

¹ An Application for Detecting Plagiarism in University Theses

³ **Elyah Frisco Andriantsialo¹, Volatiana Marielle Ratianantitra¹, and Thomas
⁴ Mahatody¹**

⁵ **1** Laboratory for Mathematical and Computer Applied to the Development Systems, University of
⁶ Fianarantsoa, Madagascar

DOI: [10.xxxxxx/draft](https://doi.org/10.xxxxxx/draft)

Software

- [Review ↗](#)
- [Repository ↗](#)
- [Archive ↗](#)

Editor: [Open Journals ↗](#)

Reviewers:

- [@openjournals](#)

Submitted: 01 January 1970

Published: unpublished

License

Authors of papers retain copyright and release the work under a Creative Commons Attribution 4.0 International License ([CC BY 4.0](#)).

⁷ Summary

⁸ Academic plagiarism has evolved beyond simple copy-paste text to include complex paraphrasing and the reuse of visual elements like figures and diagrams. To address this, we present a ⁹ hybrid, multimodal web application designed for contextualized plagiarism detection. The ¹⁰ system utilizes a multi-criteria approach, analyzing documents based on six distinct dimensions: ¹¹ Theme, Location, Methodology, Results, Global Content, and Images (THLME-Gre schema). ¹² Built with **Flask**, the application leverages advanced semantic models—specifically **Sentence-BERT** ([Reimers & Gurevych, 2019](#)) for textual analysis and **CLIP** (Contrastive Language-¹³ Image Pre-training) ([Radford et al., 2021](#)) for visual analysis. It employs a vector database ¹⁴ (**ChromaDB**) to perform efficient Approximate Nearest Neighbor (ANN) searches across large ¹⁵ repositories of university theses.

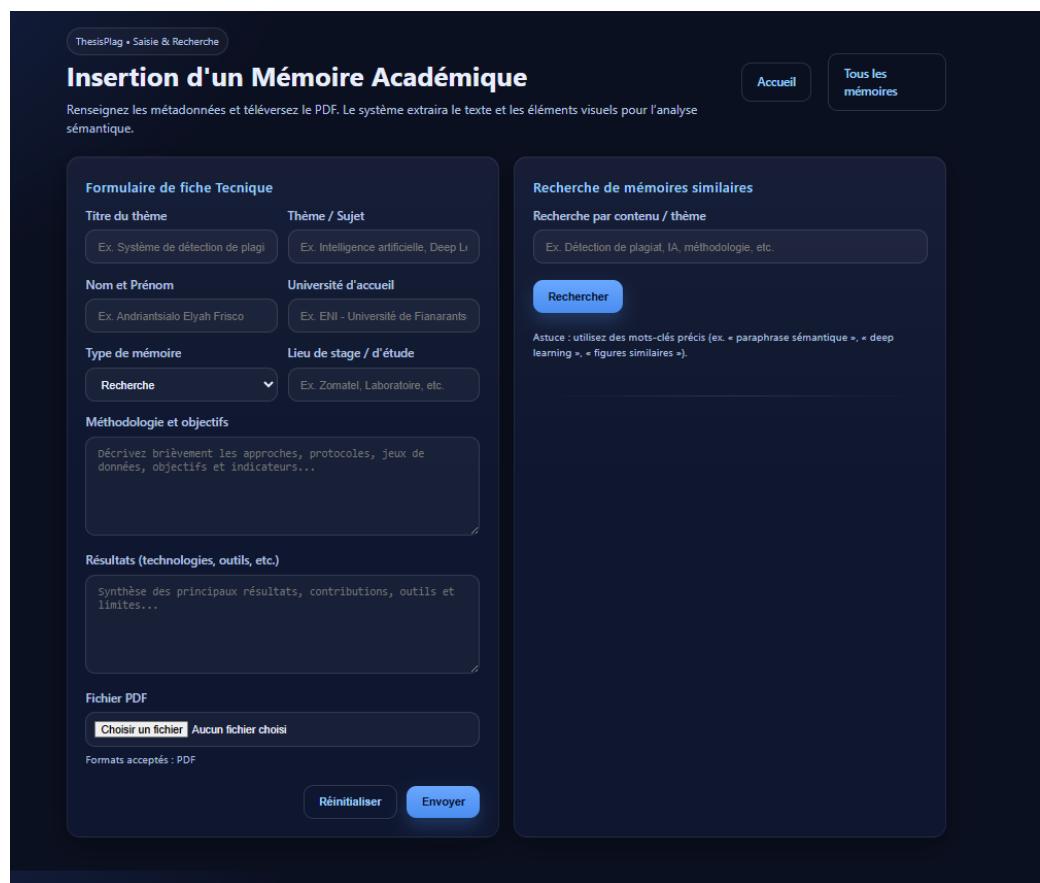


Figure 1: Screenshot of the application interface showing the dashboard and analysis results.

18 Statement of Need

19 Ensuring academic integrity is a growing challenge for higher education institutions in Madagas-
20 gar, particularly at the **University of Fianarantsoa**, which manages over 30,000 students
21 across various doctoral schools and departments. Currently, the university lacks a centralized,
22 automated institutional tool for plagiarism detection. Faculty members often rely on manual
23 verification or commercial tools that primarily focus on English content and surface-level text
24 matching.

25 These existing solutions present two major limitations for our context: 1. **Language and Context:** The majority of student theses are written in **French**. Generic tools often struggle
26 to distinguish between legitimate thematic overlap (e.g., multiple students working on “Web
27 Design” or “Digitalization”) and actual plagiarism. Our system addresses this by explicitly
28 modeling the “Study Location” and “Methodology” as separate semantic criteria, reducing false
29 positives caused by common academic jargon or shared internship locations. 2. **Multimodality:**
30 Traditional tools frequently miss “visual plagiarism,” where students might rewrite the text
31 but copy diagrams, charts, or results directly. By integrating CLIP, our application detects
32 similarities in visual content that text-only tools overlook ([Chowdhury & Chellappa, 2016](#)).
33

34 This software provides a robust, scalable, and locally deployable solution to enforce academic
35 honesty, specifically tailored to the linguistic and structural needs of Malagasy university
36 research.

37 Implementation and Architecture

38 The application follows a modular architecture. The core processing pipeline handles
 39 PDF extraction, separating text and images. - **Text** is encoded into dense vectors using
 40 SentenceTransformer to capture deep semantic meaning (Devlin et al., 2019). - **Images** are
 41 processed via CLIP to project visual data into a shared embedding space. - **Data Storage** is
 42 hybrid: metadata and structured criteria (Theme, Location, etc.) are stored in a Relational
 43 DBMS, while high-dimensional embeddings are indexed in a Vector Database for real-time
 44 retrieval.

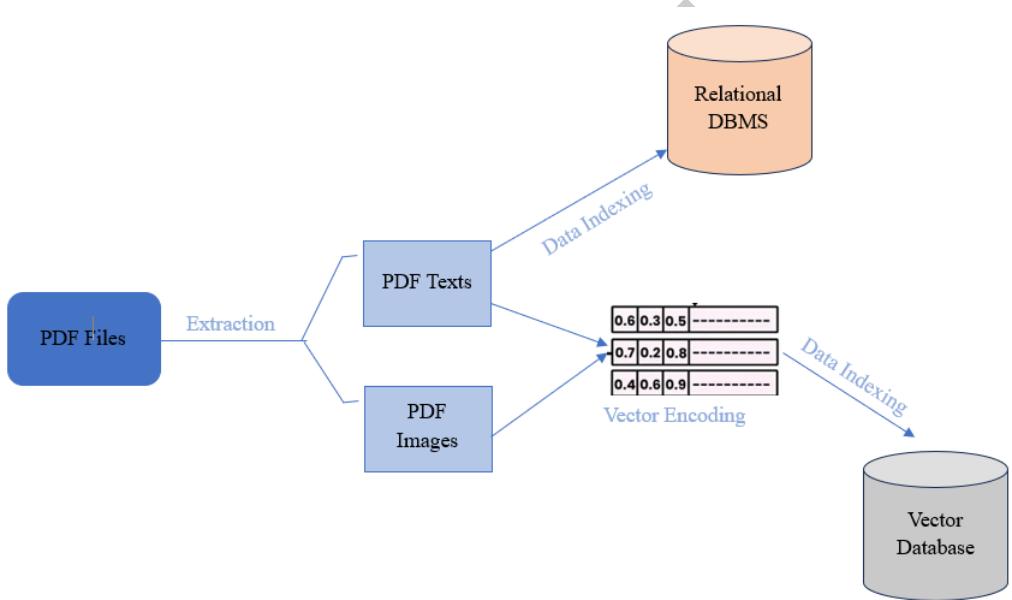


Figure 2: System Architecture: Data flow from PDF extraction to hybrid storage (Relational and Vector Database).

45 The global similarity score (S_{global}) is computed using an egalitarian weighting model, aggre-
 46 gating cosine similarities from the six defined criteria. This allows for a nuanced assessment,
 47 providing decision support thresholds (e.g., >80% for high suspicion) rather than a simple
 48 binary judgment.

49 References

- 50 Chowdhury, A. K., & Chellappa, R. (2016). Visual plagiarism: A new challenge in multimedia
 51 forensics. *IEEE Transactions on Information Forensics and Security*, 11(8), 1709–1724.
- 52 Devlin, J., Chang, M.-W., Lee, K., & Toutanova, K. (2019). BERT: Pre-training of deep bidi-
 53 rectional transformers for language understanding. *Proceedings of NAACL-HLT*, 4171–4186.
 54 <https://arxiv.org/abs/1810.04805>
- 55 Radford, A., Kim, J. W., Hallacy, C., Ramesh, A., Goh, G., Agarwal, S., Sastry, G., Askell,
 56 A., Mishkin, P., Clark, J., & others. (2021). Learning transferable visual models from
 57 natural language supervision. *International Conference on Machine Learning*, 8748–8763.
 58 <https://arxiv.org/abs/2103.00020>
- 59 Reimers, N., & Gurevych, I. (2019). Sentence-BERT: Sentence embeddings using siamese
 60 BERT-networks. *Proceedings of the 2019 Conference on Empirical Methods in Natural
 61 Language Processing*. <https://arxiv.org/abs/1908.10084>