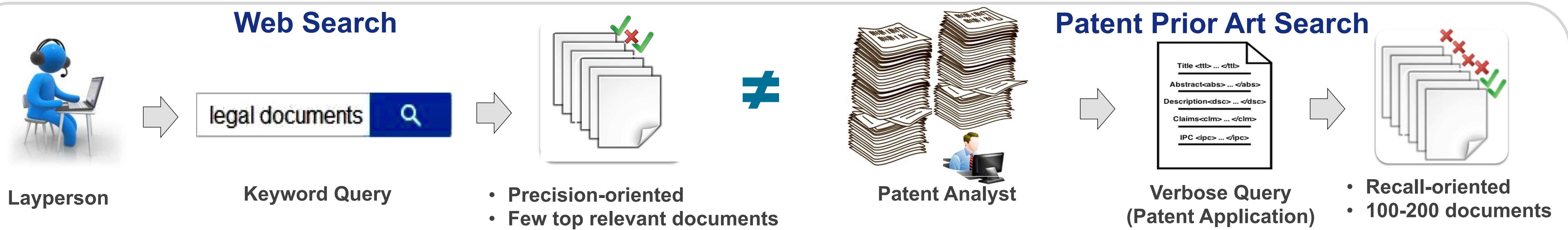


On Term Selection Techniques for Patent Prior Art Search

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Introduction



Patent Prior Art Search: Finding previously granted patents or any published work relevant to new patent application.

Data Collection: English subset of CLEF-IP 2010 [1].

PATATRAS: Highly engineered system and top CLEF-IP 2010 competitor [2].

Previous Work: Mainly focused on different query expansion techniques to cope with significant term mismatch [3].

Problem Definition: Generic IR techniques (e.g., different query reformulation) are ineffective for patent prior art search!

Oracular Term Selection

1. Relevance Feedback Score

Relevance feedback (RF) score for each term ($t \in \{\text{top-100}\}$):

$$\text{RF}(t, Q) = \text{Rel}(t, Q) - \text{Irr}(t, Q) \quad (1)$$

where

$\text{Rel}(t) \rightarrow \text{Avg. Term Frequency in Rel. Docs.}$

$\text{Irr}(t) \rightarrow \text{Avg. Term Frequency in Irr. Docs.}$

2. Oracular Query Formulation

Formulate two oracular queries:

- ① Oracular Query = $\{t \in \text{top} - 100 | \text{RF}(t, Q) > \tau\}$
- ② Oracular Patent Query = $\{t \in Q | \text{RF}(t, Q) > \tau\}$

Take Home Message

- Sufficiency of terms in baseline query
- Precise methods to eliminate poor query terms (query reduction)

3. Baseline vs. Oracular Query

Table.1: Performance for the Baseline Query, two variants of the Oracular Query, and PATATRAS.

		Baseline	PATATRAS	Oracular Query	Oracular Patent Query
LM	MAP	0.112	0.226	0.482	0.414
	Recall	0.416	0.467	0.582	0.591
BM25	MAP	0.123	0.226	0.492	0.424
	Recall	0.431	0.467	0.584	0.598

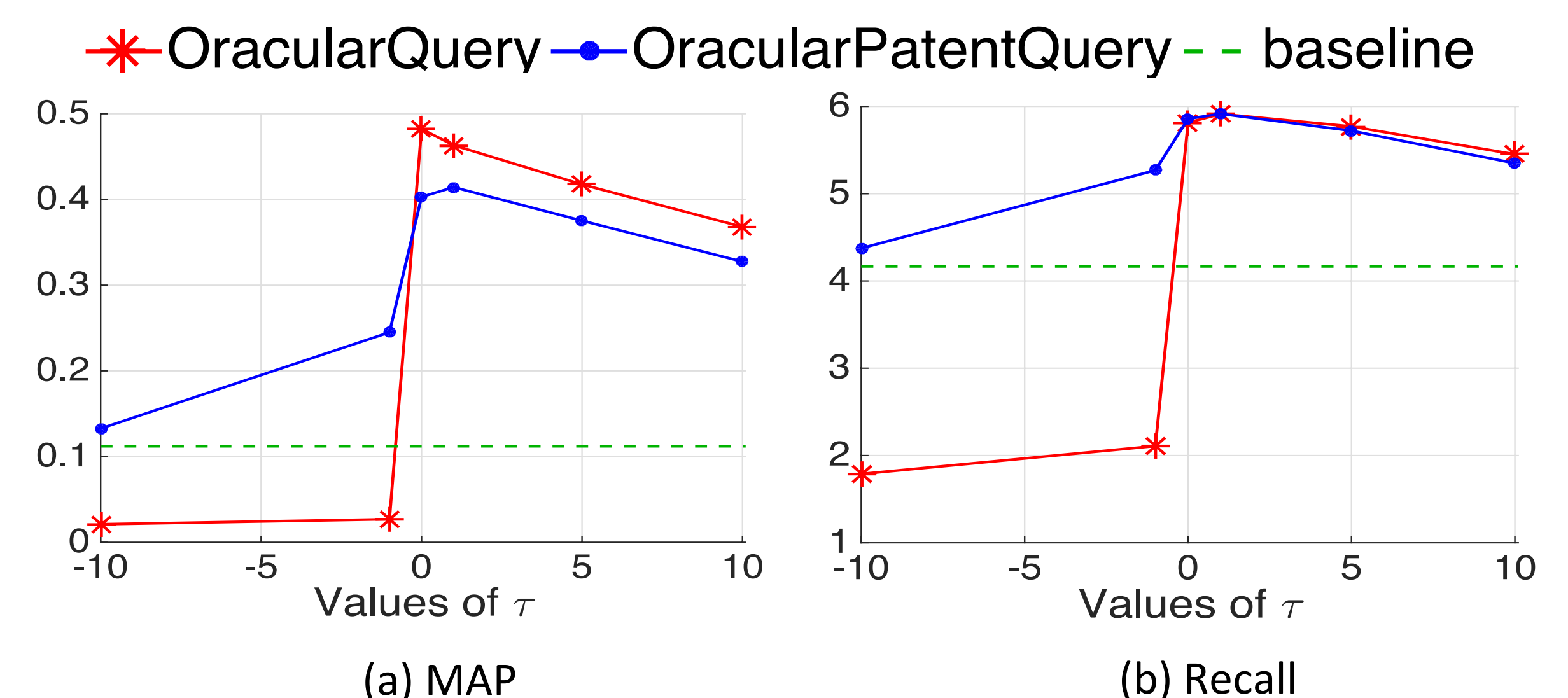


Fig.1: Comparing the performance of two different oracular queries.

Query Reduction (QR): Approximating Oracular Query

1. Automated Reduction

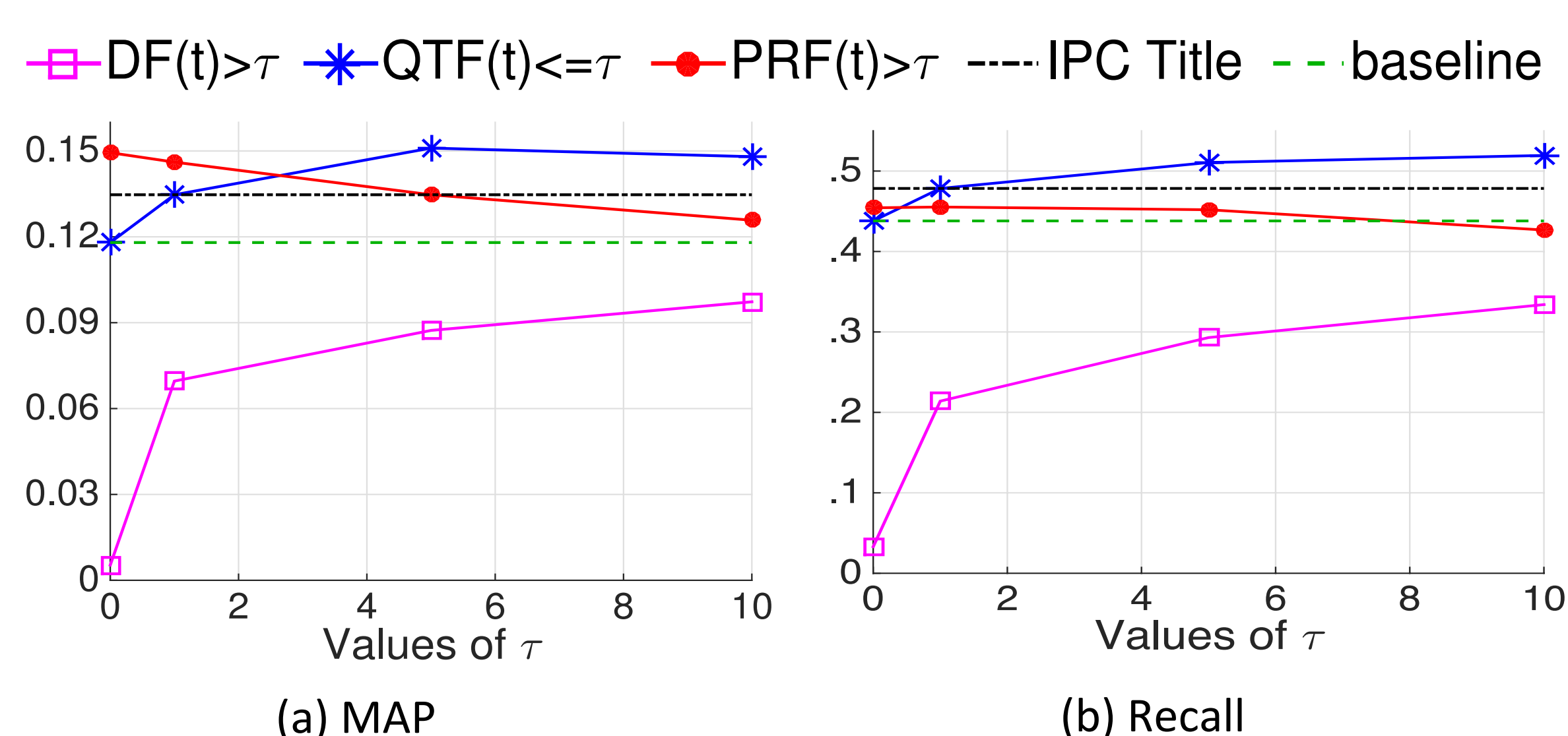


Fig.2: System performance vs. the threshold τ for four QR approaches.

QR Approaches:

- Pruning document frequent (DF) terms ($DF(t) > \tau$).
- Pruning query infrequent terms ($QTF(t) \leq \tau$).
- Pseudo relevance feedback term selection ($PRF(t) > \tau$).
- Pruning IPC title general terms.

Take Home Message

- Proposed QR methods **fail** to approximate oracular query.
- They cannot **discriminate** between useful and noisy terms.

2. Semi-automated Interactive Reduction

Table.2: Performance of an Oracular Patent Query derived from only the top-k ranked relevant documents identified in the search results. We assume that the remaining documents in the top-100 are irrelevant.

	Baseline	PATATRAS	Oracular Patent Query (k=1)	Oracular Patent Query (k=3)
MAP	0.112	0.226	0.289	0.369
Avg. Recall	0.416	0.467	0.484	0.547

- MAP **Doubles** over the baseline ($0.112 \rightarrow 0.289$)
- Outperforms** PATATRAS ($0.226 \rightarrow 0.289$)

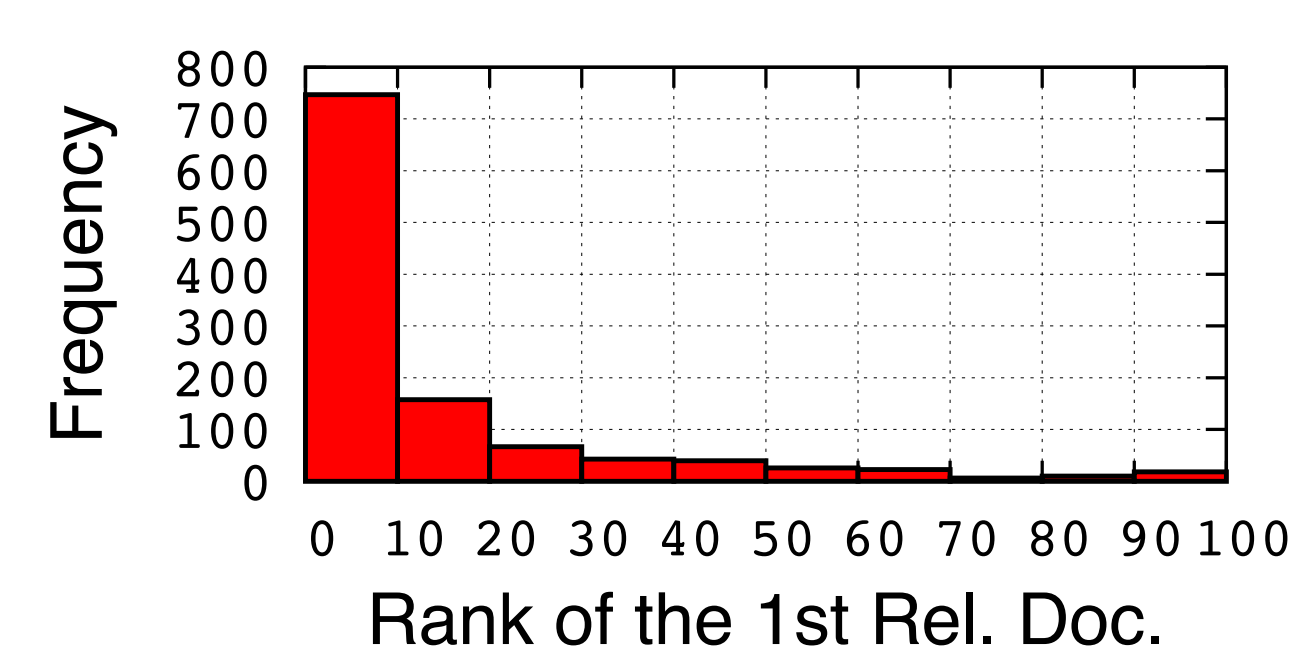


Fig.3: The distribution of the first relevant document rank over test queries.

- Baseline returns first rel. patent
- ✓ 80% of time in top 10 results,
- ✓ 90% of time in top 20.
- Minimum user effort

Take Home Message

- Interactive methods** offer a promising avenue for simple but effective term selection in prior art search.

Acknowledgement

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