



Supporting a Data-Driven World through Data Integration and Data Cleaning

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وزارة التخطيط التنموي والإحصاء
Ministry of Development Planning and Statistics

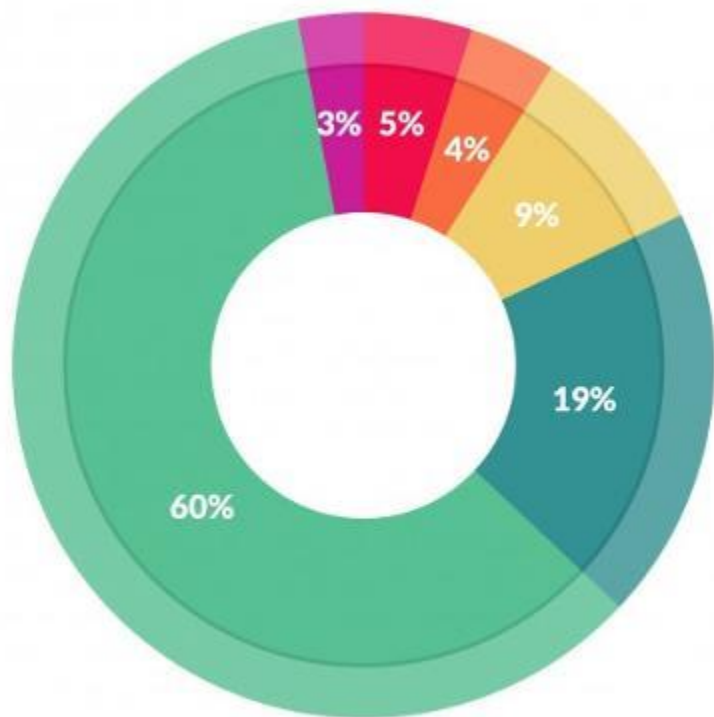
ورشة عمل بشأن تحديث الإحصاءات الرسمية في دولة قطر
Workshop on Modernization of Official Statistics in Qatar

الاثنين ١١ ديسمبر ٢٠١٧ - فندق روتانا، سيتي سنتر

Monday 11 December, 2017 - Rotana City Center Hotel - Doha

Agenda

- Why is this an important problem?
- Data Civilizer - An end-to-end system
- Overview of some key components



What data scientists spend the most time doing

- Building training sets: 3%
- Cleaning and organizing data: 60%
- Collecting data sets; 19%
- Mining data for patterns: 9%
- Refining algorithms: 4%
- Other: 5%





What's the least enjoyable part of data science?

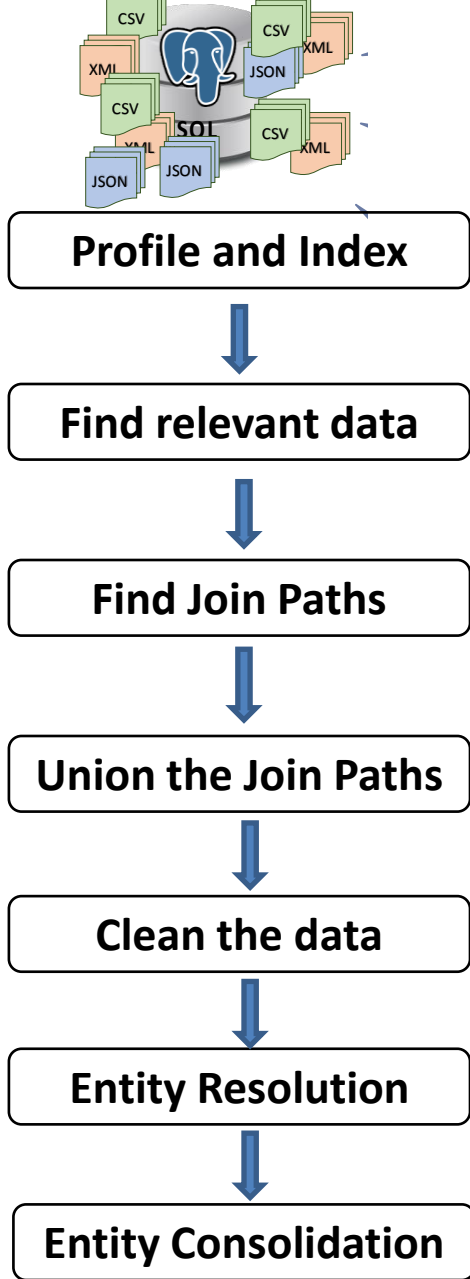
- Mark Schreiber (Merck) reports that his data scientists spend 98% of their time, i.e. 39 hours/week, in grunt work and only 1 hour/week doing the job for which they were hired
- For Big-Data Scientists, ‘Janitor Work’ Is Key Hurdle to Insights (The New York Times <https://www.nytimes.com/2014/08/18/technology/for-big-data-scientists-hurdle-to-insights-is-janitor-work.html>)
- Nobody reports less than 80% grunt work

We're building Data Civilizer to help ...

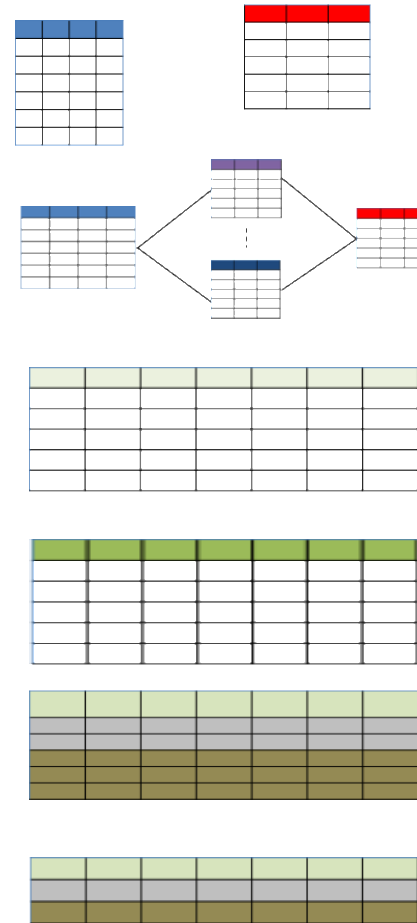
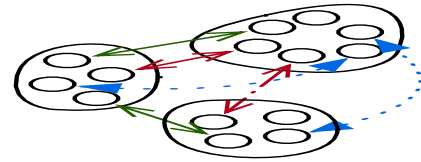
- ✓ **discover** data of interest from large numbers of data sets;
- ✓ **link** and **enrich** relevant data sets;
- ✓ **deduplicate** and **consolidate** the data;
- ✓ **clean** the data; and
- ✓ **iterate** through these tasks using a workflow system.

Algorithms do the grunt work (80% of the pain) while data scientists can do what *they* are good at

A Typical Data Civilizer Pipeline

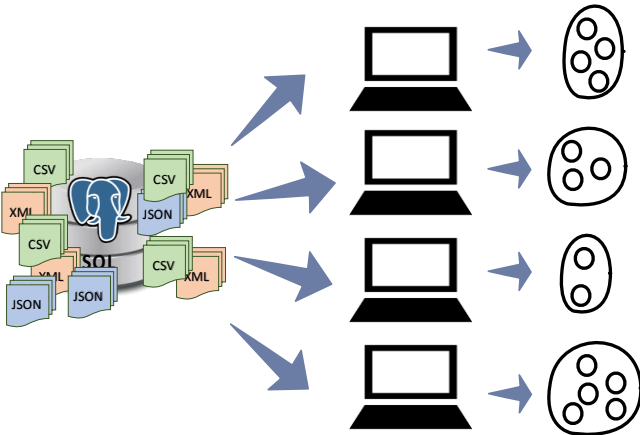


Enterprise Knowledge Graph



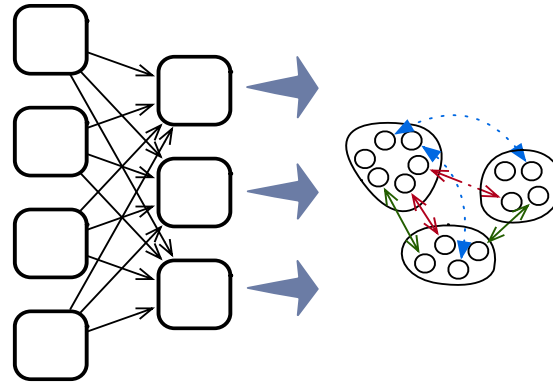
Data Discovery

Profiler
Create Summaries



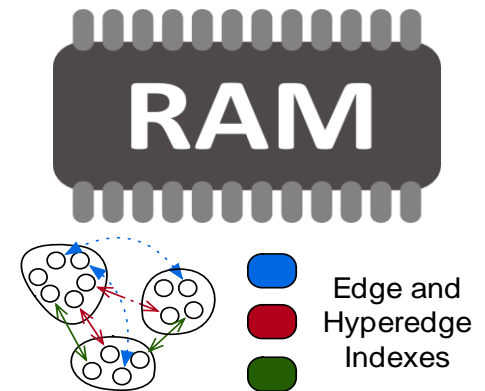
Distributed architecture to scale data summarization

Graph Builder
Connect Summaries



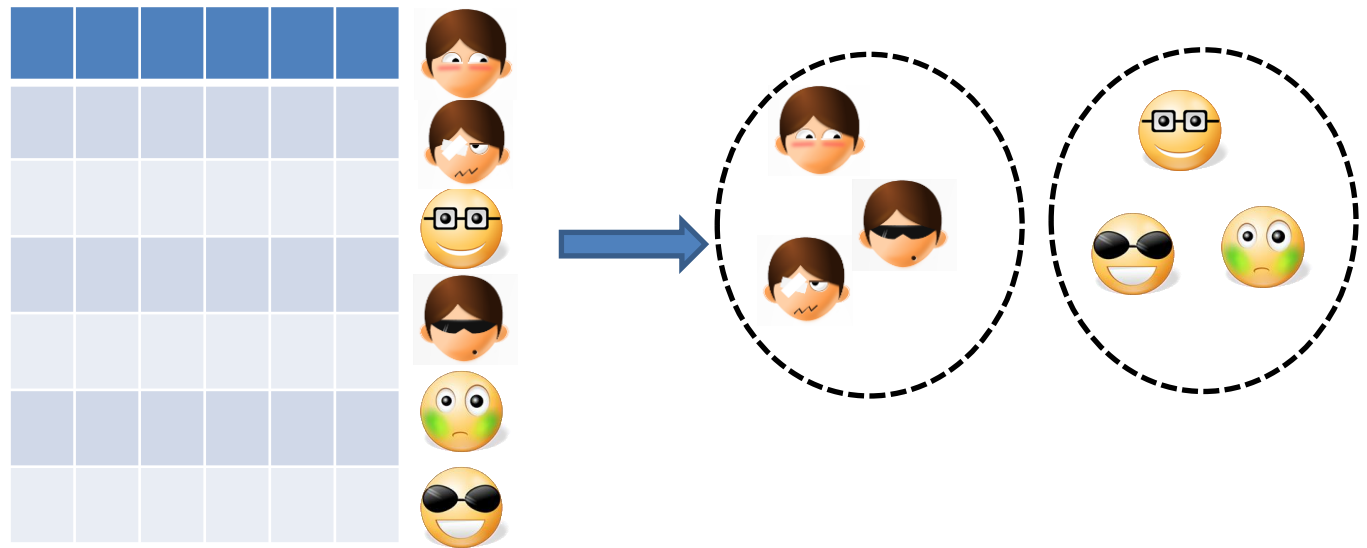
Scalable all-pairs comparison of multiple data types

SRQL Query Processing
Find relevant data



Concise in-memory indexes for interactive query answering

Entity Resolution using Deep Learning



A turn key solution using distributed representation (DR) and deep learning (DL)

- Tuples \rightarrow high dimensional vectors where (semantically) similar tuples have a high (cosine) similarity
- Using pre-trained DR dictionaries (e.g., GloVe which is trained on a corpus of 840B tokens) \rightarrow no need for manual feature engineering
- Much less training data
- Competitive or superior results wrt prior state-of-the-art methods
- Locality Sensitive Hashing-based blocking
 - automated and semantic blocking based on the entire tuple
 - no need for blocking functions from domain experts



Entity Consolidation



From clusters of duplicate records to Golden Records

① Clusters

9 St, 02141 Wisconsin 9th St, 02141 WI 9 Street, 02141 WI
3 E Avenue, 33990 CA 3rd E Ave, 33990 California 5th Str, 22701 New York

② Matchings

9 ↔ 9th WI ↔ Wisconsin 9th St ↔ 9 Street St ↔ Street 3rd ↔ 3 Avenue ↔ Ave California ↔ CA

③ Groups

Wisconsin ↔ WI California ↔ CA
9th ↔ 9 3rd ↔ 3
Street ↔ St Avenue ↔ Ave
9 Street ↔ 9th St

④ Interactions

Yes ✓ No ×
Yes ✓ No ×
Yes ✓ No ×

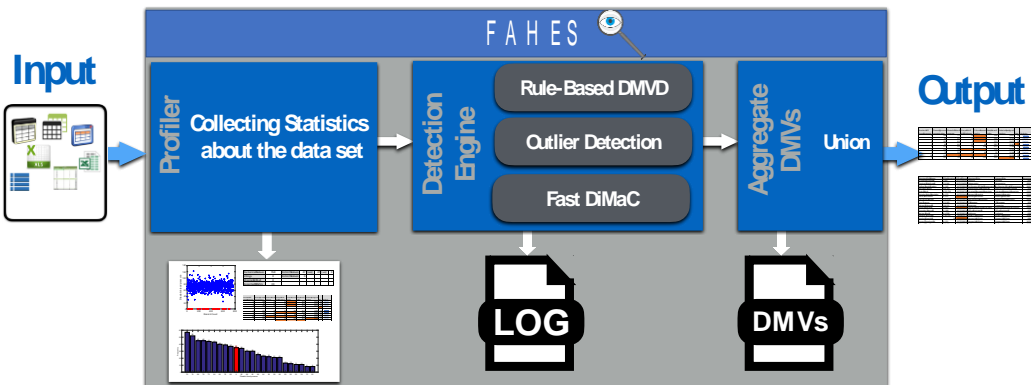


Cluster duplicates, detect matchings and group them, and ask a human

Detecting Disguised Missing Values

Source	Table Name	Column Name	DMV
UCI Machine Learning Repository	Pima Indians Diabetes	Diastolic Blood Pressurs	0
	adult	workclass	?
		education	Some-college
U.S. Food and Drug Administration	Adverse Event Reporting System (AERS)	EVENT_DT	20010101, 20030101
data.gov	Alleghency County WIC Vendor Location	Ref_ID	-1
data.gov	Graduation Outcomes - School Level - Classes of 2005 - 2011 - SWD	Advanced Regents Num	s, -
data.gov.uk	Accidents 2015	Junction Control	-1

DMV in different databases



- Rules to detect DMVs with special patterns, e.g., strings with repeated substrings
- Outlier detection algorithms
- A fast algorithm for detecting DMVs following a missing at random model

The Civilizer Studio – Gluing Things Together

The image displays two screenshots of the Civilizer Studio interface. The top screenshot shows a workflow diagram with three main components: 'CollectionSourceDWH' at the top, 'DataDiscovery-Aurum' in the middle, and 'JoinPathDiscovery' at the bottom. The bottom screenshot shows a more detailed workflow diagram with 'JoinPathDiscovery' at the top, branching into 'CleaningByDMVS' and 'CleaningByARS'. Below these are 'EntityMatching' and 'EntityConsolidation-GR'. A dialog box titled 'Select Constructor Signature' is open on the right side of the bottom screenshot, listing two constructor signatures for the 'JoinPathDiscovery' component. The interface includes a menu bar with 'Editor' and 'Monitor' tabs, and a sidebar with options like 'New', 'Save', 'Load Another Plan', 'Build', and 'Execute'. A 'Source Operators' list is visible on the left side of the bottom screenshot.

Civilizer STUDIO Editor Monitor

CollectionSourceDWH

DataDiscovery-Aurum

JoinPathDiscovery

Civilizer STUDIO Editor Monitor Show Execution Plan

New Save

Load Another Plan

Build Execute

Source Operators

JoinPathDiscovery

CleaningByDMVS

CleaningByARS

EntityMatching

EntityConsolidation-GR

Select Constructor Signature

- (civilizer.JoinDiscovery.Aurum, civilizer.JoinDiscovery.Aurum.EKGPath, civilizer.JoinDiscovery.Aurum.EKGPath.Index)
- (civilizer.JoinDiscovery.ML4JoinPathSelection, civilizer.JoinDiscovery.ML4JoinPathSelection.Model)

ErrorDetection

EntityConsolidation

DataRepairing

DataVisualization

DataAggregator

Next Steps ...

- Open-source release (ver 0.1)
- Get our technology in as many users' hands as possible
- Run tutorials in Spring 2018



شكراً
أسئلة؟

Thank You
Questions?