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Understanding and Optimizing the Environmental Footprint of Computing

Carole-Jean Wu & Udit Gupta

Meta Al

Meta (formerly Facebook), Arizona State University, Harvard University: Udit Gupta, Young Geun Kim, Sylvia Lee, Jordan Tse, Hsien-Hsin S. Lee, David Brooks, Gu-Yeon Wei, Bilge Acun, Kim Hazelwood



NOVEMBER 9-10, 2021

Computing Industry is Growing







Growing # of devices and data center capacity Emerging applications demanding more compute resources

Further efficiency improvements challenging

Global AI Market Growth 29.86 billions (2020) to 299.64 billions (2026)

https://www.fnfresearch.com/artificial-intelligence-market





Growing Importance in the Research Community

Architecting a Sustainable Planet

Srilatha (Bobbie) Manne Principal Hardware Engineer Microsoft

Green AI

Roy Schwartz^{* \diamond} Jesse Dodge^{* $\diamond \clubsuit$} Noah A. Smith^{$\diamond \heartsuit$} Oren Etzioni^{$\diamond \Rightarrow$}

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July 2019

DTCO including Sustainability: Power-Performance-Area-Cost-Environmental score (PPACE) Analysis for Logic Technologies

M. Garcia Bardon¹, P. Wuytens¹, L.-Å. Ragnarsson¹, G. Mirabelli¹, D. Jang¹, G. Willems¹, A. Mallik¹, A. Spessot¹, J. Ryckaert¹, B. Parvais^{1,2} ¹ imec, Kapeldreef 75, 3001 Leuven, Belgium; ² Vrije Universiteit Brussel (VUB), Brussels, Belgium <u>marie.garciabardon@imec.be</u>



Zero Carbon Cloud Sustainable Computing **University of Chicago**

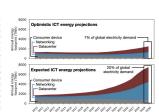
Energy and Policy Considerations for Deep Learning in NLP

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Abstract-Given recent algorithm, software, and hardware innovation, computing has enabled a plethora of new applications. As computing becomes increasingly ubiquitous, however, so does its environmental impact. This paper brings the issue to the attention of computer-systems researchers. Our analysis, built on industry-reported characterization, quantifies the environmental effects of computing in terms of carbon emissions. Broadly, carbon emissions have two sources: operational energy consumption, and hardware manufacturing and infrastructure. Although carbon emissions from the former are decreasing thanks to algorithmic, software, and hardware innovations that boost performance and power efficiency, the overall carbon footprint of computer systems continues to grow. This work quantifies the carbon output of computer systems to show that most emissions related to modern mobile and data-center equipment come from hardware manufacturing and infrastructure. We therefore outline future directions for minimizing the environmental impact



Abstract

Recent progress in hardware and methodology for training neural networks has ushered in a new generation of large networks trained on abundant data. These models have obtained notable gains in accuracy across many NLP tasks. However, these accuracy improvements depend on the availability of exceptionally large computational resources that necesistate similarly substantial energy consump-

Consumption	CO2e (lbs)
Air travel, 1 passenger, NY↔SF	1984
Human life, avg, 1 year	11,023
American life, avg, 1 year	36,156
Car, avg incl. fuel, 1 lifetime	126,000

Training one model (GPU)

NLP pipeline (parsing, SRL)	39
w/ tuning & experimentation	78,468
Transformer (big)	192

Abstract

The computations required for deep learning research have been doubling every few months, resulting in an estimated 300,000x increase from 2012 to 2018 [2]. These computations have a surprisingly large carbon footprint [40]. Ironically, deep learning was inspired by the human brain, which is remarkably energy efficient. Moreover, the financial cost of the computations can make it difficult for academics, students, and researchers, in particular those from emerging economics, to engage in deep learning research.

This position paper advocates a practical solution by making efficiency an evaluation criterion for research alongside accuracy and related measures. In addition, we propose reporting the financial cost or "price tag" of developing, training, and running models to provide baselines for the investigation of increasingly efficient methods. Our goal is to make AI both greener and more inclusive—enabling any inspired undergraduate with a laptop to write high-quality research papers. Green AI is an emerging focus at the Allen Institute for AI.

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Chasing Carbon: The Elusive Environmental Footprint of Computing

> Udit Gupta^{1,2}, Young Geun Kim³, Sylvia Lee², Jordan Tse², Hsien-Hsin S. Lee², Gu-Yeon Wei¹, David Brooks¹, Carole-Jean Wu²

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Computing's Environmental Footprint

Applications



Hardware Manufacturing





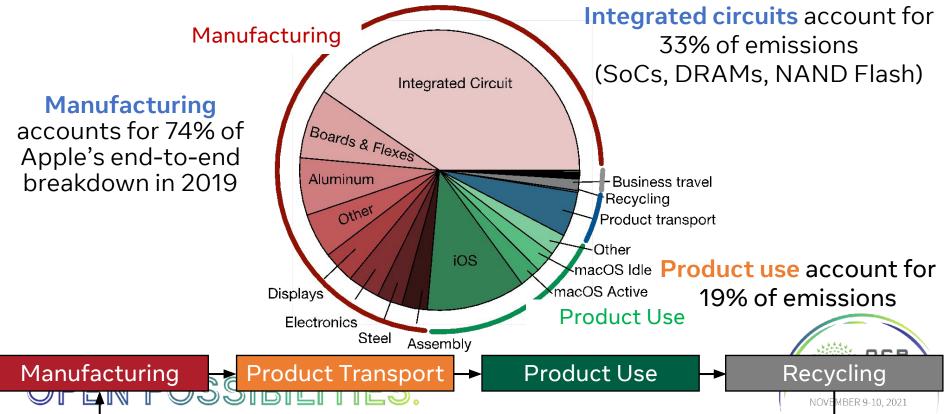
Embodied vs. **Operational** CO₂

Chasing Carbon: The Elusive Environmental Footprint of Computing. Udit Gupta, Young Geun Kim, Sylvia Lee, Jordan Tse, Hsien-Hsin Lee, Gu-Yeon Wei, David Brooks, Carole-Jean Wu. In Proceedings of the International Symposium on HPCA. 2021

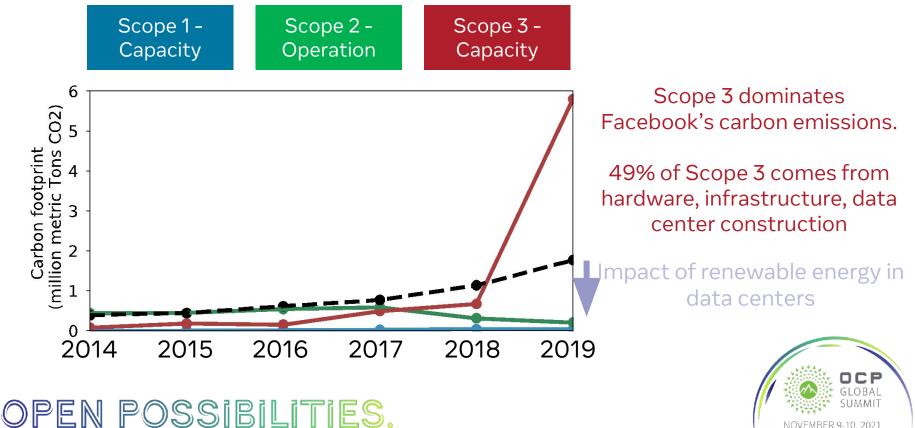


DCP GLOBAL SUMMIT NOVEMBER 9-10, 2021

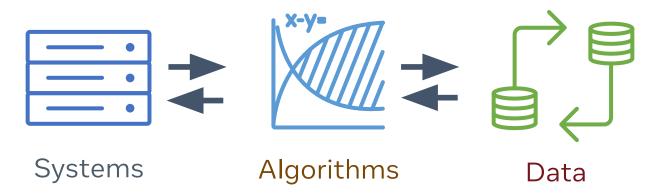
Manufacturing Dominates the Environmental Footprint of Client Computing



Historical analysis of Facebook's carbon footprint



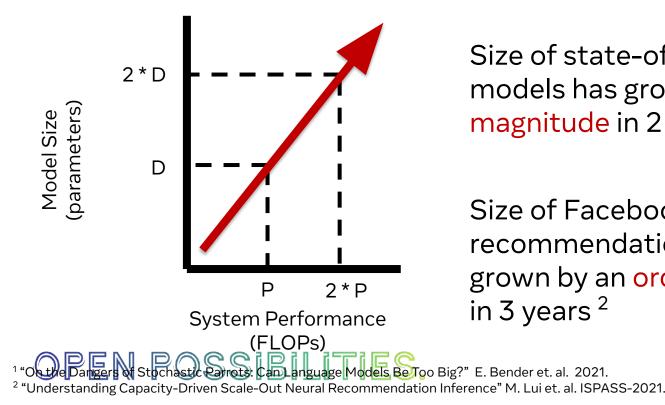
Efficiency optimization is a must





But, efficiency alone is not enough!

Benefits of higher efficiency overshadowed by higher application demands



Size of state-of-the-art NLP models has grown by **3 orders of** magnitude in 2 years ¹

Size of Facebook's production recommendations models has grown by an order of magnitude in 3 years ²

Environmentally-sustainable computing infrastructures







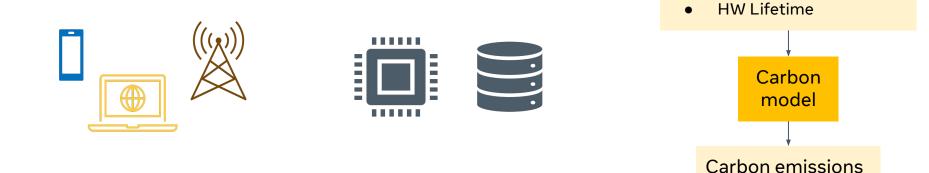
Metrics and Accounting

Optimization at Scale CO2 Footprint Amortization



Metrics and Accounting

Carbon cost modeling



End-to-end product life cycle analyses

PEN POSSIBILITIES.

Component-level carbon costs



Inputs

Renewable energy Fab characteristics

Workload HW design

Recycling

Metrics and Accounting

Carbon cost modeling

Carbon cost accounting/reporting



MLPerf



TOWARDS THE SYSTEMATIC REPORTING OF THE ENERGY AND **CARBON FOOTPRINTS OF MACHINE LEARNING**

A WORKING PAPER

Peter Henderson[†], Jieru Hu[‡], Joshua Romoff[¢] Emma Brunskill[†], Dan Jurafsky[†], Joelle Pineau^{‡,} [†]Stanford University, [‡]Facebook, [°]Mila, McGill University



Optimization at Scale

Datacenter scale load shaping

Cloud-edge computation scheduling

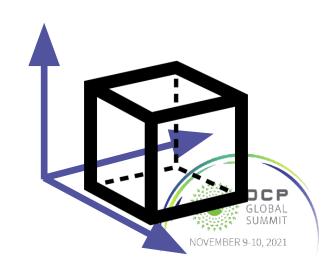


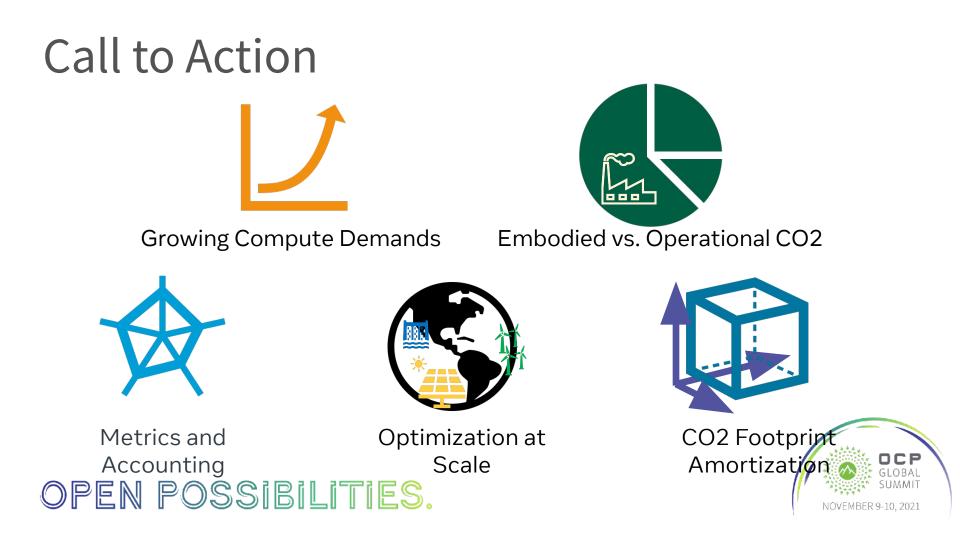
Carbon Footprint Amortization

Utilization **†**

- Virtualization, multi-tenancy, workload consolidation
 Hardware Lifetime
- Modular infrastructure design
- Disaggregated infrastructures
- Resilient hardware/software systems

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