

OpenSearch v3: A New Era of Search Innovation

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OpenSearch is the trusted open source platform for AI-powered search, observability, and analytics with built-in security, high performance, and a flexible architecture for modern applications.



OpenSearch Software Foundation



OpenSearch by the numbers

1.6B+

project downloads

1.25M

*monthly page views
For opensearch.org*

100+

solution providers

3K+

active contributors

400+

active organizations

29

*new releases since
project launch*

140+

GitHub repositories

4K+

*Slack workspace
members*

7K+

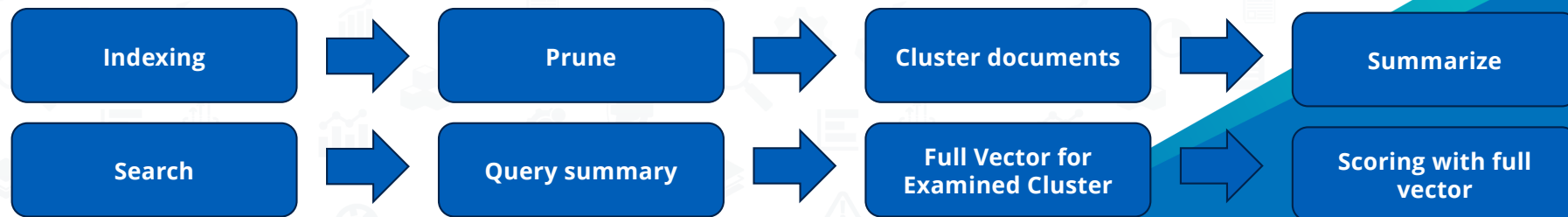
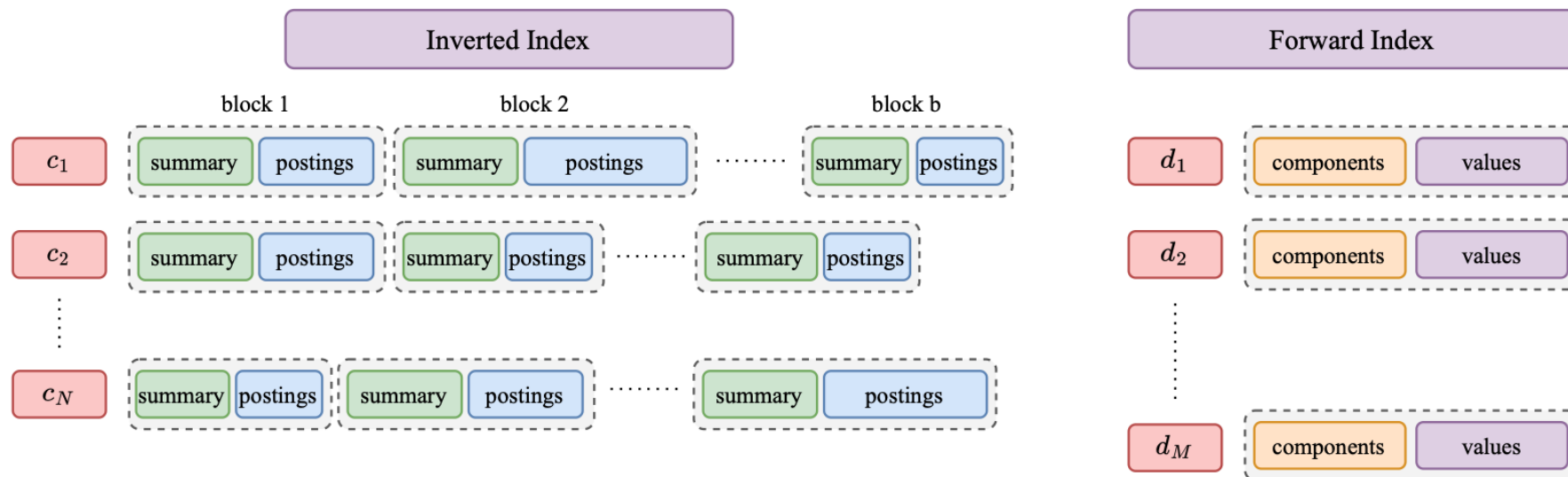
user forum members

Next major release: OpenSearch v3!

- ✅ Upgrade to **Apache Lucene v10** and JDK 24
- ✅ Pull-based ingestion
- ✅ Reader-Writer separation
- ✅ Native **MCP** support
- ✅ Expanded **PPL** queries, backed by **Apache Calcite**
- ✅ and much more...



Neural Sparse ANN – SEISMIC

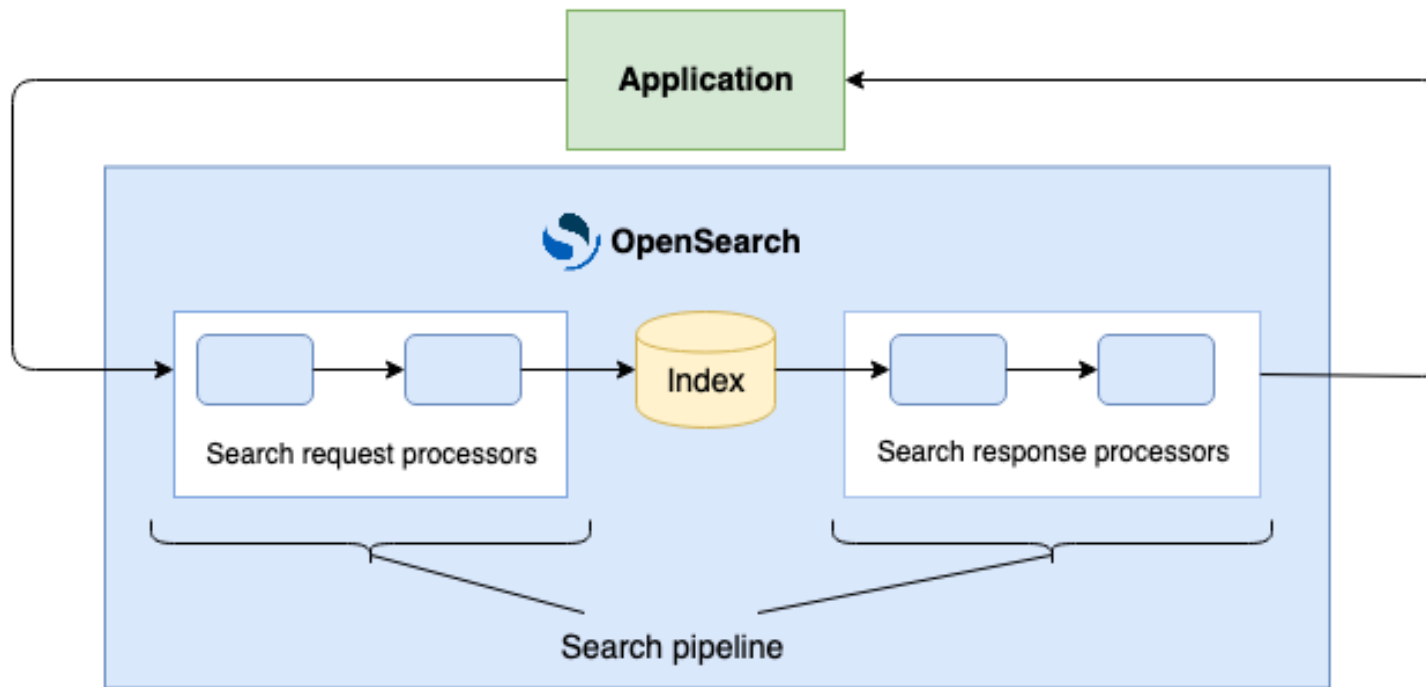


Meet OpenSearch's SOTA model

Search latency on billion docs (v3-gte, OpenSearch v3.3)

Metrics		Neural sparse	Neural sparse two phase	BM25	SEISMIC
Recall @ 10 (%)		100	90.483	N/A	90.209
Single-threaded	Average latency (ms)	125.12	45.62	41.52	11.77
	P50 latency (ms)	109	34	28	11
	P90 latency (ms)	226	100	90	16
	P99 latency (ms)	397.21	200.21	200.21	27
	P99.9 latency (ms)	551.15	296.53	346.06	50.02
Multithreaded	Mean throughput (op/s)	26.35	82.05	85.86	158.7

Sys generated Search Pipeline



AI Search flow

> Get started

Manage workflows New workflow

Create a workflow using a template

Import workflow

Search

Agentic Search

EXPERIMENTAL

Build a search application that leverages an agent to convert natural language to search queries.

Create

Custom Search

Build a custom workflow tailored to your specific use case without using a template.

Create

Hybrid Search

Build an application that searches using a combination of vector and lexical search.

Create

RAG with Hybrid Search

Build a search application that uses retrieval-augmented generation (RAG) to retrieve relevant documents using hybrid search, pass them to large language models, and synthesize answers.

Create

Multimodal Search

Build an application that searches both text and image data using multimodal embedding models.

Create

Semantic Search

Build an application that interprets the meaning and context of user queries to deliver more relevant and accurate search results.

Create

Semantic Search using Sparse Encoders

Build a flow that allows you to search by text and rank results by semantic similarity, to improve search quality. This template uses *Neural Sparse*, a sparse encoder, to convert text into sparse vectors. This implementation is potentially more cost efficient than the dense (k-NN) vectors for smaller indexes (< 10M documents).

Create

RAG with Vector Retrieval

Build a search application that uses retrieval-augmented generation (RAG) to retrieve semantically similar documents using vector search, pass them to large language models, and synthesize answers.

Create

demo-workflow

Flow overview

▼ Ingest flow

Sample data

Sample data ingested

↓

Transform data

ML Inference Processor

Amazon Bedrock - Titan Text Embedding

+ Add processor

↓

Index

knn_index_ftdbced

▼ Search flow

Sample query

↓

Transform query

ML Inference Processor

Amazon Bedrock - Titan Text Embedding

+ Add processor

↓

Run query

↓

Transform results

ML Inference Processor

Amazon Bedrock - Claude 3 Sonnet

+ Add processor

ML Inference Processor

Model

Amazon Bedrock - Titan Text Embedding

Inputs

Model input	Transformation type	Value
inputText	Data field	review

Outputs

Model output	Transformation type	New document field(s)
embedding	Data field	my_embedding

Advanced settings

Model Config

1 {}

☐ Full Response Path

☐ Ignore Missing

☐ Ignore Failure

Max Prediction Tasks

10

Tag

Description

Inspect

Test flow

Ingest response

Errors

Resources

Preview

Query and result transformations

Run test

Query

Query samples

Parameter	Type	Value
query_text	Text	Value

```
1- {
2-   "query": {
3-     "match": {
4-       "review": {
5-         "query": "{{query_text}}"
6-       }
7-     }
8-   }
9- }
```

Results

Use your sample query or write another one to test out your search flow

Search Relevancy Workbench

Search Relevance Workbench

Experiments

- Single Query Comparison
- Query Set Comparison
- Search Evaluation
- Hybrid Optimizer
- Query Sets
- Search Configurations
- Judgments

Experiments

Manage your existing experiments and create new ones. Click on a card to create an experiment.



Single Query Comparison

Test two search configurations with a single query. View side-by-side results to find the best performer.



Query Set Comparison

Perform a comparison across an entire set of queries. Determine differences across your complete use case.



Search Evaluation

Calculate search quality metrics to evaluate specific search configuration.



Hybrid Search Optimizer

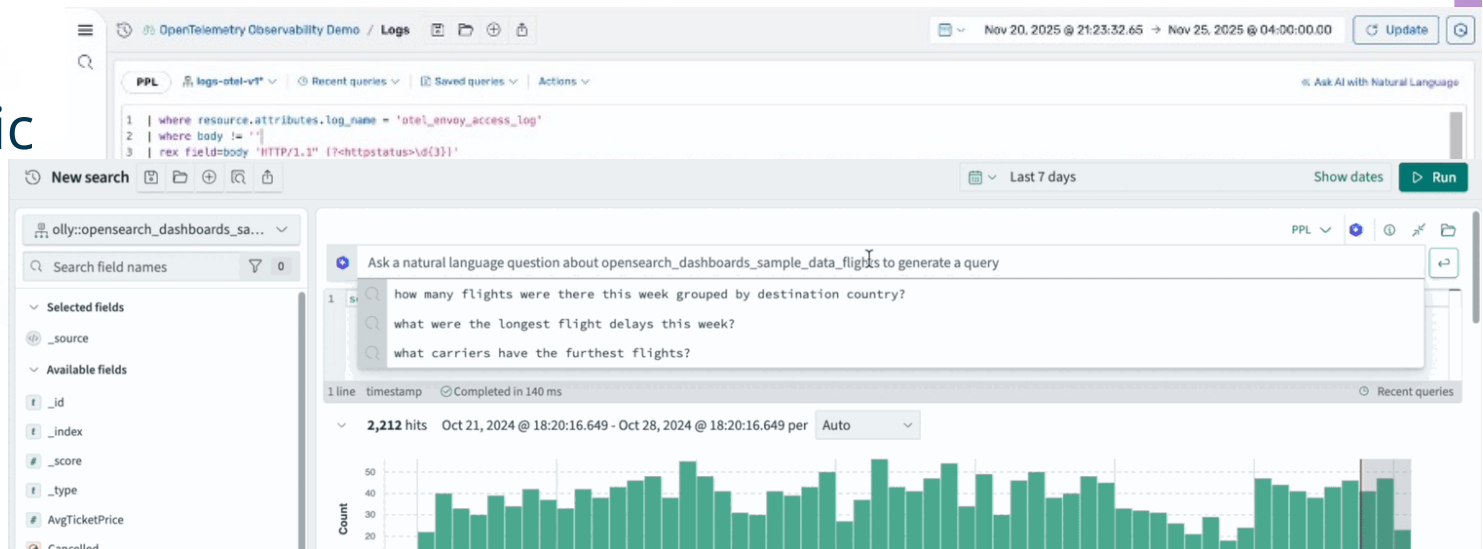
Find the best balance between neural and lexical hybrid search configuration.



Agentic Search

Demo

- PPL – Build Viz via Natural Language
- BQ – ADC & RR for improving recall
- Batch Inference Semantic Highlighting
- Late interaction
- Persistent Agentic Memory
- Radial Search, RRF, zscore, MMR



```

4 source=logs-otel-v1*
5 | where resource.attributes.log_name = 'otel_envoy_access_log' and body != ''
6 | rex field=body 'HTTP/1.1' (?<httpstatus>\d{3})'
7 | eval status_class = case(httpstatus >= '200' and httpstatus < '300', 'HTTP 2xx', httpstatus >= '300' and
8 | httpstatus < '400', 'HTTP 3xx', httpstatus >= '400' and httpstatus < '500', 'HTTP 4xx', httpstatus >= '500' and
  httpstatus < '600', 'HTTP 5xx', httpstatus >= '600', '0ther', true, 'Unknown')
  | stats count() as `Request Count` by `attributes.url.path`, status_class

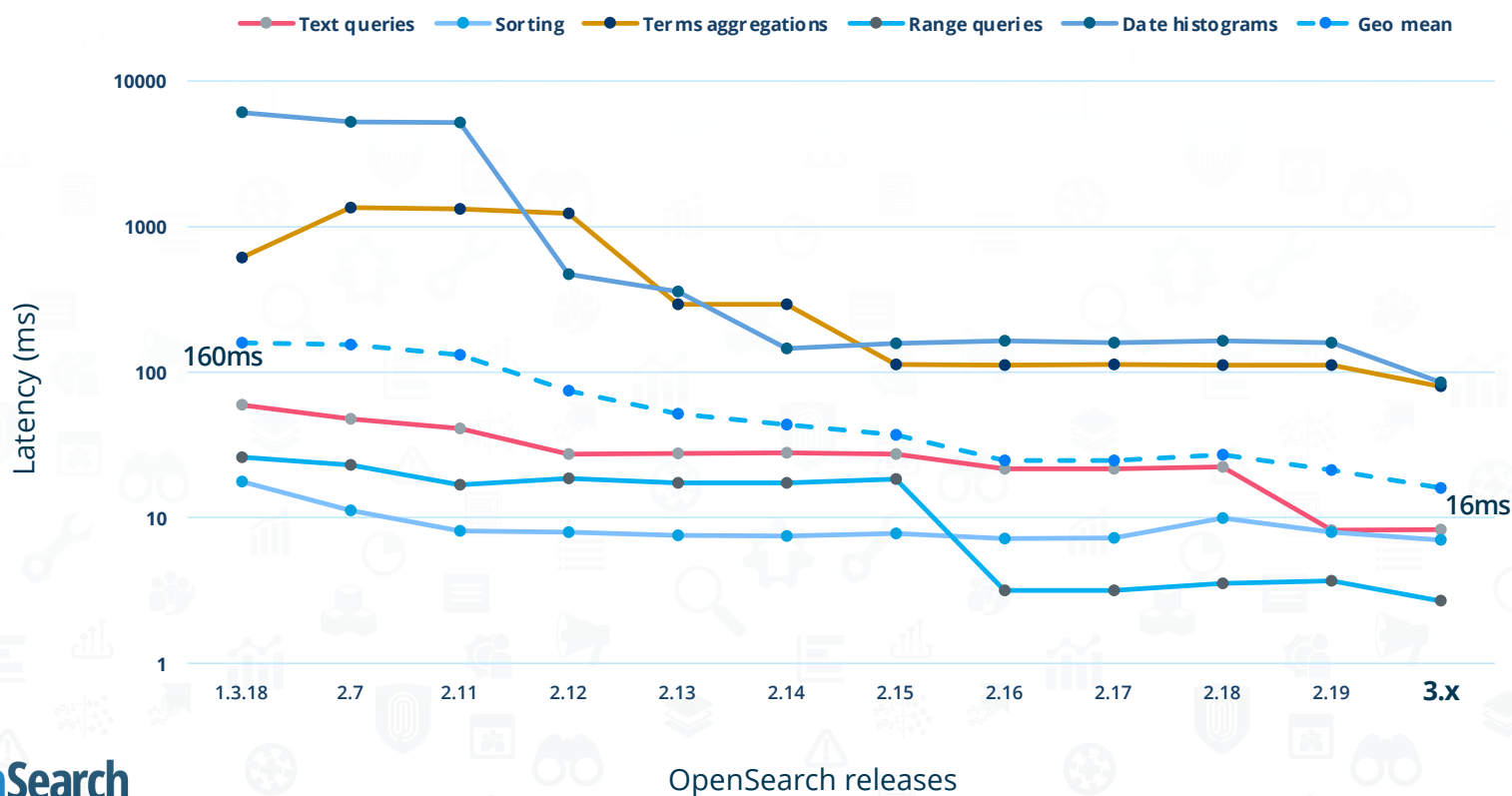
```

Performance improvements

OpenSearch v3

PERFORMANCE IMPROVEMENTS

"Big 5" areas - Latency (Log10 logarithmic scale)



Performance improvements

- **GPU acceleration:** 9.3x indexing speed, reducing costs by 3.75x
- **Lucene on FAISS:** nearly **doubles QPS** at 32x quantization
- **gRPC/protobuf transport:** ~50% reduction in client-side processing time, and ~20% higher throughput for vector search workloads
- **Apache Arrow support:** eliminate serialization overhead
- **Pull based ingestion** for Streaming service like Apache Kafka
- **Reader-Writer separation:** decouple indexing and search workloads for predictable performance
- **Derived Source:** 2x storage savings
- **Star-tree index** for complex aggregation over large set of data

OpenSearch 3.5 — coming in February

- Skip list support for aggregations
 - github.com/opensearch-project/OpenSearch/issues/19384
- New Application Performance Monitoring experience
 - github.com/opensearch-project/dashboards-observability/issues/2545
- Built-in agent observability
 - Experimental in 3.5, looking for feedback
 - github.com/opensearch-project/dashboards-traces/blob/main/GETTING_STARTED.md

Find us at



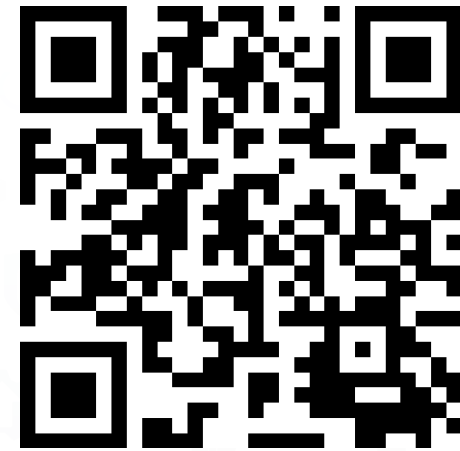
Github



Events



Blog



Slack



Forum



User groups

Thank you!

Dotan Horovits



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