

Harnessing the Power of LLMs to Construct Structured Datasets from Unstructured Sources

David Zarruk

Disclaimer: All views expressed are my own.

Empirical research stages

1. Dataset construction:

- Access to raw files:
 - Available online
 - Administrative or bureaucratic process
 - Go to archives and find historical documents



Unavoidable

- Data processing:
 - Readily available excel files
 - Press releases, news articles, speeches
 - Historical archives (pdfs, physical paper, etc)
 - Manually-filled survey forms



Easy processing
Can take a significant amount
of time (50%? 90%?)
Difficult/slow to process

2. Data analysis
3. Paper writing

Some examples of costly data processing

- Papers that study impact of speeches/press releases:
 - The Importance of Fed Chair Speeches as a Monetary Policy Tool – Swanson (2023)
 - Shocking language: Understanding the macroeconomic effects of central bank communication – Hansen and McMahon (2016)
- Datasets based on historical archives:
 - The Colonial Origins of Comparative Development: An Empirical Investigation – Acemoglu, Johnson and Robinson (2001)
 - History, Institutions, and Economic Performance: The Legacy of Colonial Land Tenure Systems in India – Banerjee and Lakshmi (2005)

LLMs to the rescue

- LLMs excel at reading and understanding text and images
- Can easily transcribe images into text and extract relevant information
- Can analyze a text, find relevant parts, and analyze sentiments

2 examples

1. Construct a dataset with European Central Bank bulletins

- Hard data (inflation, unemployment, etc)
- Qualitative features:
 - Risks
 - Labor market conditions: improving, deteriorating, etc.
 - Financial conditions: tightening, easing, etc



	report_date_YYYY_MM	hicp_current	hicp_core_current	unemployment_rate_latest	m3_growth_latest	covid_mentions_count
0	2020-01	1.3	1.3	7.5	5.6	0
1	2020-03	1.2	1.2	7.4	5.2	19
2	2020-05	0.4	0.9	7.4	7.5	8
3	2020-06	0.1	0.9	7.3	8.3	48
4	2020-07	0.3	0.8	7.4	8.9	28
5	2020-09	-0.2	0.4	7.9	10.2	35
6	2020-10	-0.3	0.2	8.1	10.4	15
7	2020-12	-0.3	0.2	8.4	10.5	29
8	2021-01	-0.3	0.2	8.3	11.0	18
9	2021-03	0.9	1.1	8.1	12.5	12
10	2021-03	1.3	0.9	8.3	12.3	2
11	2021-06	2.0	0.9	8.0	9.2	8
12	2021-07	1.9	0.9	7.9	8.4	18

2 examples

2. Construct a dataset based on hand-written forms

- Elections in Colombia

7310430010102
Var 01 Pág 1 de 2
ACTA DE SCRUTINIO DE LOS
JUICIOS DE VOTACION
ELECCIONES TERRITORIALES
OCTUBRE 29 DE 2023
ALCALDE
DEPARTAMENTO: 16 - BOGOTA, D.C.
MUNICIPIO: 991 - BOGOTA, D.C.
ZONA: 04 PUESTO: 01 MESA: 091
LUGAR: LAS BRISAS

X 5-32-51-35 X
REGISTRO DE LA MESA
TOTAL VOTANTES FORMULARIO E-11 1 5 2
TOTAL VOTOS DE ALCALDE EN LA URNA 1 5 1
TOTAL VOTOS INCUMPLIDOS

CANDIDATO	AGRUPACIÓN	VOTACIÓN
1 JORGE LUIS VARGAS VALENCIA	CS	• 3
2 DIEGO ANDRES MOLANO APONTE	CD	• 2
3 RODRIGO JARA RESTREPO	CD	• 3
4 NICOLAS RAMOS BAROSA	CD	• 1
5 CARLOS FERNANDO GALAN PACHON	CD	1 0 3
6 GUSTAVO BOLIVAR MORENO	CD	• 3 1
7 RAFAEL ALFONSO QUINTERO ROMA	CD	
8 JUAN DANIEL OVIDIO ARANGO	CD	• 2 3
9 JORGE ENRIQUE REBOLLO CASTILLO	CD	• 1 1



department	municipality	zone	puesto	mesa	form_code	category	votes	computed_total_votes	total_votos_alcalde_en_la urna
16 - BOGOTA D.C.	001 - BOGOTA, D.C.	01	01	001	X 7-72-59-31 X	CARLOS FERNANDO GALAN PACHON	119	149	149
16 - BOGOTA D.C.	001 - BOGOTA, D.C.	01	01	002	X 8-67-73-32 X	CARLOS FERNANDO GALAN PACHON	145	194	194
16 - BOGOTA D.C.	001 - BOGOTA, D.C.	01	01	003	X 5-25-26-30 X	CARLOS FERNANDO GALAN PACHON	30	88	88
16 - BOGOTA D.C.	001 - BOGOTA, D.C.	01	01	004	X 5-47-38-32 X	CARLOS FERNANDO GALAN PACHON	157	223	223
16 - BOGOTA D.C.	001 - BOGOTA, D.C.	01	01	005	X 7-66-09-36 X	CARLOS FERNANDO GALAN PACHON	94	209	209
16 - BOGOTA D.C.	001 - BOGOTA, D.C.	01	01	006	X 4-07-62-32 X	CARLOS FERNANDO GALAN PACHON	102	137	137

Tips for good prompt engineering

1. Enclose sections in the prompt using <> keys
2. Define a set of rules that define boundaries for the LLM:
 - “Do not hallucinate”
 - “Return a JSON – nothing else”
3. Give examples to the LLM!
4. Ask the LLM to return citations or the thought process that support the output
5. ITERATE: writing a good prompt is an iterative process – usually the LLM never gets it right on the first attempt