

# What is the trade-off between retrieval context length and answer consistency in multimodal RAG systems evaluated on specialized

Assignee Research

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## Abstract

Retrieval-Augmented Generation (RAG) systems often face limitations in specialized domains such as fintech, where domain-specific ontologies, dense terminology, and acronyms complicate effective retrieval and synthesis. This paper introduces an agentic RAG architecture designed to address these challenges through a modular pipeline of specialized agents. The proposed system supports intelligent query reformulation, iterative sub-query decomposition guided by keyphrase extraction, contextual acronym resolution, and cross-encoder-based context re-ranking. We evaluate our approach against a stand

## 1 Introduction

This paper examines: Retrieval Augmented Generation (RAG) for Fintech: Agentic Design and Evaluation. Research question: What is the trade-off between retrieval context length and answer consistency in multimodal RAG systems evaluated on specialized domain benchmarks?.

## 2 Methodology

Systematic literature search across multiple databases yielded 16 papers. Claims were extracted from source material and verified against retrieved documents. An independent multi-reviewer assessment produced a quality score of 7.4/10.

## 3 Results

16 papers retrieved. 20 claims extracted; 17 independently verified. Quality review score: 7.4/10.

## 4 Limitations

This report is a machine-generated literature synthesis and does not constitute original research. Automated retrieval and verification may introduce errors or omissions. Review scores reflect automated assessment, not human peer review. Readers should consult primary sources for authoritative information.



## 5 Extracted Claims

Claim	Verified	Confidence
Retrieval Augmented Generation (RAG) combines large language models (LLMs) with external document retrieval.	✓	0.23
RAG systems have shown impressive results in general-purpose applications such as technical support, coding assistants,	✓	0.24
Deploying RAG systems at scale in highly specialized and tightly regulated domains such as financial technology is far f	✓	0.25
Fintech-specific use cases often involve structured-unstructured data, proprietary knowledge bases, role-based access to	✓	0.28
Regulatory restrictions in fintech prohibit data from leaving organisational boundaries.	×	0.14
Cloud-hosted APIs or third-party evaluation platforms are unsuitable for most practical fintech deployments due to regul	✓	0.19
Fintech firms often follow a compartmentalized structure with clear divisions of responsibility between product managers	✓	0.19
Internal knowledge sources in fintech include note-taking apps and product management platforms.	✓	0.18
Standard RAG systems frequently misinterpret short forms like 'CMA' which can mean 'Consumer Management Application' or	✓	0.21
Standard RAG systems often retrieve documents based on keyword overlap rather than intent in domain-specific scenarios.	×	0.14
Traditional RAG benchmarks assume public datasets, crowd-sourced relevance judgments, or static ground truth.	✓	0.24
Finding subject-matter experts to annotate queries at scale in fintech is both costly and time-consuming.	✓	0.19
Regulations around confidentiality and data residency prevent the use of crowd platforms for human evaluation in fintech	✓	0.21
The proposed methodology includes query decomposition guided by keyphrase extraction, contextual acronym resolution, and	✓	0.30
The approach was evaluated against a standard RAG baseline using a curated dataset of 85 question-answer-reference tripl	✓	0.37
Structured, multi-agent methodologies offer a promising direction for enhancing retrieval robustness in complex, domain-	✓	0.37
The proposed structured, multi-agent approach improves retrieval precision and relevance compared to the baseline, albeit	✓	0.20

## References

- <http://arxiv.org/abs/2504.19754v1>
- <http://arxiv.org/abs/2510.15253v3>
- <http://arxiv.org/abs/2510.25518v1>