Department of Environmental Science

### Zoran Steinmann Estimating missing data How to fill data gaps in life cycle inventories?

#### LCA Discussion forum, Zürich

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Radboud University Nijmegen

#### **Problem setting**

LCA: environmental impact of a process/product Requires considerable amount of data

# Common problem: lack of data in inventory and impact assessment

Typically make assumptions to address the issue

Questions

- What can we do to fill these data gaps?

- How much data do we really need in impact assessment?

#### Plant specific carbon footprints of fossil power plants

Dominant source (67%) of global electricity generation

#### Relatively high CO<sub>2</sub> emissions per kWh generated



Image Source: http://news.thomasnet.com/green\_clean/wp-content/uploads/2011/12/Lifecycle\_CO2\_Parliament\_Small.png

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### (Lack of) data availability; coal plants

- Emissions data available for 445 plants (495 GW) worldwide
- Sources: government databases, private utilities, operators

Distribution of available emissions data, by plant capacity [as a fraction of total installed capacity in each country]



# Solution $\rightarrow$ Model based on readily available data



#### **Regression model**

Generalized linear regression (GLM) model:

 $log(CF) = \beta_0 + \beta_1$  · Plant age +  $\beta_2$  · Plant capacity +  $\beta_3$  · Fuel type + ...

Data on specific power plants:

WEPP (World Electric Power Plants) database: almost all operational power plants >100 MW in the world

Power plant design details such as: location, capacity, number of generators, year of construction, et cetera...

After fitting and model validation, the model was applied to 2538 gas oil fuelled power plants and 764 coal fuelled power plants

### Plant specific predictions for oil and gas



 $R^2 = 0.81$  (validation set)

Type of fuel most important predictor by far

Smallest uncertainty in prediction for Combined Cycle Natural Gas

Largest uncertainties for Single Cycle oil plants

## Plant specific predictions coal fuelled power plants



 $R^2 = 0.49$  (validation set)

Steam pressure and lignite/non-lignite most important predictors

Model cannot reliably predict very high carbon footprints (>1.5 kg  $CO_2/kWh$ )

#### Direct and modelled data availability; coal plants



# **Application: world-wide distributions of carbon footprints**



Carbon footprint (kg CO2eq/kWh) gas and oil fuelled power plants

Carbon footprint (kg CO<sub>2</sub>/kWh) coal fuelled power plants

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#### How much data do we really need?

# Do we need data on 1500+ interventions and 100+ indicators for 15 different impact categories?

Impact assessment method	Number of indicators	Category
CML2001	49	Midpoint indicators
EDIP 2003	21	
Impact 2002	14	
ReCiPe	31	
TRACI	9	
Ecolndicator99	3	Endpoint indicators
Ecological scarcity2013	1	
EPS2000	1	
Impact 2002	3	
ReCiPe	3	
Fossil energy footprint	1	Resource footprints
Water footprint	1	
Material footprint	1	
Land footprint	1	

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#### How much data do we really need?

PCA: Principal Component Analysis

Based on the rank correlations between indicators

Variance rewritten as a set of linear transformations of the original data

The first component covers the maximum amount of variance, consecutive components cover less

First couple components cover the majority of the variance  $\rightarrow$  potential for reduction



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To explain >90%:

3 principal components

But: all original indicators are needed to calculate these....

So: How many original indicators?

→ use regression on PCA scores

#### How much data do we really need?



To explain >90%: 3 principal components

OR

Best 4 Indicators:

- Climate change
- Ozone depletion
- Land footprint
- Marine ecotoxicity

3 Resource footprints: Energy, Land, Material **83.8%** 

### **Discussion**

Regression approaches are a strong tool to fill data gaps Other application possibilities, e.g. farm specific or waste water treatment facility specific emissions

Model uncertainty is introduced through the regression model

We can reduce the number of impacts, but:

This is a purely statistical tool, however RACER (Relevant, Accepted, Credible, Easy and Robust) criteria are also important

### Thank you for your attention

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References:

Steinmann, Zoran JN, et al. 2014. How to address data gaps in life cycle inventories: A case study on estimating CO2 emissions from coal-fired electricity plants on a global scale. *Environmental science* & *technology*, 48(9): 5282-5289.

Hauck, M. et al. 2017. Estimating the Greenhouse Gas Balance of Individual Gas-Fired and Oil-Fired Electricity Plants on a Global Scale. *Journal of industrial ecology*, 21(1):127-135

Steinmann, Zoran JN, et al. 2016. How many environmental impact indicators are needed in the evaluation of product life cycles? *Environmental science & technology*, 50(7): 3913-3919.