

Adjustable Data Cleaning Towards Extracting Statistical Information

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Abstract. Each device, organization, or human, is affected by the effects of Big Data. Analysing these vast amounts of data can be considered of vital importance, surrounded by many challenges. To address a portion of these challenges, a Data Cleaning approach is being proposed, designed to filter the non-important data. The functionality of the Data Cleaning is evaluated on top of Global Terrorism Data, to furtherly create policies on how terrorism is affecting national healthcare.

Keywords. Data Cleaning, Health Policies, Global Terrorism Data

1. Introduction

Every device or human, is both producing and surrounded by the term of data, falling under three (3) categories: Fast, Large and Complex data [1], creating numerous challenges. Since Data Cleaning improves the data quality and the analysis outcomes, this paper presents a way of effectively performing cleaning actions through four (4) steps, which are adapting their functionality based on the domain that the cleaning actions must be performed.

For Data Cleaning the authors in [2] are filling the missing values and smoothing out the noise, while in [3] they create materialized views (MVs) that pre-process and avoid complex resource intensive calculations. Furthermore, in [4] they proposed the NADEEF architecture that allows users to specify multiple types of data quality rules.

Regarding the rest of this paper, Section 2 depicts the overall proposed approach, while Section 3 includes the evaluation of the derived results, presenting our concluding remarks.

2. Proposed Approach

The Data Cleaning workflow comprises of four (4) steps. However, the initial basic action of the mechanism is to ingest the data from the various data sources and identify their domain, by discovering and analysing the semantics of the ingested data, and thus the Data Cleaning actions to be performed are adjusted according to the domain's nature. Afterwards, the rest of the services (Fig. 1) initiate their tasks for accomplishing all the actions related to Data Validation, Data Cleaning, Data Verification, and Data Logging.

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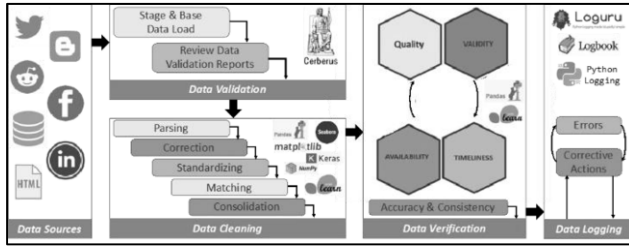


Figure 1. Data Cleaning Workflow.

3. Evaluation and Discussion

To evaluate the Data Cleaning approach, the described data cleaning services have been applied to the Global Terrorism Database (GTD) [5] (Fig. 2), in order to gather insights and policies regarding how global terrorism is affecting the healthcare systems of countries with increased or decreased terrorism attacks accordingly.

	eventid year month day approxdater extended resolution country
1	19700000000 1970 7 21 0 8 Dominican Republic 1 Central America
2	19700000000 1970 0 0 0 130 Mexico 1 North America Federal Mexico
3	19700000000 1970 0 0 0 140 Philippines 5 Southeast Asia Tarlac
4	19700100000 1970 1 0 0 78 Greece 1 Western Europe Attica Athens
5	19700100000 1970 1 0 0 101 Japan 4 East Asia Fukouka Fukouka 33
6	19700100000 1970 1 1 0 21 United States 1 North America Illinois
7	19700102000 1970 1 2 0 21 Uruguay 3 South America Montevideo
8	19700102000 1970 1 2 0 21 United States 1 North America Califor
9	19700102000 1970 1 2 0 21 United States 1 North America Wiscon
10	19700102000 1970 1 2 0 21 United States 1 North America Wiscon
11	19700103000 1970 1 3 0 21 United States 1 North America Wiscon
12	19700103000 1970 1 1 0 21 United States 1 North America Wiscon
13	19700106000 1970 1 6 0 21 United States 1 North America Colo

(a)

	eventid year month day approxdater extended resolution country
1	19700102000 1970 1 2 0 21 Japan 4 East Asia Fukouka Fukouka 33
2	19700102000 1970 1 2 0 21 United States 1 North America Wisco
3	19700105000 1970 1 1 0 21 United States 1 North America Wisco
4	19700108000 1970 1 8 0 98 Italy 8 Western Europe Lazio Rome 41
5	19700109000 1970 1 9 0 21 United States 1 North America Parr
6	19700110000 1970 1 1 0 65 Ethiopia 1 Sub-Saharan Africa Dnkn
7	19700118000 1970 1 18 0 21 Uruguay 3 South America Montevideo
8	19700119000 1970 1 19 0 21 January 19-20, 1970 0 21 United States 1
9	19700120000 1970 1 20 0 83 Guatemala 2 Central America & Carib
10	19700121000 1970 1 21 0 169 Philippines 5 Southeast Asia Metro
11	19700122000 1970 1 22 0 22 Venezuela 3 South America Caracas
12	19700122000 1970 1 22 0 21 United States 1 North America Nebr
13	19700122000 1970 1 22 0 21 United States 1 North America Nebr

(b)

Figure 2. (a) Ingested Data, (b) Cleaned Data.

Several rows of the GTD were dropped, where, out of the 5,000 different instances, 1,563 instances were dropped after three (3) iterations of the provided mechanism. It is within our next goals to perform additional evaluation, to deploy the services in such a way to run in parallel and, respecting privacy and security concerns [6].

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