Search Quality Evaluation in the Era of Large Language Models: Dataset Generator

Speaker: Alessandro Benedetti

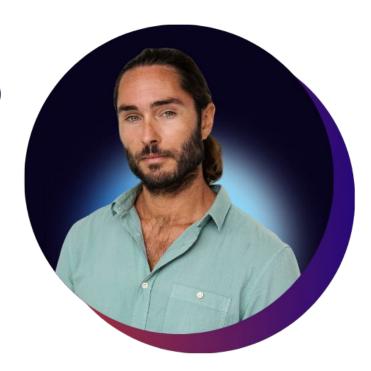
Search Solutions - 26th November 2025

Who are we

ALESSANDRO BENEDETTI

Director + R&D Software Engineer @ Sease

- Born in **Tarquinia** (ancient Etruscan city in Italy)
- Master Degree in Computer Science
- Program Committee member for ECIR, SIGIR and Desires
- Apache Solr Chair of the PMC + Apache Lucene/Solr committer
- Elasticsearch/OpenSearch expert
- Semantic search, NLP, Machine Learning technologies passionate
- Beach Volleyball player and Snowboarder



- Headquarter in London/distributed
- Open-source Enthusiasts
- Apache Lucene/Solr experts
- Elasticsearch/OpenSearch experts
- Community Contributors
- Active Researchers



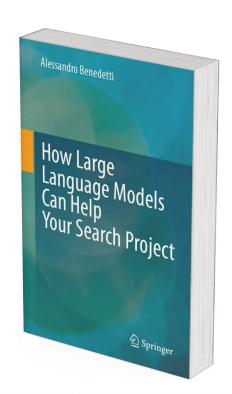
Hot Trends:

- Large Language Models Applications
- Vector-based (Neural) Search
- Natural Language Processing
- Learning To Rank
- Document Similarity
- Search Quality Evaluation
- Relevance Tuning





How Large Language Models Can Help Your Search Project



COMING SOON





Overview

- Search Quality Evaluation Problem
 - **2** The Dataset Generator Introduction
 - How the Dataset Generator works
- **A** Next Steps



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Why measure search quality?

- You <u>can't improve</u> what you <u>can't</u> <u>measure</u>
- It opens a tangible <u>comparison between</u> <u>more query designs</u>
- It <u>reveals</u> dataset and model <u>weaknesses</u>
- It's the foundation for long-term iteration
- It unlocks <u>continuous improvement</u> through measurement



Implicit and Explicit feedback

<u>Implicit</u>

- Feedback inferred <u>automatically</u> from user behavior.
- Clicks on search results, dwell time, scroll depth, query reformulations, skipped results.
- Easy to gather at scale.
- Reflects real-world behavior.
- Noisy and ambiguous (e.g., click ≠ satisfaction)

Explicit

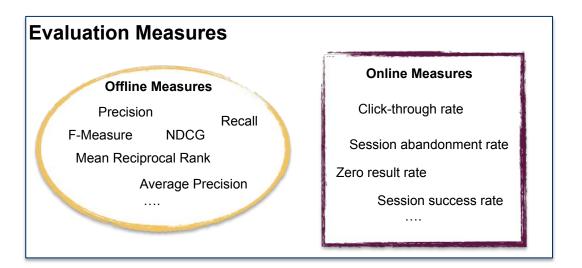
- Feedback directly provided by users in a <u>deliberate</u> way.
- Ratings (e.g., 1–5 stars), relevance judgments, thumbs up/down, written comments, likes.
- High precision and interpretability.
- Expensive and time-consuming.
- Hard to scale and may suffer from subjectivity.

Evaluation Measures

<u>Evaluation measures</u> for an information retrieval system formalise how well a search system satisfies its user information needs.

Measures are generally split into two categories: online and offline measures.

We'll focus on <u>offline</u> measures.



Offline measures

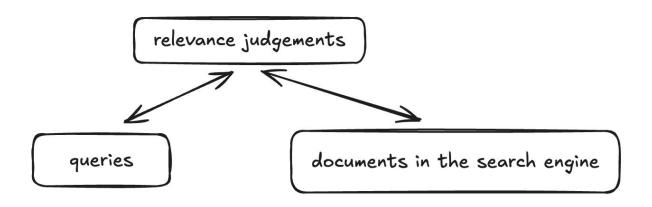
Advantages:

- Find anomalies in data, like: weird distribution of the features, strange collected values, ...
- Check how the <u>ranking function performs</u> before using it in production: implement improvements, fix bugs, tune parameters, ...
- Save <u>time</u> and <u>money</u>. Put in production a bad ranking function can worse the user experience on the website.

Offline search quality evaluation data

In the scenario of offline search quality evaluation, the main obstacle is the lack of <u>queries</u> and <u>relevance judgements</u>.

Without labeled <u>query-document pairs</u>, it becomes difficult to assess retrieval quality.



The Challenge of Obtaining Labeled Data

- Expensive & time-consuming: Requires expert annotators.
- Accuracy: Labels become less accurate due to the fact that labeling is a boring job and requires focus for a long period of time.
- <u>Subjective:</u> Relevance may depend on context, user intent, or domain.
- <u>Scalability issue:</u> Hard to label enough data for large corpora or many queries.



Why LLMs change the game for synthetic dataset generation



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Dataset Generator

Tutorial material

rated-ranking-evaluator repo:



<u>Ilm-search-quality-evaluation</u> repo:



Ilm-search-quality-evaluation-tutorial repo:



The team behind the Dataset Generator



DANIELE ANTUZIR&D SOFTWARE ENGINEER
SEARCH CONSULTANT



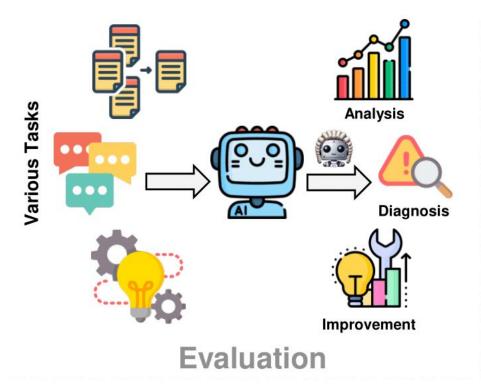
NICOLÒ RINALDI SOFTWARE ENGINEER/DATA SCIENTIST



NAZERKE SEIDAN IR/ML SOFTWARE ENGINEER



LLM-as-a judge for evaluation



Assumption: you have only the documents indexed



Generate queries

Generate relevance labels (with rationale)

Functionalities



Supported search engines (input)









Main Inputs

- Documents (quantity, filters)
- (optional) existing Queries
- LLM configurations
- Search engine connection parameters (URL, collections...)



Interpretability

- Possibility to enable LLM "explainability"
- Possibility to enrich the set of document scored for each query using a query template and retrieve more documents directly from your search engine

Dataset Generator Output formats

quepid

mteb

RRE



quepid

```
query,docid,rating
What is the nature and purpose of the mission the helicopter was on?,d1,2
How did the helicopter crash in the Colombian jungle?,d1,2
What is the involvement of U.S. technology or personnel in the Colombian drug war?,d1,2
Why did the police use a Taser on the python?,d2,2
What is the nature and purpose of the mission the helicopter was on?,d2,0
How did the helicopter crash in the Colombian jungle?,d2,0
What is the involvement of U.S. technology or personnel in the Colombian drug war?,d2,0
Why did the police use a Taser on the python?,d1,0
```



mteb

```
{"id": "d1", "title": "Helicopter Crashes in Colombian Drug War, Kills 20", "text": "BOGOTA, ... army said."}
{"id": "d2", "title": "Police Use Taser on Python to Free Man", "text": "UNIONTOWN, Pa. ... not let go."}
{"id": "d3", "title": "Zidane apologizes for head butt", "text": "French soccer ... and sister."}
{"id": "d4", "title": "Iraqi PM Says Sticking to January Election Plan", "text": "Iraqi Prime ... insurgent."}
{"id": "d5", "title": "Ga. Crematory Operator to Plead Guilty", "text": "Relatives ... guilty plea."}
{"id": "d6", "title": "Long-driving Kuehne finally breaks out of 2004 slump", "text": "Hank Kuehne ... Kuehne"}
{"id": "q1", "text": "What is the nature and purpose of the mission the helicopter was on?"}
{"id": "q2", "text": "How did the helicopter crash in the Colombian jungle?"}
{"id": "q3", "text": "What is the involvement of U.S. technology or personnel in the Colombian drug war?"}
{"id": "q4", "text": "Why did the police use a Taser on the python?"}
{"query_id": "q1", "doc_id": "d1", "rating": 2}
{"query_id": "q2", "doc_id": "d1", "rating": 2}
{"query_id": "q3", "doc_id": "d1", "rating": 2}
{"query_id": "q4", "doc_id": "d2", "rating": 2}
{"query_id": "q1", "doc_id": "d2", "rating": 0}
{"query_id": "q2", "doc_id": "d2", "rating": 0}
```



RRE

```
"index": "testcore",
"id_field": "id",
"query_placeholder": "$query",
"query_groups": [
    "name": "What is the nature and purpose of the mission the helicopter was on?",
    "queries": [
        "template": "template_solr.json",
        "placeholders": {
          "$query": "What is the nature and purpose of the mission the helicopter was on?"
    "relevant_documents": {
      "2": ["d1"],
      "1": ["d2", "d32", "d107", "d13"],
      "0": ["d175", "d189", "d162", "d88", "d126", "d65", "d184" ]
```

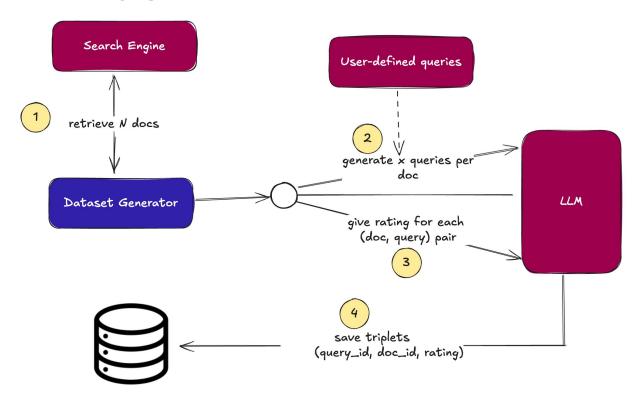


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First step: query generation





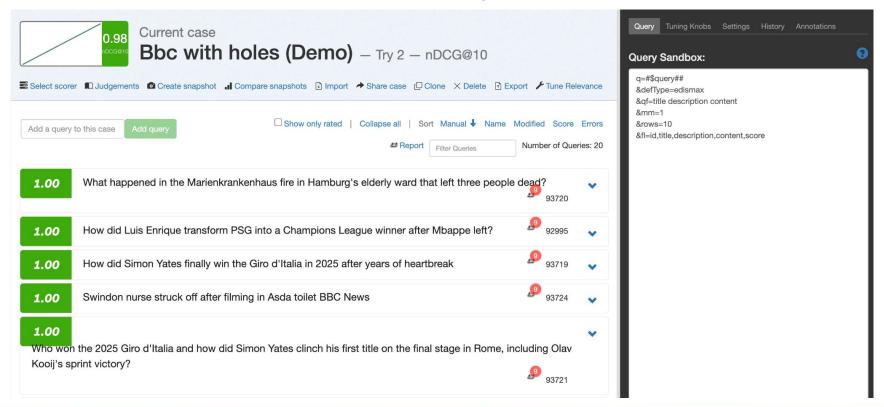
About cartesian product



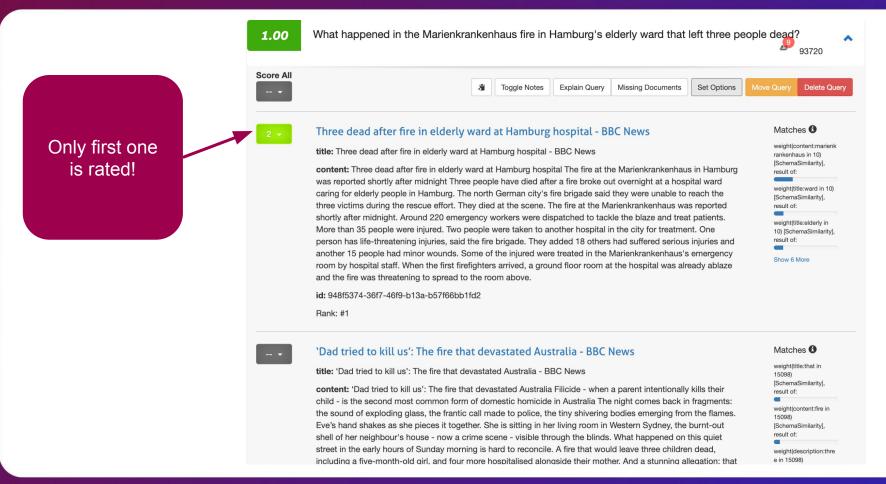
		91	92	93	
d1		91,d1	q2,d1	93,d1	
d2		91,d2	92,d2	93,d2	
		•••		•••	

<u>Recommendation</u>: disable this if not strictly needed, since it increases consistently the number of LLM calls!

Second step: retrieval with query template







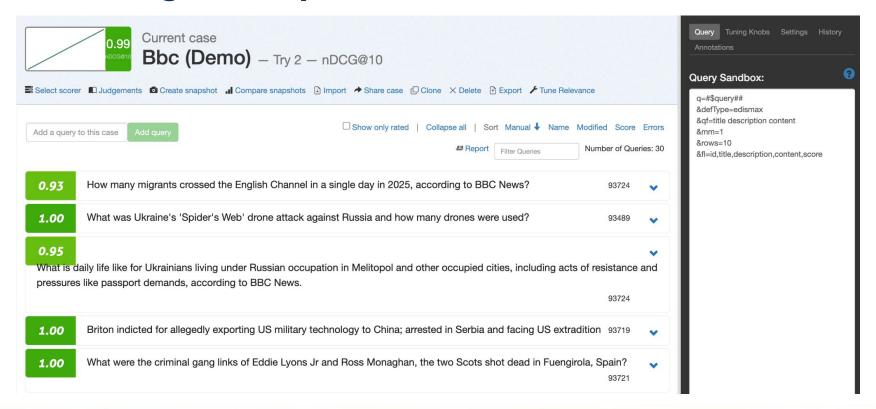
35

Why using a template is important

- Documents in input may be a subset of entire corpus
- Generated queries may retrieve un-rated documents
- You can run the queries and use the LLM to fill the gaps

In this way, you can also visualize your results with Quepid for example.

After using the template: all retrieved docs are rated!





Configuration file

```
query_template: "templates/template_solr.json"
                                                         <- be careful: this file must follow <a href="RRE guidelines">RRE guidelines</a>
search engine type: "solr"
                                                   <- support: "solr", "vespa", "opensearch", "elasticsearch"</pre>
collection name: "testcore"
search_engine_url: "http://localhost:8983/solr/"
                                                        <- instance MUST be up and running
documents_filter:
                                                         # if any
  - genre:
      - "horror"
number of docs: 100
doc_fields:
  - "title"
  - "description"
queries: "queries.txt"
                                                         # if anv
generate_queries_from_documents: true
                                                         <- if not set: generation enabled
num_queries_needed: 10
relevance scale: "graded"
                                                         \leftarrow graded ({0, 1, 2}), binary ({0, 1})
llm_configuration_file: "configs/dataset_generator/llm_config.yaml"
max_query_terms: 5
                                                         <- if not set: let LLM decide
output format: "quepid"
                                                         <- "rre", "quepid", "mteb"
output_destination: "resources"
save_llm_explanation: true
llm_explanation_destination: "resources/rating_explanation.json"
datastore autosave every n updates: 50
                                                        <- if not set: autosave disabled
enable_cartesian_product: false
                                                         <- if not set: cartesian product enabled
```

Query template file structure

```
"queries": [
    "template": "only_q.json",
    "placeholders": {
      "$query": "fender"
 "template": "filter_by_language.json",
 "placeholders": {
   "$query": "Fender",
   "$lang": "eng"
```

```
only q.json
  "q" : "$query"
filter by language.json
  "q" : "$query",
  "fq": "language:$lang"
```

Example of Ilm_config.yml file

```
# OpenAI LLM
name: openai  # supported: openai, gemini

# Chat model name
model: gpt-5-nano-2025-08-07

# Maximum number of tokens the model may return
max_tokens: 2000

# Environment variable where LLM API key is stored
api_key_env: OPENAI_API_KEY
```

How to run RRE Dataset Generator

```
Step 1: ensure that your search engine is up and running
Step 2: clone rated-ranking-evaluator repo
Step 3: cd rre-tools
Step 4: prepare configuration file
Step 5: uv run dataset_generator
That's it!
```



Results: explanation

Doc: Ukrainian forces outnumbered Russians by eight to one in last week's counter-attack in the Kharkiv region, Russia's top occupation official there says.

Vitaly Ganchev told Russian TV that Ukraine's army had taken villages in the north and broken through to the Russian border.

Ukraine says it has regained control over 3,000 sq km (1,158 sq miles) of territory in a potential breakthrough in the six-month war.

The BBC cannot verify these figures.

The Ukrainian army says it took back 20 villages in the past 24 hours alone, in its continued counter-offensive in the north-east of the country.

It also said its forces have taken control of around 500 sq km in the southern Kherson region of the country. UK defence officials say the Ukrainian army's successes will have "significant implications" for Russia's overall operational design.

Kremlin spokesman Dmitry Peskov was, however, undeterred, saying operations in Ukraine would continue "until all the tasks that were initially set" had been fulfilled.

Query: Life under occupation in Ukraine

Explanation: The article is war news about counter-attacks and territorial changes and includes a mention of an occupation official in the Kherson region, which signals an occupation context. However, it does not discuss daily life under occupation, civilian experiences, restrictions, or living conditions. Because it touches on occupation only superficially and not the core topic of "life under occupation," it is only marginally relevant (score 1) to the query.



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Next steps

- Documents cluster for query generation
- Use input queries as few shots for the LLM
- Use some existing ratings as few shots for the LLM
- New tools (Vector Search Doctor!)

Vector Search Doctor

1. Did I pick the right embedding model?

2. Did I pick the right vector search implementation ?



THANK YOU









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