

# Tackling Climate Change with Machine Learning

## Parakeet: Emission Factor Recommendation for **Carbon Footprinting** with Generative AI\*

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\* The methods discussed are for research purposes only, and are not indicative of Amazon's business use cases for carbon footprinting.

# Urgent need to de-carbonize the economy

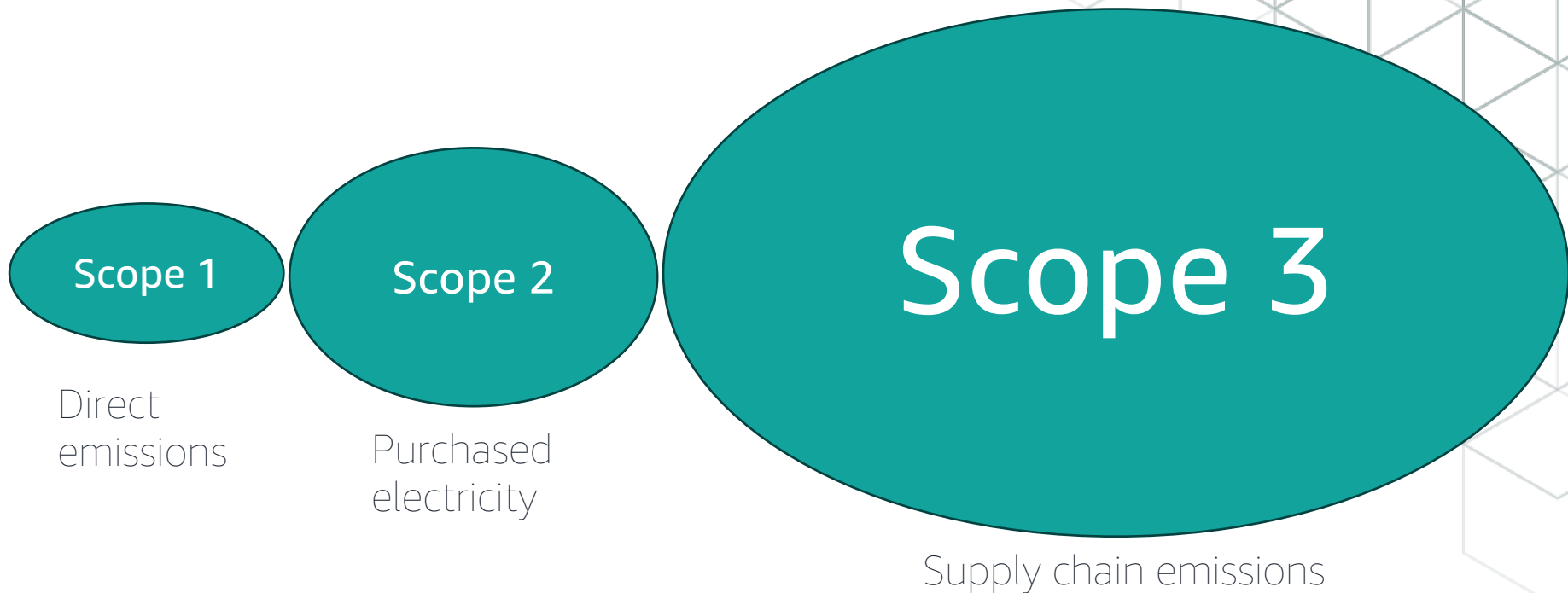
1.5/2C requires “rapid and deep and, in most cases, immediate GHG emissions reductions in all sectors this decade”

- IPCC AR6 Synthesis Report

# Rapid Decarbonization (for corporations)



# Scope 3 emissions dominate for most businesses



Life cycle assessment (LCA) – science of estimating impacts across the supply chain

Blanco, C., Caro, F. and Corbett, C.J., 2016. The state of supply chain carbon footprinting: analysis of CDP disclosures by US firms. Journal of Cleaner Production, 135, pp.1189-1197.

# Carbon footprint of household products



Brussel sprouts  
2.1 kgCO<sub>2</sub>e



Butter  
7.5 kgCO<sub>2</sub>e

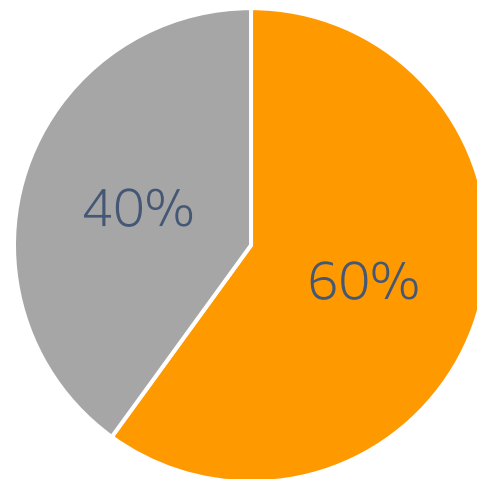


Vase  
14.1 kgCO<sub>2</sub>e



Plastic comb  
2.7 kgCO<sub>2</sub>e

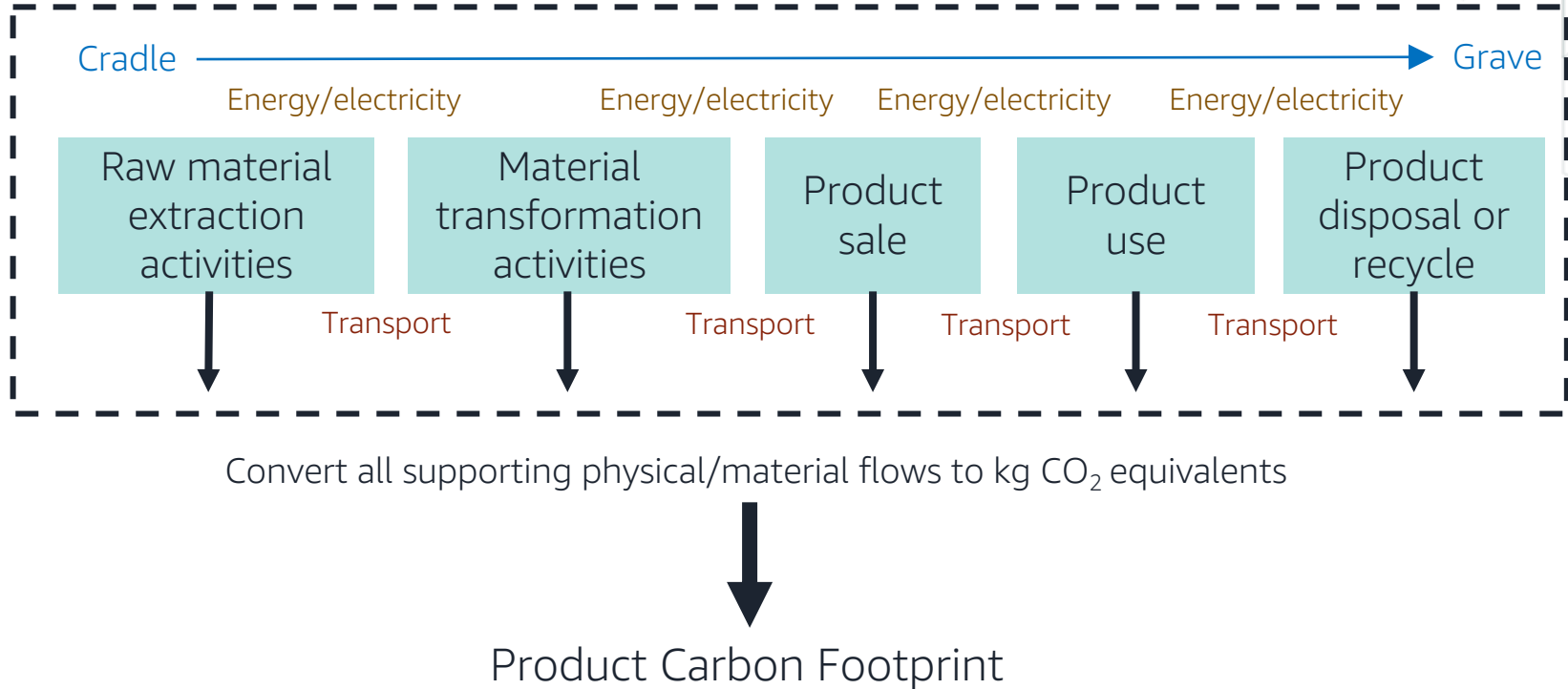
Global carbon footprint<sup>1</sup>



- Household products
- Others

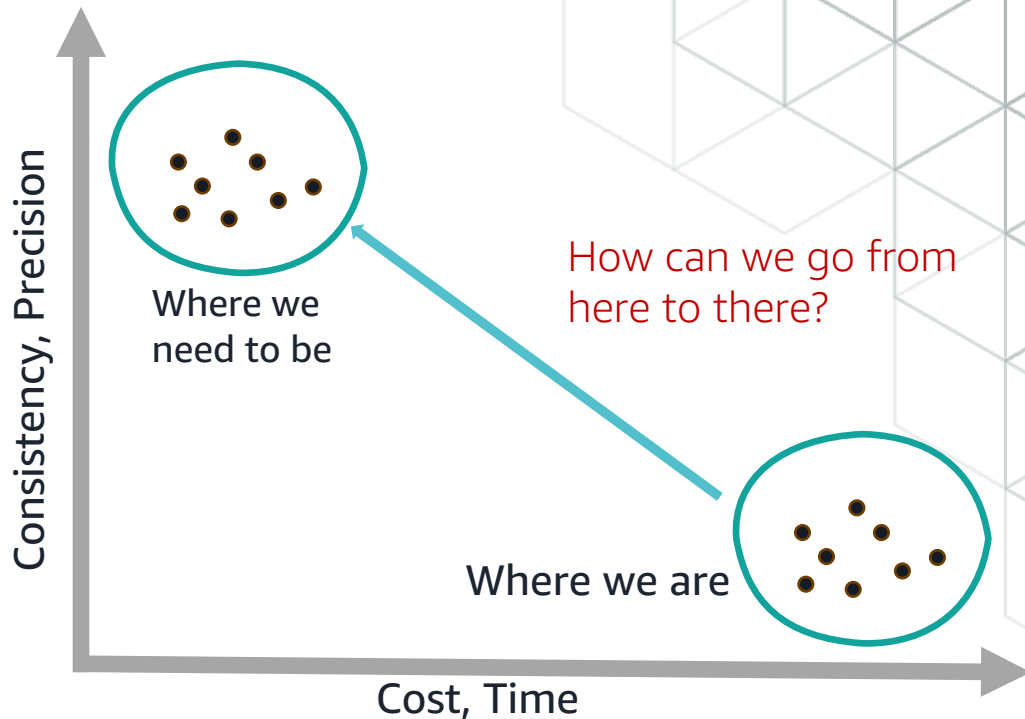
1. Ivanova, Diana, et al. "Environmental impact assessment of household consumption." Journal of Industrial Ecology, 2016.
2. Images licensed under [CC BY](#) [CC BY-NC](#)

# Life Cycle Assessment (LCA)



# Challenges in Scaling LCA

- Expensive
- Time consuming
- Expert dependent
- Extensive data collection



Balaji, Bharathan, Geoffrey Guest, Venkata Sai Gargeya Vunnava, Jared Kramer, Aravind Srinivasan, and Michael Taptich. "Scaling carbon footprinting: Challenges and opportunities." In Proceedings of the AAAI Symposium Series, vol. 2, no. 1, pp. 35-39. 2023.

# Spend-based Carbon Footprint Estimate

Emission Factor

$$\text{Manufacturing footprint of Sneaker} = (0.06 \text{ kg CO}_2\text{eq}/\$) \times (\$60) = 3.60 \text{ kgCO}_2\text{e}$$

USEPA/USEEIO

US Environmentally-Extended Input-Output  
Modeling Framework



\$60



# Example of Emission Factor Selection for Spend-based Estimate

The screenshot shows the NAICS Association website with search results for 'shaving cream'. The search bar contains 'shaving cream' and the results table lists various NAICS codes and titles. The entry for '325620 Toilet Preparation Manufacturing' is circled in red.

NAICS ASSOCIATION

Home Data Enrichment Business Lists Market Research Products Company Lookup NA

### NAICS/SIC SEARCH RESULTS

shaving cream NAICS Search

Enter Keyword(s) SIC Search

| NAICS  | NAICS Title  | Common Keywords   | SIC Crosswalk            |
|--------|--|---|--------------------------|
| 311520 | Ice Cream and Frozen Dessert Manufacturing             | Ice cream specialties manufacturing                                       | <a href="#">View SIC</a> |
| 311512 | Creamery Butter Manufacturing                          | Butter, creamery and whey, manufacturing                                  | <a href="#">View SIC</a> |
| 325620 | Toilet Preparation Manufacturing                       | Shaving preparations (e.g., creams, gels, lotions, powders) manufacturing | <a href="#">View SIC</a> |
| 423990 | Other Miscellaneous Durable Goods Merchant Wholesalers | Wood products (e.g., chips, posts, shavings, ties) merchant wholesalers   | <a href="#">View SIC</a> |
| 333517 | Machine Tool Manufacturing                             | Shaving machines, metalworking, manufacturing                             | <a href="#">View SIC</a> |
| 424210 | Drugs and Druggists' Sundries Merchant Wholesalers     | Shaving preparations merchant wholesalers                                 | <a href="#">View SIC</a> |
| 321113 | Sawmills   | Sawdust and shavings (i.e., sawmill byproducts)                           | <a href="#">View SIC</a> |

After identifying, the NAICS code was then mapped to the BEA sector “Toilet Preparation Manufacturing”

# Selection with Retrieval Augmented Generation

## Given Activity Input (real example)

FAC.WRC.OAL0508IN9.GRU - Combina o chave, 5/8 pol., 9-1/2 pol. OAL (chave de extremidade aberta)



## AI Paraphrase

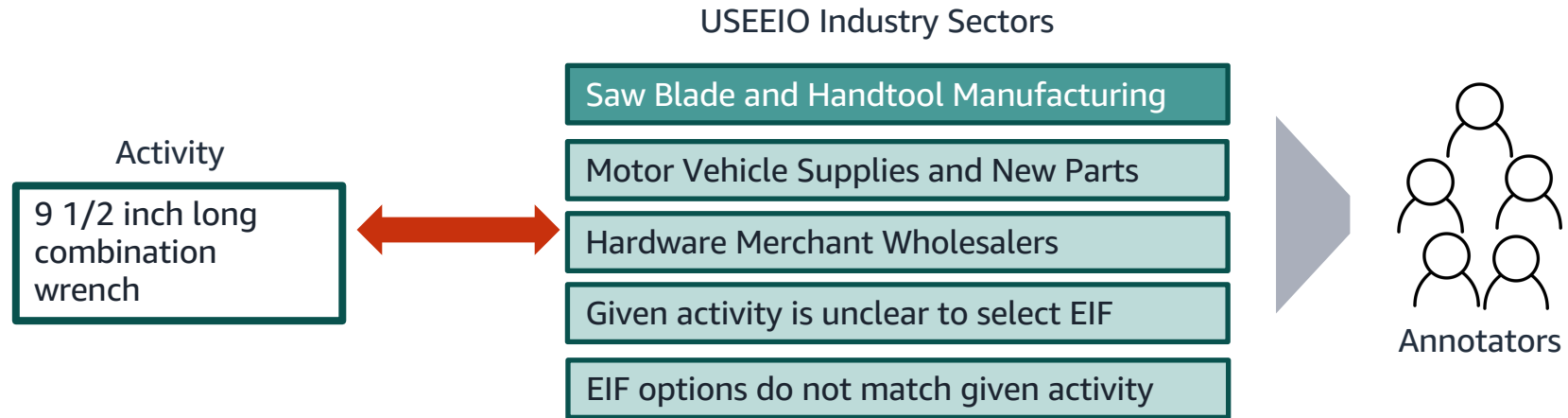
The item is a 9 1/2 inch long combination wrench for use in warehouses and with basic tools.



## AI Recommendation

NAICS title is 'Saw Blade and Handtool Manufacturing'. This covers the manufacturing of nonpowered handtools like wrenches, which is what the product is.

# Human In The Loop Validation



- Top-1 accuracy of 87% and Top-K accuracy of 98%
- 1000x cheaper than fully manual solutions

# Team



Bharathan  
Balaji



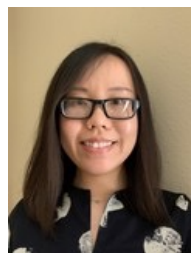
Fahimeh  
Ebrahimi



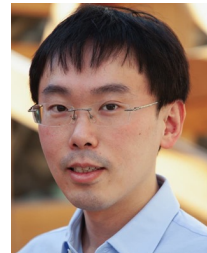
Jeremie  
Hakian



Shankar  
Gurumurthy



Anran  
Wang



Qingshi  
Tu



Zaid  
Thanawala



Gargeya  
Vunnava



Nina  
Domingo



Kellen  
Axten



Stephen  
Yoshida



Ethan  
Roday



Kommy  
Weldemariam



Jared  
Kramer

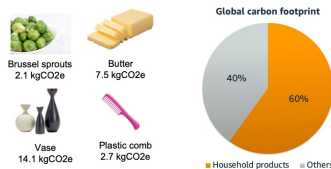
# Poster # 52

## Parakeet: Emission Factor Recommendation for Carbon Footprinting with Generative AI

Bharathan Balaji, Fahimeh Ebrahimi, Nina Gabrielle G Domingo, Venkata Sai Gargeya Vunnava, Abu-Zaher Faridee, et al.\*



### Carbon Footprinting of Products



Household products constitute >60% of global carbon footprint. Measuring the impact of each product is a key step towards reducing their associated carbon emissions.

### Life Cycle Assessment

#### EIO LCA

Production of Pears = (0.5 kgCO<sub>2</sub>e/\$) x (\$2) = 1 kgCO<sub>2</sub>e

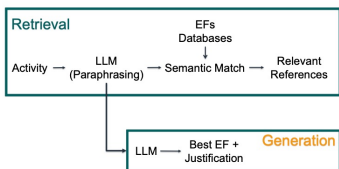


#### Process-based LCA

Production of Pears = 0.48 kgCO<sub>2</sub>e ← "Market for Pear"



### Parakeet



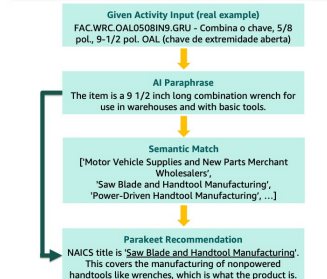
1) A business activity description get converted into a plain English text. 2) Top K similar reference products or NAICS titles to the business activity get retrieved using an embedding model. 3) All this information fed into an LLM model and ask for best impact factor that fits this activity. The model also return a human readable justification for its choice.

Dataset Precision@1 Precision@K Dataset Size

#### EIEIO LCA EF Recommendation

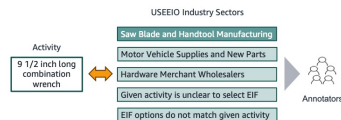
|                               |      |      |      |
|-------------------------------|------|------|------|
| Govt of Austin Invoices       | 93.5 | 98.8 | 2159 |
| Katana ML Invoices            | 97.1 | 100  | 1121 |
| Procurement Products          | 90.9 | 98.1 | 3980 |
| Heavy Equipment Invoices      | 82.2 | 93.5 | 1803 |
| Process LCA EF Recommendation |      |      |      |
| Food.com Ingredients          | 71.0 | 72.9 | 1956 |
| Grocery Packaging             | 82.2 | 89.2 | 195  |

The results of Parakeet performance on six datasets. We use the General Text Embedding (gte-large) model for semantic text matching, and Claude 3 Sonnet as our LLM.



### Human in the loop

We provide the query, paraphrased text, the recommended emission factor, and top-ranked list of emission factors to an annotator. The annotator can choose to override the Parakeet recommendation, indicate input data or EFs provided are inappropriate, or that they are uncertain to make a choice.



#### References

- Balaji, Bharathan, et al. (2023). "GAIL: Carbon footprinting of household products with zero-shot semantic text similarity." In: Proceedings of the ACM Web Conference 2023, pages 4004–4014.
- Ding, Zhi, et al. (2023). "Anupop: Efficient product carbon footprint accounting with large language models." In: arXiv preprint arXiv:2303.02614.
- Balaji, Bharathan, et al. (2023). "Flamingo: Environmental impact factor matching for life cycle assessment with zero-shot machine learning." In: ACM Journal on Computing and Sustainable Sciences, 1(2):1–2.

# Link to Paper





# Thank you!

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