# DATA MANAGEMENT FRAMEWORK FOR MONITORING AND ANALYZING THE ENVIRONMENTAL PERFORMANCE

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Abstract: Monitoring the environmental performance of a product is recognised to be increasingly important. The stakeholders are pressuring the manufacturers for improved information about the environmental burden caused by the manufacturing of a product. However, there are problems to accurately quantify the environmental burden of an individual product because the supply chains are dynamic. In this paper we present a model that enables calculating and monitoring the environmental performance of products at an item level in a dynamic supply chain and performing multidimensional analysis of environmental data.

### **1 INTRODUCTION**

A physical product has its own supply history represented as a supply chain. In practice, the history of the product is the history of its parts composed in the supply chain. The precision of traceability of products depends on how detailed the history of the components can be traced. This paper deals with the tracing of the emissions and resources of products and their components. We give a logical framework for tracing and analyzing the emissions and resources even single products but as well larger patches.

The supply chain can be defined as a network of autonomous or semiautonomous business entities collectively responsible for procurement, manufacturing and distribution activities of a product. In recent times the importance of environmental aspects has been widely recognised. The valuation of environmental impacts caused by the production of products and services is becoming more and more important.

The problem with measuring the environmental impact caused by a product at the item level is that supply chains are dynamic. A manufacturer can use various subcontractors and supply various end manufacturers or retailers in different countries. For example a product that is transported from another continent to a supermarket is bound to have different environmental impact than another product that is transported to a supermarket from a nearby producer.

However the common method of calculating the environmental impact on a product is to measure the resources used, emissions and production in some time period and calculate the average environmental impact on the product. This does not take the dynamic nature of the supply chains into account.

To be able to track the objects through the dynamic supply chain, the products must be identified. The development of an auto identification enables us to identify an object moving in the supply chain. This means that we can connect the physical world objects with their virtual counterparts in databases. With the traceability we can track the relationships among properties of processes, in this case the environmental burden caused by processes, and actual product instances.

In this paper we demonstrate a model which can be used to allocate the environmental burden to individual products. Unlike existing methods our model enables analyzing environmental impact on the product level – not only average values. The model supports for monitoring emissions (e.g.  $CO_2$ ) and resources (e.g. Energy) in any precision level only depending on how precisely physical products

Sirkka A. and Junkkari M. (2010). DATA MANAGEMENT FRAMEWORK FOR MONITORING AND ANALYZING THE ENVIRONMENTAL PERFORMANCE. In *Proceedings of the Multi-Conference on Innovative Developments in ICT*, pages 57-62 DOI: 10.5220/0003046400570062 Copyright © SciTePress and patches can be identified and monitored. Further, our approach enables multidimensional analyses of data associated with the emissions and resources of products and their components.

The rest of paper is organised as follows. Section 2 deals with environmental accounting. In Section 3 we introduce the *traceability graph* which is the basis of our model. The implementation and usage of the traceability graph is presented in Section 4. Then we demonstrate how information associated with the traceability graph can be used in multidimensional analysis called the *traceability cube* in Section 5. Finally, the conclusions are given in Section 7.

## 2 ENVIRONMENTAL PERFORMANCE MONITORING

Nowadays the Environmental performance is being monitored in most organizations at the company level, resulting a total impact for the whole company. The most used methodology is the Greenhouse Gas Protocol which is a guideline for estimating the greenhouse gas emission of an organization. This kind of total organisation value for greenhouse gas emission can't be used for measuring an environmental impact for a certain product or service because all the emissions are calculated together and the emissions are not correctly allocated to products. Also the total emissions for the life cycle of a product are not calculated.

The product level measuring of environmental performance is under development and many different approaches are used for calculating the environmental impact of a product or service. There are many studies made about the different approaches (Usva, Hongisto et. al 2009 and Dada et. al 2009, 2010). The most important of these are the international standards of life cycle assessment (LCA) (ISO 14040 series) and eco-labels (ISO 14020) and verification (ISO 14064) and Publicly Available Specification (PAS) 2050 which builds on ISO 14040 and 14044 standards by specifying requirements for the assessment of the greenhouse gas emission. The international standards organization has also started a subcommittee for developing the standards for Quantification and Communication of Carbon footprint of product (ISO 14067).

The LCA has four main phases. In the first phase, the goal and boundaries of the life cycle assessment are defined. This means that we define the processes that we will perform the study on. In the second phase, called life cycle inventory analysis, input and output flows of the underlying processes are defined, collected and calculated. If a process produces more than one product, an allocation is also needed. The third phase of the LCA is impact assessment. First, the relevant impact categories (e.g. Climate change, Ozone depletion or Acidification) are selected. Then, the results of life cycle inventory analysis are assigned to the selected impact categories. For example, the carbon dioxide is a greenhouse gas and is thus assigned to the Climate change category. The last main phase is interpretation where the conclusions of the analysis are made.

In this paper we present the model for tracing and storing the life cycle data about product manufactured in dynamic supply chain. Unlike the existing methods our model enables analyzing resources and emissions at the single product level – not only average values. This is achieved by allowing gathering real monitored activity data from supply chain processes.

## **3 TRACEABILITY GRAPH**

The traceability graph is used to model the supply processes of physical products and resources and emissions associated with the products and their components. The traceability graph has the ability to manipulate products and the transformations of the products. For example a product may be composed from many parts or a product may be manufactured using masses of raw materials. The traceability graph has also the ability to manipulate the properties of processes and to allocate them to products that are handled in that process.

The traceability graph can be presented using nodes and edges and their properties. A node is used to describe a supply chain process. An edge is used to describe a product flow between processes.

#### 3.1 Supply Chain

The supply chain can be viewed as supply processes following each other in a partial order. A manufacturing process is an event that transforms the input elements (raw material, energy) into output elements (product, waste and emissions).

In a traceability graph processes can be grouped

based on their process types, i.e. similar processes are instances of a process type. Within a process type the specific properties of processes, such as timing, placing etc., may vary.

In Figure 1 there are seven process types (A, B, ..., G). Process type A has four instances. These nodes have no predecessor which means that the traced objects have been created in these nodes. The objects are transferred forward in the graph. For example, objects from Nodes A1 and A2 are transferred to Node B1. Now objects are not changed but object sets (product portions) of A1 and A2 are unionized in B1. This also means that the resources and emissions of A1 and A2 are cumulated to the new set of objects. The objects from B1 are transferred to the Node C1 where products are classified and sent to one of the D processes.

In the D processes objects are divided into several objects. A double-headed arrow illustrates this. For example a physical object is decomposed or divided into parts. Then, parts may be classified and sent to forthcoming processes. The E nodes receive product portions consisting of these parts. In an E process they are refined and sent to Node F1 which is a shared process for all products. The products of F1 are components for the G processes, i.e. in G1 and G2 objects are composed from the objects that F1 yields. A shared start arrow illustrates this.



Figure 1: A sample traceability graph.

In a traceability graph it is possible to trace the supply chain of an object, i.e. to find all the preceding processes where the object at hand has participated. This also means that all the information related to those processes can be attached to the object. Given the running example, let us assume that we are interested in an object that belongs to Node G1. Then, the processing history of the object is a subgraph of the main graph. In Figure 2 the colored nodes are processes in which the underlying object, its part or a whole related to the parts has participated. In the example, parts of the underlying object have gone through F1, E2, E3 and D2, whereas the larger objects consisting of the parts have gone through C1, B1, A1 and A2. This subgraph is also the supply chain of the underlying object.

The traceability graph can also be used for analyzing different aspects on processes. For example a process type can be selected and we can see how much some process causes the environmental burden. Further, this analysis can be done in a supply chain of one object or a set of objects.



Figure 2: The supply chain of an objects belonging to Node G1.

#### 3.2 Data Management Primitives

Next, we introduce the properties of edges and nodes of the traceability graph. For the reason of the limited space the detailed and exact definitions of the primitives associated with the traceability graph are not given explicitly.

An *object* is the unit of tracing in a phase of the related supply chain. This can be a single product or a patch depending on the precision of tracing in the underlying supply system.

A process node contains the identity of a process, the set of product portions and the set of attributes associated with the process. A *product portion* involves the quantity of products, the identifiers of objects and the ratio of the emissions and resources compared with the total ones in the process node. The ratio is calculated by an application specific method. It can be based on the portion of mass or used time of machines, for example. Product portions of a process.

An *attribute* of a node determines information associated with a process. In terms of input attributes it can be described the resources of a process whereas output attributes can be used for determining the emissions of a process. Each attribute has two values: one for the underlying process and the other for containing the cumulated values from the previous nodes. A cumulated value is calculated based on the ratios of product portions and quantity that is sifted from the previous nodes via edges.

Via edges, products are sifted from a process to another more precisely from a product portion to another. An edge also determines the mapping of objects between two processes. The mapping can be:

- 1. Equivalence: Objects from a start node of a product portion are sifted to the related product portion of the end node.
- 2. Subsetting: Only some objects are sifted to the related product portion of the end node.
- 3. Supersetting: All the objects are sifted to the related product portion of the end node but the product portion of the end node contains similar objects from another process node.
- 4. Division: Objects of a start node are divided into smaller objects. If an object represents a single product, this is portioned.
- 5. Composition: Products of the start nodes are components for the end node.

In 1-3 the objects maintains their identities but in 4 and 5 the identities must be changed. In case 4 the identity of a product is mapped with the identities of parts that are produced from the product. In case 5 several objects are needed used for one composition, i.e. the identities of components are mapped with the identity of the related composition. It is worth noting that a product of an end node may contain components from several start nodes.

Through an edge the information of sifted products from a node to another node is transferred to an end node following the mapping of objects. An edge involves those objects that are sifted from a start node to the end node (only some products of a product portion may be selected from other processes). This part of the product portion of the start node is called a *sifted product portion*. In transferring products from a process to another, the attributes must be re-calculated for corresponding to the sifted product portion. This is based on the ordinary and *derived attributes*. The derived attribute is associated with an edge and it determines the amount of an ordinary attribute that is related to the sifted product portion.

### 4 IMPLEMENTING TRACEABILITY GRAPH

The traceability graph is mapped to the relational database as presented in Figure 3. We selected the

relational database because the standard OLAP (Online Analytic Processing) methods (Chaudhuri and Dayal, 1997) are used to further analyze the huge amount of data that is a result for tracing the individual objects. In Figure 3 PK means primary key and FK means foreign key.

The information of the traceability graph is stored into eight relations:

- Node relation is used to store the identities of process nodes.
- Attributes (e.g. raw materials, energy) are stored into Attribute relation
- Product types are stored into Product relation.
- The relation NodeAttribute is used to store the process (Node) specific attributes. For example (Process#1, Electricity, 100 kWh) specifies the use of electricity of Process#1.
- The relation NodeProduct is used to store the information about product portions of a specific process (Node). The column Ratio is used to allocate the environmental burden between the portions of products and by-products. For example (Process#1, Product#1, 0.6) specifies that Product#1 is an end product of Process#1 and the related ratio is 0.6.
- The relation Object is used to store the object specific information like physical code of the object and its volume.
- The relation ObjectRelation is used to store the object mapping when object identities are changed. The column Transformation function is used to calculate the cumulated attribute values.
- The route of the objects through a supply chain is realised by the Route relation. This corresponds to the sifted product portion.



Figure 3: Database schema for the traceability graph.

The relation model in Figure 3 can be easily extended to include more product and supply chain specific information. For example, we can implement an organisational hierarchy by creating Process, Site and Organisation relations (Node $\rightarrow$ Process  $\rightarrow$  Site $\rightarrow$  Organisation). This kind of extension enables analysis of environmental data by using the hierarchy as a dimension in multidimensional OLAP model.

## **5 TRACEABILITY CUBE**

To be able to use the OLAP type operations for analyzing the information of the traceability graph we must combine the previous tables as a data cube. In this work we will use the multidimensional data model "MD" that is presented in (Torlone, 2003). In Figure 4 the dimensions are presented as roundcornered boxes, the facts are presented as boxes and the measures as circles. The circles with drawn with dashed line presents calculated measures.



Figure 4: Traceability cube scheme.

Figure 4 presents the traceability cube with some example dimensions. The Process dimension can be used to compare the environmental impact between manufacturing sites and manufacturers. The Object dimension is used to aggregating the environmental data for different product groups. The measure Flow Amount is the attribute amount for an object. The measure Volume specifies the volume of an object. In Table 1 some sample instances over the traceability cube are presented.

The EPC column is the unique identity of an object.

Table 1: A sample instance over the traceability cube.

EPC	Process	Site	Product	Day	Month	Year	Flow	Amount	Volume
1	Kiln Drying	Mill#1	Timber	1	1	2010	Electricity	0,5 kWh	0,30 m <sup>3</sup>
2	Kiln Drying	Mill#1	Timber	1	1	2010	Electricity	0,5 kWh	0,35 m <sup>3</sup>
3	Kiln Drying	Mill#1	Timber	1	1	2010	Electricity	0,5 kWh	0,33 m <sup>3</sup>
1	Transporting	Truck#1	Timber	2	1	2010	Lorry Transport	100 km	0,30 m <sup>3</sup>
2	Transporting	Truck#1	Timber	2	1	2010	Lorry Transport	100 km	0,35 m <sup>3</sup>
3	Transporting	Truck#1	Timber	2	1	2010	Lorry Transport	100 km	0,33 m <sup>3</sup>
4	Kiln Drying	Mill#1	Timber	1	1	2010	Electricity	0,53 kWh	0,40 m <sup>3</sup>
5	Kiln Drying	Mill#1	Timber	1	1	2010	Electricity	0,53 kWh	0,42 m <sup>3</sup>
6	Kiln Drying	Mill#1	Timber	1	1	2010	Electricity	0,53 kWh	0,38 m <sup>3</sup>
4	Transporting	Train#1	Timber	2	1	2010	Rail Transport	200 km	0,40 m <sup>3</sup>
5	Transporting	Train#1	Timber	2	1	2010	Rail Transport	200 km	0,42 m <sup>3</sup>
6	Transporting	Train#1	Timber	2	1	2010	Rail Transport	200 km	0,38 m <sup>3</sup>

The Flow Amount is used for calculating the calculated measures - amount of emissions (e.g. carbon dioxide, methane, nitrous oxides) and amount of key environmental performance indicators (see e.g. Lim and Park, 2009). The emission amount is the amount of emissions caused when using elementary flow (raw material or energy) in some process. For example, carbon dioxide emissions when using electricity from Tampere electricity station in Finland were 194 g / kWh in the year 2008. There are many environmental databases that comprise life cycle inventory data from different supply chain processes. For example, the ELCD core database by European Commission - DG Joint Research Centre -Institute for Environment and Sustainability comprises more than 300 process datasets (e.g. key materials, energy carriers, transport, and waste management).

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EPC	Process	Site		Product		U	IVI	rear	FIOW		volume	
2	Transp.	Truck#1		Timber		2	1	2010	Lorry Transport		100 km	0,35 m³
4	Transp.	p. Truck#2		Tin	nber	2	1	2010	Rail Transport		200 km	0,40 m <sup>3</sup>
$\checkmark$												
								Impact				
EPC	Emissic	n	Amou	ınt	Unit			Categ	ory	Amount	Unit	
2	CO2		19,91	.02	kg	-	$\rightarrow$	GW	Ρ	19,9102	as kg of	CO₂ eq.
2	CH₄		0,196	36	g	$\rightarrow$		GWP		0,004909	as kg of CO <sub>2</sub> eq.	
2	N₂O		0,15401		g	-	÷	GWP		0,045895	as kg of $CO_2$ eq.	
4	CO2		22,11	.93	kg		$\rightarrow$	GWP		22,1193	as kg of CO₂ eq.	
4	CH₄		0,139	33	g	-	$\rightarrow$	GWP		0,003483	as kg of CO₂ eq.	
4	N <sub>2</sub> O		8,638	72	g	-	$\rightarrow$	GW	Ρ	2,574338 as kg of C		CO₂ eq.

In Table 2 the emissions and impact category for objects with code 2 and 4 are presented. Key environmental performance indicators are calculated based on the emissions. In this example we use the global warming potential which is one commonly used an environmental key performance indicator. It is calculated based on carbon dioxide, methane, nitrous oxide and several other emissions. The measurement unit for the global warming potential is kg of carbon dioxide equivalent which means that all the other emissions are converted by using a conversion factor. For example the conversion factor of Methane is 25. Full list of emissions and factors can be found from PAS 2050 (Carbon Trust 2008).

The analytics capabilities of the traceability cube can be used for analyzing the environmental data.



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Figure 5: Using the traceability cube.

For example, environmental data can be summed up to create the total environmental impact for the whole life cycle of the product. The data can also be used for comparing the performance between different manufacturers or manufacturing sites as illustrated in Figure 5. The possibility to analyze the supply chain on the process and item level allows the end users to select a product which creates least environmental burden. This creates pressure for the manufacturers to improve the eco-efficiency of their supply chains.

#### 6 **DISCUSSION**

The precision of traceability of the resources and emissions depends on the underlying data model and ability how strictly physical products and their components can be identified. Our model can be applied to any granularity of tracing. For applications, it is required physical identity mechanism that can be mapped to their logical counterparts in the database.

One option for marking the objects is Radio-Frequency Identification (RFID) technology which can be compared to the bar code identification: an identification code is embedded to an object. In the RFID technology the identification process does not require a clear line of sight. The potential of the RFID technology to monitor the carbon footprint of products is demonstrated in (Dada et al., 2009, 2010; Ilic et al., 2009).

### 7 CONCLUSIONS

We presented a model how emissions and resources can be monitored from the data management perspective. The model can be mapped to any precision level of physical tracing. At the most precise level, even a single physical object and its components can be analyzed. This, of course, demands that the related objects and their components are identified and mapped to the database. From the opposite perspective our model also supports rough level analysis of products and their histories. We showed how multidimensional analysis can be applied for OLAP analysis based on the traceability graph.

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