Translation or Multilingual Retrieval? **Evaluating Cross-Lingual Search Strategies for Traditional Chinese Financial Documents**

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MOTIVATION

Traditional Chinese is underrepresented in CLIR, especially in finance where precise terminology matters. Dense models often miss key terms, returning semantically similar but legally incorrect results. This motivates revisiting translation-based retrieval—aligning queries and documents to improve term-level matching.

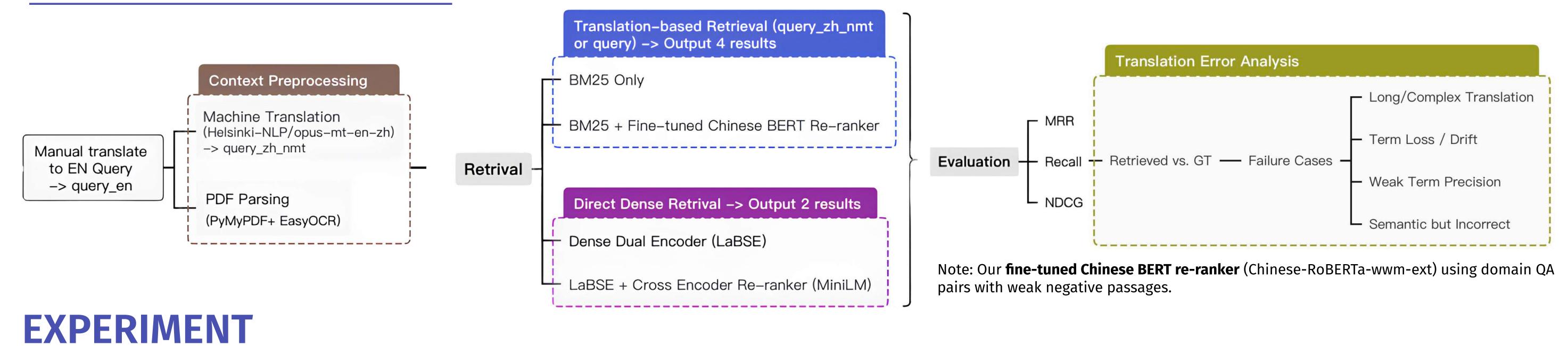
DATASET

We evaluate six retrieval pipelines using the **AI Cup 2024 dataset:**

• 1,600+ Traditional Chinese documents from financial reports, insurance contracts, and FAQs

We evaluate its advantage over multilingual models on financial QA tasks. This final version integrates all 11 reviewer suggestions to inform robust CLIR in high-stakes, terminology-rich domains.

- OCR fallback for scanned or image-heavy PDFs
- 150 manually translated English queries covering regulatory and operational intent
- All documents and query pairs are in the finance/legal domain, where retrieval failure carries real-world risk.



METHODOLOGY

Retrieval Setup	F
6 pipelines tested on 150 queries (ZH/NMT/EN).	BM
BM25+Rerank yields best MRR@10.	BM
Cross-encoder built on LaBSE top-100.	BM25+
All rerankers = fine-tuned Chinese RoBERTa.	
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Pipeline	Query Type	MRR@10	R@100	Latency (s/query)	(
BM25	ZH	0.1067	0.5667	0.105	- (
BM25+Rerank	ZH	0.1535	0.5800	19.211	(
BM25+NMT	NMT	0.0423	0.3533	0.105	
BM25+NMT+Rerank	NMT	0.0755	0.5133	19.211	MRR@10
LaBSE	EN	0.0526	0.3400	0.274	MR
Cross-Encoder	EN	0.0125	0.1600	3.004	

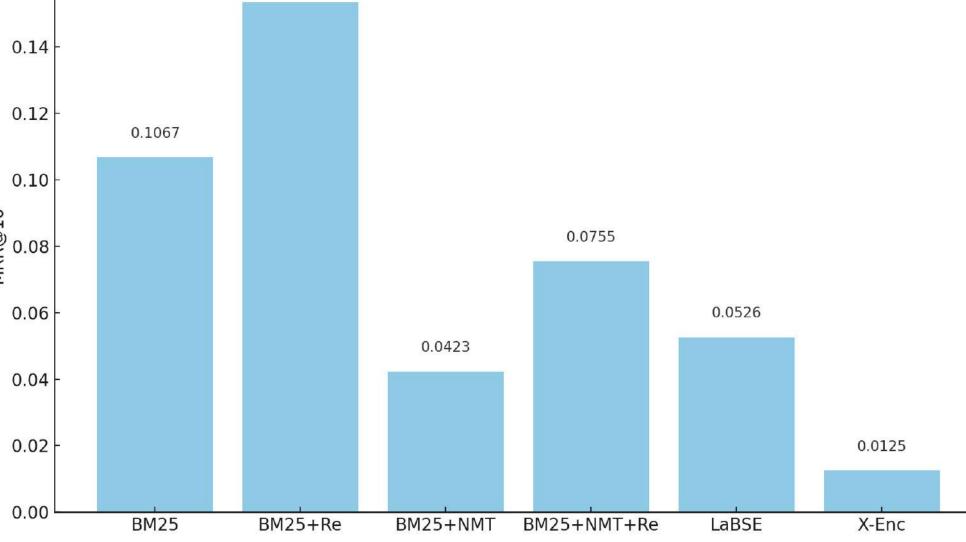
- BM25+Reranker gives best accuracy (MRR@10 = 0.1535) but is slow (~19s/query).
- LaBSE offers faster, balanced trade-off (0.27s/query).
- Reranking boosts MRR 2× on both native and translated queries.
- Translation excels at regulatory keywords; dense models help with verbose concepts.

TRANSLATION ERROR ANALYSIS

We manually compared retrieved passages to gold answers and categorized translation-based failures into four types: (1) Long or complex translations dilute key info, (2) Term loss or drift from mistranslated terminology,

CONCLUSION

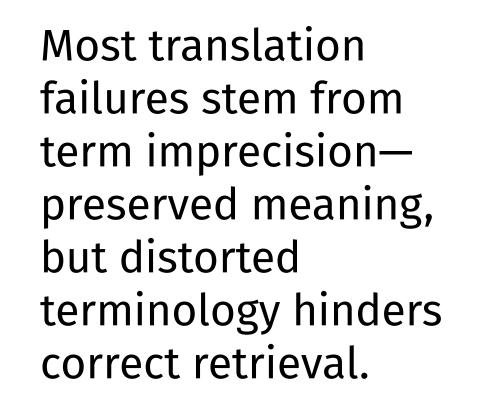
Translation-based retrieval, combining query translation, BM25, and a domain-adapted reranker, outperforms multilingual dense models on terminology-heavy financial documents.

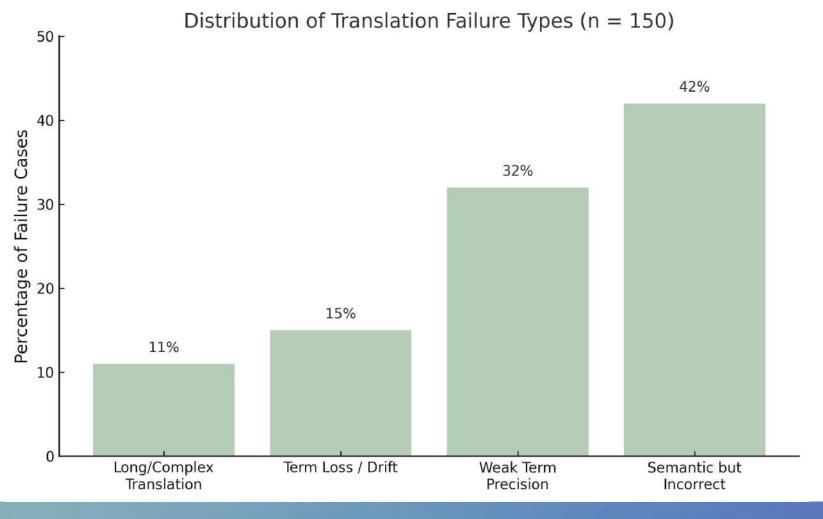


MRR@10 Across Pipelines

0.1535

(3) Weak term precision leads to partial but insufficient matches (4) Semantically plausible but incorrect results.





These results highlight the importance of precise terminology alignment in specialized domains such as finance and insurance.

Multilingual encoders provide broad semantic coverage, but often underperform without fine-tuning.

Moving forward, we aim to design hybrid pipelines that combine multilingual embeddings for coverage with translated reranking for precision. We also plan to fine-tune cross-encoders, and lightweight on financial QA pairs and extend evaluation to compliance-related domains.



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GitHub (Code & Models) Open-source at: github.com/Eric0801/NTCIR-18-CLIR-pipeline Fork, replicate, build on top.