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Digital Personalized Health and Medicine

# Exploring Patient Path Through Sankey Diagram: A Proof of Concept

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Abstract. Managers, physicians and researchers need to study patient's path for purposes of management, quality of care and research. We present the proof of concept of the use of a flow diagram, the Sankey diagram, to visualize the trajectory of a population that experienced an event. This representation was tested with two case studies in populations from the anesthesia data warehouse of Lille University Hospital. For the 551 patients undergoing a pancreaticoduodenectomy, Sankey diagram helped us identify atypical care paths of patient being transferred too late in an intensive care unit. For 473953 patients who have had anesthesia procedure, Sankey diagram highlighted that mortality and re-operation rates increase with the number of operations. This preliminary work has been well received by end-users and allowed managers, physicians and researchers to visualize the paths of patients and to provide visualization support for research questions. This work will be followed by generalization.

Keywords. Data Visualization, Data Reuse, Patient Path, Sankey Diagram

## 1. Introduction

To display complex information in a most effective way, graphical representations are more appropriate than tables and figures [1, 2]. In healthcare facilities, to have access to visual representations of patients' path would be of great value for assessing quality of care (e.g. to detect atypical care paths) and for management purpose [3]. From a research perspective, studying patient flows may generate hypotheses about patient care that can be assessed later with classical biostatistics methods.

Designing and providing clear and efficient visualization of patients' paths faces several issues:

• Hospital Information Systems (HIS) generate a huge amount of data and the patient path is documented by many variables (admission and discharges modes, type of units, succession of interventions and re-interventions, medical acts, diagnostics, etc). This makes it difficult to extract relevant information;

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• Patients' stays differ in complexity and may represent a large number of patterns (e.g. one patient went through two care units and underwent one operation while another went through 6 units and underwent two operations).

Several kinds of diagrams are commonly used to represent flows. From a practical point of view, a flow is a succession of steps, each one defined by an initial state, a final state, and a quantity. From a graphical point of view, it is represented by links carrying information about quantity from start nodes to end nodes. Table 1 compares 8 types of flows diagrams [4]. Most of these representations enable to visualize only one step of a process. On the contrary, the Sankey diagram enables to represent several steps along with the information about flows' volumes, and to manage the complexity of the patient paths.

This work aims to assess whether the Sankey diagram is feasible to represent patient paths, in a way that would be readable and understandable by clinicians. With this goal in mind, two cases were used: (i) paths of patients with the pancreaticoduodenectomy and (ii) association between re-operation and mortality.

Table 1. Comparison of flow diagrams

The connections may be	
The connections may be	It addresses only one step of
directed and weighted.	the whole path.
The same as Network	It addresses only one step o
diagram.	the whole path and present
	reading difficulties.
It displays successive	It displays as many final node
	as paths, which overwhelm
	the diagram.
	The number of steps must b
1 1	limited. At each step, patien
,	with the same criteria a
00 0	displayed on different parts
every steps.	the circle, which makes the reading harder.
It displays successive	For clarity sake, the number of
	steps must be limited.
	It does not synthesiz
for each patient.	information on the who
	population.
	It represents only one step
between distant places.	the path and does not consider
	the volume of flows and the
Deedebiliter	temporal sequence.
Readability.	Only one final sub-population
	The same as Network diagram. It displays successive steps of the path with detail about the count. It displays successive steps of the path with detail about the count, and aggregate links at every steps.

### 2. Materiel and methods

We worked on two different cases to illustrate the design of the Sankey diagram used to represent patient paths. Each case aimed to answer a question: (i) in the digestive surgery department, what is the patient volume in each care unit after surgical operation for pancreaticoduodenectomy and what are the atypical patient paths? (ii) how mortality is related to the number of surgical operations during an hospital stay?

To design the Sankey diagrams and representing answers to those questions, we used data from the anesthesia data warehouse of Lille University Hospital. This data warehouse stores and organizes data about surgery and anesthesia procedures (drugs, vital signs, steps of surgery) with administrative data (hospital stay characteristics, medical unit, entry and discharge modes, medical and surgical acts, diagnoses) [5,6].

The Sankey diagrams were designed following three stages. The first stage is crucial and consists in determining information needed to answer the question. The information are summarized in Table 2.

Source node	Case study 1	Case study 2
Population	Patients who undergo a	Patients who undergo an
	duodenopancreatectomy (HNFA007 code)	operation
Event under scrutiny	Admission to unplanned critical care unit after operation	Death and re-operation rates
Steps of the path	Successive medical units after operation	Discharge status after each operation

Table 2. Aggregated data to upload in the graphical library

The second stage consists in aggregating the data and compute the count of patients in each step between two successive states of the path. For each link, we had to provide a source node, a target node and the quantity between the two nodes. Finally, these aggregated data are loaded into graphical libraries (D3.js and D3sankey [7]).

The resulting Sankey diagrams were presented to clinicians to get their feedback on the readability, the understandability of this visual representation and its impact.

#### 3. Results

### 3.1 Case study 1 : Patient flow for pancreaticoduodenectomy

Operations of pancreaticoduodenectomy were selected based on the code HNFA007 of the French medical act classification (CCAM, Classification Commune des Actes Médicaux). For operations between 2010 and 2018, 551 patients were included. We selected two significant states of the hospital stay for pancreaticoduodenectomy: the post-operative unit and the last unit before discharge from the hospital. These two states were characterized with the type of unit (conventional, continuous care or intensive care unit) and final discharge status (mortality) was reported. The Figure 2 is the Sankey representation for this population.

Two managers and two physicians examined the Sankey diagram. They were able to describe patient flows in each unit after operation. Moreover, physicians highlighted atypical paths: some patients were transferred from conventional care unit to critical care after operation, and some patient died even thought they did not go through critical care beforehand. They reported that this Sankey diagram was intuitive to read, more understandable than a table but that they needed contextual information when the graph is accessed for the first time. Physicians will now investigate why these patients were not treated earlier in an appropriate care unit.

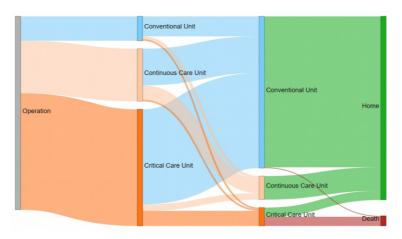


Figure 1. Patient flow for pancreaticoduodenectomy

#### 3.2 Case study 2: Mortality rates after re-operation

Between 2010 and 2018, 473953 patients underwent operation. For each patient, we computed the number of operations, and final discharge status (death). In order to get the most out of the diagram, two key points were relevant. First, we had to define a maximum number of operations to be covered. Secondly, we had to display results in percentages to compare modalities with small numbers (3<sup>rd</sup> and 4<sup>th</sup> surgical procedures) with larger modalities (1<sup>st</sup> and 2<sup>nd</sup> surgical procedures). With the Sankey diagram in Figure 2, we can see that mortality and re-operation rates increase with the number of operations performed during the same hospital stay. This diagram has been shown to 2 managers and 2 physicians, and they all agreed on the intuitiveness of the information displayed. They only asked for contextual information to know the source of the data. According to them, this result will be the starting point of research program about how going through multiple operations could impact mortality rates.

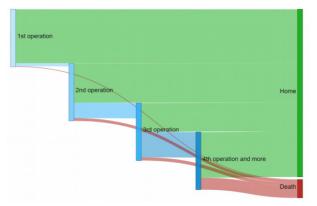


Figure 2. Sankey diagram representing the mortality and re-operation rates according to the number of operations during a hospital stay

### 4. Discussion

In this paper, we offer a way of visualizing patients' path with the Sankey diagram. This representation allowed managers, physicians and researchers to visualize the paths of patients, to detect atypical patient flows between units and to provide visualization support for research questions. We will now investigate which signals in operating room and post-anesthesia care unit could be use to detect patients prone to complications and refer them to an intensive care unit as soon as possible.

Usually, end-users are provided with data in a table format, on a case-by-case basis, or for a single step of the patient path. With the Sankey diagram, the benefits reported by users are that this representation is easily understandable, gives a useful overview of patient path and offers the opportunity to identify atypical paths that they may not have seen otherwise. It could also be used as a way to validate their hypotheses and further investigate.

However, the implementation of such diagrams might be challenging as patient paths have more than two steps and are complex. Thus, the data aggregation must be adapted to the specific issue to deal with. Morevover, the steps displayed must be selected to highlight important facts and avoid overload of information.

A limitation of the Sankey diagram, with its current implementation, is that it is not possible to follow an individual path from the first to the last node. The next step is to improve the current Sankey representation by adding a traceable multi-level feature. This would represent a more precise way to highlight a specific path.

Another perspective would be to offer an online Sankey visualisation tool, in which the users would select the targeted population (e.g. a surgical act) and predefined steps (e.g. surgical complication) to have a display of the corresponding patients' path.

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