

Comparison of Greenhouse Gas databases using FoodEx2 codes

Parallel thematic session: Resilience, vulnerability, human and planetary health

From: 28 June 2022, 15:10 to 16:25 BST British Summer Time

Christian Reynolds, Jacqueline Tereza da Silva, Josefa Maria Fellegger Garzillo, Angelina Frankowska, Alana Kluczkovski, Diego Rose, Berill Takacs, Victoria Padula de Quadros, Bridget Anna Holmes, Ximena Schmidt Rivera, Sarah Bridle

Dr Christian Reynolds

Centre for Food Policy, City, University of London



@sartorialfoodie

christian.reynolds@city.ac.uk



Who am I?

Senior Lecturer at the Centre for Food Policy



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Routledge Handbook of Food Waste

Edited by Christian Reynolds, Tammara Soma, Charlotte Spring and Jordon Lazell





Public Health Nutrition: 22(8), 1503-1517

University of South Australia



The University Of Sheffield. Institute for Sustainable Food.



Focus: healthy sustainable diets and food consumption (including waste)



Healthy and sustainable diets that meet greenhouse gas emission reduction targets and are affordable for different income groups in the UK

Christian J Reynolds¹, Graham W Horgan², Stephen Whybrow¹ and Jennie I Macdiarmid^{1,*} ¹The Rowett Institute University of Aberdeen, Aberdeen AB25 2ZD, UK: ²Biomothematics & Statistics Scotland, Aberdeen, UK

Previously: Food waste politics/history, social sciences approaches

Shameless plug for FLW text book – if you want open access let me know $\ensuremath{\textcircled{\odot}}$

Part of ongoing research...



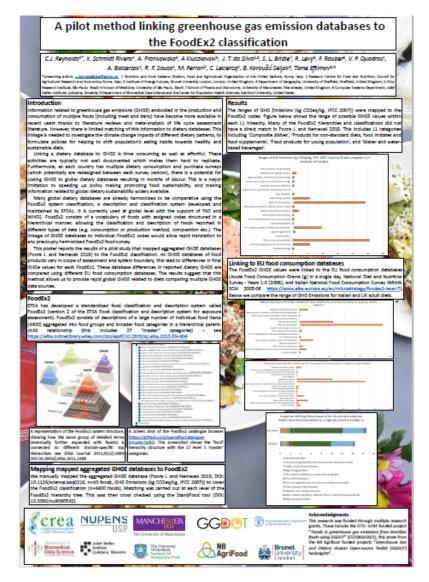
PERSPECTIVE published: 23 February 2021 doi: 10.3389/frai.2020.621577



Using Natural Language Processing and Artificial Intelligence to Explore the Nutrition and Sustainability of Recipes and Food

Marieke van Erp¹*[†], Christian Reynolds^{2†}, Diana Maynard³, Alain Starke⁴, Rebeca Ibáñez Martín⁵, Frederic Andres⁶, Maria C. A. Leite⁷, Damien Alvarez de Toledo⁶, Ximena Schmidt Rivera⁸, Christoph Trattner⁴, Steven Brewer⁹, Carla Adriano Martins¹⁰, Alana Kluczkovski¹⁰, Angelina Frankowska¹⁰, Sarah Bridle¹⁰, Renata Bertazzi Levy¹¹, Fernanda Rauber¹¹, Jacqueline Tereza da Silva¹⁰ and Ulbe Bosma¹²

https://doi.org/10.3389/frai.2020.621577



http://dx.doi.org/10.13140/RG.2.2.15990.34889

The problem: lack of comparable GHGE data

- Multiple Greenhouse Gas Emissions (GHGE) databases exist (Each describes the impacts of different agricultural production systems around the world).
- There is a growing need to capture the environmental impacts of dietary choices.
- Direct matching of GHGE databases to dietary databases is very time consuming.
- However, there are standards for comparing dietary databases one of these is FoodEx2.

Can a harmonised dietary classification system be used to compare/allocate GHGE impacts to food categories?

In this presentation, we aim to assess the reliability of the linking a GHGE database to FoodEx2, by comparing it to similar databases.

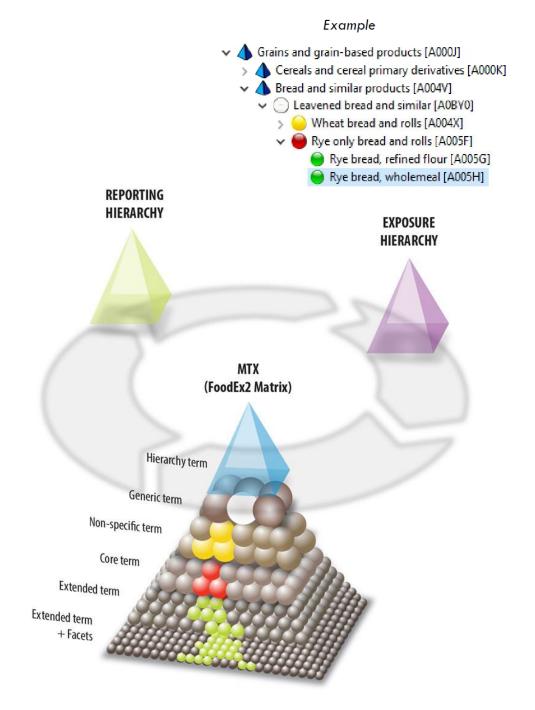
What is FoodEx2?

- A comprehensive food classification and description system
- A common language
- Developed and maintained by EFSA
- Clearly defined
- Hierarchical structure
- A food fits in one group only
- For every food there is a group

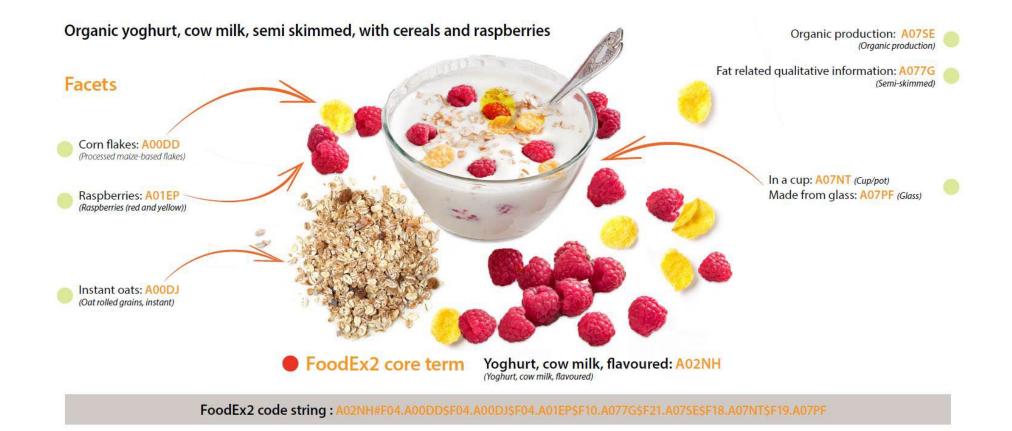
21 Food groups in total for 4558 FoodEx2 codes

At least 56 food consumption databases have been coded with FoodEx2. see https://www.globaldietarydatabase.org/





Example of FoodEx2 coding



Source: https://www.hapih.hr/wp-content/uploads/2019/11/Ioannidou_FoodEx-2-klasifikacija-hrane.pdf

FoodEx2 is linked to many global dietary datasets

33 countries via FAO/WHO GIFT https://www.fao.org/gift-individual-food-consumption/en/

21 countries via The EFSA Comprehensive European Food Consumption Database <u>https://www.efsa.europa.eu/en/data-report/food-consumption-data#the-efsa-comprehensive-european-food-consumption-database</u>

407 data sets via https://www.globaldietarydatabase.org/

FoodEx2 offers an opportunity to link many datasets to environmental impacts in a quick and comparable manner.

FAO/WHO GIFT | Global Individual Food consumption data Tool

Overview Data and indicators Inventory of surveys Resources Methodology





FAO/WHO GIFT: Better data, better policies, better diets Providing tailored answers for health, nutrition and agriculture policies.

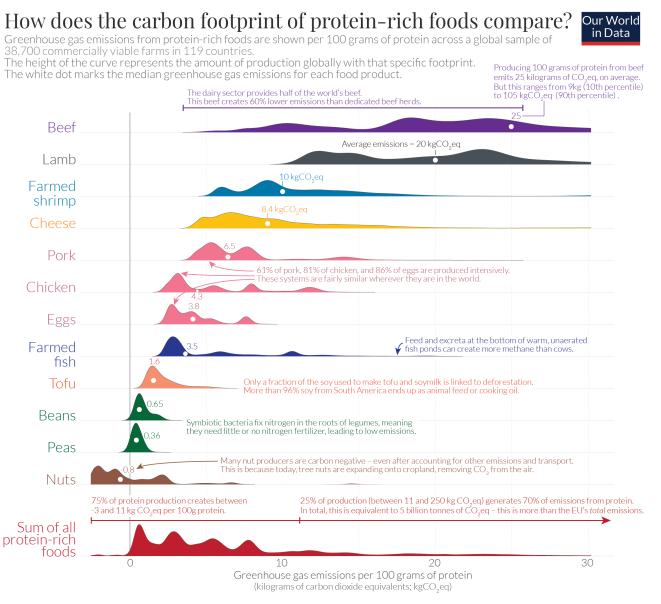


The advantage of Poore and Nemecek (2018)

The Poore and Nemeck (2018) database provides 5% and 95% confidence intervals as well as **mean global impacts**

43 food categories meta-analysis comparing various types of food production systems.

Impact can vary 50-fold among producers of the same product, creating substantial mitigation opportunities



Note: Data refers to the greenhouse gas emissions of food products across a global sample of 38.700 commercially viable farms in 119 countries. Emissions are measured across the full supply-chain, from land use change through to the retailer and includes on-farm, processing, transport, packaging and retail emissions. Data source: Joseph Poore and ThomasNemecek (2018). Reducing food's environmental impacts through producers and consumers. *Science*. **OurWorldinData.org** – Research and data to make progress against the world's largest problems. Licensed under CC-BY by the authors Joseph Poore & Hannah Ritchie.

Matching P&N (2018) to FoodEx2

43 food categories matched to 4558 FoodEx2 code (Kg of Co2e per 100g)

All products were matched by hand, using the closest raw product; if it was a product with multiple ingredients, we took the largest ingredient by weight. GHGE Values corrected for hydration and processing.

0	Р	Q
L7_FoodEx2_desc 🗸	level 💌	Category 🔍 is
Grains and grain-based products	1	Wheat & Rye (Bread)
Cereals and cereal primary derivatives	2	Wheat & Rye (Bread)
Cereal grains (and cereal-like grains)	3	Wheat & Rye (Bread)
Mixture of grains	4	Wheat & Rye (Bread)
Barley and similar-	4	Barley (Beer)
Barley grains	5	Barley (Beer)
Barley grain, pearled	6	Barley (Beer)
	Grains and grain-based products Cereals and cereal primary derivatives Cereal grains (and cereal-like grains) Mixture of grains Barley and similar- Barley grains	Grains and grain-based products1Cereals and cereal primary derivatives2Cereal grains (and cereal-like grains)3Mixture of grains4Barley and similar-4Barley grains5

N	0	Р	Q
	L7_FoodEx2_desc 🗸	level 💌	
AUDEE	Fotatoes and similar	3	Folaloes
A00ZT	Potatoes	4	Potatoes
A011P	Potato boiled	5	Potatoes
A011R	Potato baked	5	Potatoes
A00ZX	Main-crop potatoes	5	Potatoes
A00ZV	New potatoes	5	Potatoes
AODPM	Andigena	4	Potatoes
A00ZY	Tropical root and tuber vegetables	3	Cassava
A04JX	Cassava roots and similar-	4	Cassava
A00ZZ	Cassava roots	5	Cassava

GHGE Databases matched to FoodEx2

Reducing food's environmental impacts through producers and consumers

J. POORE D AND T. NEMECEK

SCIENCE • 1 Jun 2018 • Vol 360, Issue 6392 • pp. 987-992 • DOI: 10.1126/science.aag0216

https://doi.org/10.1126/science.aaq0216

Data Article SHARP-Indicators Database towards a public database for environmental sustainability

Elly Mertens ^a R 🖻, Gerdine Kaptijn ^a, Anneleen Kuijsten ^{a, b}, Hannah van Zanten ^c, Johanna M. Geleijnse ^{a, b}, Pieter van 't Veer ^{a, b} 🖻

https://doi.org/10.1016/j.dib.2019.104617

Carbon footprint of self-selected US diets: nutritional, demographic, and behavioral correlates 3

Donald Rose 🖾, Martin C Heller, Amelia M Willits-Smith, Robert J Meyer

The American Journal of Clinical Nutrition, Volume 109, Issue 3, March 2019, Pages 526–534, https://doi.org/10.1093/ajcn/nqy327

https://doi.org/10.1093/ajcn/nqy327

Footprints of foods and culinary preparations consumed in Brazil Josefa Maria Fellegger Garzillo, Priscila Pereira Machado, Maria Laura da Costa Louzada, Renata Bertazzi Levy, Carlos Augusto Monteiro, https://doi.org/10.11606/9788588848405 "City"

"Garzillo"

43 food categories matched to 4558 FoodEx2 code matched by authors

"SHARP" 945 food categories matched to FoodEx2

"Rose/Heller" 608 food categories 357 categories linked to FICD to National Health and Nutrition Examination

Survey (NHANES), this resulted in 608 linked to FoodEx2 (using Global Dietary Database concordance).

329 food categories

linked to the Brazilian Food Consumption Survey which was matched to FoodEx2 $\,$

(All databases normalised to kg of Co2e per 100g)

Correlations

Database	n	Spearman correlation	p-value
Sharp	945	0.699	< 0.001
Rose/Heller	608	0.572	< 0.001
Garzillo	329	0.610	< 0.001

Table 1. Correlation between "City" database to other databases

Visualisation of matches

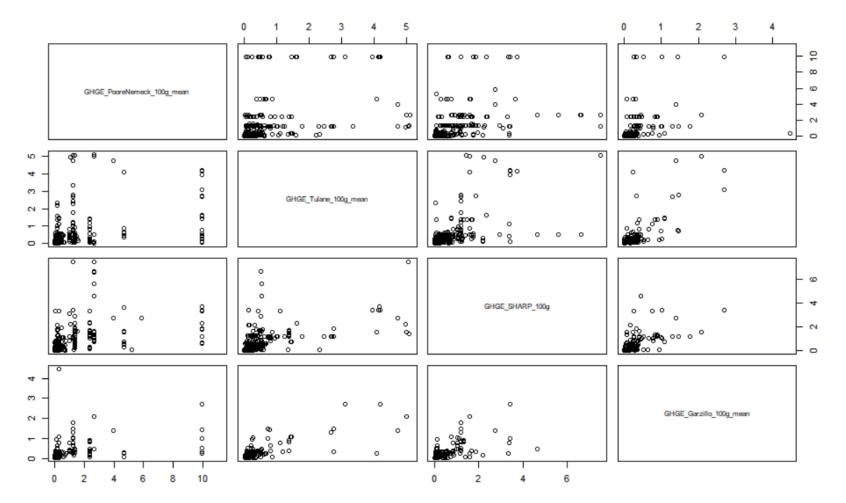


Figure 1. Scatterplot displaying the relationship between GHGE from multiples sources (Reynolds/Takacs, Rose/Heller, Sharp, Garzillo)

Differences in "City" and SHARP

Of the 945 food items with GHGE in "City" and SHARP, 50% (n = 476) were ranked in the same quintile. The kappa statistics was 0.536 (p < 0.001).

Of the 469 food items not ranked into the same quintiles,

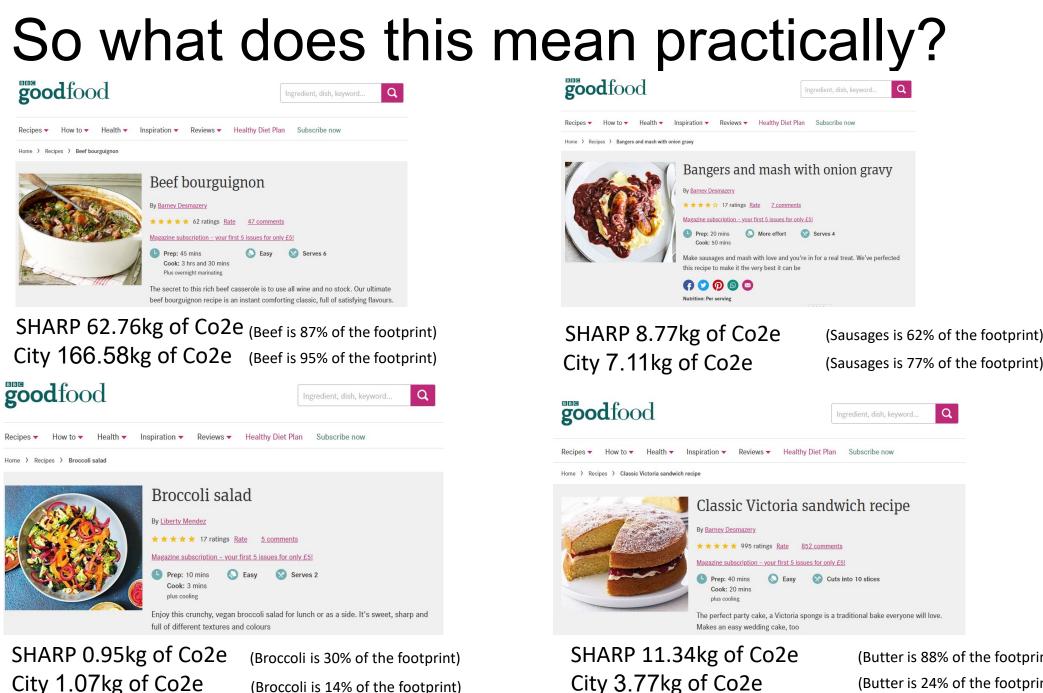
44% (n=206) were within p5 and p95 confidence interval values of City

31% (n=144) were lower than the p5 confidence interval values of City

25% (n= 119) were higher than p95 confidence interval values of City.

The food items with the biggest differences between mean values for "City" and SHARP are wheat and rye; fish and seafood; pig meat; fruits; nuts and pulses.

These food items will be further investigated in the next update of the data, aiming to increase reliability to estimate GHGE from food consumption.



(Broccoli is 14% of the footprint)

(Butter is 88% of the footprint) (Butter is 24% of the footprint)

Many thanks to all the co-authors

Dr Christian Reynolds

Centre for Food Policy, City, University of London

e @sartorialfoodie <u>christian.reynolds@city.ac.uk</u>

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https://www.city.ac.uk/prospective-students/courses/postgraduate/food-policy

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