

Re-thinking significant term discovery

Mark Harwood 25/2/26
@elasticmark

Some background

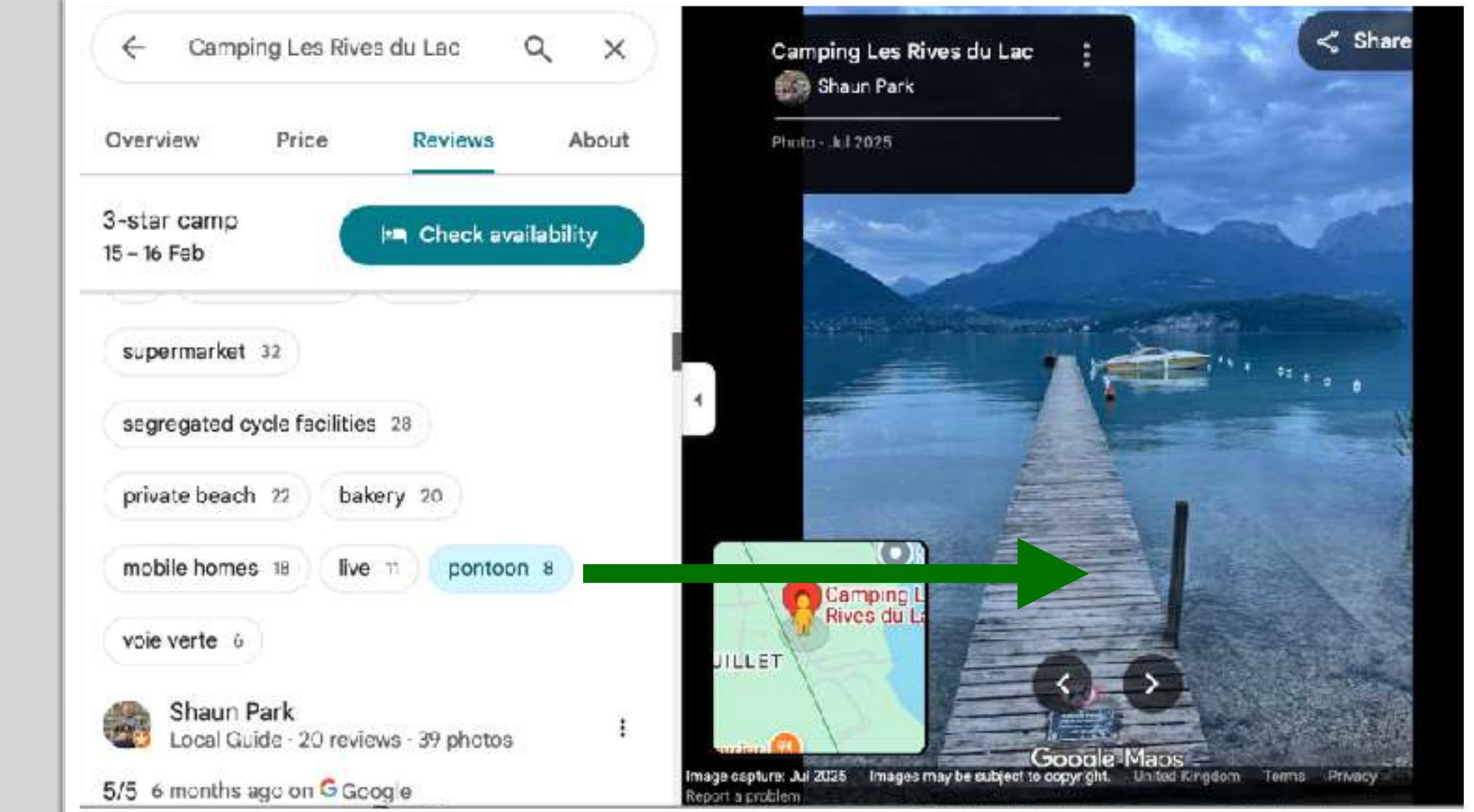
- About me:
 - Early Lucene contributor 
 - Early elastician (employee #30) 
 - Currently tinkering on hybrid search interfaces

Text summarisation with significant terms

News cluster labels

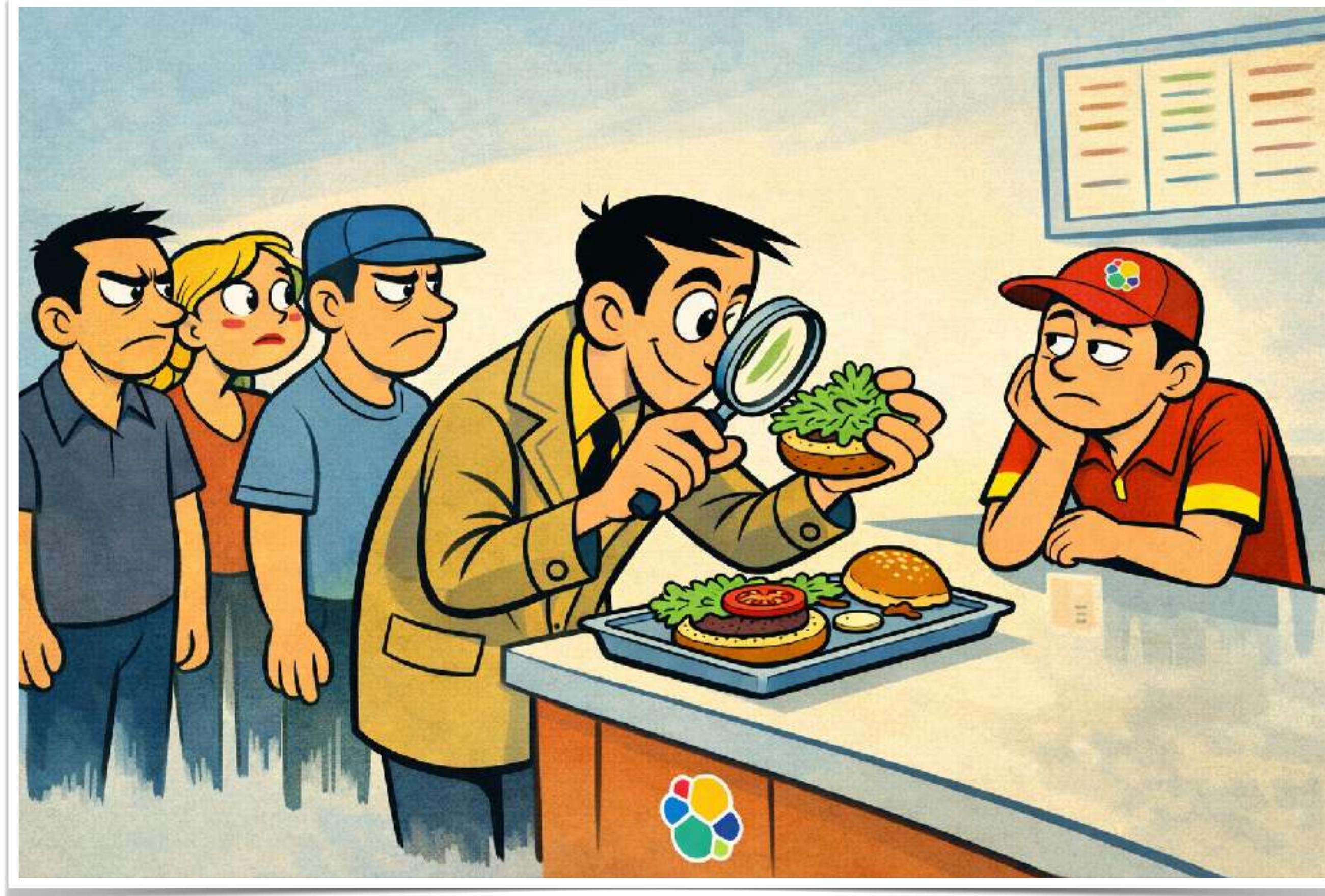


Google Maps reviews



Elasticsearch's text analytics

Significant term discovery can be expensive



Expensive aggregations can cause problems

Client-side text analytics

Same discovery algorithms (and more...)



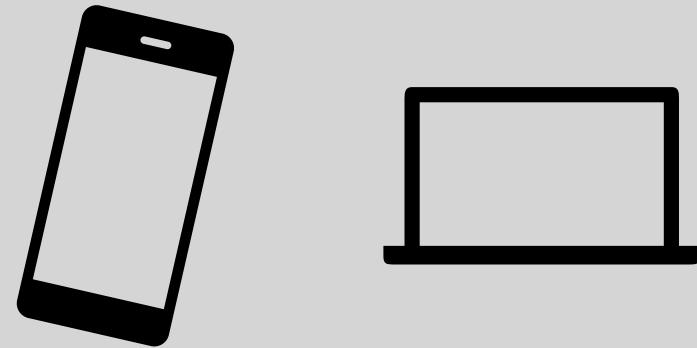
Same analysis, better location

Text summarisation options

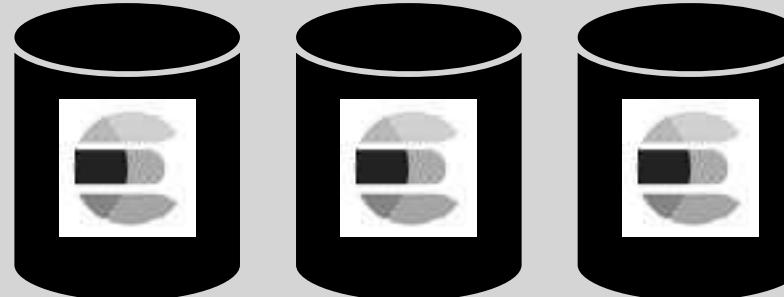
Where to compute?



On your clients?



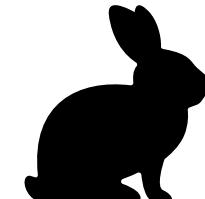
In your
elasticsearch
cluster?



Cloud icon followed by the text "LLM Provider?"

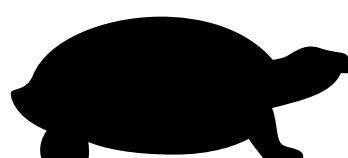


FREE!



Significant keywords
and phrases

\$\$



Significant keywords

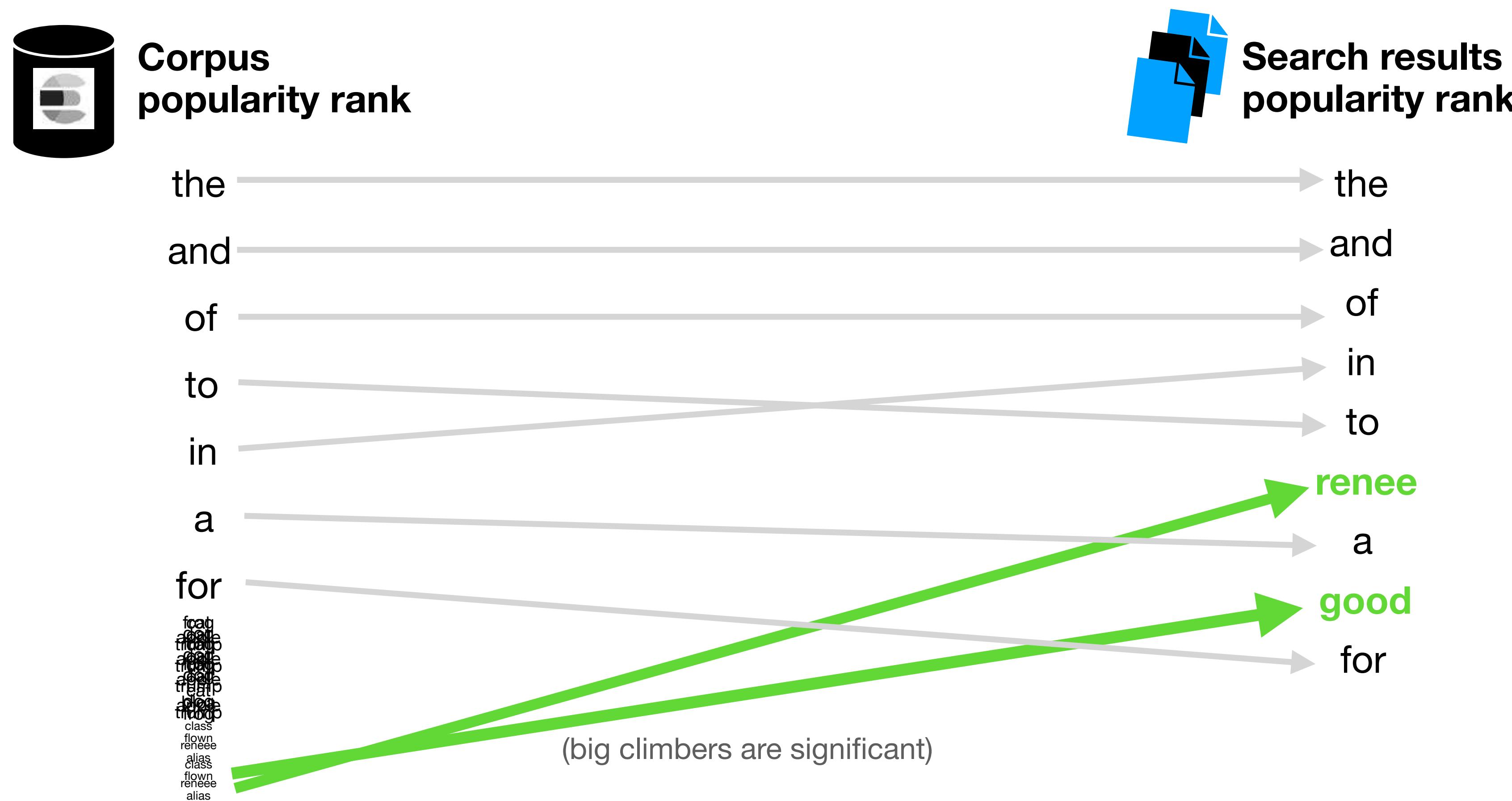
\$\$\$



Unstructured and
structured summaries

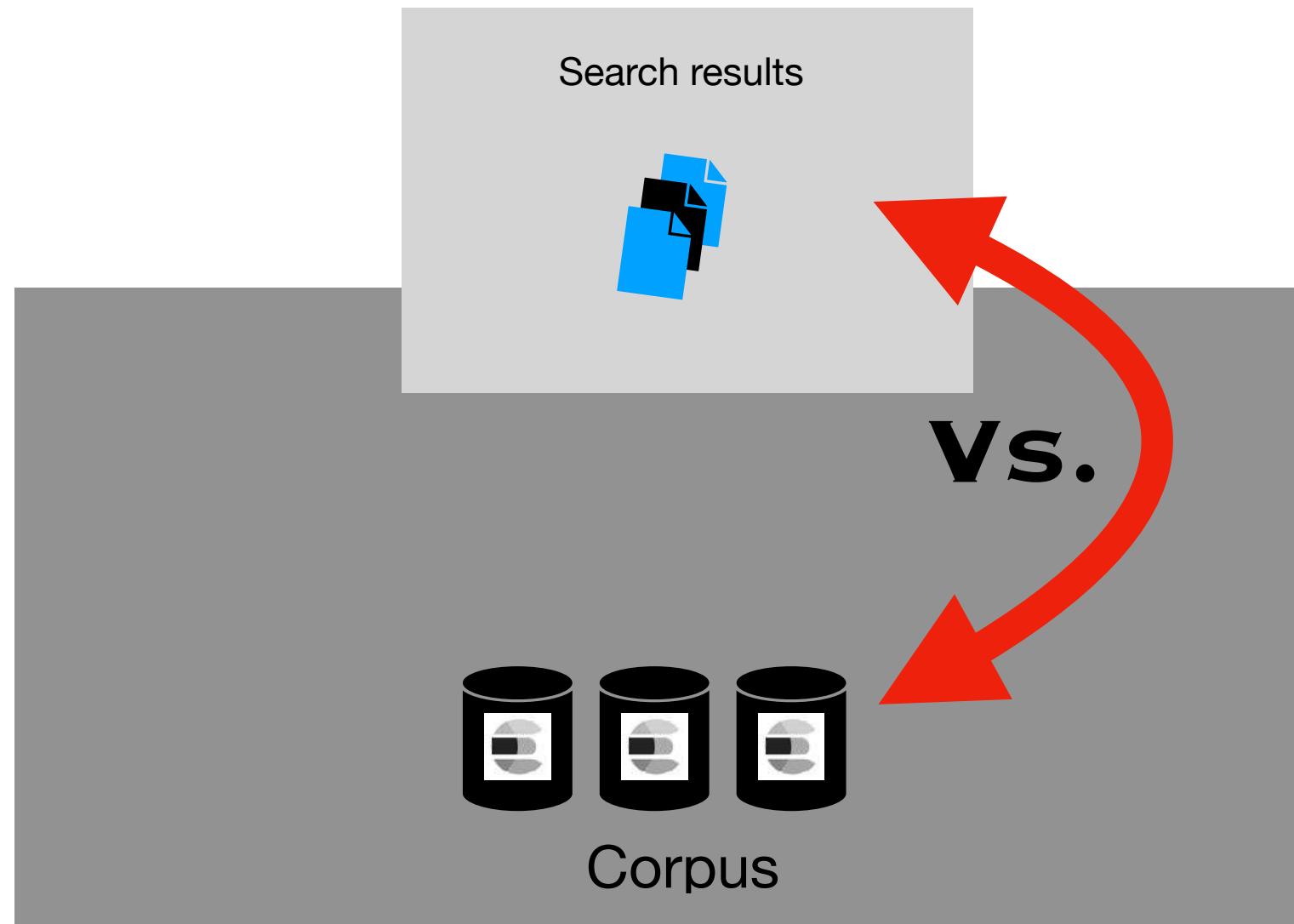
Recap: how elasticsearch determines significance

Foreground vs background word popularity



Comparison issues in elasticsearch

What makes this particular analysis expensive?



- **Tokenisation costs**

the top search results have to have their text re-analyzed to discover the foreground terms

- **Fragmented background data**

Each term's background frequency is expensive to derive as it is spread across indexes → shards → on-disk segments. Date-based indices actually make it impossible to answer “what is trending today (compared to previously)?”

- **Memory pressure**

large dictionaries / heaps push JVM heap; circuit-breakers and GC churn kick in.

- **IO and network**

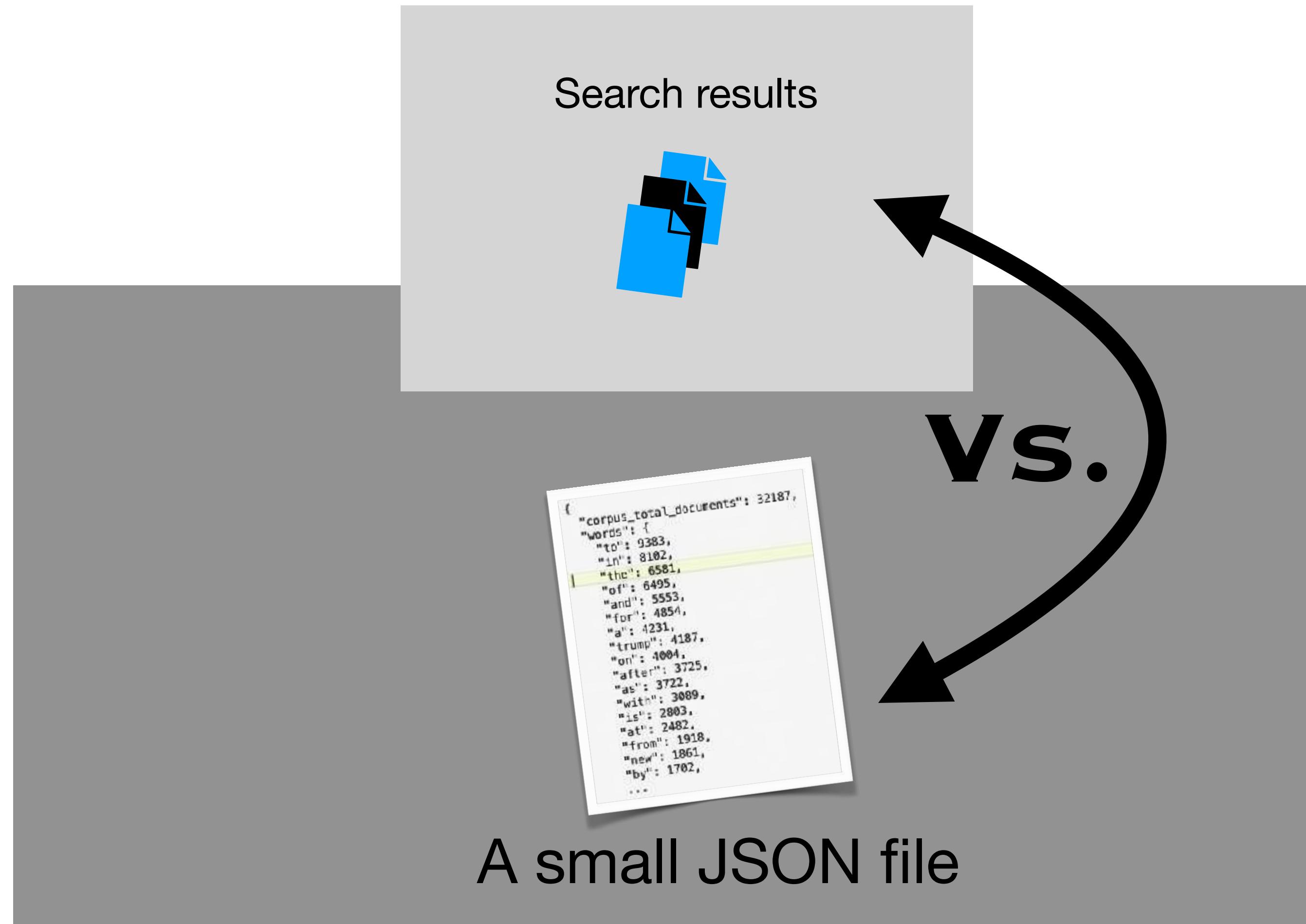
segment seeks, shard fan-out, and cross-node reduces add latency and cost.

- **Shared resources**

heavy BG lookups run alongside everyone else's queries, competing for CPU/heap/I/O. Risk of hot shards, cache churn, and GC pauses can reduce cluster stability

Clientside alternative

Foreground vs background word popularity...

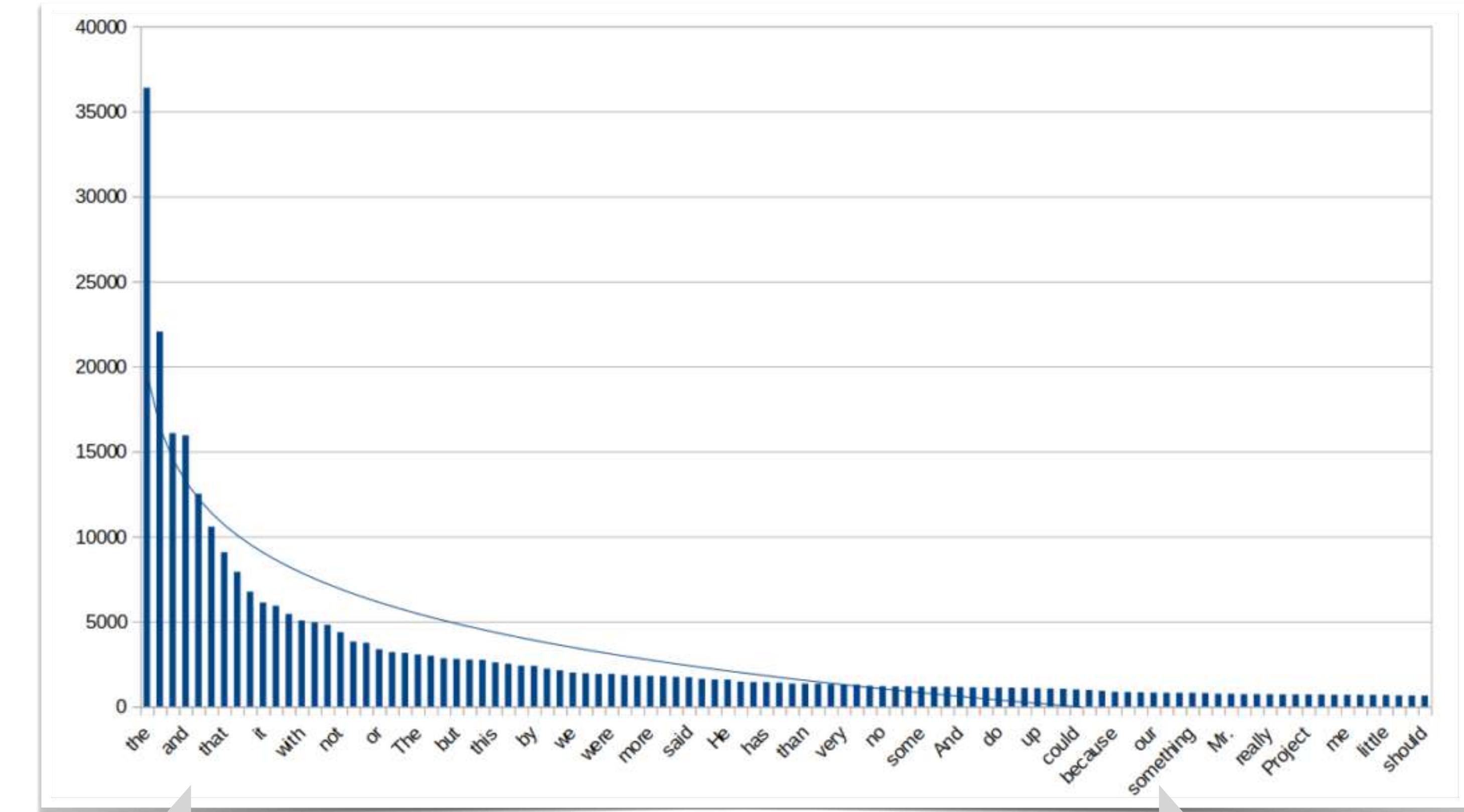


How much data is needed for a background?

Zipf's law works in our favour...

```
{  
  "corpus_total_documents": 32187,  
  "words": {  
    "to": 9383,  
    "in": 8102,  
    "the": 6581, highlighted  
    "of": 6495,  
    "and": 5553,  
    "for": 4854,  
    "a": 4231,  
    "trump": 4187,  
    "on": 4004,  
    "after": 3725,  
    "as": 3722,  
    "with": 3089,  
    "is": 2803,  
    "at": 2482,  
    "from": 1918,  
    "new": 1861,  
    "by": 1702,  
    ...  
  }  
}
```

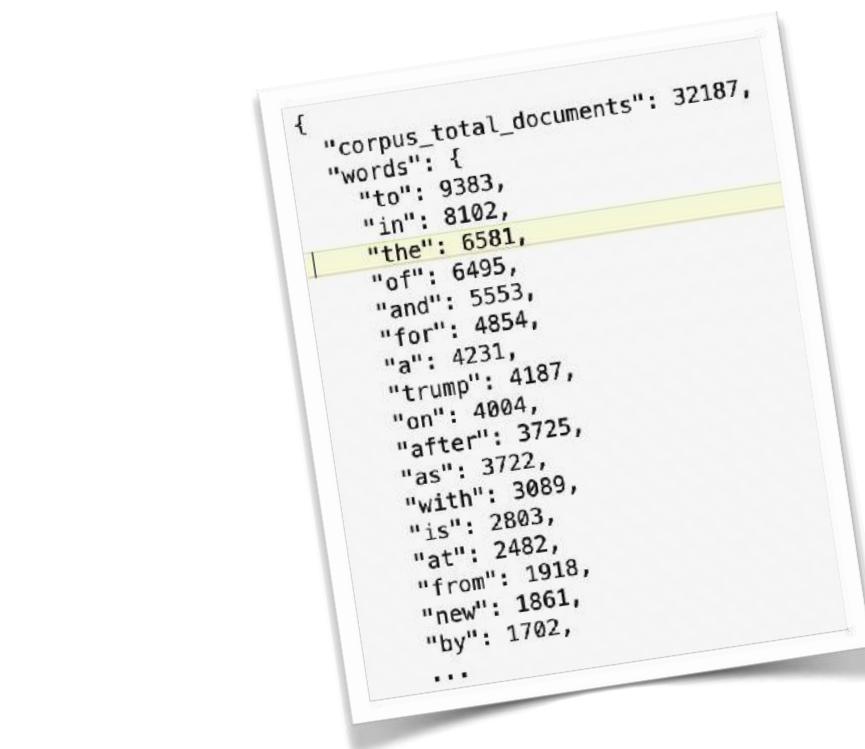
(<100kb zipped)



(~20k top words)

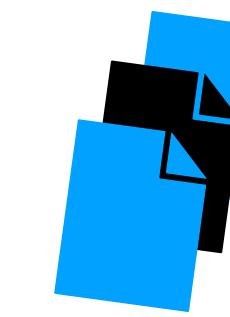
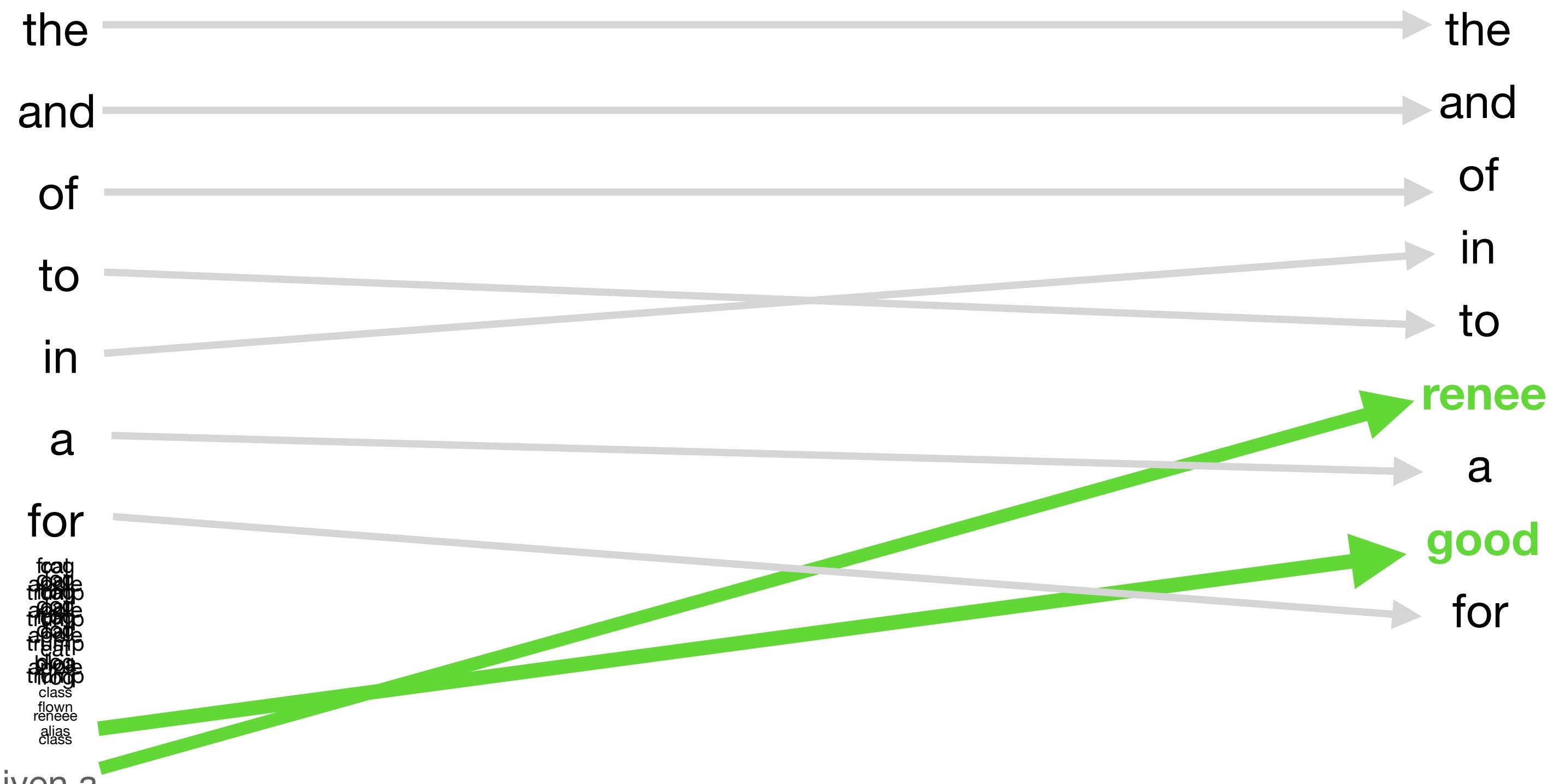
Does it really work?

Can a small background sample produce quality results?



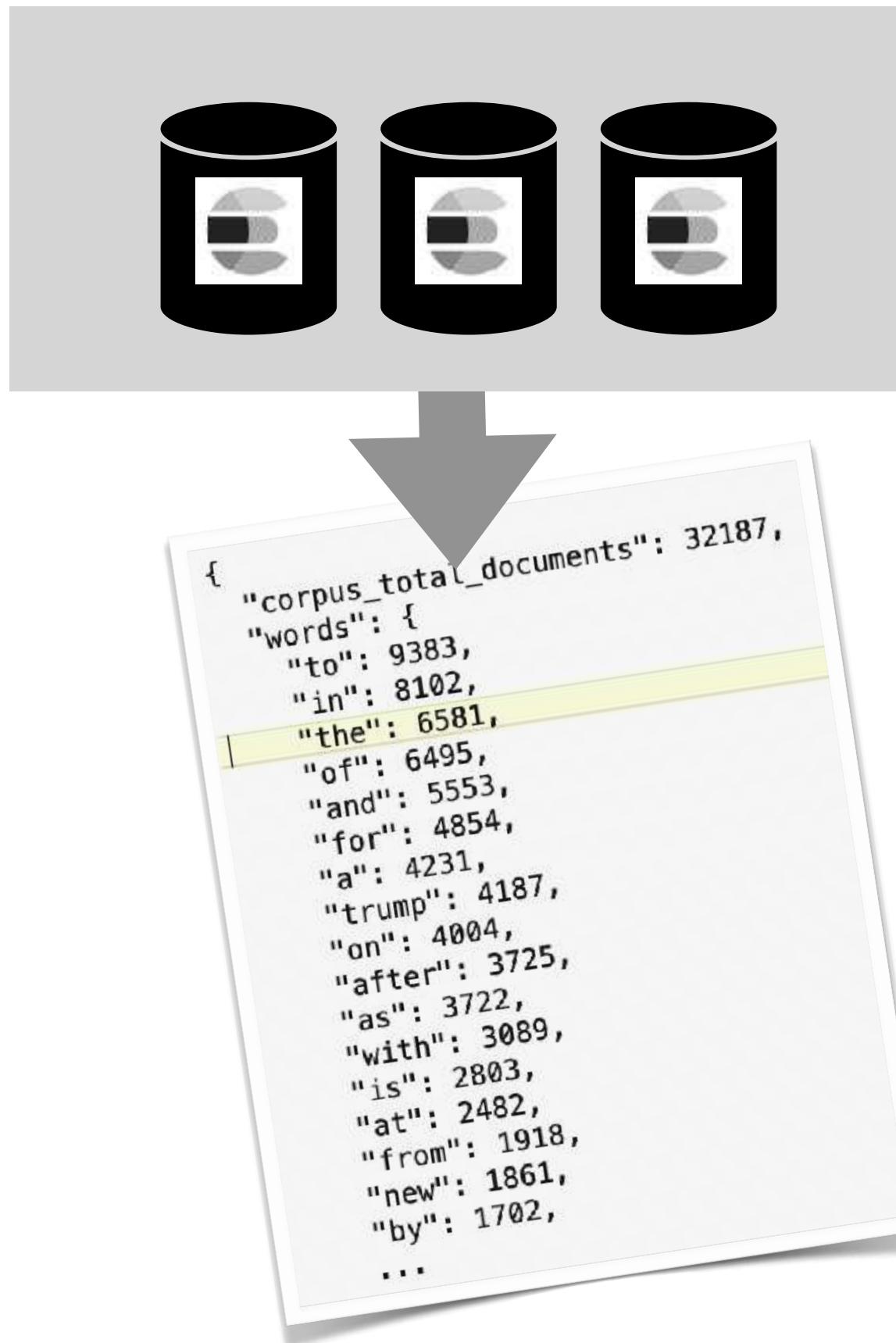
A JSON file of ~20k popular words and frequencies

Sampled
background
popularity rank



Search results
popularity rank

Exporting word popularities from your index



```
GET myindex/_search  
{  
    "query": {  
        "function_score": {  
            "query": {  
                "match_all": {}  
            },  
            "random_score": {}  
        }  
    },  
    "size": 0,  
    "aggs": {  
        "sample": {  
            "sampler": {  
                "shard_size": 50000  
            },  
            "aggs": {  
                "keywords": {  
                    "significant_text": {  
                        "field": "headline",  
                        "size": 100000,  
                        "include": "[a-zA-Z]\\S*",  
                        "script_heuristic": {  
                            "script": {  
                                "lang": "painless",  
                                "source": "params._superset_freq * 1"  
                            }  
                        }  
                    }  
                }  
            }  
        }  
    }  
}
```



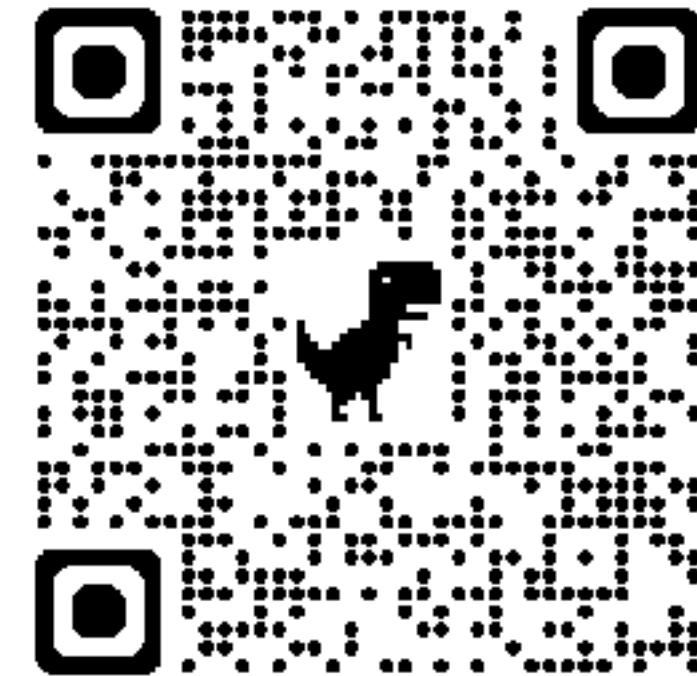
Open source JS algorithms for use in browsers

Install

```
> npm i @andorsearch/significant-terms
```

Repository

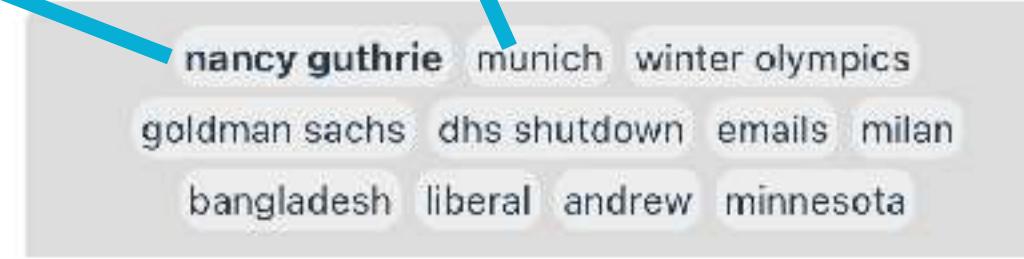
❖ github.com/markharwood/significant-terms



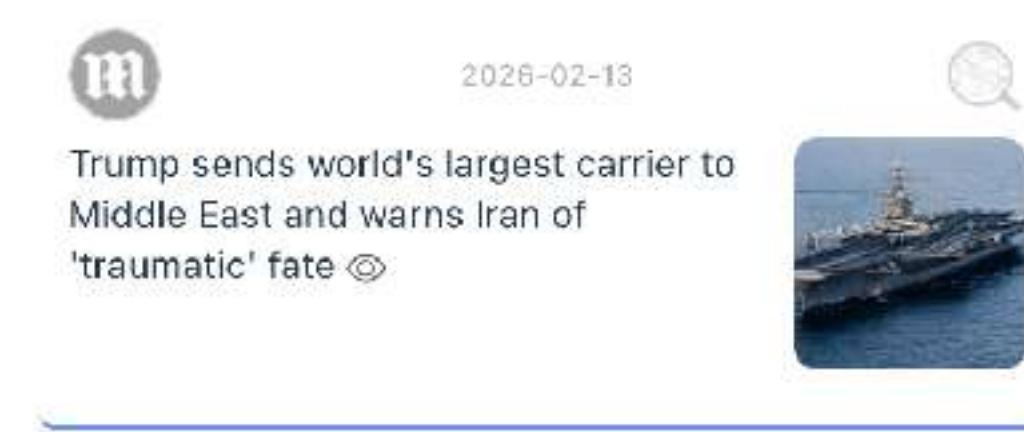
Significant terms detection example

Detect single words *and* phrases e.g. people names

```
function findSignificantWordsOrPhrases(searchResultTexts: string[]) {  
    // Tokenise the text found in search results  
    let tokenStreams = searchResultTexts.map((textValue) => simpleTokenizer(textValue))  
  
    //Find the significant words compared to the background  
    const significantWords = computeSignificantTerms(  
        tokenStreams,  
        backgroundWordStats,  
        backgroundCorpusSize  
    );  
    // Optionally, examine how the significant words are placed in the text to identify word p  
    let significantWordsOrPhrases = detectAndSortSequences(significantWords, tokenStreams)  
}
```



Trending in today's news:



Comparing approaches

Pros and cons

Trade-off	Javascript	Elasticsearch aggregation
Content scope	All raw text under consideration must be sent to the client	Aggregations can summarise content without sending raw original text to clients
Compute cost	Mostly offloaded to client	Can tie-up your cluster
Readability	Can discover readable terms and phrases	Can be unintelligible (e.g. stemmed words used in the search index)
Score correctness	Dictated by choice of background sample	Dictated by physical arrangement of data (number of indices/shards/shared tenants)
Dependencies	Works with any database backend	Works with elasticsearch and indexed fields

Demo