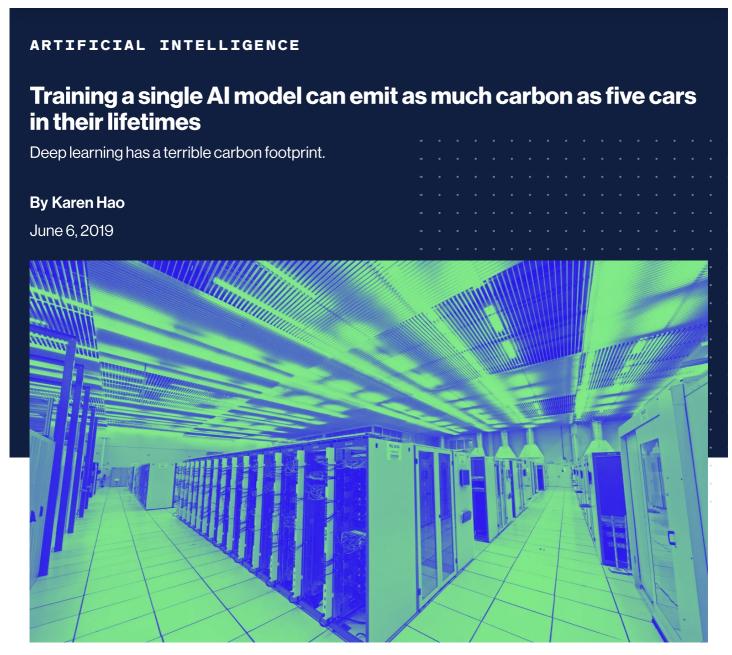


Subscribe



A data center

DEAN MOUHTAROPOULOS | GETTY; EDITED BY MIT TECHNOLOGY REVIEW

The <u>artificial-intelligence</u> industry is often compared to the oil industry: once mined and refined, data, like oil, can be a highly lucrative commodity. Now it seems the metaphor may extend even further. Like its fossil-fuel counterpart, the process of deep learning has an outsize environmental impact.

In a <u>new paper</u>, researchers at the University of Massachusetts, Amherst, performed a life cycle assessment for training several common large AI models. They found that the process

MIT Technology Review

Common carbon footprint benchmarks

in lbs of CO2 equivalent

Roundtrip flight b/w NY and SF (1 passenger)

Human life (avg. 1 year)

American life (avg. 1 year)

US car including fuel (avg. 1 lifetime)

Transformer (213M parameters) w/ neural architecture search

Chart: MIT Technology Review • Source: Strubell et al. • Created with Datawrapper

It's a jarring quantification of something AI researchers have suspected for a long time. "While probably many of us have thought of this in an abstract, vague level, the figures really show the magnitude of the problem," says Carlos Gómez-Rodríguez, a computer scientist at the University of A Coruña in Spain, who was not involved in the research. "Neither I nor other researchers I've discussed them with thought the environmental impact was that substantial."

The carbon footprint of natural-language processing

THE DOWNLOAD	×
Sign up for your daily dose of what's up in emerging technology.	
Enter your email	
	Sign up

By signing up, you agree to our Privacy Policy.

The paper specifically examines the model training process for <u>natural-language processing</u> (NLP), the subfield of AI that focuses on teaching machines to handle human language. In the last two years, the NLP community has reached several noteworthy performance milestones in machine translation, sentence completion, and other standard benchmarking tasks. <u>OpenAI's infamous GPT-2 model</u>, as one example, excelled at writing convincing fake news articles.

But such advances have required training ever larger models on sprawling data sets of sentences scraped from the internet. The approach is computationally expensive—and highly energy intensive.

MIT Technology Review

Subscribe

single GPU for up to a day to measure its power draw. They then used the number of training

hours listed in the model's original papers to calculate the total energy consumed over the complete training process. That number was converted into pounds of carbon dioxide equivalent based on the average energy mix in the US, which closely matches the energy mix used by Amazon's AWS, the largest cloud services provider.

The estimated costs of training a model once

In practice, models are usually trained many times during research and development.

	original consumption footpri		Carbon footprint (lbs of CO2e)	Cloud compute cost (USD)			
Transformer (65M parameters)	Jun, 2017	27	26	\$41-\$140			
Transformer (213M parameters)	Jun, 2017	201	192	\$289-\$981			
ELMo	Feb, 2018	275	262	\$433-\$1,472			
BERT (110M parameters)	Oct, 2018	1,507	1,438	\$3,751-\$12,571			
Transformer (213M parameters) w/ neural architecture search	Jan, 2019	656,347	626,155	\$942,973-\$3,201,722			
GPT-2	Feb, 2019	-	-	\$12,902-\$43,008			

Note: Because of a lack of power draw data on GPT-2's training hardware, the researchers weren't able to calculate its carbon footprint.

Table: MIT Technology Review • Source: Strubell et al. • Created with Datawrapper

They found that the computational and environmental costs of training grew proportionally to model size and then exploded when additional tuning steps were used to increase the model's final accuracy. In particular, they found that a tuning process known as neural architecture search, which tries to optimize a model by incrementally tweaking a neural network's design through exhaustive trial and error, had extraordinarily high associated costs for little performance benefit. Without it, the most costly model, BERT, had a carbon footprint of roughly 1,400 pounds of carbon dioxide equivalent, close to a round-trip trans-America flight

MIT Technology Review

Subscribe

"Training a single model is the minimum amount of work you can do," says Emma Strubell, a PhD candidate at the University of Massachusetts, Amherst, and the lead author of the paper.

In practice, it's much more likely that AI researchers would develop a new model from scratch or adapt an existing model to a new data set, either of which can require many more rounds of training and tuning.

To get a better handle on what the full development pipeline might look like in terms of carbon footprint, Strubell and her colleagues used a model they'd produced in a <u>previous paper</u> as a case study. They found that the process of building and testing a final paper-worthy model required training 4,789 models over a six-month period. Converted to CO₂ equivalent, it emitted more than 78,000 pounds and is likely representative of typical work in the field.

The significance of those figures is colossal—especially when considering the current trends in AI research. "In general, much of the latest research in AI neglects efficiency, as very large neural networks have been found to be useful for a variety of tasks, and companies and institutions that have abundant access to computational resources can leverage this to obtain a competitive advantage," Gómez-Rodríguez says. "This kind of analysis needed to be done to raise awareness about the resources being spent [...] and will spark a debate."

"What probably many of us did not comprehend is the scale of it until we saw these comparisons," echoed Siva Reddy, a postdoc at Stanford University who was not involved in the research.

The privatization of AI research

The results underscore another growing problem in AI, too: the sheer intensity of resources now required to produce paper-worthy results has made it increasingly challenging for people working in academia to continue contributing to research.



The Gender issue

Subscribe and learn how science and technology have long pushed the boundaries of sex and gender.

See offers

MIT Technology Review

Subscribe

"This trend toward training huge models on tons of data is not feasible for academics—grad students especially, because we don't have the computational resources," says Strubell. "So

there's an issue of equitable access between researchers in academia versus researchers in industry."

Strubell and her coauthors hope that their colleagues will heed the paper's findings and help level the playing field by investing in developing more efficient hardware and algorithms.

Reddy agrees. "Human brains can do amazing things with little power consumption," he says. "The bigger question is how can we build such machines."

by Karen Hao









Continue reading more stories

Subscribe now for unlimited access.

Subscribe

DEEP DIVE

ARTIFICIAL INTELLIGENCE

MIT Technology Review

MIT Technology Review

Why business is booming for military Al startups

The invasion of Ukraine has prompted militaries to update their arsenals—and Silicon Valley stands to capitalize.

By Melissa Heikkilä

2022/8/29 17:48

MIT Technology Review

Subscribe

7/9

Get the latest updates from MIT Technology Review

Discover special offers, top stories, upcoming events, and more.

Enter your email										
										\rightarrow
Privacy Policy		-		٠	٠	٠		•	٠	•

MIT Technology Review

Our in-depth reporting reveals what's going on now to prepare you for what's coming next.

Subscribe to support our journalism.

About us Help & FAQ

Careers My subscription

Custom content Editorial guidelines

Advertise with us Privacy policy

International Editions Terms of Service

Republishing Write for us

MIT News Contact us

MIT Technology Review

© 2022 MIT Technology Review