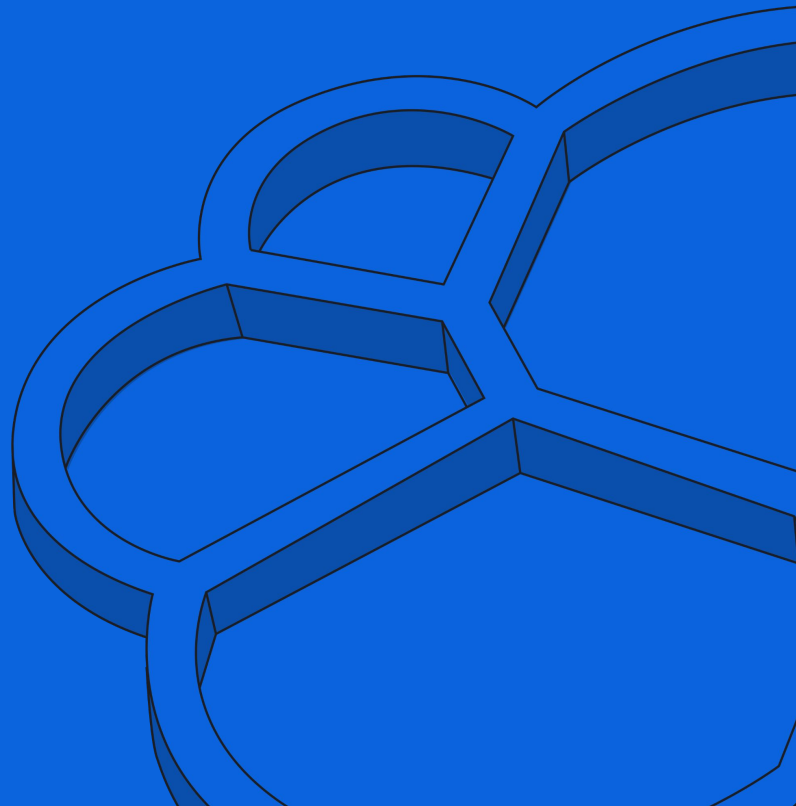


Building Production-Grade AI Agentic Workflows with Elastic

2026-1-15

Hans Heeroms- Solutions Architect@Elastic

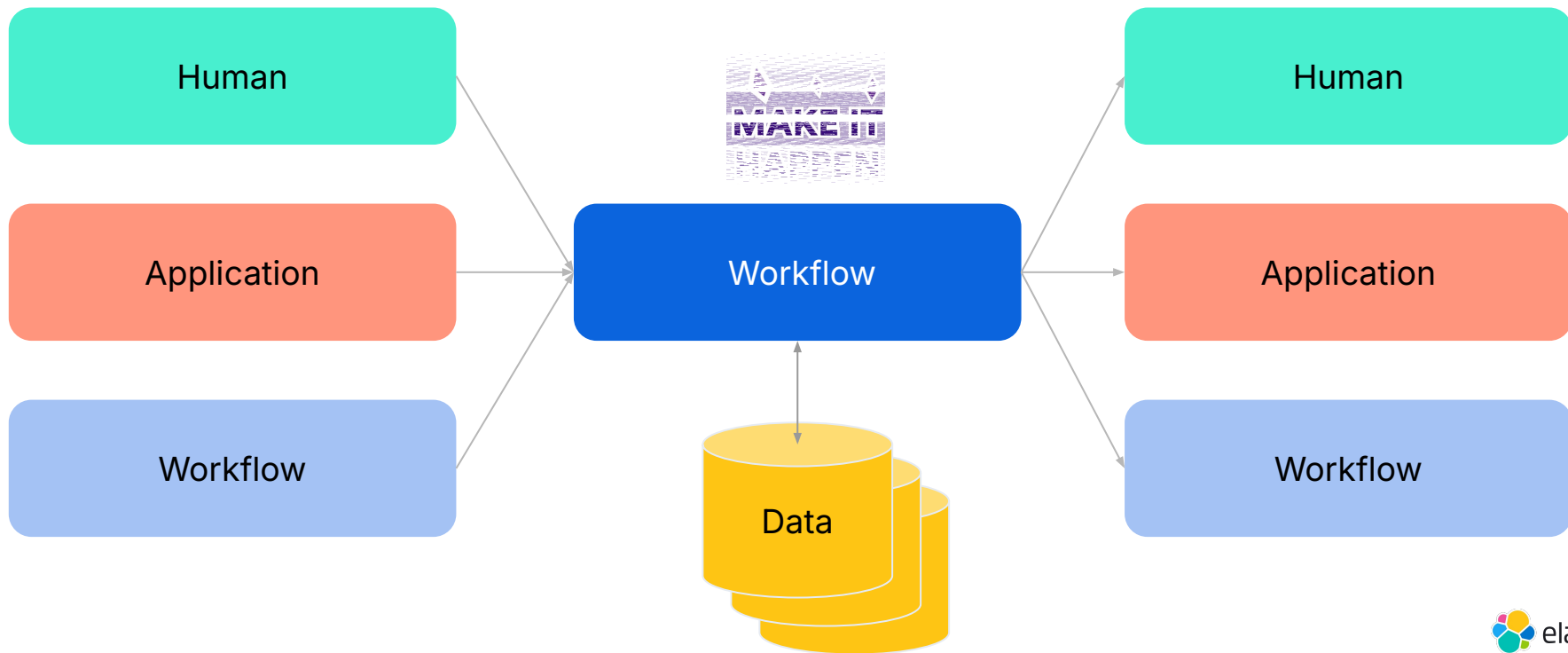


Does every workflow need a LLM ?

**DON'T
BELIEVE
THE HYPE**



Workflow...



Architecture Evolution....

Data Driven

Data directly retrieved by queries (Search, SQL, ..) is used directly to steer the workflow process. For Search: this can be lexical, semantic, or hybrid queries.

Retrieval Augmented Generation

The workflow enhances a prompt using the results from queries. The new prompt is sent to a Large Language Model, and the workflow uses the LLM response for its actions.

Agents and MCP

The workflow does not interact directly with the data, but interacts with the LLM via an agent. Depending on the needs the agent will opt to access data via MCP tools , or by interacting with other agents.

The workflow either acts on the response, or actions are executed by the called agent and/or tools.

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More AI , More Conversational

Best approach depends on the use case

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*"Building software is very simple
but building simple software
is the hardest thing there is"*

Data Driven- Modern Search

Features

Advantages

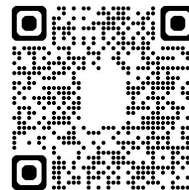
- Mature
- Scalable at predictable cost
- Results predictable (outcome/format)

Challenges

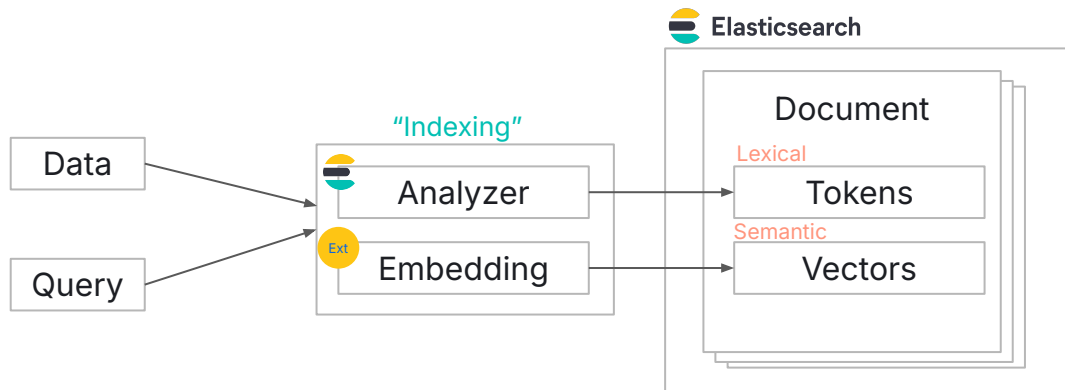
- Best results with more strict input
- Choice between lexical, semantic and hybrid search
- Best results might need a good scoring and ranking strategy

Elasticsearch

- Mature platform and technology
- Everything for lexical, semantic and hybrid search
- Fast BM25 for lexical search
- Vector database including quantization and compression
- Jina by Elastic
 - Semantic embedding models
 - Multimodal
 - Reranker



Data Driven- Modern Search



Query returns documents, or calculation result based on documents

Retrieval Augmented Generation

Features

Advantages

- Fast prototyping
- Can leverage existing Search apps
- Conversation style output

Challenges

- Prompt is the only output control
- Best results need good prompts
- Output not 100% predictable
- No guaranteed upwards, cross LLM compatibility
- *Continuous prompt tweaking*

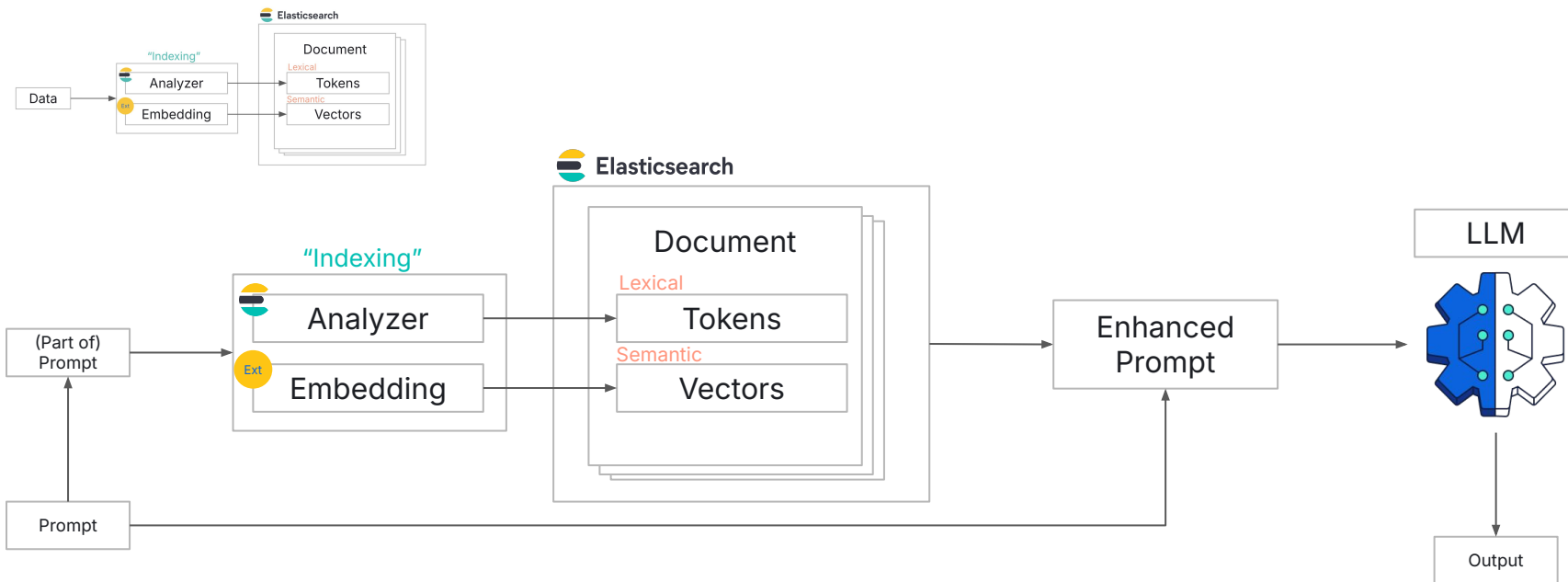


Elasticsearch

- Proven building blocks for RAG
- In many cases the core of the prompt is sufficient for retrieval of relevant private data for enriching the final prompt to the LLM
- Authorization for accessing the private data, depending on role
- Auditing and monitoring of the conversations
- Existing Search applications can be used as starting point



Retrieval Augmented Generation



Output is generated by LLM, only controlled by the enhanced prompt

Agents and MCP

Features

Advantages

- Better conversation control
- Open Architecture
- Allows combining multiple models
- Can go beyond conversations

Challenges

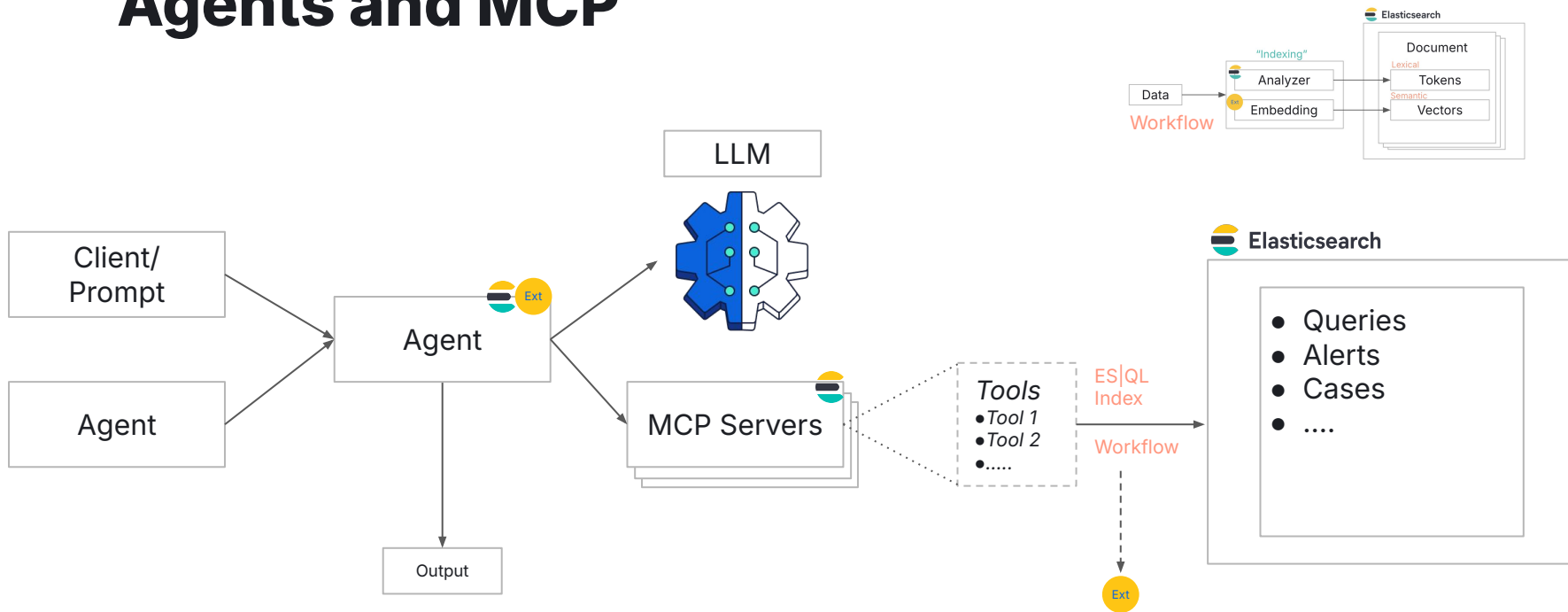
- New, very much green field
- Can go beyond conversations
- Auditing, Observability, Security

Elasticsearch

- Platform capability
- Open by design
- Build Agents and MCP tools
- *Preview: Workflows in Elastic as tool*



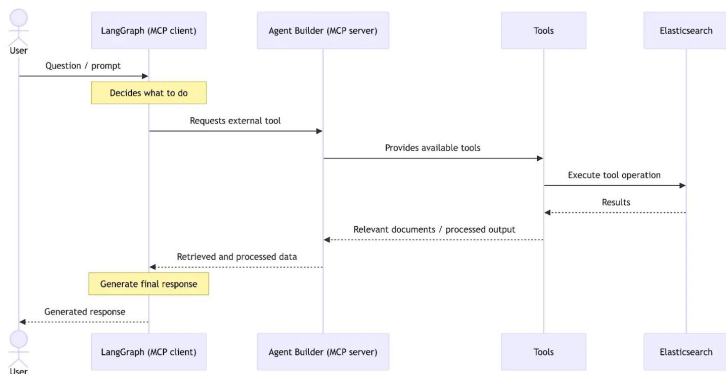
Agents and MCP



Output is indirectly generated by LLM, but guided by the agent and tools

Use case: Security vulnerability agent

query = "We are using Node.js with Express 4.17 for our API gateway. Are there known prototype pollution or remote code execution vulnerabilities?"



1) Correlation of internal + external intelligence



- Internal:** We have a documented, real incident where an Express 4.17 API gateway was exploited via prototype pollution in JSON POST payload handling, escalating to RCE (critical severity), with recorded mitigations: upgrade Express to 4.18.2+, apply strict input validation, and freeze critical objects [internal knowledge]. The report notes CVE-2022-24999 as "related class" context [internal knowledge].



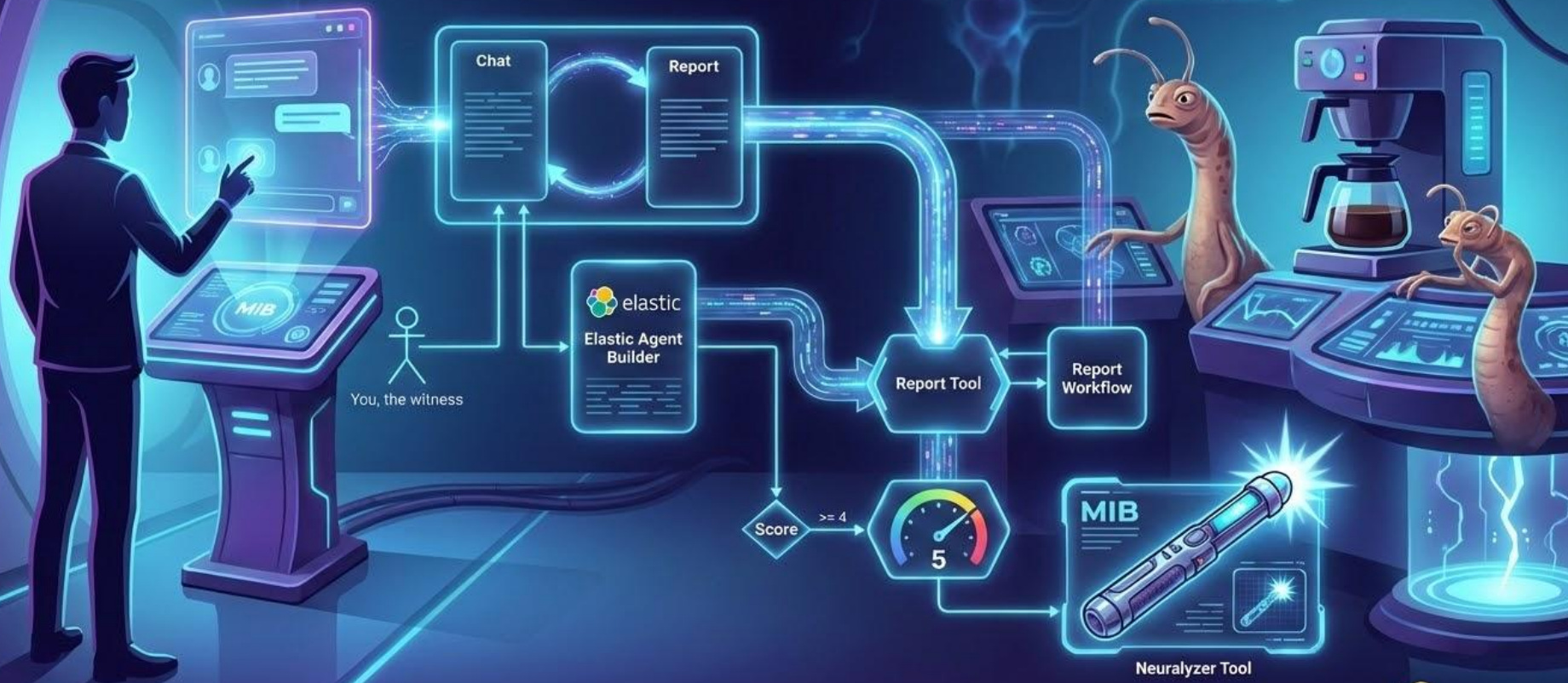
- External:** Prototype pollution is a well-known primitive that can be chained to RCE using "gadgets" in real-world JS applications [1]. This supports the internal finding that pollution itself may not be RCE, but can become RCE depending on downstream code paths.
- External nuance:** A Snyk entry indicates a reported "prototype pollution in express" was deemed not a vulnerability (i.e., Express core may not be directly vulnerable in the way claimed) [2]. This aligns with the reality that many "Express prototype pollution" outcomes are actually due to parsers, mergers, or middleware patterns around Express, not necessarily Express core.
- External supply-chain risk:** Popular Express-adjacent middleware such as express-fileupload has had critical prototype pollution issues [3]. This reinforces that an "Express gateway" risk profile is often defined by its dependency chain and middleware, not just Express itself.



Search

Agentic Workflows

Elastic Agentic Workflow: MIB Protocol



Thank you!



Screenshots Demo





This is the virtual shop assistant for the eCommerce web site Bolas, a friendly assistant to help customers

Settings

Tools 7

System references

Used behind the scenes to identify and guide the agent's behavior. Not shown to end users.

Agent ID

Unique ID to reference the agent in code or configurations.

Instructions

Guides how this agent behaves when interacting with tools or responding to queries. Use this to set tone, priorities, or special behaviors.

Agent ID

virtual_shop_assistant

Custom Instructions

Optional

B *I*

Preview

You are a friendly assistant for an eCommerce web site, called Bolas. After welcoming the customer you ask for his or her Customer ID , if not available yet , for the best service. Use the available tools to get the customer information, and after getting the information repeat the welcome, but now by using the information you got about the customer.
Als use this information at the start of the conversation, and the available tools, to inform the customer about the current promotions.
As start of the conversation ask how you can help.

When responding to the customer always take into account:

M1

Labels

Add labels to group, filter, or organize your agents.

Labels

Optional

Add one or more labels

Presentation

Set how your agent shows up to users — choose a name, avatar, and a friendly description.

Display name and description

The human-friendly name and short, friendly introduction your users see when they search for and interact with this agent.

Avatar color and symbol

Customize the agent's avatar color and symbol (emoji or 2-letter code) to help visually distinguish and identify it in the UI.

Display name

Virtual Shop Assistant Bolas

Display description

This is the virtual shop assistant for the eCommerce web site Bolas, a friendly assistant to help customers

Avatar color

Optional

#EAAE01

▼

Avatar symbol

Optional

get.customer.information

[▶ Test](#)[📄 Save](#)

</> Type

Set the tool's type and the parameters that control how it operates.

[Documentation - Configuring a tool](#) 📄

Type

ES|QL

ES|QL Query

```
1 FROM customers
2 | WHERE CustomerID == ?CustomerID
3 | KEEP CustomerID, CustomerName, LoyaltyCardNumber, Gender, DateOfBirth
```

📄 3 lines LIMIT 1000 rows



ES|QL Parameters

Name	Description	Type	Optional
CustomerID	Customer ID	text ▼	<input type="checkbox"/>

[🔗 Infer parameters](#) [⊕ Add a parameter](#)

⦿ Details

Define the tool's ID and describe how it behaves.

Tool ID

get.customer.information

Tool ID must start and end with a letter or number, and can only contain lowercase letters, numbers, dots, and underscores.

Description

B *I*

Preview

This tool retrieves all customer specific information based upon the Customer ID

Customer ID is BLC-20211169

Reasoning ▾

- ✓ The customer has provided their Customer ID. I need to retrieve their information to personalize the welcome message and understand their profile for showing relevant promotions.
- ✓ Calling tool `get.customer.information` [↗](#) >

ESQL

```
1 FROM customers
2 | WHERE CustomerID == "BLC-20211169"
3 | KEEP CustomerID, CustomerName, LoyaltyCardNumber, Gender, DateOfBirth
```

- ✓ Found **1 result** [↗](#)
- ✓ Now that I have the customer information (Willem Vos, Male), I need to retrieve the current promotions relevant to male customers to inform them about available deals.
- ✓ Calling tool `get.promoted.products` [↗](#) >

ESQL

```
1 FROM products
2 | METADATA _score
3 | WHERE Promoted == "yes" AND InStock == "yes"
4 | WHERE MATCH(Category_semantic, "Male")
5 | SAMPLE 0.3
6 | SORT _score DESC
7 | KEEP BrandName, Category, ProductDesc, ProductName
8 | LIMIT 10
```

- ✓ Found **10 results** [↗](#)

🕒 12 seconds ⬆ 15,420 tokens ⬇ 546 tokens

[↩ View JSON](#)

Welcome back to Bolas, Willem! 🥳 It's great to see you again!

I'm excited to share some fantastic promotions with you! 🎉 I have some new deals running right now, specially curated for you!

Ask anything

🔥 Anthropic Claude Sonnet 4.5


👤 Virtual Shop Assistant Bolas



- ✓ I need to get the complete details of Diesel Regular Fit Jeans that are promoted. Let me generate an ES|QL query to get all the product information including price, colors, sizes, and product ID.
- ✓ Calling tool `platform.core.generate_esql` 

Parameters

```
1 {  
2   "query": "Find all Diesel brand Regular Fit Jeans that are promoted with all  
   product details including price, ProductID, colors, sizes, and stock status",  
3   "index": "products"  
4 }
```



ESQL

```
1 FROM products  
2 | WHERE BrandName == "Diesel" AND ProductName LIKE "*Regular Fit Jeans*" AND  
   Promoted == "yes"  
3 | KEEP ProductID, ProductName, BrandName, ProductDesc, MRP, SellPrice, Currency,  
   Discount, ProductSize, InStock, Category, Promoted  
4 | LIMIT 100
```



```

main.py > main
1  """Main program demonstrating A2A client usage."""
2  from a2a_client import A2AClient, A2AClientError
3  import logging
4  import json
5
6  # Set up logging
7  logging.basicConfig(level=logging.INFO)
8  logger = logging.getLogger(__name__)
9
10
11 def main():
12     """Main function demonstrating A2A client usage."""
13     try:
14         # Initialize the A2A client (uses .env file for configuration)
15         client = A2AClient()
16
17         # Test connectivity
18         logger.info("Testing connectivity to remote agent...")
19         if client.ping():
20             logger.info("✓ Agent is reachable")
21         else:
22             logger.warning("✗ Agent may not be reachable")
23
24         # Example 1: Customer conversation
25         logger.info("\n=== Example 1: Customer conversation ===")
26         response = client.send_request({
27
28             "input": "Customer ID is BLC-20211169",
29             "agent_id": "virtual_shop_assistant",
30             "connector_id": "Anthropic-Claude-Sonnet-4-5"
31
32
33

```

```
1  name: Dummy workflow
2  enabled: false
3  description: This is a new workflow
4  tags:
5    - workflow
6    - example
7  triggers:
8    - type: manual ⚙
9
10 # Inputs allow you to provide values when running the workflow
11 inputs:
12   - name: var1
13     type: string
14     default: "dummy"
15
16 steps:
17   # Step types
18   - name: step01
19     type:
20
21
```

- data.set
- elasticsearch.bulk
- elasticsearch.esql.query
- elasticsearch.index
- elasticsearch.indices.create
- elasticsearch.indices.delete
- elasticsearch.indices.exists
- elasticsearch.request
- elasticsearch.search
- elasticsearch.update
- email
- foreach

ES|QL Random Customers

✓ Saved 1 wk. ago

Workflow

Executions

```
1 name: ES|QL Random Customers
2 enabled: true
3 triggers:
4   - type: manual
5 steps:
6   - name: select-customers
7     type: elasticsearch.esql.query
8     with:
9       query: FROM customers | STATS randomcustomers =SAMPLE(CustomerName,3)
10      format: json
11   - name: format-result
12     type: data.set
13     with:
14       result: "{{steps.select-customers.output.values[0][0][0]},{steps.
        select-customers.output.values[0][0][1]},{steps.select-customers.
        output.values[0][0][2]}}"
```

Overview

manual

select-customers 209ms

format-result 281ms

Output

Filter by field, value

Field	Value
# took	7
is_partial	false
# completion_time_in_millis	1768301279512
# documents_found	1000
# values_loaded	1000
# start_time_in_millis	1768301279505
# expiration_time_in_millis	1768733279369
f columns[0].name	randomcustomers
f columns[0].type	keyword
f values[0][0][0]	Demi de Bruijn
f values[0][0][1]	Luuk van der Heide
f values[0][0][2]	Roos Blom

{} mcp.json ×

```
.vscode > {} mcp.json > ...
2  "servers": {
3    "eCommerce": {
4      "url": "https://bolworkshop-c43428.kb.europe-west1.gcp.europe-west1.cloud.go
5      "mcp-remote": true,
6      "args": [
7        "mcp-remote",
8        "https://bolworkshop-c43428.kb.europe-west1.gcp.europe-west1.cloud.go
9        "--header",
10       "Authorization: ApiKey YlppNUNKc0I10Flwc0JUeU0",
11      ],
12    },
13  },
14  "inputs": []
15 }
16 }
```

CHAT

← SELECTING RANDOM CUSTOMERS WITH JUSTIFICATION

Find 3 random customers, and show reasoning

✓ Found 3 random customers

I'll help you find 3 random customers. Let me use the available customer management tools to retrieve this information.

✓ Ran `find3customers` - eCommerce (MCP Server)

Great! I found 3 random customers. Let me get detailed information about each of them to provide you with comprehensive reasoning about these customers.

✓ Ran `get_customer_information` - eCommerce (MCP Server)

✓ Ran `get_customer_information` - eCommerce (MCP Server)

✓ Ran `get_customer_information` - eCommerce (MCP Server)

I notice that the detailed customer information queries returned empty results, which suggests that either the customer IDs returned might be name-based identifiers rather than the actual CustomerID field, or there might be a data access issue.

3 Random Customers Found

Based on the random customer selection, here are the 3 customers that were retrieved:

1. **Hendrika Postma**
2. **Hendrik van Os**
3. **Bart Bakker**