A satellite view of Earth, showing the Middle East, North Africa, and parts of Europe and Asia. The image is dark, with the text overlaid in white.

Is AI good or bad for the climate? ...it's complicated

David Rolnick

McGill University
Mila – Quebec AI Institute
Climate Change AI

Climate change

- Increasingly severe effects
- Disproportionate impact on disadvantaged communities
- How bad it gets depends on what we do now
- Need net-zero greenhouse gas emissions by 2050 (IPCC)
- Action encompasses both **mitigation** (reducing greenhouse gas emissions) and **adaptation** (resilience to consequences)



**AI applications relevant
to climate change
mitigation / adaptation**

**AI applications that
increase greenhouse gas
emissions**

**AI applications with
uncertain or systemic
impacts**

**Emissions impacts of
AI computation
and hardware**

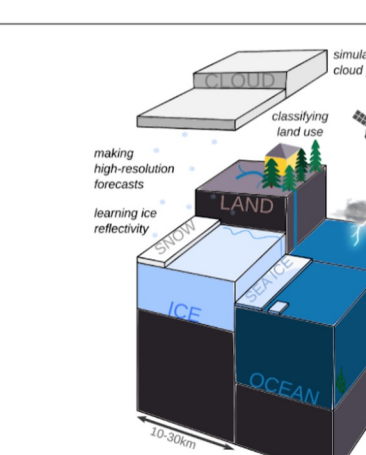
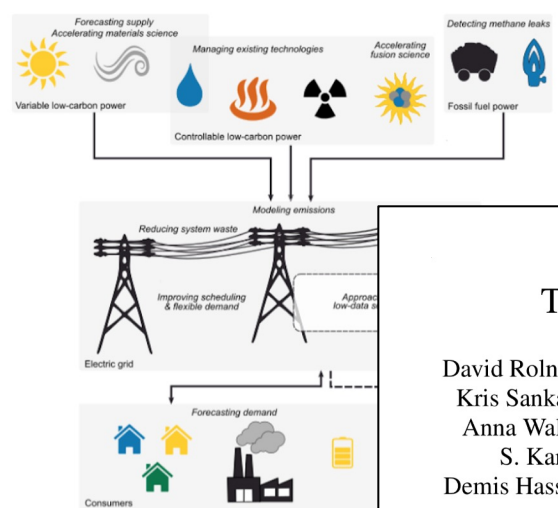
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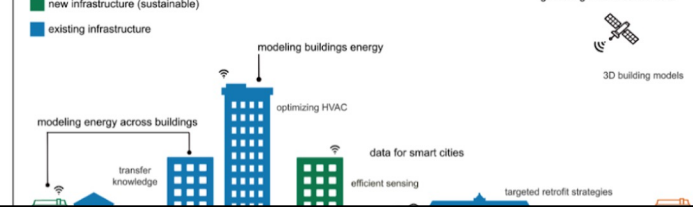
**Emissions impacts of
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Electricity systems



Climate prediction

Buildings

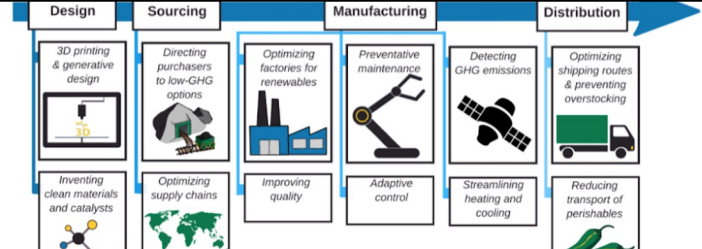


Industry

Tackling Climate Change with Machine Learning

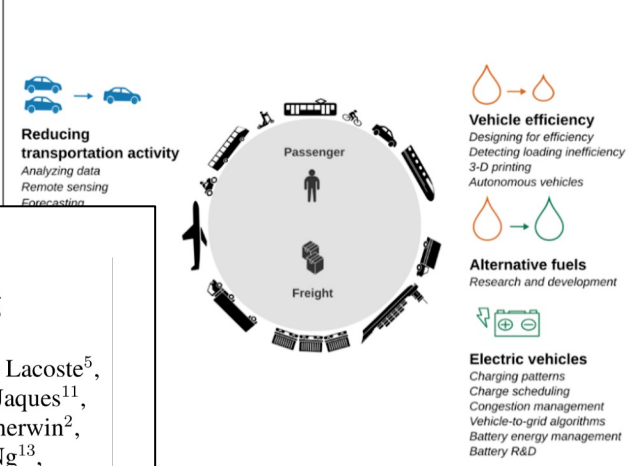
David Rolnick^{1*}, Priya L. Donti², Lynn H. Kaack³, Kelly Kochanski⁴, Alexandre Lacoste⁵, Kris Sankaran^{6,7}, Andrew Slavin Ross⁸, Nikola Milojevic-Dupont^{9,10}, Natasha Jaques¹¹, Anna Waldman-Brown¹¹, Alexandra Luccioni^{6,7}, Tegan Maharaj^{6,7}, Evan D. Sherwin², S. Karthik Mukkavilli^{6,7}, Konrad P. Kording¹, Carla Gomes¹², Andrew Y. Ng¹³, Demis Hassabis¹⁴, John C. Platt¹⁵, Felix Creutzig^{9,10}, Jennifer Chayes¹⁶, Yoshua Bengio^{6,7}

¹University of Pennsylvania, ²Carnegie Mellon University, ³ETH Zürich, ⁴University of Colorado Boulder, ⁵Element AI, ⁶Mila, ⁷Université de Montréal, ⁸Harvard University, ⁹Mercator Research Institute on Global Commons and Climate Change, ¹⁰Technische Universität Berlin, ¹¹Massachusetts Institute of Technology, ¹²Cornell University, ¹³Stanford University, ¹⁴DeepMind, ¹⁵Google AI, ¹⁶Microsoft Research

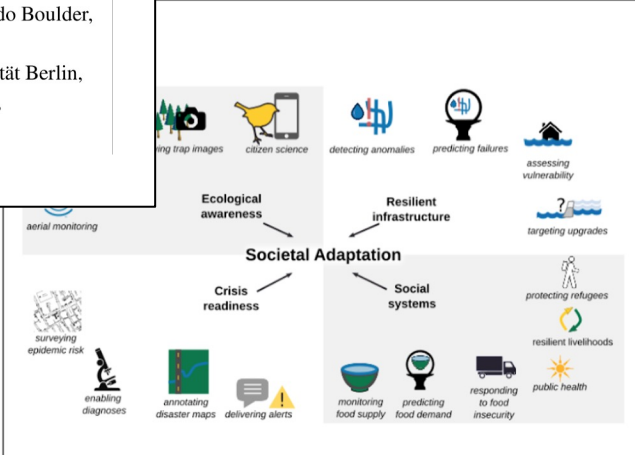


Industry

Transportation

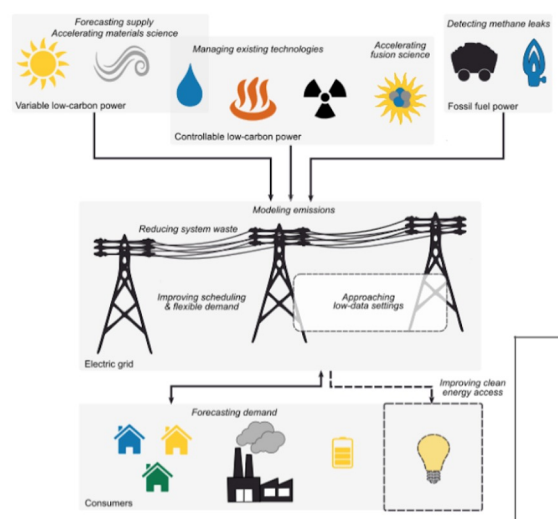


Societal adaptation

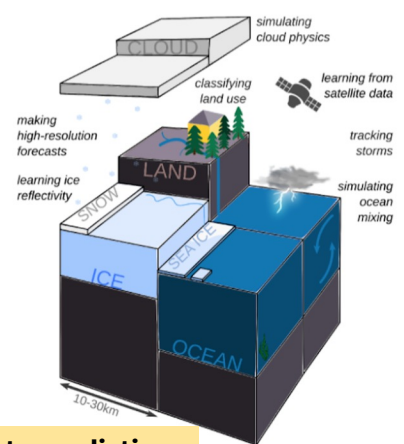


Societal adaptation

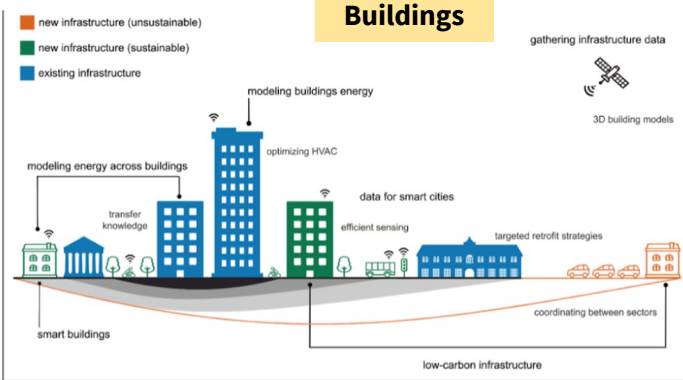
Electricity systems



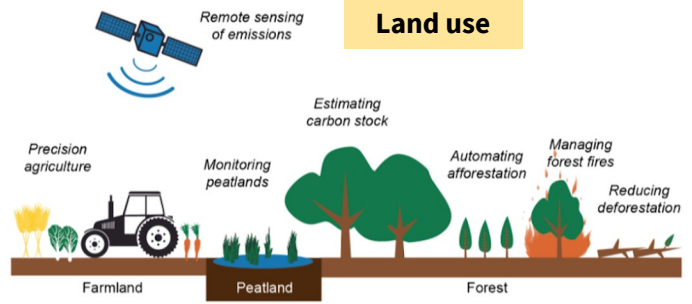
Climate prediction



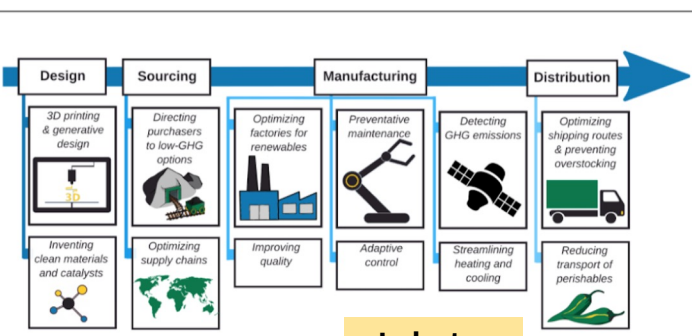
Buildings



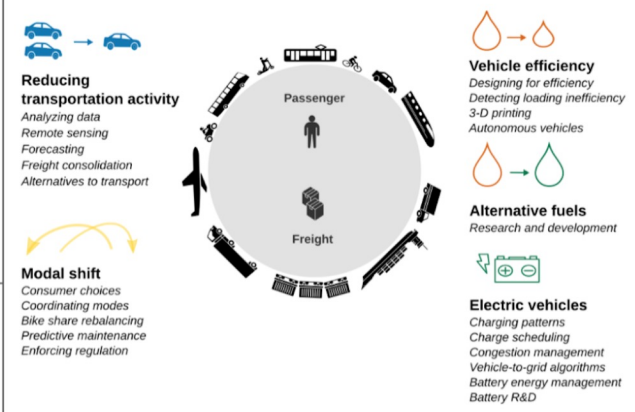
Land use



Industry



Transportation



Societal adaptation



Opportunities for AI in climate action

- ▶ Improving operational efficiency
e.g. optimizing HVAC control or steel/cement manufacture
- ▶ Gathering information
e.g. estimating carbon stock or parsing financial disclosures
- ▶ Forecasting
e.g. nowcasting electricity supply or predicting demand
- ▶ Speeding up simulations
e.g. emulating parts of climate models or grid planning models
- ▶ Accelerating scientific discovery
e.g. suggesting materials for use in batteries and perovskites

See: Tackling Climate Change with Machine Learning, ACM Computing Surveys 2022.

Opportunities for AI in climate action

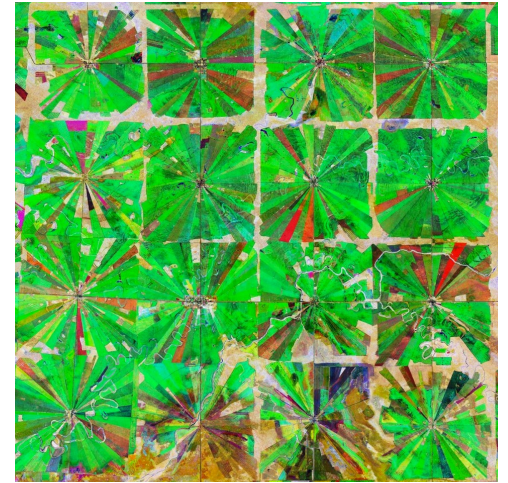
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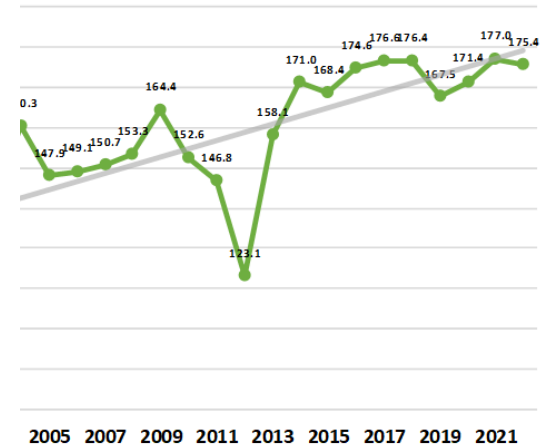
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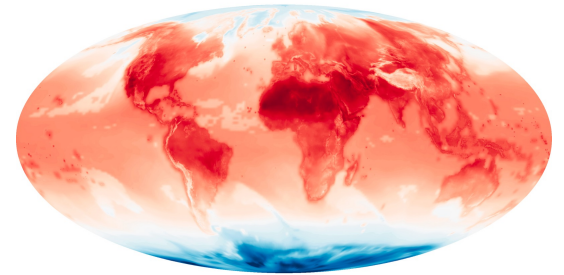
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Efficient operation of electrical grids

AC Optimal Power Flow (ACOPF): nonconvex optimization problem to determine power to produce at each generator in a power grid

Exact solutions take too long, so grid operators simplify the problem and waste power, especially w/ solar & wind

Naïve AI algorithms can approximately solve ACOPF fast, but violate power flow constraints, risking blackouts

We show how to enforce these constraints, solving ACOPF problems 10-100x faster than traditional methods without violating power flow

$$\begin{aligned} & \underset{p_g \in \mathbb{R}^b, q_g \in \mathbb{R}^b, v \in \mathbb{C}^b}{\text{minimize}} && p_g^T A p_g + b^T p_g \\ & \text{subject to} && p_g^{\min} \leq p_g \leq p_g^{\max} \\ & && q_g^{\min} \leq q_g \leq q_g^{\max} \\ & && v^{\min} \leq |v| \leq v^{\max} \\ & && (p_g - p_d) + (q_g - q_d)i = \text{diag}(v) \overline{W} \overline{v}. \end{aligned}$$

$$\begin{aligned} & \min_y f_x(y) \\ & \text{s.t.} \quad g_x(y) \leq 0 \\ & \quad \quad h_x(y) = 0 \end{aligned}$$

P. Donti, D. Rolnick, Z. Kolter, *DC3: A learning method for optimization with hard constraints*, ICLR 2021.

Gathering information on biodiversity

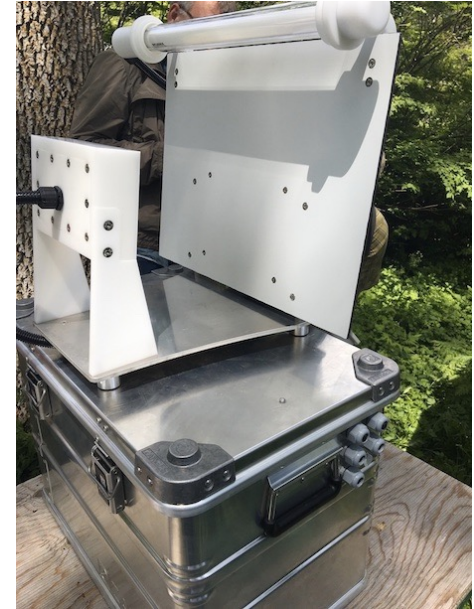
Ecosystems are collapsing, but evaluating biodiversity requires specialized experts

AI can help scale up ecological monitoring

We're developing automated sensors to monitor insect populations, with a coalition of partner ecologists

Solar-powered device attracts & photographs insects, and AI algorithms identify them

Data then sent to experts for interpretation and proof-reading



Partners include: Aarhus University, Montreal Insectarium, eButterfly, UK Centre for Ecology & Hydrology, Naturalis, Université de Sherbrooke, Université Laval, Natural Resources Canada

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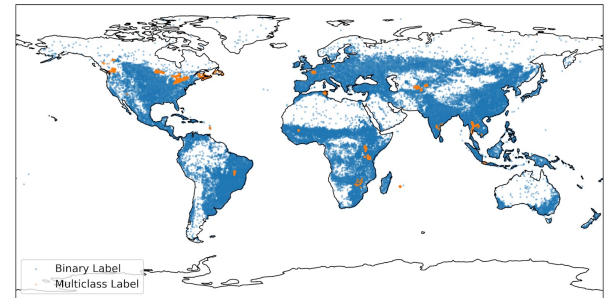
Forecasting crop yield

Forecasting crop yield is essential to averting food insecurity in a changing climate

AI can help predict yield from satellite imagery

But it needs labeled data – and data can be scarce and uneven across different locations

In work with NASA Harvest, we develop meta-learning algorithms for crop yield prediction that can quickly adapt with minimal new data, by leveraging metadata on location and crop type



Gabriel Tseng, Hannah Kerner, David Rolnick, *TIML: Task-Informed Meta-Learning for agriculture*, preprint arXiv 2202.02124.

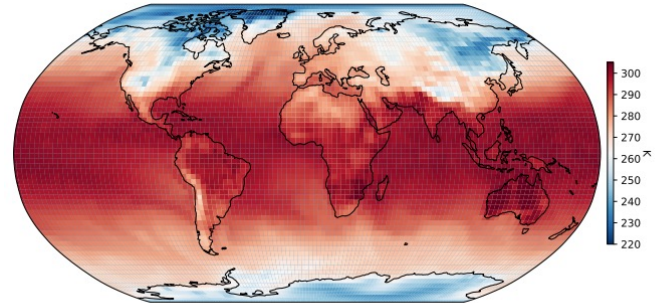
Speeding up climate simulations

Climate simulations are accurate but can be slow
(e.g. months even on a supercomputer)

This makes it harder to get localized predictions that help in adapting to climate change

With Environment & Climate Change Canada, we use AI to quickly approximate radiative transfer computations, an especially slow part of standard climate models

Our algorithms incorporate known physical relations to improve accuracy



Salva Rühling Cachay*, Venkatesh Ramesh*, Jason N. S. Cole, Howard Barker, David Rolnick, *ClimART: A Benchmark Dataset for Emulating Atmospheric Radiative Transfer in Weather and Climate Models*, NeurIPS 2021, and forthcoming work.

Key considerations

AI is never a silver bullet and is only relevant sometimes

Partnership between stakeholders with complementary expertise is crucial

High-impact applications are not always flashy

Even when working with data, sometimes simple methods work

Technosolutionism can be counterproductive or contribute to greenwashing

Equity considerations

- ▶ Empowering diverse stakeholders
- ▶ Selecting and prioritizing problems
- ▶ Ensuring data is representative

AI applications relevant
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Emissions impacts of
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AI applications that increase greenhouse gas emissions

- ▶ AI used to accelerate fossil fuel exploration and extraction
 - AI and advanced analytics estimated to yield \$400B+ additional profit for fossil fuel companies by 2025
 - Many leading technology companies have partnered on such uses
- ▶ Direct facilitation of other high-emissions activities, e.g. fast fashion
- ▶ AI is often used to optimize systems, but sometimes optimizing for cost \neq optimizing for emissions (e.g. if labor costs outweigh energy costs)

Further reading: Greenpeace “Oil in the Cloud.”



AI applications relevant
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Emissions impacts of
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AI applications with systemic impacts

Many AI applications have systemic impacts on climate action which are poorly quantified but likely significant

Consumer behavior:

- ▶ Advertising recommender systems designed to boost consumption may be greatly increasing GHG emissions
- ▶ Autonomous vehicles – may reduce emissions in public transportation context but increase miles driven (and emissions) for personal transport. *Depends on goals as the technology is developed.*



Credit: Grendelkhan, Wikimedia Commons

AI applications with systemic impacts

Many AI applications have systemic impacts on climate action which are poorly quantified but likely significant

Rebound effects

- ▶ Efficiency gains in a sector may be partially counterbalanced by corresponding increases in use
- ▶ E.g. lower energy used in making some consumer products

Lock-in effects

- ▶ Applications facilitated by technology may become more entrenched (can be positive or negative)

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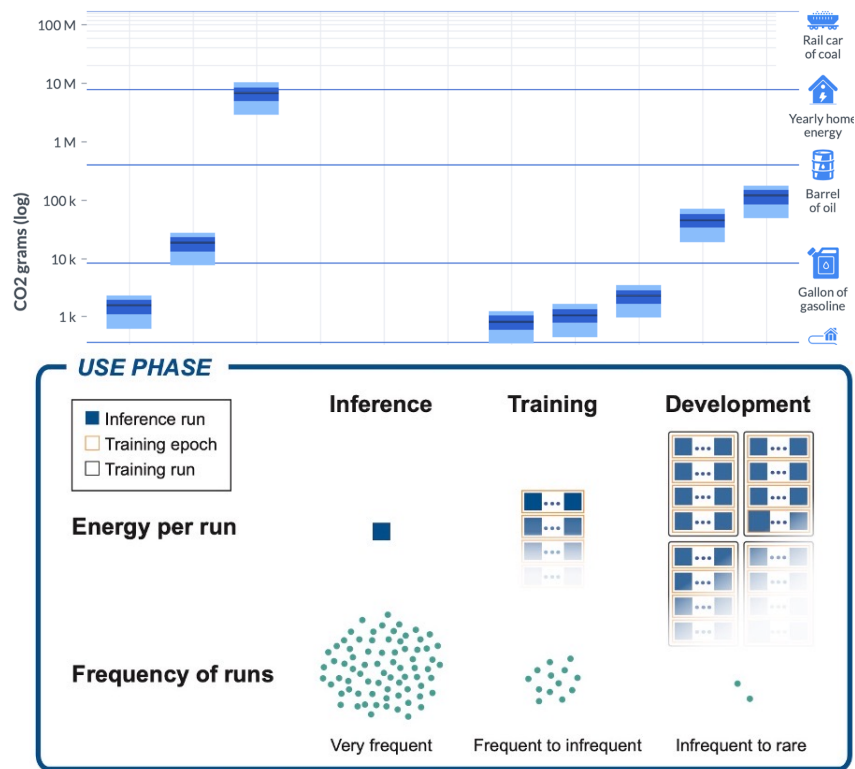
**Emissions impacts of
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Emissions impacts of AI computation and hardware

The ICT sector contributed 1 to 4 % of global GHG emissions in 2020, with two thirds from operational energy and one third from hardware

AI is some fraction of ICT; Google reports AI is 15% of server energy use
Highly variable between algorithms, the *biggest* AI algorithms are getting bigger (300,000x since 2012)

Machine learning algorithm lifecycle of development, training, inference



Emissions impacts of AI computation and hardware

Impact assessment needed for computation-related emissions, including cloud compute

That said, these effects are likely significantly smaller than application-related negative impacts

Major tech players may have incentive to focus attention on efficient computation, rather than also re-evaluating what algorithms *do* (Scope 1 & 2 vs Scope 3 emissions)



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Failures in the AI innovation landscape

AI continues to rely on benchmarks like ImageNet-1k to evaluate models and pre-train for applied settings.

Such benchmarks are often derived from Internet data, chosen & labeled without relevant experts in the room.

Example: We worked with ecologists to analyze the 27% of ImageNet-1k that is wild animals.

- 12% of the images are wrong, 12% of categories are contradictory.
- Species heavily biased towards United States.

Exemplary of AI innovations developed without stakeholders relevant to societal impact.



Alexandra Sasha Luccioni and David Rolnick,
Bugs in the Data: How ImageNet Misrepresents Biodiversity, preprint arXiv 2208.11695.

AI and climate change

- AI can either help or hinder climate action, depending on how it is used
- Overall impacts poorly understood and complex to measure, but may be significant and *can be shaped*
- Consideration of impacts and inclusion of relevant stakeholders must be part of AI innovation
- Aligning AI with climate action means more than adding “AI for Good” applications on top of business as usual – implicit choices matter

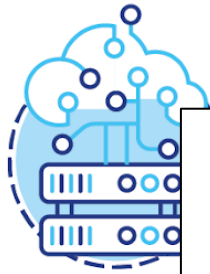
CLIMATE CHANGE AND AI

Recommendations for
Government Action

Global Partnership on AI Report

In collaboration with Climate Change AI and
the Centre for AI & Climate





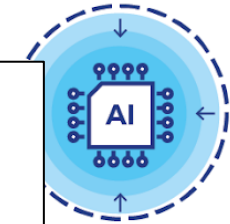
Data & digital infrastructure

Data, simulation environments, testbeds, libraries, computational hardware



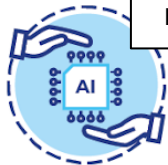
Selected policy recommendations

- Improve **data standards & data sharing** via task forces & platforms
- Ensure **impact-driven funding** for AI research & innovation
- Develop **cross-sectoral innovation centers** with private & public stakeholders to incubate projects & facilitate collaboration
- Build AI **capacity & literacy** in climate-relevant industries, government, & civil society, via upskilling & secondment programs
- Establish best practices for responsible & **participatory design**
- Consider potential **positive/negative climate impact** in shaping new technology development



Reducing AI's negative impacts on the climate

Application and mitigation-related impacts



Responsible AI



Capacity building



International collaboration



Impact assessment



Climate Change AI

Catalyzing impactful work at the intersection of climate change & ML

Digital resources

Reports with opportunities for researchers, practitioners, and policymakers

New community-driven Wiki w/ datasets & additional resources

+ Forecasting supply and demand

High Leverage

+ Improving scheduling and flexible demand

Conferences & events

Workshop series

- ▶ Attend @ NeurIPS '22
- ▶ Mentorship programs
- ▶ www.climatechange.ai/papers

Summer school (multiple tracks)



Funding programs

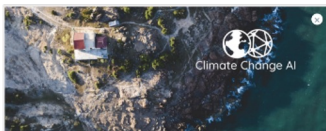
Global research funding for impactful projects

Climate Change AI Innovation Grants

Announcing a **\$1.8M grants program** for projects at the intersection of AI and climate change

- Funding of up to \$ 150K for **year-long** research projects
- Supporting projects involving AI or machine learning that address problems in **climate change mitigation, adaptation, or climate science**
- Focus on fostering **pathways to impact** and the creation of catalytic **datasets**

Newsletter, blog, & community



Welcome to the Climate Change AI community!

We are excited to have you here!

This is a place to connect, share and discuss all things related to climate change & machine learning 🌍🤖

If this is your first time here, you might want to head over to the [Hello](#) channel and introduce yourself.



Calls for Submissions



Funding



Projects & Courses



Readings



Jobs

Webinars & happy hours

Webinar series (monthly)

Virtual happy hours (biweekly)

Climate Change AI June 2021

Spatial planning of low-carbon cities with machine learning

Cities represent the lion's share of the world's energy use and GHG emissions, requiring rapid mitigation



Speakers

Dr. Jason Cao
Professor
Humphrey School of Public Affairs at the University of Minnesota

Learn more & join in:
www.climatechange.ai

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