

# Large Scale Data Integration Project **Prof. Dr. Ziawasch Abedjan** Muaid Mughrabi, Binger Chen **Data Integration and Data preparation** Group (D2IP) ww.bifold.berlin





## **LSDIPRO** Team

- Muaid Mughrabi
  - muaid.mughrabi@tu-berlin.de
  - Seminar Organization and Topic Supervision
- Dr. Binger Chen
  - chen@tu-berlin.de
  - Topic supervision
- Prof. Dr. Ziawasch Abedjan
  - abedjan@tu-berlin.de
  - Here to enjoy











## **Project Goal and Grading Criteria**

## Soft Goals:

- 1. Reproduce a given paper
- 2. Create a competitive submission for the BTW Data Science challenge

We look at:

Code

Presentation

Individual contribution sheet (1A4 page)

## Hard Goals (relevant for grading):

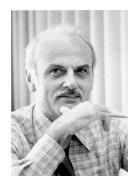
- 1. Clean and documented code repository
- 2. Systematic evaluation of generated solutions
- 3. Scaling solutions to the size of underlying datasets
- Clearly describing, visualizing and presenting the results in a presentation





## **Database Research in the 70s-90s**

• The Coddfather spoke:



Managing Data shall be Science! Memory hierarchy Portability Optimization User concurrency Distribution Usability ....



Jim Gray



Make it happen!







Michael Stonebraker



Donald Chamberlin





McKinsey&Company

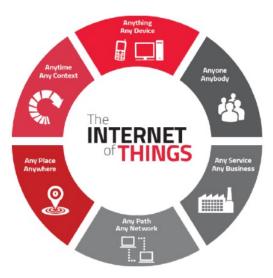
2011

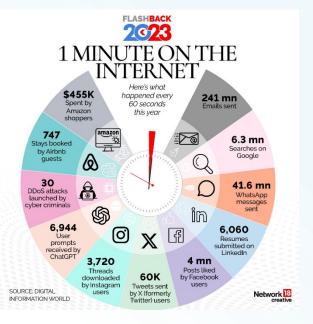
Big data: The next frontier for innovation, competition, and productivity





## Where do we find Big Data ?







## **Big Data?**

Gartner Volume Velocity Variety





Yuri Demchenko

Volume Velocity Variety Veracity Value





DATA

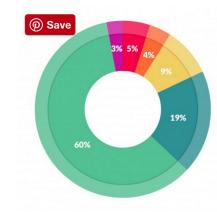
### Harvard Business Review

## Data Scientist: The Sexiest Job of the 21st Century

by Thomas H. Davenport and D.J. Patil

FROM THE OCTOBER 2012 ISSUE

**Data preparation** accounts for about 80% of the work of data scientists

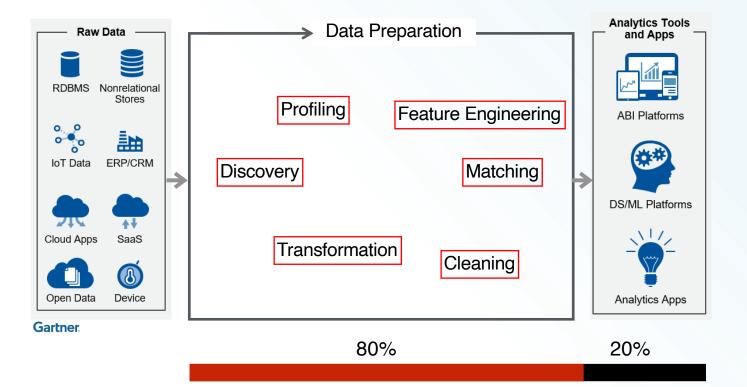


What data scientists spend the most time doing

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

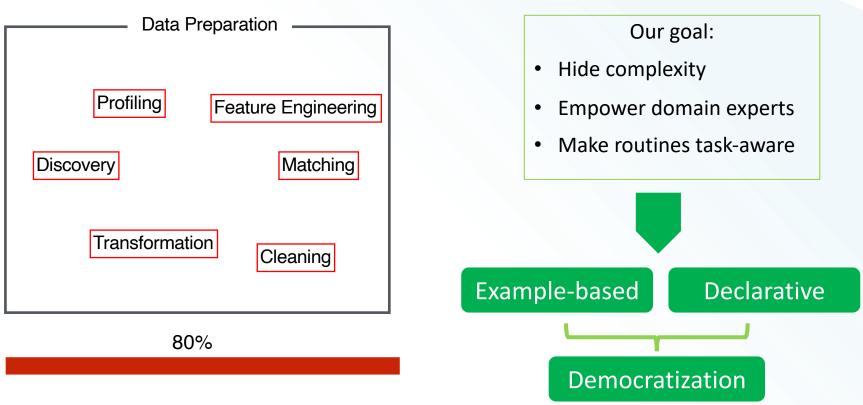






## **DIP** Focus of our research









## **Topics**

1.BTW 2025 Data Science Challenge Integrating various sources for improving forecasting
2.Transformation Discovery
Using Web resources to find transformation functions via input output examples
3.Table Question Answering
Use the concept of chain of thoughts on an LLM to generate functions that retrieve the answer for a question in natural languages
4.AutoTables:
Synthesizing tables without example
5.Finding related tables
Given an input table and a corpus of tables, find tables that enrich your table





### https://btw2025.gi.de/data-science-challenge

## **BTW 2025**

- Supervisor: Muaid Mughrabi
- Goal: Predict hourly day-ahead energy prices for Germany on February 18, 2025
- Problem: Not enough data and features available for accurate prediction



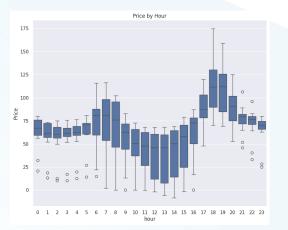


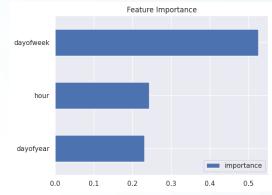


## **BTW 2025**

- Data collection
- Data curation
- Prediction
- Explanation and Visualization
- Be Creative!











## BTW 2025: Tasks

- Gathering Domain Knowledge & Data Sources
  - Data crawling from DMARD.DE
  - Weather information?
  - Special events and how to account for them: pandemic, war, news, etc.
- Data Preparation
  - Data cleaning, removing anomalies if needed, investigating variable relationships, and creating statistical summaries.
  - Training, Testing, and Validation splits (without harming temporal information)
- Visualization
  - Plot trends and forecasts, sanity check model performance visually, and explain your findings
- Predictive Modelling:
  - Develop a forecasting model, compare it against a reasonable baseline, and iterate on it.





### https://cs.uwaterloo.ca/~ilyas/papers/AbedjanICDE16.pdf

## **Transformation Discovery with DataXFormer**

Supervisor: Muaid Mughrabi Problem:

- Tables often have missing values/ Representations need to be unified
- Using external resources, such as tables, functions, catalogues, one can fill in the gaps
- DataXFormer swifts through relational tables finding tables that contain examples of one-to-one and many-toone mappings

Name	Team	Height	League		Lastname	Teamname	НТ
Messi	FCB	1.70	La Liga		Neuer	FC Bayern Munich	6'4''
Neuer	FCB	1.93	Bundesliga		Ronaldo	Real Madrid	6'1"
Pirlo	Juve		Serie A		Pirlo		5'10"
Sourc	e 1		$\leq$	1		- Sou	rce 2

Lastname Team		League	нт	Height	
			Teamname		
Neuer	FCB	Bundesliga	FC Bayern Munich	6'1"	1.93
Messi	FCB	La Liga	?		1.70
Pirlo	Juve	Serie A	?	5'10"	?
Ronaldo		?	Real Madrid	6'4"	?





## **Task: Reproduce DataXFormer**

### - Implement DataXformer

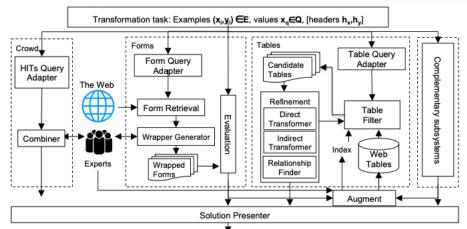
- Table retrieval
- Transformation Discovery
- EM Algorithm

### - Evaluate your implementation

- Correctness and Runtime
- Dataset coverage

### Improve one aspect!

- Partial Strings inside Cells
- Be creative!



Ranked transformations with scores





#### https://arxiv.org/pdf/2401.04398

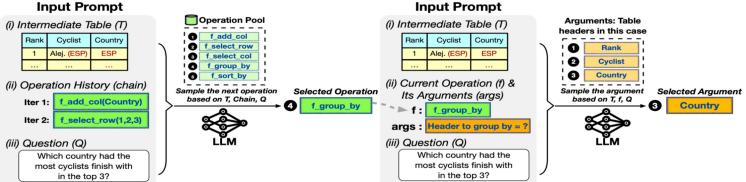
## **Chain of Tables**

Supervisor: Muaid Mughrabi

Goal: Generating programs that extract the answers to a natural language question from a table

#### Problem:

- How to represent tables? \_
- How to prompt the LLM? -
- How to keep the chain of operations? -







## **Chain of Tables: Tasks**

- Implement Chain of Tables using small LLM
  - Table Actions implementation
  - Prompting LLM to generate plan and action args
  - Ensure that Table actions are performed correctly
- Evaluate your implementation
  - Against different LLMs
  - You need to start with smaller model with your notebooks
  - Performance and Runtime
- Improvement suggestion!
  - Instead of generating one plan, generate multiple plan candidates
    - Implement an evaluator to score each state, traverse graph based on given scores
  - Be creative!





Peng Li, Yeye He, Cong Yan, Yue Wang, Surajit Chaudhuri: Auto-Tables: Synthesizing Multi-Step Transformations to Relationalize Tables without Using Examples. Proc. VLDB Endow. 16(11): 3391-3403 (2023)

https://www.vldb.org/pvldb/vol16/p3391-he.pdf

## **AutoTables**

Supervisor:	Binger	Chen
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Goal: Discovering the operators and transforming nonrelational tables to relational tables Transforming operators (choose 2):

- Stack: transforming homogeneous columns into rows. (Pandas API: melt)
- Transpose: transforming rows to columns and vice versa. (Pandas API: transpose)
- Wide-to-long: transforming repeating column groups into rows. (Pandas API: wide)
- Pivot: transforming repeating row groups into columns. (Pandas API: pivot)

	A	В	C	D	E	F	G	Н	1	J	K	L	М
1	Product	Product Category	Store	19-Oct	20-Oct	21-Oct	22-Oct	23-Oct	24-Oct	25-Oct	26-Oct	27-Oct	28-Oct
2	Huffy 18 in. Boys Bike	Sports	s_sk_101	5	3	4	15	19	2	5	11	3	9
3	Kent 18 In. Boy's BMX Bike	Sports	s_sk_101	8	5	11	12	8	14	7	5	9	9
4	HP 11 in. Chromebook 16W64	Electronics	s_sk_102	17	9	14	5	19	18	17	18	10	7
5	Mainstays Computer Desk	Furniture	s sk 103	6	4	1	16	8	7	6	9	8	20

Stack

	A	В	С	D	E
1	Product	Product (	Ca Store	Date	Unit Sold
2	Huffy 18 in. Boys Bike	Sports	s_sk_101	19-Oct	5
3	Huffy 18 in. Boys Bike	Sports	s_sk_102	20-Oct	3
4	Huffy 18 in. Boys Bike	Sports	s_sk_103	21-Oct	4
5	Huffy 18 in. Boys Bike	Sports	s_sk_104	22-Oct	15
5					
7	Kent 18 In. Boy's BMX Bike	Sports	s_sk_101	19-Oct	8
3	Kent 18 In. Boy's BMX Bike	Sports	s sk 102	20-Oct	5





## AutoTables: Tasks

- Generate a training dataset from relational table dataset

- Method: leverage inverse operators
- Relational dataset source: https://relational-data.org/
- Implement an Input-only model
  - Model input: non-relational table.
  - Model output: operator used for transformation.
  - Feature extraction based on CNN from computer vision as the pattern is visual.
- Implement Input/Output re-reranking model
  - Model input: top-k transformed tables from input-only model.
  - Model output: ranking score of these tables.
- Similar feature extraction as Input-only model - Scaling(optional): Implement data augmentation:
  - (1)Cropping (2)Shuffling





Anish Das Sarma, Lujun Fang, Nitin Gupta, Alon Y. Halevy, Hongrae Lee, Fei Wu, Reynold Xin, Cong Yu: Finding related tables. SIGMOD Conference 2012: 817-828

https://amplab.cs.berkeley.edu/wpcontent/uploads/2012/06/finding-related-tables.pdf

## **Finding Related Tables**

Supervisor: Binger Chen

Goal: Finding Related Tables based on Entity Complement - Find Columns that contain semantically similar entities but complement each other

	Anday, Rankings by Country: 2010 ▼ All Countries			ditional Standings: p 100   ▼
Rank	Name & Nationality	Points	Position Moved	Tournaments Played
1	Nadal, Rafael (ESP)	12,450	0	20
2	Federer, Roger (SUI)	9,145	0	21
3	Djokovic, Novak (SRB)	6,240	0	21
4	Murray, Andy (GBR)	5,760	0	19
5	Soderling, Robin (SWE)	5,580	0	24
6	Berdych, Tomas (CZE)	3,955	0	26
7	Ferrer, David (ESP)	3,735	0	24
8	Roddick, Andy (USA)	3,665	0	21
9	Verdasco, Fernando (ESP)	3,240	0	25
10	Youzhny, Mikhail (RUS)	2,920	0	24

### Entity complement

101 - 2	101 - 200							
As of Monday, Rankings by Country: Additional Standings: 27.12.2010 V All Countries V 101-200 V								
Rank	Name & Nationality	Points	Position Moved	Tournaments Played				
101	Gil, Frederico (POR)	551	0	29				
102	Phau, Bjorn (GER)	551	0	31				
103	Beck, Karol (SVK)	549	0	26				
104	Brands, Daniel (GER)	541	0	28				
105	Falla, Alejandro (COL)	540	0	23				
106	Dimitrov, Grigor (BUL)	536	0	20				
107	Bolelli, Simone (ITA)	532	0	29				
108	Devvarman, Somdev (IND)	526	0	27				
109	Darcis, Steve (BEL)	521	0	23				
110	Zeballos, Horacio (ARG)	517	0	32				





## **Finding Related Tables: Tasks**

- Measure the relatedness of two tables:

- entity consistency: they have entities from the same class, such as all tennis players
- schema similarity: they have same table schema
- Leverage signals from the web source to infer entity groups
  - Knowledge bases providing signals: WebIsA, DBpedia, etc.
  - Example of signals: (Entity: Paris, Class: City)
- Leverage signals from the web source to infer entity groups
- Scaling(optional): implement all above on big dataset





## **Important Dates**

17.10. Topic selection, group formation

24.10. Weekly meetings

Identify the main idea, the problem being solved, who/what is involved, understood paper, find datasets, get familiar with repository

31.10.Weekly meetings

#### Present a plan

Per person: what is going to be implemented/ tested and what was done the week before.

07.11. Weekly meetings (development)

- 14.11. Weekly meetings (first pipeline of the software should e ready, development)
- 21.11. Weekly meetings (development)

28.11. Weekly meetings (development)

#### 05.12. Expert review

A different group will test your system

- 12.12. System improvement, experiments.
- 19.12. Weekly meetings (experiments)
- 09.01. Weekly meetings (experiments/Start writing the report)
- 16.01. Weekly meetings (experiments, visualization)
- 23.01. Weekly meetings (experiments, visualization)
- 30.01. Weekly meetings (Wrap up documentation, and presentation slides)
- 06.02. Final presentations.
- 13.02. Final presentations.
- 14.02. One A4 page describing individual contributions on the project.





## TODOS

- Identify group members via ISIS
- Send Muaid Mughrabi until October 21st 5pm your chosen top 3 topics