

# Information Extraction From Fiscal Documents Using LLMs

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## 1. Motivation

- Governments periodically release fiscal data that is critical for research
- The data format (PDF) is difficult to analyze
- How can LLMs understand these formats?
- How can we create research-ready datasets at scale?

## 2. Challenge

Large Language Models (LLMs) are great at language, yet poor at multi-page tabular extraction

- Table structure has multi-level, complex hierarchy
- Documents are too big, larger than context window
- Documents are in Indic languages, where LLMs do worse
- Tables are inconsistently rendered, cells are split or merged

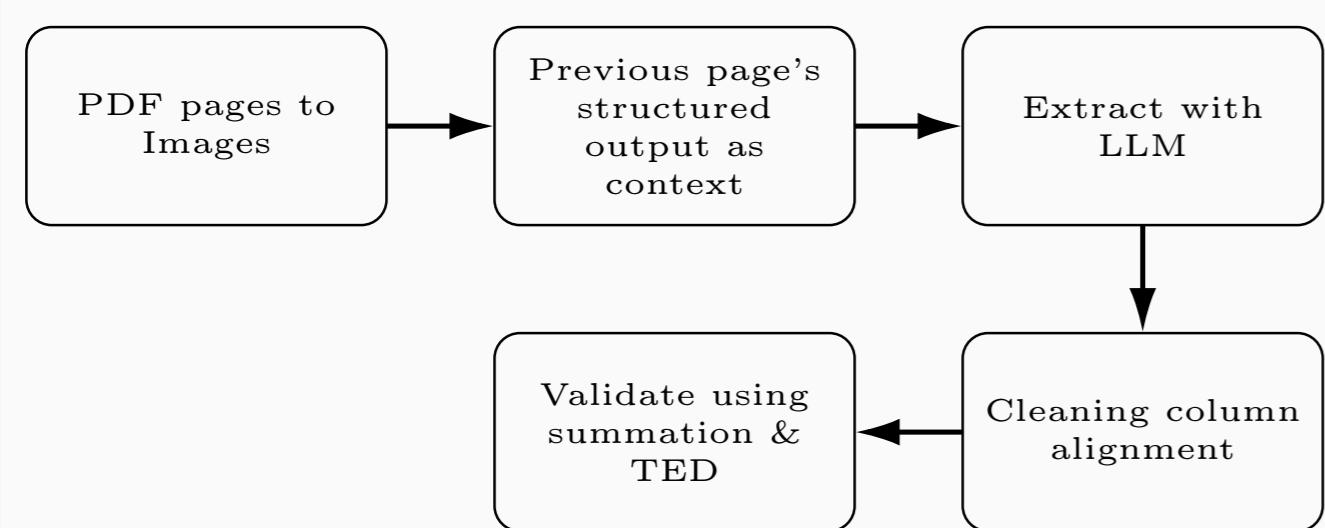
## Sources of complexity

- Tables span multiple pages; tables have varying structures
- The files are large: roughly hundreds of pages (84-241 pages in the 7 volumes). Since tables span multiple pages, page-by-page extraction missed vital information
- Regional languages mix with English in inconsistent patterns
- Units and formatting vary across documents

## 3. Our approach

Our improvements to a naïve LLM document extraction

- Image-based processing:** Convert PDF pages to high-resolution JPGs (300 DPI). This improves LLM comprehension
- Sequential context:** Provide previous page's structured output as an aide to understanding the current page.
- Multi-level validation:** Use column-sums to ensure numerical consistency. Use hierarchy to validate structural consistency
- Meta-prompts:** Provide domain context, get LLM to write the extraction prompt
- Intelligent cleaning:** Post-process to improve column alignment



## Input PDF to output table

(clockwise) Two pages that capture complexity of the PDF; the output table extracted from the first page

સુધી બજેટ ડેમ્યુન્ડ ડાયલ 2049 વિધુ દાયક 2049 Interest Payments						
કોર્સ - 1 ફાફન્દાન નંબર : 29 VOLUME - 1 Demand No : 29 ( રૂ વિધુ ) ( રૂ લખ )						
દિન કોર્સ અનુભવ કરીને નાનું હશે	Heads of Account, Dem No & Dept Code	દિન Accounts 2018-19	બજેટ Budget 2019-20	રસ્તો Revised 2019-20	બજેટ Budget 2020-21	દિન Budget 2020-21
1	2	3	4	5	5	5
01 અભિયાન માટે ચીની વિધુ	Interest on Internal Debt	C 1235918.78	1541652.00	1499527.55	1829267.00	
03 સ્કોર કરીને આપણી વિધુ	Interest on Small Savings, Provident Funds Etc	C 235754.02	287656.00	287656.00	320909.00	
04 સ્કોર કરીને આપણી વિધુ	Interest on Loans & Advances from Central Government	C 70610.04	76710.00	76604.00	67400.00	
05 સ્કોર કરીને આપણી વિધુ	Interest on Reserve Funds	C 8.33	12.00	12.00	4882.00	
60 અભિયાન માટે ચીની વિધુ	Interest on Other Obligations	C ...	1.00	305.83	...	
સાથી	Total 2049	C 1542291.17	1906031.00	1864105.38	2221639.00	
સાથી	TOTAL V+C	C 1542291.17	1906031.00	1864105.38	2221639.00	
સાથી વિધુ	GRAND TOTAL	C 1542291.17	1906031.00	1864105.38	2221639.00	

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1	2	3	4	5	5	5
01 અભિયાન માટે ચીની વિધુ	Interest on Internal Debt	C 1024916.40	1343614.00	1303940.00	164445.00	
101 અભિયાન માટે ચીની વિધુ	Interest on Market Loans	C ...	...	...	5000.00	
102 સ્કોર કરીને આપણી વિધુ	Interest on Loans	C ...	...	...	4.55	10.00
115 સ્કોર કરીને આપણી વિધુ	Interest on Ways & Means Advances from Reserve Bank of India	C 178961.95	163527.00	163527.00	147784.00	
123 સ્કોર કરીને આપણી વિધુ	Interest on Special Securities issued to NSP of the Central Government by State Government	C 28891.49	31050.00	28595.00	27694.00	
200 સ્કોર કરીને આપણી વિધુ	Interest on Other Internal Debts	C 3148.94	3461.00	3461.00	4334.00	
305 સ્કોર કરીને આપણી વિધુ	Total 01	C 1235918.78	1541652.00	1499527.55	1829267.00	
03 સ્કોર કરીને આપણી વિધુ	Interest on Small Savings, Provident Funds Etc	C 116439.48	139473.00	139473.00	163084.00	
104 સ્કોર કરીને આપણી વિધુ	Interest on State Provident Funds	C 53690.16	62520.00	62511.00	56046.00	
107 સ્કોર કરીને આપણી વિધુ	Interest on Trusts and Endowment	C ...	1.00	1.00	1.00	
108 સ્કોર કરીને આપણી વિધુ	Interest on Insurance and Reinsurance	C 119314.54	146881.00	146881.00	155704.00	
115 સ્કોર કરીને આપણી વિધુ	Interest on Other Obligations	C ...	1.00	1.00	1.00	
117 સ્કોર કરીને આપણી વિધુ	Interest on Defined Contribution Pension Scheme	C ...	1300.00	1300.00	1300.00	
104 સ્કોર કરીને આપણી વિધુ	Total 03	C 235754.02	287656.00	287656.00	320909.00	
101 સ્કોર કરીને આપણી વિધુ	Interest on Loans & Advances from Central Government	C 527.51	483.00	474.00	412.00	
104 સ્કોર કરીને આપણી વિધુ	Interest on Loans for State / Union Territory Plan Schemes	C 16364.26	13677.00	13588.00	10900.00	

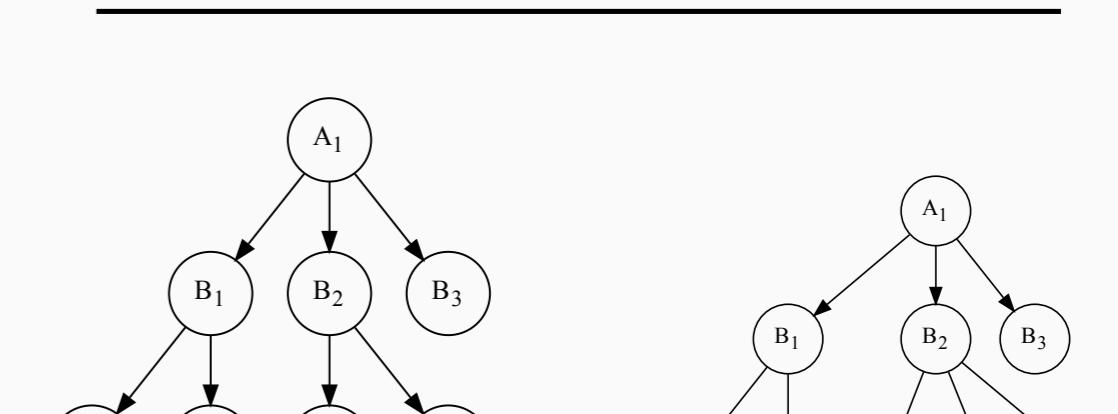
## 4. Validation results

**Numerical Consistency:** Verify budget head sums within and across schema

Volume	Checks	Passed	Pass Rate %
Volume 1	528	463	88%
Volume 2	463	402	87%
Volume 3	289	233	81%
Volume 4	249	206	83%
Volume 5	390	316	81%
Volume 6	155	134	86%
Volume 7	237	198	84%
All	2311	1952	84%

**Structural Consistency:** Use Tree Edit Distance Similarity to verify tabular structure

Volume	Pages	Accuracy
Volume 1	227	95.24%
Volume 2	179	73.68%
Volume 3	139	88.00%
Volume 4	142	83.33%
Volume 5	181	96.77%
Volume 6	74	91.30%
Volume 7	112	79.17%



## 5. Advantages of our method

- Can handle arbitrarily long PDF files. Extract multipage tables from 200+ page PDFs
- Resilient against inconsistent Indic character encoding
- Works in the absence of ground-truth data. Information extraction is at 74%-95% accuracy
- Identifies extraction failure source to page & table location
- Create research-ready datasets for states finances
- Can also create parallel PDF & structured (CSV/JSON) corpus for LLM training