T-reX: Quantifying waste and material footprints in current and future Life Cycle Assessment (LCA) databases

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# Abstract

The quintessential principle of the circular economy is to minimise material consumption and waste generation. Thus, identifying and quantifying waste and material flows is critical. Life Cycle Assessment (LCA) is a powerful quantitative method for this, especially given its capacity to expose the details of an activity’s entire life cycle, often guiding the implementation of circular principles to where they could be most effective. Pursuant to these goals the authors have developed ‘T-reX’, an analytical tool for LCA, written in Python. T-reX extends on the existing computational ecosystem centred on Brightway and Premise and can facilitate the quantification of user-defined supply chain demands for activities in current and prospective scenarios. T-reX streamlines database manipulation for LCA practitioners, integrating methods to aggregate and analyse demand inventories that could expose hotspots of hidden risk. With a simple case study using lithium-ion batteries, we demonstrate some of T-reX’s potential by detailing and exploring aspects of their waste and material footprints. Since such footprints are often linked to negative externalities, T-reX could support sustainable decision-making and contribute to the development of the circular economy by promoting the understanding of our material consumption and waste generation.

Table 1

*List of abbreviations*

CCS Carbon Capture and Storage

CF Carbon Footprint

EF Ecological Footprint

CML Centrum voor Milieuwetenschappen (Institute of Environmental Sciences)

CRM Critical Raw Material

CPC Cooperative Patent Classification

CSI Crustal Scarcity Indicator

CSP Crustal Scarcity Potential

EDIP Environmental Design of Industrial Products

EEIOA Environmentally Extended Input-Output Analysis

EOL End-of-Life

FOEN Swiss Federal Oﬀice for the Environment

IAM Integrated Assessment Model

IE Industrial Ecology

LCA Life Cycle Assessment

LCI Life Cycle Inventory

LCIA Life Cycle Impact Assessment

LFP Lithium Iron Phosphate

Li-ion Lithium-ion

LiMn2O4 Lithium Manganese Oxide

MF Material Footprint

MFA Material Flow Analysis

NCA Nickel Cobalt Aluminium

NMC Nickel Manganese Cobalt

PWF Product Waste Footprint

RCP Representative Concentration Pathway

ReCiPe A standard LCIA method set

SDG Sustainable Development Goals

SSPs Shared Socioeconomic Pathways

T-reX The Tool for analysing re-X in LCA

UBP Umweltbelastungspünkte (Environmental Impact Points)

UN United Nations

WF Waste Footprint

# Introduction

## Introduction to T-reX

The development of a circular economy has become a central focus in the pursuit of sustainability objectives toward curtailment of our environmental footprint within planetary boundaries (European Commission, 2019, 2020; Government of the Netherlands, 2023, 2016; Pardo and Schweitzer, 2018; Ellen MacArthur Foundation, 2015).

Fundamental to this development is a decrease in primary material consumption and a reduction of life cycle waste through the implementation of ‘re-X’v strategies (e.g., refuse, rethink, design for—and implementation of— repair, remanufacturing and recycling) (Reike et al., 2018; European Commission, 2022; Alfieri et al., 2022; Parker et al., 2015). In addition to circular economy goals, contemporary geopolitical tensions in an ever more globalised world have highlighted the vulnerability of many advanced economies to intentional supply disruptions—wrought as an act of competition or even outright hostility (Carrara et al., 2023; Hartley et al., 2024; Berry, 2023).

The concept of the ‘footprint’ as an environmental sustainability indicator began with the Ecological Footprint (EF) (Wackernagel, 1994) and after being popularised by the Carbon Footprint (CF) (Čuček et al., 2015) the ‘footprint family’ has adopted many additional metrics— albeit without yet coalescing into a coherent or consistent framework (Giampietro and Saltelli, 2014; Vanham et al., 2019; Ridoutt and Pfister, 2013). More recently, the footprint collection has been extended to include the Material Footprint (MF) (Wiedmann et al., 2013), which is often encountered in the ‘macro methods’ of Industrial Ecology (IE), such as environmentally Extended Input-Output Analysis (EEIOA) (Lenzen et al., 2021) and Material Flow Analysis (MFA) (Schaffartzik et al., 2013). Whether at the level of products, entire industries, nations, or even continents, the material footprint aims to quantify the total material consumed in supply chains. It has been shown that the MF can be ‘highly representative of damage to human health and biodiversity’ (Steinmann et al., 2017) and indeed, this metric was recently beatified by the United Nations (UN), becoming the ‘core official indicator’ for targets 8.4 and 12.2 of the Sustainable Development Goals (SDGs) (Lenzen et al., 2021).

Though steadily emerging (Laurenti et al., 2016; Demirer, 2019; Guillotreau et al., 2023), the ‘Waste Footprint’ (WF) metric is less well-developed. While it seems obvious that reducing life cycle waste is critical to the development of the circular economy (Towa et al., 2020; Ellen MacArthur Foundation, 2015), the WF remains largely overlooked— especially in LCA models where waste itself is seldom apportioned any inherent environmental significance aside from the emissions related to its treatment (Laurenti et al., 2023). Such neglect strikes the authors as unjustified, given that it has been repeatedly demonstrated that WFs have a strong association with environmental damage (Laurenti et al., 2023; Doka, 2024; Ridoutt et al., 2010; Jiao et al., 2013). Additionally, it has been shown that the most vulnerable communities suffer disproportionately from the social and ecological impacts created by wasteful behaviours and substandard end-of-life treatment (N. Pellow, 2023; Akese and Little, 2018).

To date, Life Cycle Assessment (LCA) has been the most widely-used method for assessing environmental footprints of supply chains. LCA can comprehensively evaluate impacts across the entire life cycle of products and services—from ‘cradle to grave’— often identifying critical hotspots and guiding prioritisation of action (Guinée et al., 2010). The standard approach is to apply Life Cycle Impact Assessment (LCIA) methods (such as ReCiPe (Huijbregts et al., 2016) and CML (Guinée et al., 2002)), which convert emissions and resource consumption inventory data into a set of impact scores based on the sum of the elementary flows (those between the technosphere and biosphere). These scores can then be aggregated into one metric for each impact category that can be compared across products and processes. Several LCIA methods include, to some extent, waste generation (Swiss Eco-Factors, EDIP and EN15804) (Swiss Federal Office for the Environment (FOEN), 2021; Hauschild and Potting, 2004; CEN (European Committee for Standardization), 2019) and material consumption (Crustal Scarcity Indicator (CSI) and Swiss Eco-Factors (Arvidsson et al., 2020; Swiss Federal Office for the Environment (FOEN), 2021)). These methods, however, are generally limited in their scope (especially for waste), do not allow for flexible quantification of specific waste and material footprints, and often provide results in characterised units that are abstract or difficult to interpret (e.g., *Ümweltbelastungspunkte* (UBP) [English: environmental impact points] in the case of the SwissEco-Factors) (Su, 2020).

If included within an LCA framework, WF and MF, the two metrics of focus in the study, can provide a comprehensive assessment of highly relevant aspects of sustainable supply chain management which are distinct but complementary to existing LCA environmental impact indicators. These two additional metrics encapsulate both (critical) material resource use and waste generation, allowing more targeted analysis to address undesired consequences that are not captured by the conventional suite of LCA indicators. Two examples of the shortfalls that can be addressed by MF and WF are (i) specific concerns around the availability and supply risks of specific materials or groups of materials like CRMs (the existing “resource depletion” category does not allow such distinctions), and (ii) other consequences of waste management such as landscape impacts, safety, odour annoyance and other aforementioned social impacts. Thus, to reduce the negative externalities of human consumption and improve supply chain resilience, it is essential to uncover, disaggregate, and quantify the material and waste footprints of human activities in as much detail as possible (Bisinella et al., 2024; Towa et al., 2020; Berger et al., 2020; Sonderegger et al., 2020). An additional benefit of having WF and MF side-by-side conventional LCA indicators is that trade-offs can be identified, e.g. where material efficiency measures may not necessarily result in environmental benefits such as climate change impact reduction.

To facilitate the quantification of waste and material flows in LCA, we developed a Python package to extend the Brightway open-source LCA framework (Mutel, 2017a) and track these exchanges by translating them into inventory indicators and ‘pseudo’ LCA impact (LCIA) categories. T-reX is ‘T[ool for] re-X’ (reduce, reuse, recycle, etc.) and enables LCA practitioners to manipulate their databases to allow them to easily aggregate the mass and volume of any desired exchange, and to create flexible categories that differentiate between material categories, waste types, and treatment or EOL handling.

Integration with the premise Python package (Sacchi et al., 2022)—which connects the projections of integrated assessment models (IAMs) with current LCA databases— enables the user to create and manipulate prospective LCA databases. Frustratingly, the current utility of prospective databases—in general, and in particular for the waste sector—is constrained by the fact that, to date, the sectoral coverage of future life cycle inventories (LCIs) is largely confined to energy, steel, cement, and transport (Sacchi et al., 2023). Despite the ever more critical need to reliably model and future waste management systems, Bisinella et al. (2024) reports an alarming lack of coherent development in this field. premise offers an excellent structure for the integration of future waste and material flow modelling in LCA databases, but its potential contribution to the development of the circular economy cannot be realised if the scientific resources for data collection and validation are not forthcoming. The authors believe that there is an urgent need for research investment in this area.

The purpose of T-reX is not to quantify the environmental impacts of material consumption and waste production (the requisite data and impact modelling remain lacking), but rather to quantify the material and waste flows themselves, even those that are finally consumed by waste treatment processes. It provides, thus, not an impact assessment in the traditional sense, but an accounting of the material consumed and waste generated by a product or service inside of the technosphere, regardless of the end-of-life fate of these flows. By definition, the development of the ‘circular economy’ necessitates the reduction and ultimate elimination of waste (though whether this objective is thermodynamically impossible has long been the subject of debate by Ayres (1999), Reuter and van Schaik (2012) and many others). In any case, minimising primary material consumption and the generation of waste is of paramount importance. By allowing LCA practitioners to easily classify and quantify these exchanges, T-reX provides a practical means to identify hotspots and potentially highlight opportunities for waste reduction and material efficiency. This is T-reX’s key strength and novelty: by enabling identification of waste and material consumption hotspots throughout entire value chains, LCA practitioners can pinpoint specific processes which would be key levers for decision-makers to substantially improve resource efficiency while controlling for environmental impacts.

With the continual improvement of LCA databases and the development of prospective databases, T-reX can produce ever more realistic estimates, enhancing its actionable utility derived from the exploration of the potential impacts of future waste management systems and the material demands of emerging technologies. The better the data, the more specific and accurate the results of T-reX analysis, and, thus, the more useful the tool becomes. The authors hope that T-reX will be a valuable asset to the LCA community and contribute to the development of a more sustainable and circular economy.

## Background and need

### Waste in LCA

Though often described simply as a ‘material with a negative economic value’ (Guinée et al., 2004), waste is a nebulous concept, and one whose definition is poorly delineated and highly variable across space and time. From a systems perspective, the notion of waste is anathema to the ‘circular economy’, they cannot co-exist. From a practical economic viewpoint, the viability of an extensive waste processing system is highly dependent on precise knowledge (or at least, reasonable predictions) of material and waste flows, in addition to energy costs and the relative prices of virgin and secondary materials. Thus, as Bisinella et al. (2024) argues, waste must be ‘thoroughly characterised’ and ‘modelling [of its management] must be physically based’. The reliable, robust models needed to guide the development of the circular economy can be built only on a foundation of high-quality data.

Accurate and detailed information about waste and waste systems is, by definition, essential for understanding the ‘circularity’ of a human activity and for predicting its life cycle externalities. There remains, however, a knowledge gap regarding WFs and their associated environmental impacts (Laurenti et al., 2023).

Conventional LCA database models consider waste as a ‘service’ (accounting for the treatment, not the material) (Guinée and Heijungs, 2021) and typically use generic waste processing models (Beylot et al., 2018) that break the causal link between the functional unit and the waste associated impacts. In LCA, waste flows are (almost exclusively) not considered as fundamental biosphere exchanges, but rather, as technosphere flows within the human economy. Waste produced by an activity is transferred to a relevant waste treatment activity where it is accepted ‘burden-free’ and transformed into a combination of emissions and other waste ‘products’ (Guinée and Heijungs, 2021). There can be several treatment steps in this pathway leading, ultimately, to a mass of material being deposited in a landfill. In this system of waste accounting, the impacts apportioned to the waste-producing activity are a sum of those incurred by the transport, treatment, and final disposal of the waste into terrestrial or aquatic environments. In particular, the extensive work of Doka (2024) has contributed significantly to understanding the environmental impacts of waste treatment processes and the modelling of the long-term impacts of disposal.

A considerable portion of a product’s total waste is generated during earlier stages of its supply chain such as resource extraction, transportation, and manufacturing, thus, often remaining ‘invisible’ to conventional LCA accounting practices (Laurenti et al., 2016). This oversight in measuring and communicating a cradle-to-grave product waste footprint (PWF) highlights a gap in circular economy indicators. Traditional LCA does not typically view waste as having environmental significance *per se*, considering instead only emissions and resource use resulting from waste treatment (Bisinella et al., 2024; Laurenti et al., 2023). The environmental significance of waste and its correlation with other indicators has been the subject of extensive research. Studies have shown that popular resource footprints can cover a significant portion of environmental impact variance between activities (Steinmann et al., 2017; Laurenti et al., 2023). Correlations between various environmental indicators are not always consistent, however, as seen with the carbon footprint, which often does not correlate well with other impact assessment scores (Laurent et al., 2012). The aggregation of waste in PWFs raises concerns among LCA experts, regarding the uncertainties introduced, as well as the potential misrepresentation of environmental performance due to differences in waste types (Chen et al., 2021; Huijbregts et al., 2010).

Some existing LCA methodologies offer direct indicators at the impact assessment level, but generally, there is little attention given to the impacts of waste in LCA (Oers et al., 2020). This limitation becomes particularly evident when attempting to identify waste generation hotspots within a product’s life cycle. Addressing these hotspots is crucial for advancing ‘circularity’, but there is a lack of convenient and flexible methods to calculate waste flows in LCA and a pressing need for more comprehensive methods that can effectively quantify waste flows and, therefore, contribute to a greater understanding of a product’s total environmental footprint.

Laurenti et al. (2023) developed a method to calculate the waste footprint of a product or service based on solving the demand vectors of the activities, while also proposing simple measures to quantify waste ‘hazardousness’ and ‘circularity’ with this data. The method presented, as the authors explain, suffers from errors due to double counting due to the way that exchanges are identified and ‘tagged’. In T-reX, the source database is deconstructed into a list of separate exchanges, and only those explicitly defined as hazardous are marked as such. The ‘T-reX’ Python software tool presented herein provides a more flexible, transparent, and user-friendly approach to quantifying waste flows in LCA. Moreover, the utility of T-reX is not limited to waste, it can be used to categorise and aggregate any technosphere exchange of interest (or customised grouping thereof), such as water, gas, and critical raw materials.

### Material demand in LCA

In the context of geopolitical tension and a mineral-hungry ‘renewable energy transition’, greater attention is being given to the security of material supply—especially for those considered ‘critical raw materials’ (CRMs) (European Commission and DG-GROW, 2023; Hool et al., 2023; Mancini et al., 2013; Carrara et al., 2023; Hartley et al., 2024; Salviulo et al., 2021; IEA, 2023a,b). While LCA seeks to model the technosphere (a.k.a. the anthroposphere) and its exchanges with the biosphere, its focus is often on the environmental impacts of the system rather than the primary material flows themselves.

The Crustal Scarcity (CSI) LCIA method was developed in order to introduce an assessment of the long-term global scarcity of minerals in LCA (Arvidsson et al., 2020). The CSI introduced crustal scarcity potentials (CSPs), which are measured in kg–silicon–equivalents per kg–element and derived from crustal concentrations. CSPs, provided for 76 elements, reflect the long-term global elemental scarcity based on crustal concentration proxies. The CSI, calculated by multiplying CSPs with extracted masses, effectively gauges the impact of elemental extraction.

While useful for its intended purpose, the CSI presents its results in an abstract unit (kg-Si eq.) that is difficult to interpret and compare with other impact categories. Furthermore, the CSPs are not available for all elements (or more complex materials), and the method does not allow for the quantification of material demands in terms of mass or volume.

# Methodology

This section is divided into two parts. In subsection 2.1, we describe the T-reX program, and in subsection 2.2, we describe the methodology used to calculate the waste and material inventory footprints in a case study of five Li-ion batteries, highlighting inconsistencies in LCA methodology and probing departures from projections in prospective models.

## T-reX

### Computational framework

Developed in the Python programming language, T-reX extends the *Brightway* LCA framework, utilising especially the components bw2data, bw2calc, and bw2io (Mutel, 2017a). Additionally, the wurst package—which can deconstruct databases into a list of exchanges—is used to facilitate searching and data transformation at the exchange level (the individual supply chain flows) (Mutel, 2017b). Integration with the premise package (Sacchi et al., 2022)—which integrates the projections of integrated assessment models (IAMs) into current LCA databases—enables the user to easily create and manipulate prospective LCA databases. T-reX is also compatible with *ActivityBrowser* (Steubing et al., 2020)—an open-source graphical user interface for LCA— and after running T-reX, the manipulated databases and the ‘pseudo-LCIA’ methods created by T-reX can be used in the project in the accustomed way. T-reX is installable via the Python Package Index (PyPI) (McDowall and Lanphear, 2023b) and is fully open-source under the CC-0 licence. The full source code for T-reX is indexed on Zenodo (McDowall and Lanphear, 2023a) and under further development in the GitHub repository (McDowall, 2024). T-reX is designed to be used with ecoinvent databases (Wernet et al., 2016), but could be adapted to other databases by changing the search criteria. Currently, it has been tested with all available system models of ecoinvent versions 3.5–3.10.

T-reX can be used directly from the command line, or imported as a Python package with which the user can access the individual functions and modules. In the simplest case, the user can run the program with the default settings, which will calculate the waste and material footprint of the ecoinvent database. The user can adapt the settings of TreX as desired, to calculate alternative or additional waste and material inventory footprints, use a custom database, or a ‘prospective’ database based on future scenarios by implementing the premise with one of T-reX’s internal modules.

The Supplementary Material item 1 contains metadata of T-reX, along with a list of the constituent modules, a description of their functions, and a detailed computational workflow. Further details can be found in the GitHub repository and in the package documentation (McDowall, 2024, 2023).

### Functionality and purpose

T-reX is a Python package that enables one to produce LCA databases—both current and prospective—that are manipulated to facilitate the calculation of waste and material inventory footprints in the supply chain of any activity in the same way as one would apply standard LCIA methods. We refer to the T-reX methods as ‘pseudo-LCIA’ methods, as they represent an accounting of an activity’s technosphere inventory without reference to impact modelling, as in standard LCIA.

If desired, prospective databases can be defined by the user or constructed with the projections of the integrated assessment models such as IMAGE (Stehfest et al., 2014) and REMIND (Aboumahboub et al., 2020), which offer a range of options aligned with the Shared Socioeconomic Pathways (SSPs) (Meinshausen et al., 2020) and can be paired with a variety of mitigation scenarios.

The deconstruction of the databases by T-reX into lists of exchanges allows the relevant flows of material and waste to be identified and categorised by the search functions. The queries facilitating this are tailored to the specific database and the user can easily modify them to suit their needs. The categories defined in the configuration are used to create T-reX’s ‘pseudo LCIA’ methods that are indicators of aggregated technosphere demand. The exchange editing function of T-reX then takes each list of exchanges and appends to the relevant activity a copy of the technosphere exchange as a mirrored ‘pseudo-biosphere’ exchange that matches the ‘pseudo-LCIA’ method.

In the default configuration of T-reX, there are 10 waste categories which are then further divided by their unit of measurement (kilograms and cubic meters) to create a total of 20 waste methods. The waste categories include aggregations such as ‘total waste’ ‘hazardous waste’ as well as, those for ‘waste sent to treatment’ by, for example, incineration or recycling. The full details are listed in the Supplementary Material in item 3. One advantage of T-reX is that it can identify ‘hidden’ waste exchanges that would otherwise be ‘consumed’ by a treatment process and not leave the technosphere.

In addition to these waste categories, T-reX can incorporate any number of material demand categories, the defaults being based on the EU Critical Raw Materials (CRM) list for 2023 (European Commission and DG-GROW, 2023), which contains 30 materials considered essential to the EU economy and also at risk of supply disruption. Especially valuable in the case of CRMs is T-reX’s ability to uncover and quantify these threads of background consumption, which can be inconspicuously embedded in products or consumed in seemingly distant supply chain activities. The identity of the materials considered, and their categorical groupings, are easily customisable by the user. The full list of 59 materials included in the default configuration is provided in the Supplementary Material in item 3.

Following the logic of the standard LCA models Wernet et al. (2016); Guinée et al. (2004), the identification of material exchanges with T-reX differs from that used to identify waste exchanges. That is, the search queries are based on the names of the relevant ‘market activities’ for the material of interest rather than keywords in the exchange’s name. A useful feature of T-reX is that, in cases where there are several markets for one material or material grouping, the program can aggregate these flows. For example, exchanges with markets for the rare-earth-elements (REEs) ‘market for cerium’, ‘market for dysprosium’, ‘market for erbium’, etc. can be aggregated into a single indicator category for REEs. Similarly, the total demand for all critical raw materials (CRMs) can be calculated in the same manner.

In these methods, the material demand is calculated based on the total mass that is extracted from the environment, thus, their focus is essentially solely on the mining-related exchanges that bring these materials from the biosphere into the technosphere. T-reX, in contrast, uses accounting for material demand and waste generation based on exchanges solely within the technosphere, a material basis. This offers a different perspective, allowing for the estimation of overall supply chain material demands that consider the entire life cycle of an activity, including non-direct effects on the market, such as a ‘negative consumption’ of other materials due their coproduction with the functional unit. Consider a demand for an activity containing a metal, for example; while the existing material use methods allow one to calculate the total mass of that metal that is extracted from the environment, T-reX can provide insight into the broader supply chain impacts of the demand for this metal. If the production of other materials is attributed to the production of this metal, these would appear as negative material demands in the T-reX results—supply chain pressure for one material can result in lessening of supply chain pressure for another. For example, in the results of the Li-ion battery case study, this is indeed the case for nickel demand, which, in the final inventory because of such co-production, is counterintuitively negative (due to co-production and substitution in the LCA system models) despite the presence of nickel in the final products.

### The workflow of T-reX

The workflow of T-reX is divided into several modules, each performing a separate function. The modules are designed to be used in a particular order, but the user can also use them individually to perform specific tasks. The standard workflow is as follows:

* + - 1. Configuration of waste and material exchange categorisation (optional)
      2. Generation of prospective LCA databases (optional)
      3. Database expansion—to create a list of all exchanges in the database
      4. Identification and categorisation of exchanges
      5. Creation of ‘pseudo-biosphere’ databases
      6. Creation of ‘pseudo-LCIA’ methods to calculate waste and material inventory footprints
      7. Exchange editing—whereby the technosphere exchange is mirrored as a ‘pseudo-biosphere’ exchange
      8. Database verification—to ensure that T-reX has manipulated the database correctly

This workflow creates a copy of the original *Brightway* project containing the original biosphere database, a TreX ‘pseudo-biosphere’ database along with one or more manipulated technosphere databases that can be used to calculate the waste and material inventory footprints of activities in the same way as standard LCIA calculations.

An overview of the T-reX workflow is presented in Figure 1. The Supplementary Material in item 3 provides more detailed information.

A diagram of a method

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Figure 1: The workflow of T-reX presented in a generalised (a) and computational (b) format. Application of T-reX to one or more LCA databases creates a *Brightway* project with customised methods and activities that can be used to calculate waste and material inventory footprints of the activity’s supply chain.

## Case study methodology

To provide an example, we investigated five types of Li-ion batteries in the ecoinvent 3.9.1 ‘cutoff’ LCA database (Wernet et al., 2016), each represented by their unamended global market inventories. The system boundaries. These following markets were used for the example calculations and unaltered by the authors:

* Li-ion, LFP, rechargeable, prismatic
* Li-ion, LiMn2O4, rechargeable, prismatic
* Li-ion, NMC111, rechargeable, prismatic
* Li-ion, NMC811, rechargeable, prismatic
* Li-ion, NCA, rechargeable, prismatic

In addition to the waste and material inventory footprint methods created by T-reX, the following standard LCIA methods were applied for comparison:

* ReCiPe 2016 v1.03, midpoint (I) (Huijbregts et al., 2016)
* Environmental Footprint (EF) v3.1 (no long-term emissions) (Andreasi Bassi, et al., 2023)
* Spatial differentiation in life cycle impact assessment—the EDIP 2003 no long term emissions method (Hauschild and Potting, 2004)
* The crustal scarcity indicator method—CSI 2020 (Arvidsson et al., 2020)

Additional to the present-day databases, T-reX created prospective database sets based on ecoinvent 3.9.1 using the functionality of premise with the REMIND model and the baseline scenario SSP2 ‘middle of the road’ with the following Representative Concentration Pathways (RCPs):

* SSP2-base: representing an approximate 3.5℃ increase in global temperatures to 2100
* SSP2-PkBudg500: representing the achievement of Paris climate goals, ca. 1.3℃ increase to 2100

For each pathway, databases (at five-year time intervals) were created with premise (Sacchi et al., 2022) and processed with T-reX for the period between 2020 and 2100.

### LCA calculations

For each combination of activity, method, and database, a total summation was calculated along with details of the top contributing processes. Additionally, for the T-reX methods, a contribution analysis was performed with the *bwa.compare\_activities\_by\_grouped\_leaves* function from the *Brightway2\_analyzer package* (Mutel, 2016), a component of the *Brightway* ecosystem. This function performs one subjective method of graph traversal on the impact matrix of the LCA object to a specified cutoff. The groups of the resulting ‘leaves’ are collected by their Cooperative Patent Classification (CPC) codes [giving a representation of industry sector involvement]. T-reX provides, essentially, a tabulation of internal exchanges and an insight into the products and sectors in the supply chain of the activity that carry the most substantial responsibility for the final waste generation or material demand footprints.

# Results

## T-reX tool

An example of the output from the application of T-reX has been included in the Supplementary Material item 5. The manipulated ecoinvent databases (which are the main product of T-reX) can be recreated by following the instructions in the package documentation (McDowall, 2023).

## Case study: Li-ion batteries

As described in subsection 2.2, this case study calculated the waste and material footprints (with a variety of other indicators) for the unaltered background inventories of five Li-ion batteries with the functional unit being 1 kg of the battery at the global market. The purpose of this simple case study was to test, verify, and demonstrate the functionality and limitations of T-reX. This section includes some highlights of the results, with full results included in the Supplementary Material (item 7 and item 8). Because the ‘pseudo-LCIA’ methods created by T-reX are integrated into the user’s *Brightway* project as if they were standard LCIA methods, the footprint calculations can be performed in the customary way. In the Supplementary Material, there are screenshots of selected results obtained using the *ActivityBrowser* software, including contribution analysis and a Sankey diagram that dis-aggregates the final footprint result over the activities in the supply chain.

### Sankey visualisation of flow inventory footprints

Figure 2 presents a heavily abridged Sankey diagram depicting the supply chain flows of ‘total solid waste’ attributed to the NMC 811 Li-ion battery at the global market in the ecoinvent database version 3.9.1. As the supply chains are expansive, this visualisation has been greatly simplified to be legible for print, multiple versions are included in the Supplementary Material in item 8. As is evident in the figure, the largest contributors to the waste footprint are the extraction and processing of the raw materials, with the largest flows coming from the extraction of cobalt, lithium, and nickel. Due to the nature of LCA, the waste flows do not represent the ‘real’ physical flows directly, but rather an abstraction thereof based on, in this case, economic allocation—the market price of the functional unit produced by the activity (Guinée et al., 2004). If an unallocated database were available, the waste flows could be calculated with T-reX in the same way and would be more directly related to the physical flows of the activities.

### Temporal and scenario variation in waste and material inventory footprints

Figure 3 (a) shows the ‘total solid waste’ inventory footprint for the five Li-ion batteries in the case study from 2020 to 2100 under the SSP2 scenario using the baseline and PkBudg500 RCPs of the REMIND model. The NMC811 battery has the largest footprint, producing over 50 kg of waste per kilogram of battery produced. The LiMn2O4 battery has the smallest footprint, producing less than 4 kg of waste per kilogram of battery. In each case, there was only a slight downward trend in the waste footprints between 2020 and 2100. This is mostly in the period between 2020 and 2040 and could be attributable to the relatively rapid decrease in fossil fuel use that is factored into the models over this time (since many produce relatively large amounts of waste during extraction and combustion). For the total waste generated by these batteries, there was very little difference observed between the baseline and PkBudg500 RCPs.

A diagram of a production process

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Figure 2: A greatly simplified Sankey diagram from *ActivityBrowser* showing the total solid waste footprint flows for the NMC 811 battery supply chain at the global market in the ecoinvent database version 3.9.1

The inclusion of carbon capture and storage (CCS) in the prospective databases is apparent in Figure 3 (b), with the rapid increase in the production of carbon dioxide ‘waste’ over the period from 2020–2040 that is not seen in the baseline scenario. This result highlights the fact that the (often) downward trends in global warming impacts calculated with prospective databases using standard LCIA methods are dependent on the assumptions made about the introduction of CCS technology. The actual deployment of these technologies was only around 37 Mt CO2/yr as of 2023 (Dziejarski et al., 2023), far short of the levels projected in many of the RCP scenarios (Sacchi et al., 2023). The authors consider that the over-representation of CCS and the under-representation of waste processing technologies in the prospective databases is one significant limitation to the validity of prospective LCA databases. Significant work is needed to improve the representation of these technologies in the databases to ensure that the results are more realistic and useful for decision-making.

A graph of a function

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Figure 3: T-reX calculated supply chain inventories for the five Li-ion batteries in the case study, showing their ‘footprints’ for:

(a) total solid waste, and (b) carbon dioxide waste (produced from carbon capture and storage (CCS) only). The footprints were modelled with prospective LCA databases from 2020 to 2100 under the SSP2 scenario using the baseline and PkBudg500 RCPs of the REMIND model.

### Contribution of ‘top-processes’ in the supply chain

Figure 4 shows the contribution of the ‘top-processes’ to the graphite footprint of the LiMn2O4 battery under the baseline scenario from 2020–2100. The total footprint is seen to more than triple, from 0.16 kg/kg in 2020 to 0.52 kg/kg in 2050 and onward to 2100. The largest contributor is not, in this case, the battery-embedded graphite electrode (which is the second top process), but ‘silicon production’ from further up in the supply chain. This result is likely a reflection of the electrification of the transport and energy sectors that is included in the REMIND model. The results of this kind of analysis with T-reX can be used to identify the most significant processes in the supply chain and to target interventions to reduce the potential environmental impact of production.

A graph of a graph showing the number of different colored bars

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Figure 4: Contribution by the ‘top-processes’ to graphite demand inventory of the LiMn2O4 battery from 2020 to 2100 under the SSP2 scenario using the baseline RCPs of the REMIND model

### Contribution of industrial sectors in the supply chain

Figure 5 shows the contribution of sectors (grouped by CPC codes) to the total natural gas footprint of the LFP battery under the PkBudg500 pathway. In this example, the relative contribution of the sector ‘47160: Electronic integrated circuits’ is seen to decrease from 11% in 2020 to 8% in 2100, while over the same period the contribution of the sector ‘46430: Parts of primary cells, primary batteries and electrodes’ increases from 32% to 38%. The method used to calculate these contributions involves traversing the supply chain branches to a certain level (max. 4, in this case), cutting a specified point (5% in this case), and grouping the value of the ‘leaves’ by their CPC code. The results, therefore, will depend on how deeply the user would like to inspect the supply chain. Additionally, the utility of these results is dependent on how well the CPC codes define the processes in the supply chain for the particular case.

A graph of different colored bars

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Figure 5: Contribution of industrial sectors to the natural gas footprint of the LFP battery from 2020 to 2100 under the SSP2 scenario using the baseline RCP of the REMIND model

### Comparison with ‘similar’ LCIA methods

A comparison of the results for ‘Coal (black)’ demand using the T-reX method with the LCIA method ‘EDIP 2003 – coal no LT’ is shown in Figure 6. In this case, both the trends and the magnitude of the scores were very similar, both demonstrating a general decrease in the use of coking coal in the battery’s footprints. Such comparability was also observed for other fossil-fuel-related demands (e.g., natural gas and petroleum), and to a lesser extent, for some of the metal demand methods (e.g., zinc and cobalt). Correlation between standard LCIA methods and T-reX methods is not generally to be expected, however, due to fundamental differences in the way that the methods are constructed. In standard LCIA methods, impact scores are derived exclusively from the magnitude of the exchanges between the biosphere and the technosphere, that is, extraction and emission. In the T-reX methods, the footprint scores are based on an accounting of either waste generated or material demand, both of which are technosphere exchanges in terms of LCA modelling. For the material demand methods especially, this distinction is critical. For example, the application of the EDIP 2003 or the CSP methods for a given metal will provide a score that is proportional to the amount extracted by mining, whereas the T-reX method provides an aggregation of the exchanges with the market for that metal. The T-reX method, therefore, considers cases of co-production, recycling, and substitution, providing a picture of the supply chain pressures that are not captured by the standard LCIA methods. This makes the T-reX methods more sensitive to the modelling choices (e.g., economic allocation) that are generally embedded in LCA databases.

A graph of different colored lines

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Figure 6: Comparison of the (a) ‘T-reX – Coal (black)’ demand footprint with the (b) conventional LCIA method ‘EDIP 2003 – coal no longterm’. In the case study from 2020 to 2100 under the SSP2 scenario using the baseline and PkBudg500 RCPs of the REMIND model for all five batteries.

### Comparison with other studies

In the most similar published method, Laurenti et al. (2023) used an alternative procedure for calculating the waste footprint of 1̃400 activities in the ecoinvent database version 3.5. Direct comparison is not possible, as this database contains only one generic Li-ion battery, ‘market for battery, Li-ion, rechargeable, prismatic’. The inventory of this battery most closely resembles that of the NMC 111 battery in this case study. Table 2 presents a comparison of the results from the two studies, where possible. For liquid, solid, and recycled waste, the results were closely aligned, however, for the hazardous waste fraction, Laurenti et al. reported 95%, whereas T-reX reported only 3%. The reason for this discrepancy is explained by the fact that in the method of Laurenti et al. a “waste flow was regarded as hazardous when the hazardousness was clearly stated in its subsequent waste treatment activity”. The authors continue: “It should be noted that the high hazardousness ratios for many products might indicate a weakness in the validity of this measure”. In T-reX, the source database is deconstructed into a list of separate exchanges, and only those explicitly defined as hazardous are marked as such.

Table 2: Comparison of the results from the T-reX battery case study (database: ecoinvent 3.9.1 REMIND SSP2 Base 2020, activity: Li-ion NMC 111) with those from Laurenti et al. (2023) (database: ecoinvent 3.5, activity:‘Li-ion’)

|  |  |  |
| --- | --- | --- |
| Indicator | T-reX | Laurenti et al. |
| Total solid waste (kg/kg) | 50.9 | 62.5 |
| Total liquid waste (kg/kg) | 3.53 | 3.63 |
| Hazardous waste (kg/kg) | 1.47 | 62.6 |
| Recycled waste (kg/kg) | 1.59 | 1.98 |

# Discussion

Waste generation and material demand are strongly associated with the environmental impacts of human activities (Laurenti et al., 2023; Steinmann et al., 2017; Demirer, 2019), and thus must be represented in LCA accounting. Although there are existing LCIA methods that provide endpoint impact scores related to material demand and waste generation, they contain convoluted formulae or subjective weighting, or their complexity and lack of transparency can make them difficult to use and interpret (Swiss Federal Office for the Environment (FOEN), 2021; Hauschild and Potting, 2004; CEN (European Committee for Standardization), 2019; Arvidsson et al., 2020; Swiss Federal Office for the Environment (FOEN), 2021).

T-reX advances the state-of-the-art in LCA by providing practitioners with a simple, flexible, and transparent way to calculate supply chain waste and material footprints, delivering results in standard units and as direct aggregations of the relevant demand inventories. Once LCA databases in the Brightway project have been processed with T-reX, the user can easily apply (and re-apply) the T-reX ‘pseudo-LCIA’ methods to calculate the waste and material inventory footprints in the same way that they would with a conventional LCIA method.

The simple case study of five Li-ion batteries presented in this paper demonstrated the utility, flexibility, and limitations of T-reX. By adjusting the user configuration for future scenarios and waste/material categorisation, we produced a set of customised versions of current and prospective ecoinvent databases along with a ‘pseudo-biosphere’ T-reX database. Then, by applying the T-reX ‘pseudo-LCIA’ methods, categorised waste and material inventory footprints for present and future supply chains were trivial to calculate. Additionally, visual exploration in *ActivityBrowser* was possible as the T-reX ‘pseudo-LCIA’ methods are structurally identical to standard LCIA methods when integrated within the Brightway project.

One limitation of T-reX is that it does not yet provide specific information in a readily accessible format on the composition of the waste generated. This information is required to thoroughly assess the potential environmental impacts of this waste. Currently, the user would need to manually explore the waste footprint inventory produced by the application of the T-reX to determine the nature of the waste flows. For example, if the waste generated could represent an actual loss of resources or potential environmental risk, or, as is often the case, simply a transfer of the ‘overburden’ in mining activity, which is classified as ‘inert waste’. A methodical classification of waste exchanges and end-of-life fates is expected to be facilitated by the increasingly detailed and disaggregated data in each successive release of ecoinvent (FitzGerald et al., 2023).

The utility of T-reX in studies of future supply chains is limited by the fact that the currently available prospective databases focus largely on changes in the energy, steel, cement, and transport sectors (Sacchi et al., 2023). As demonstrated in the results of the case study—where there were often very few scenario-temporal changes in many waste and material footprint indicators—the utility of prospective LCA is restricted if there is inadequate adoption of future background inventories. In particular, the inclusion of scenarios with future waste processing technology could greatly improve our predictions of waste and material flows and offer valuable insight into their potential impacts (Bisinella et al., 2024). A strong focus on enhancing these prospective databases is, thus, of critical importance to the future of prospective LCA, and by extension, to the development of the circular economy.

# Conclusions

T-reX is an extension to the open-source *Brightway* ecosystem that enables users to calculate the waste and material footprints of any product or service in an LCA database. It deconstructs, manipulates and reconstructs the database to identify and edit waste and material exchanges and writes matching custom ‘pseudo’ LCIA methods. These exchanges are mirrored in the *Brightway* project as ‘pseudo-biosphere’ flows and thus, the waste and material footprints of any other supply chain flow of interest, such as water, gas, or critical raw materials, can be calculated similarly to existing LCIA methods.

It is essential that we elucidate the complex relationships between material demand and waste generation and environmental and social harms. The software we present in this article enables detailed exploration of ‘footprints’ and life cycle inventories—modelled externalities of human activities—and examines their potential connections threats such as supply chain risks and environmental damage. T-reX is a step towards a better understanding of these interconnections for the practical realisation of sustainability and the circular economy.

The Supplementary Material supplied with this manuscript contains the following sections:

1. Metadata of the T-reX Python package
2. Details of the modules and user configuration
3. Description of the computational workflow
4. Modular flowchart of T-reX
5. Example of the terminal output of T-reX
6. Example of the terminal output of the case study
7. Complete tabulated results of the case study
8. Complete visualisations from the case study

# Data availability

All data used is publicly available online under the noted sources. T-reX is fully open source and dedicated to the public domain under the CC-1.0 licence. It is installable via the Python Package Index (PyPI) under the name ‘T\_reX\_LCA’ https://pypi.org/project/T\_reX\_LCA. The full source code for T-reX is archived by Zenodo at https://zenodo.org/records/10925359 and the development server is hosted by GitHub at https://www.github.com/Stew-McD/T-reX. A user guide with comprehensive documentation are presented at https://T-reX.readthedocs.io.

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# Declaration of competing interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

# CRediT authorship contribution statement

**Stewart Charles McDowall:** Methodology, Software, Validation, Formal analysis, Investigation, Data curation, Writing: original draft, Writing: review & editing, Visualization. **Elizabeth Lanphear:** Conceptualization, Methodology, Software, Validation, Writing review & editing. **Stefano Cucurachi:** Conceptualization, Writing review & editing, Supervision, Funding acquisition. **Carlos Felipe Blanco:** Conceptualization, Formal analysis, Methodology, Writing review & editing, Supervision.

A close-up of a computer screen

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Figure 7: CRediT authorship visualisation

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