

Information Extraction from Fiscal Documents using LLMs

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Motivation

- ▶ Governments produce fiscal data that is critical for machine analysis
- ▶ The data format (PDF) is difficult to analyze
- ▶ How can LLMs understand these formats?
- ▶ How can we create research-ready datasets at scale?
- ▶ Whether our approach towards validation is reasonable
- ▶ Can anything improve in the extraction that will help in the validation?

Why is this difficult?

- ▶ 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
- ▶ Tables span multiple pages

A sample of the Hierarchy

- ▶ Demand Number (2 Digit)
- ▶ Department Number (2 Digit) – Sometimes we only have the Department Name
- ▶ Major Head (4 Digit)
- ▶ Sub Major Head (2 Digit)
- ▶ Minor Head (3 Digit)
- ▶ Sub Head (1 Digit)
- ▶ Detailed Head (2 Digit)
- ▶ Object Head (3 Digit) – smallest unit in the budget (need to distinguish from 3-digit Minor Head)

The document has two parts:

- ▶ A Summary at the “Demand No., Department No. Level” (Abstract)
- ▶ A Detailed summary that covers “Minor Heads”, “Detailed Heads”, and each “Object Head”.
(The rest)

ABSTRACT

Demand No., Department No.

Major Head

→ 4 columns of totals



The Rest

REST

For each Demand No., Department No.

Major Head

Pg 1: Summary for
Minor Head

For each Minor Head;
Summary for Detailed Head

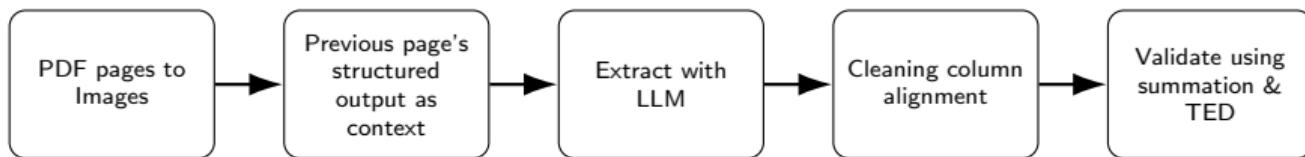
* Object Heads make the summary for “Detailed Head”

Our Approach

Our improvements to a naïve LLM document extraction

- ▶ **Image-based processing:** Converting PDF pages to high-resolution JPGs (300 DPI) 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-prompting:** Provide domain context, get LLM to write the extraction prompt
- ▶ **Intelligent cleaning:** Post processing to improve column alignment

Steps



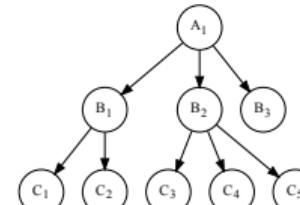
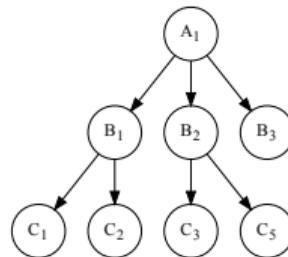
Validation Results

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% |



Advantages of our Method

- ▶ Can handle arbitrarily long PDF files
- ▶ Resilient against inconsistent Indic character encoding
- ▶ Works in the absence of ground-truth data
- ▶ Identifies extraction failure source to page & table location

Conclusion

- ▶ Extract multipage tables from 200+ page PDFs
- ▶ Information extraction is at 74%-95% accuracy
- ▶ Create research datasets for states finances
- ▶ Can also create parallel PDF & structured (CSV/JSON) corpus for LLM training