### aws summit

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AIML003

# Innovations from Elasticsearch - Drive speed, scale & relevance

The latest innovations from Elasticsearch: How to implement recent developments in search to drive speed scale, and relevance

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#### Agenda

- Relevance innovations powered by vector search
  - The power of vector search
  - How to implement in Elastic
  - Best of both worlds with hybrid scoring
  - Improving search experiences with NLP and Personalization
  - Ingestion
- Performance improvements
- Elastic solutions & differentiators



#### Want to copy your famous influencer?





#### Visual matches



₩ lyst.com Boohoo Strappy Pleated Midi



a boohoo.com Blue Dresses I Royal, Light Blu.,



Eve Mauve See through blue



lyst.co.uk Mango Polka-dot Pleated Dress...



corporacion. düzenlemek Burger devirme..



boohoo.com Strappy Pleated Midi Skater Dress \*\*\*\* (1)

♥ lulus.com

Clothing Perfec...

Royal Blue



torrid.com Plus Size - Midi Chiffon Pleated..



ASOS Tall ASOS DESIGN Tall...

Ø €45.99



orolypolyappa. V Neck Pleated Midi Sundress -...



zalando.ie Mango FORTUNY - Day dress -...







zando.co.za Womens Strap Dress | Shop &...



ninterest.com **ASOS DESIGN** 





pleated cami...



S shein.com SHEIN Solid



отт getthattrend... Playful Promises











### Vector search lets you find what you mean, with higher relevance

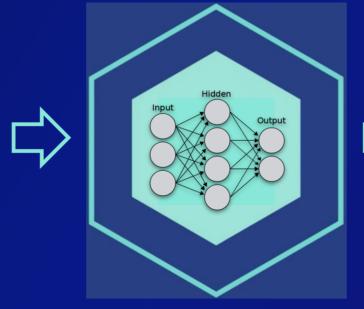
- Semantic search: understand meaning
- Multimedia: not just text, also on pictures or sounds
- Precision



#### Convert data into vector representation

UNDERLYING APPROACH OF VECTOR SEARCH

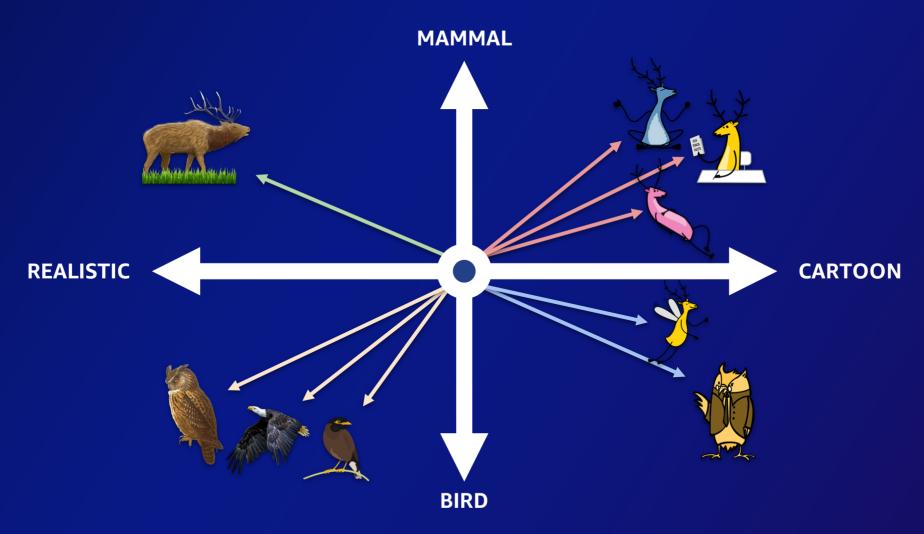
In order to stream from our service you will need a high quality connection. The required connection speed for using the service will vary depending on the quality of video and audio that you wish to stream to your device. For most customers we recommend at least...





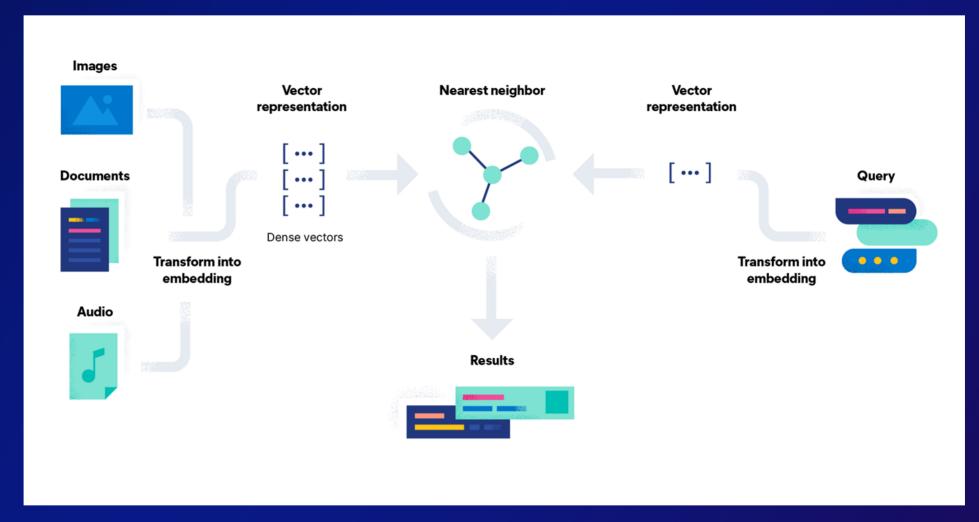


#### In vector space, similar data are nearby





### Vector search returns nearest neighbors of "vectorized" query



#### Vector search and NLP may be intimidating:



Lots of data, labelled

**Expertise** 

Scale processing



#### Elastic makes vector search easy







Just bring the data

Inherit scalability

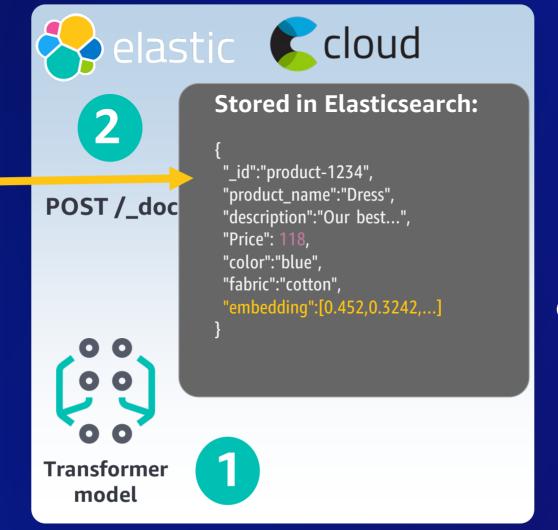
Flexibility



#### Apply vector search in 3 steps



Source data







clause

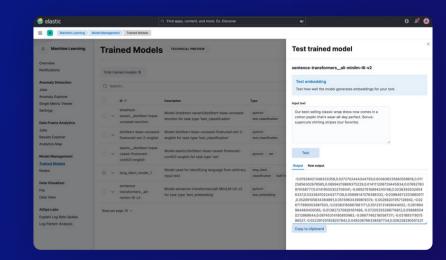


#### Step 1: Import pre-trained model



\$ eland\_import\_hub\_model

- --url https://cluster\_URL --hub-model-id BERT-MiniLM-L6
- --task-type text\_embedding --start







Load the model to the cluster





#### Step 2: Vectorize during data ingestion



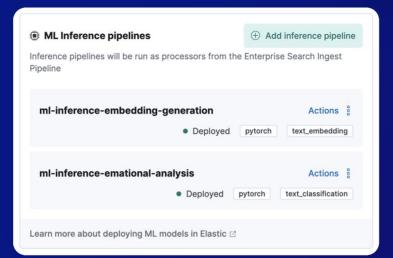
POST /\_doc

Standard field indexing for non-vector types

```
{
  "_id":"product-1234",
  "product_name":"Summer Dress",
  "description":"Our best-selling...",
  "Price": 118,
  "color":"blue",
  "fabric":"cotton",
} "desc_embedding":[0.452,0.3242,...]
}
```

POST /\_doc

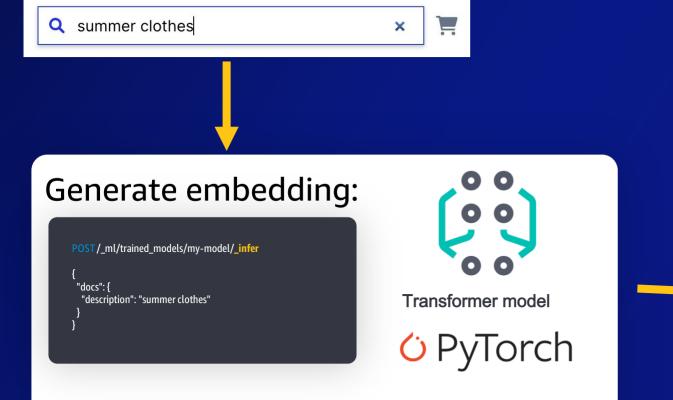
Encoding via Inference Processor





#### Step 3: Find nearest neighbor, approximately

Query is submitted to the search-powered application:

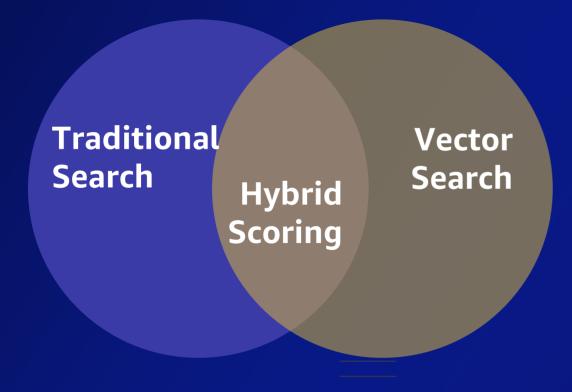


Issue knn query using the \_search endpoint

```
GET product-catalog/ search
 "knn": {
  "field": "desc_embbeding",
  "query_vector": [0.123, 0.244,...]
  "k": 5.
  "num candidates": 50.
  "boost": 0.1.
  "filter": {
   "term": {
     "department": "women"
```



### Hybrid scoring gets you the best of both worlds



```
GET product-catalog/ search
 "query": {
  "match": {
   "description": {
    "query": "summer clothes",
    "boost": 0.9
 "knn": {
  "field": "desc_embbeding",
  "query_vector": [0.123, 0.244,...],
  "num candidates": 50.
  "boost": 0.1.
  "filter": {
   "term": {
     "department": "women"
```

#### How will you apply vector search?



**Product similarity search** 

"Do you sell black v-neck shirts that look like this?"





**Answer technical support** 

"What are the troubleshooting steps for \_\_\_?"



Query medical knowledge

"Is lithium used to treat bipolar disorder?"

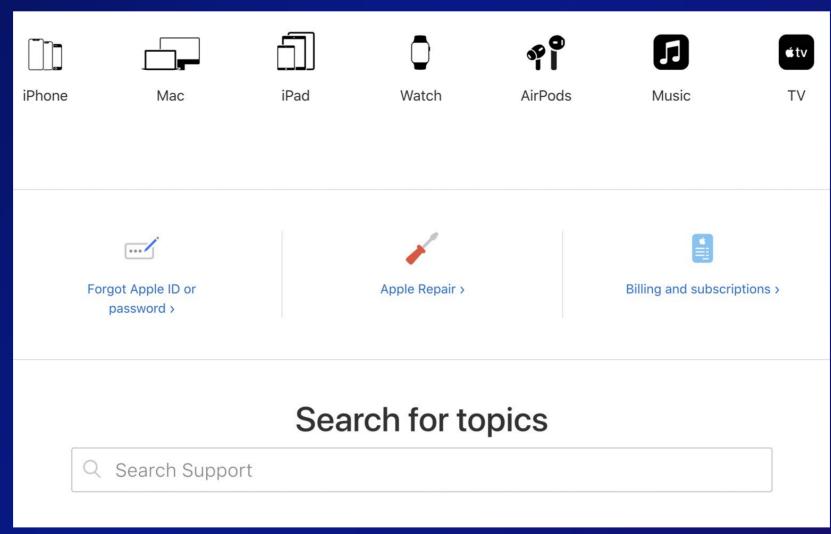


**React to user sentiments** 

Identify poor customer interactions before they lead to escalations

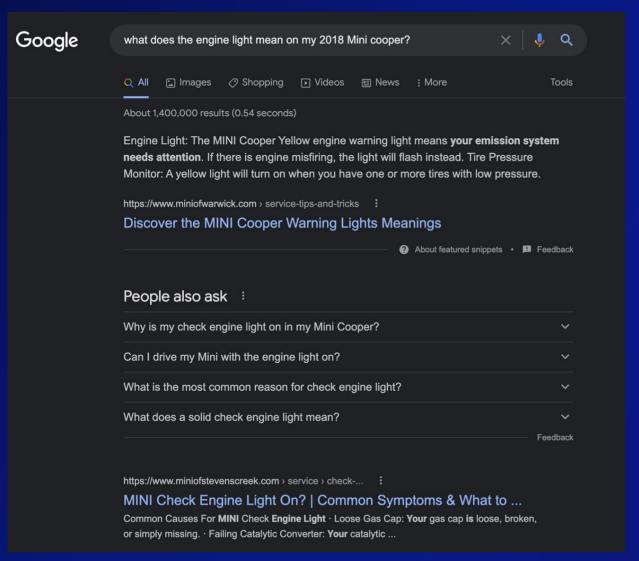


#### NLP #1: No more browsing manuals / FAQs





#### Query with question-answering model



```
POST _ml/trained_models/deepset__minilm-uncased-squad2/_infer
{
    "docs":[{ "text_field": "My name is Peter and I live in London"}],
    "inference_config": {
        "question_answering": {
        "question": "Where do I live?"
      }
    }
}
```

```
{
    "inference_results":[
        {
             "predicted_value": "London",
             "start_offset": 31,
            "end_offset": 37,
            "prediction_probability": 0.9948918266325819
        }
    ]
}
```

#### NLP #2: Prevent customer service escalations

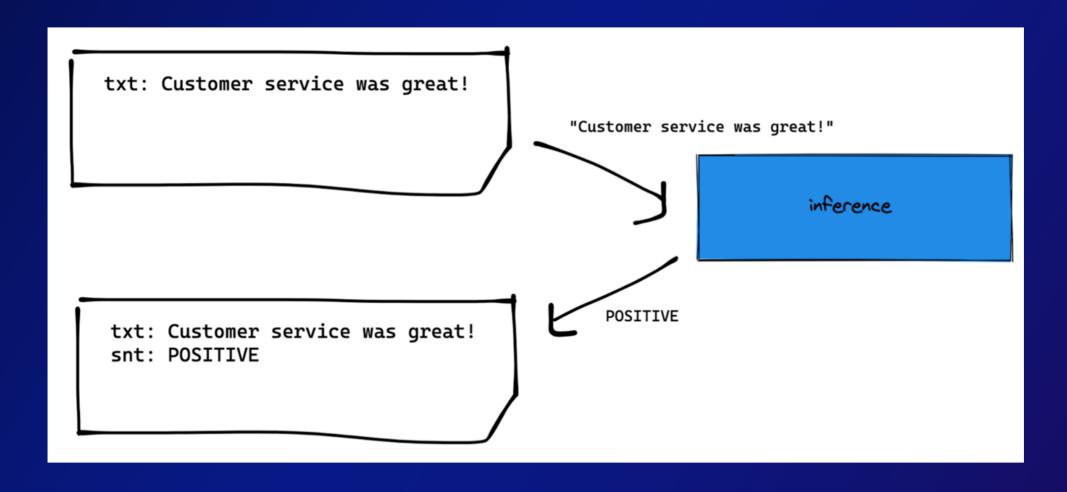
Possibly one of the worst service experiences at We checked in at 5pm and were seated for about 20 minutes and no waiter. Not busy at all. We saw a table being seated 10 minutes after us that was served right away. We had to go back to host desk to ask for our waiters name and host was flustered at best.

I received a text from Open Table saying I missed my reservation almost an hour after we checked in. On my way out I asked them to check their system and note that they would have a table ready for me shortly. I said no just check m showed them my receipt.



#### Sentiment models classify feedback

MONITOR YOUR BRAND IN SOCIAL MEDIA, GET NOTIFIED AFTER FEEDBACK!





### Personalization drives digital commerce outcomes

**Online shoppers today** 

88%

84%

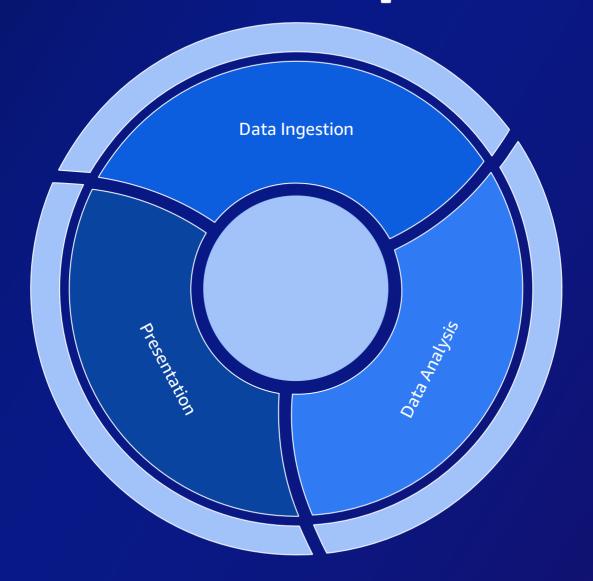
68%

are more likely to continue shopping on websites that offer a personalized experience report personalization already influences their shopping decisions

have purchased items they did not intend to initially, due to personalized recommendations



#### Personalization is a \*data\* problem





## Elastic capabilities enable all stages of a personalization implementation

**Data Ingestion** 



- Ongoing usage metrics with Search Analytics
- Prebuilt integrations with external data sources
- Custom integrations with use case specific data

**Data Analysis** 



- Visual analytics with Kibana dashboards
- Broad and flexible set of analytics capabilities
- Speed, scale, relevance

Presentation



- Built in tooling to manage search responses
- APIs to refine responses at runtime
- Accelerate development with Search UI



#### Personalization journey with elastic

Add/test with the latest in search

Extend data & analysis over time

Get started with ease

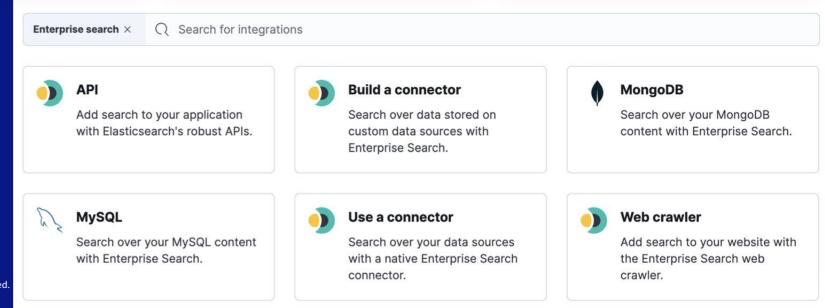
Machine learning, vector search powered recommendations Additional data sources outside Elastic, and build holistic user insights Built in usage metrics tracking and Analytics API, with a client experience built on Search UI



#### Crawl or connect searchable data

Accelerate building search experiences with ingestion flexibility and scalability

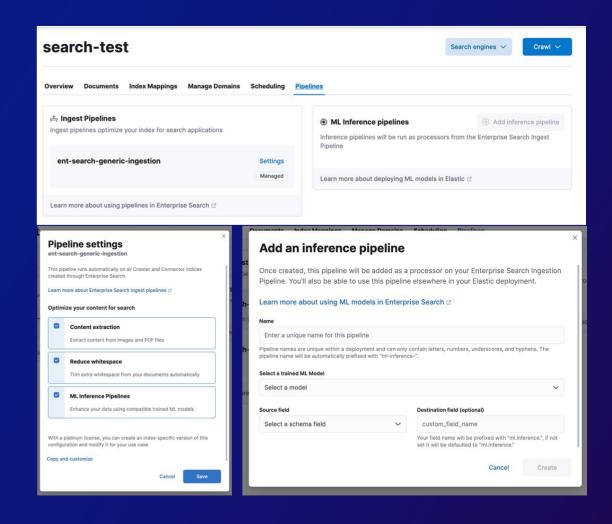
- Native web crawler with PDF extraction and simple authenticated crawls
- Native connector clients for MongoDB, Gitlab, and MySQL
- Open code connector clients in Ruby and Python
- New connectors-py framework





#### Pipeline management for ML models

- Choose what to ingest, how to transform ingested data, and ML models needed
- Managed, custom, and inference pipelines
- Tune relevance at ingest time





#### Additional features to explore

- Named entity recognition
- Zero-shot classification
- Applying NLP and vector search to Observability or Security



#### **Performance enhancements**



#### **Storage savings**

7.14 match only text Save up to 10%

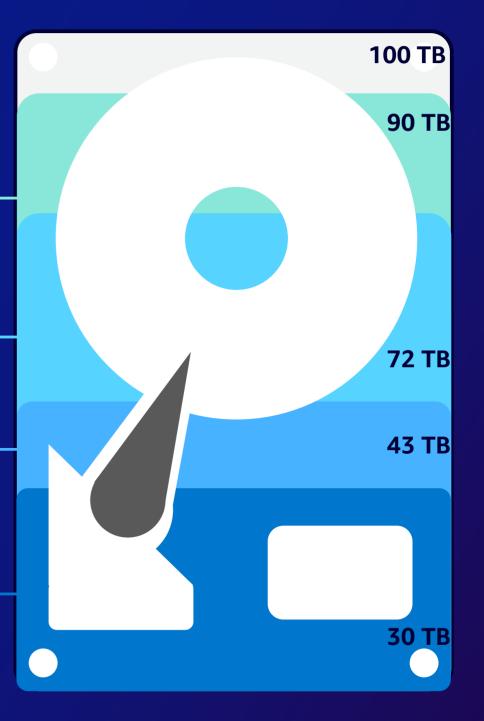
8.1 doc-value-only fields Save up to 20%

8.4 synthetic \_source Save up to 40%

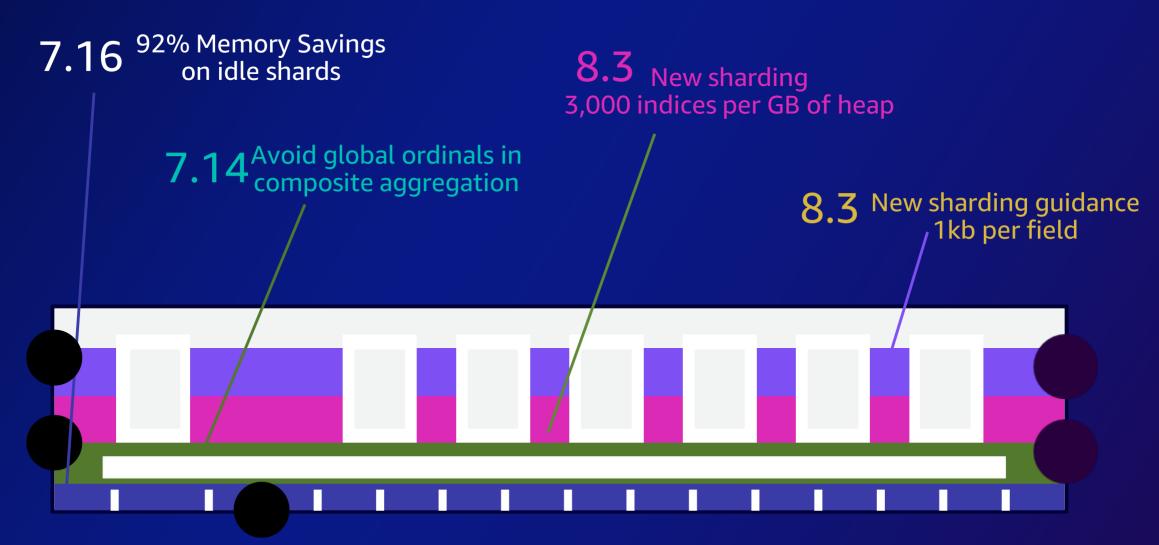
8.5 routing and sorting by TSID Save up to 30%

Version wise detailed information on enhancements is available here - https://www.elastic.co/guide/en/elasticsearch/reference/current/es-release-notes.html



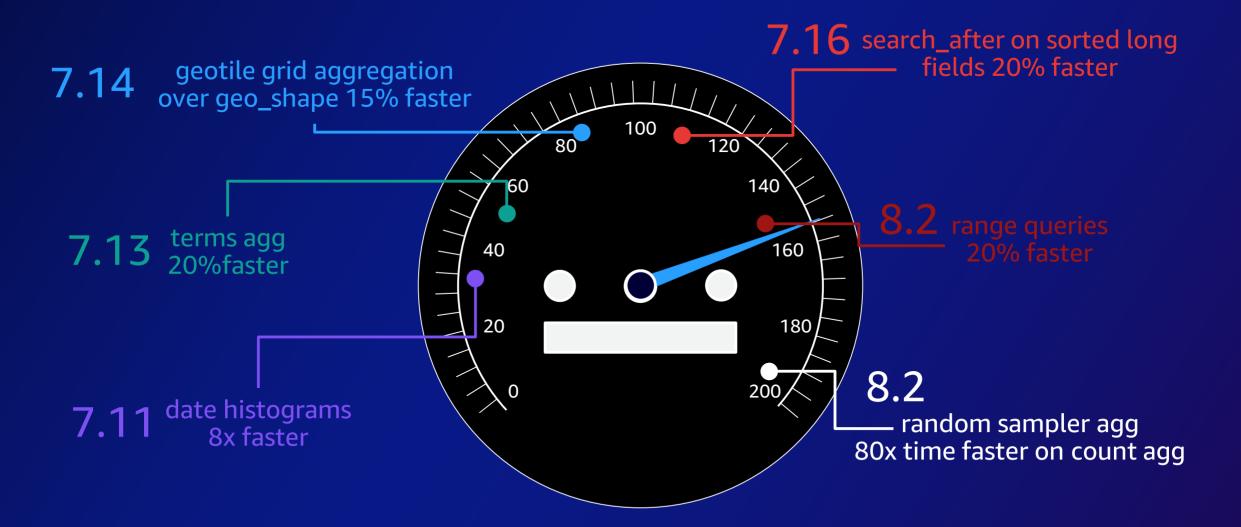


#### **Memory savings**





#### **Performance**





#### **Search 1PB in minutes stored on S3**





#### Intelligently store and search everything

**Data Usage** 

**Performance** 

Accessed Frequently

Accessed Less Frequently

Accessed Less Frequently

Accessed Intermittently

Accessed Rarely



milliseconds -



seconds - 10's seconds



seconds - 10's seconds



10's seconds - minutes

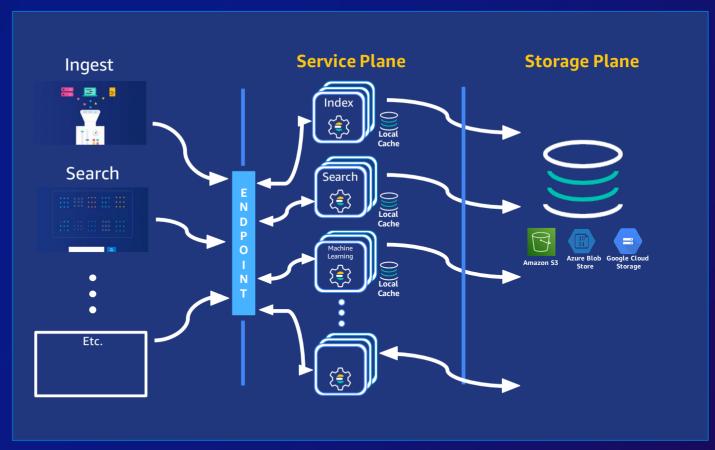


minutes -10's minutes



#### Serverless Elasticsearch

- Next generation managed service
- Scale compute & storage independently
- Consume only what you need
- Versionless
- Autoscaling index and search





#### **Elastic differentiators**

Elastic Cloud ES feature(s)	Customer benefit
Enterprise Search solution	Provides ready-to-use products and integrations for app search and workplace search
Observability solution	Provides ready-to-use log monitoring, infrastructure monitoring, APM, distributed tracing, all within the same UI; single pane of glass
Security solution	Provides ready-to-use SIEM with detection engine, rules, endpoint security, XDR, etc
Machine learning	Save time and tool bloat by using Elasticsearch to store, transform, build, test, and deploy machine learning models natively. Anomaly detection and supervised ML for sentiment analysis.
Elastic Agent and Fleet	Delivers a single unified agent with Fleet management capabilities, enabling automations for observability and security at scale from a single UI
Kibana Maps	Allows for easier analysis of geospatial data
Kibana actions	Integrates with several popular third-party systems
Breadth of integrations	Hundreds of Cloud and other native integrations



### **Community resources**

Elastic Contributor Program Elastic User Groups

**Elastic for Students and Educators** 



discuss.elastic.co

Community YouTube
Channel

Elastic Stack Community
Slack Workspace



### Thank you!

Ravindra Ramnani Principal Solutions Architect, Elastic



Please complete the session survey

