





# Implementing your own ChatGPT : developing a RAG solution with IBM Watsonx.ai

Jarlath Quinn & John McConnell







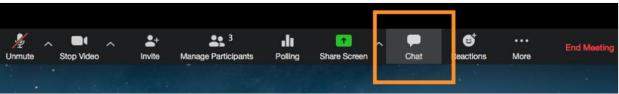
Just waiting for all attendees to join...

# Implementing your own ChatGPT : developing a RAG solution with IBM Watsonx.ai

Jarlath Quinn & John McConnell

## FAQ's

- Is this session being recorded? Yes
- Can I get a copy of the slides? Yes, we'll email links to download materials after the session has ended.
- Can we arrange a re-run for colleagues? Yes, just ask us.
- How can I ask questions? All lines are muted so please use the chat panel if we run out of time we will follow up with you.













- Gold accredited partner to IBM, Predictive Solutions and DataRobot specialising in statistics, advanced analytics & AI technologies
- Work with open-source technologies (R, Python, Spark etc.)
- Team each has 15 to 30 years of experience working in the advanced analytics industry

- Deep experience of applied data science applications across sectors
  - Retail
  - Healthcare/Pharma
  - Finance/Insurance
  - Media/Telecoms
  - Utilities
  - FMCG
  - Charity/Housing/Government



# **Agenda**

- What is a RAG solution?
- The background to ChatSPSS
- Key elements of a RAG solution
- Working with Wastonx.ai Prompt Lab
- Working with Wastonx.ai Auto Al RAG
- Deploying RAG models







# What is a RAG solution?

### What is a RAG Solution?



**Retrieval-Augmented Generation (RAG)**: Combines generative AI with a retrieval system, allowing a model to pull relevant facts and context from large databases or document collections before generating an answer



It helps to reduce hallucinations by grounding AI responses in up-to-date, verifiable data, even across vast unstructured text sources.



It makes it easy to query huge collections of data (like reports, logs, or research papers) with a single prompt, improving efficiency, accuracy, and transparency in a secure setting



# Why Build Your Own RAG?

- Data/user control & compliance
- Quality control that can be customised and measured
- Swap models and change parameters easily
- Ensure relevance, versioning and use of the most up-to-date data
- Integration & automation with your own systems

Accurate Compliant Current Controlled



# Real world applications of RAG solutions

- **Customer Support:** Powers chatbots that provide accurate, context-rich responses by retrieving information from FAQs, product guides, and support tickets.
- **Healthcare & Life Sciences:** Enables clinicians and researchers to query medical literature, patient records, data without manually sifting through unstructured text.
- Legal & Compliance: Helps lawyers or compliance officers find relevant case law, regulations, or contractual clauses efficiently.
- **Financial Services:** Retrieves and summarizes relevant regulations, policies, or market reports to assist analysts and advisors.
- **E-commerce & Retail:** Enhances product recommendation systems or answer customer questions using information drawn from product catalogues and reviews.
- **Government & Policy:** Allows policymakers and analysts to extract relevant insights from large archives of policy documents, laws, or public records.

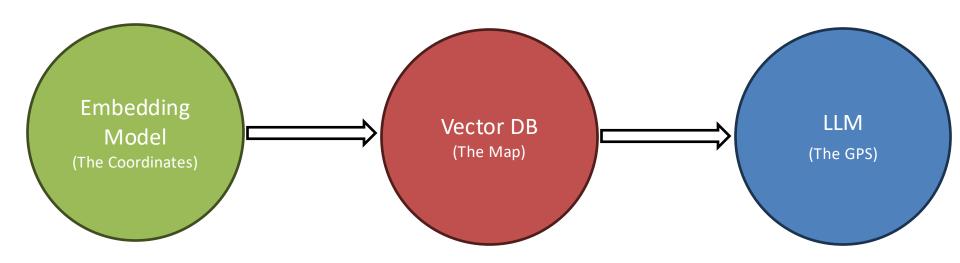


# Key elements of a RAG solution

- **Embedding Model** Instead of storing raw text, the embedding model converts documents, paragraphs, or sentences into vectors (numerical representations).
- Vector Database After an embedding model converts documents into vectors, these
  are stored in the vector database instead of plain text. You can think of a vector as a
  GPS coordinate in meaning-space instead of longitude/latitude, it's hundreds of
  numbers that describe the semantic position of a text fragment
- LLM (Large Language Model):
  - The LLM takes the query and the retrieved documents as input
  - The retriever passes a set of relevant text snippets (i.e. the context) along with the user's question to the LLM
  - The LLM then processes both the query and the retrieved passages, integrating them into its reasoning and generates a natural language response.

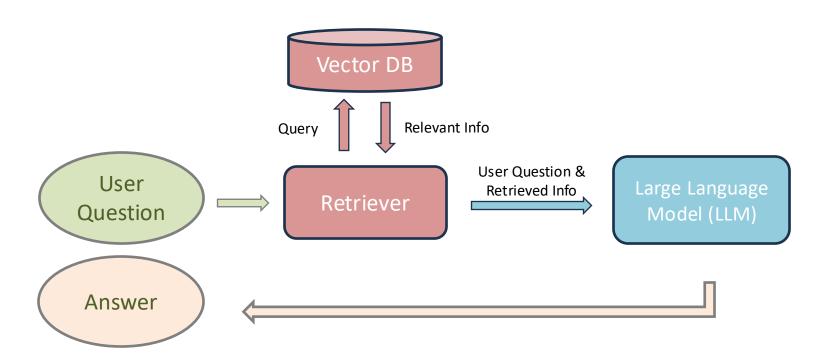


# Key elements of a RAG solution (and Generative AI)





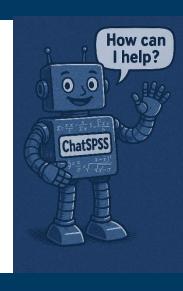
# Querying a Simple RAG solution











# The background to ChatSPSS

# The background to ChatSPSS

- Opportunity to learn more about Watsonx.ai and its RAG capabilities
- Picked a test area where we already had a lot of subject matter expertise and available data, to act as an example and demonstrator
- Decided that if it we were successful, we would deploy it to the wider community of SPSS users



# Project data: IBM SPSS Statistics User Guides

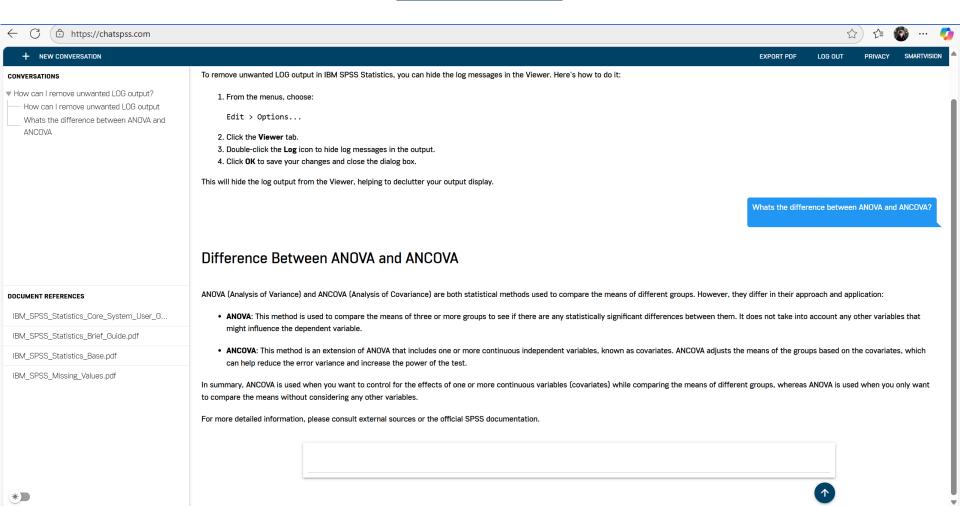
 Based on a collection of 20 pdf documents covering SPSS Statistics Base plus assorted addon modules and admin docs.

IBM_SPSS_Statistics_Brief_Guide.pdf	26/03/2025 10:47	Chrome HTML Do	5,609 KB
IBM_SPSS_Custom_Tables.pdf	26/03/2025 10:42	Chrome HTML Do	2,402 KB
IBM_SPSS_Statistics_Core_System_User_Guide.pdf	26/03/2025 10:49	Chrome HTML Do	2,237 KB
IBM_SPSS_Exact_Tests.pdf	26/03/2025 11:09	Chrome HTML Do	1,898 KB
IBM_SPSS_Statistics_Base.pdf	26/03/2025 10:40	Chrome HTML Do	1,649 KB
IBM_SPSS_Advanced_Statistics.pdf	26/03/2025 10:41	Chrome HTML Do	1,521 KB
IBM_SPSS_Regression.pdf	26/03/2025 10:44	Chrome HTML Do	1,093 KB
IBM_SPSS_Statistics_Batch_Facility_User_Guide.pdf	26/03/2025 11:12	Chrome HTML Do	586 KB
IBM_SPSS_Complex_Samples.pdf	26/03/2025 10:41	Chrome HTML Do	571 KB
IBM_SPSS_Forecasting.pdf	26/03/2025 10:44	Chrome HTML Do	571 KB
IBM_SPSS_Categories.pdf	26/03/2025 10:41	Chrome HTML Do	490 KB
IBM_SPSS_Neural_Network.pdf	26/03/2025 10:44	Chrome HTML Do	419 KB
IBM_SPSS_Data_Preparation.pdf	26/03/2025 10:42	Chrome HTML Do	408 KB
IBM_SPSS_Decision_Trees.pdf	26/03/2025 10:42	Chrome HTML Do	400 KB
IBM_SPSS_Direct_Marketing.pdf	26/03/2025 10:42	Chrome HTML Do	395 KB
IBM_SPSS_Missing_Values.pdf	26/03/2025 10:44	Chrome HTML Do	393 KB
iBM_SPSS_Conjoint.pdf	26/03/2025 10:41	Chrome HTML Do	339 KB
<ul><li>Authorized_User_License_Administrator_Guide.pdf</li></ul>	26/03/2025 11:15	Chrome HTML Do	331 KB
IBM_SPSS_Bootstrapping.pdf	26/03/2025 10:41	Chrome HTML Do	318 KB
Getting_Started_with_Installation_and_Licensing.pdf	26/03/2025 11:16	Chrome HTML Do	288 KB



# watsonx.ai™

# **ChatSPSS**







Get instant, accurate answers to your SPSS and statistics questions – anytime you need them.

Smart Vision Europe is proud to introduce **ChatSPSS**, the first Al-enabled tutor designed specifically for SPSS users. Whether you're new to SPSS or an experienced analyst, ChatSPSS helps you get the very most out of your software, saving you time and boosting your confidence with data analysis.

### What can ChatSPSS do for you?

- ☑ Guide you through loading and preparing data in SPSS
- Show you how to summarise results and run statistical tests
- Explain outputs from regression models and other advanced techniques
- Answer almost any SPSS or statistics-related question, on demand

### Why use ChatSPSS?

Unlike generic AI tools, ChatSPSS has been optimised with expert-verified SPSS resources. That means you get reliable, relevant, and accurate answers tailored to your analysis.

### Ready to try it?

register now at ChatSPSS.com and start getting better answers from your data today. ChatSPSS is completely free to use and, like any large language model, is all about improvement so we welcome any feedback or comment you may have once you've used it.





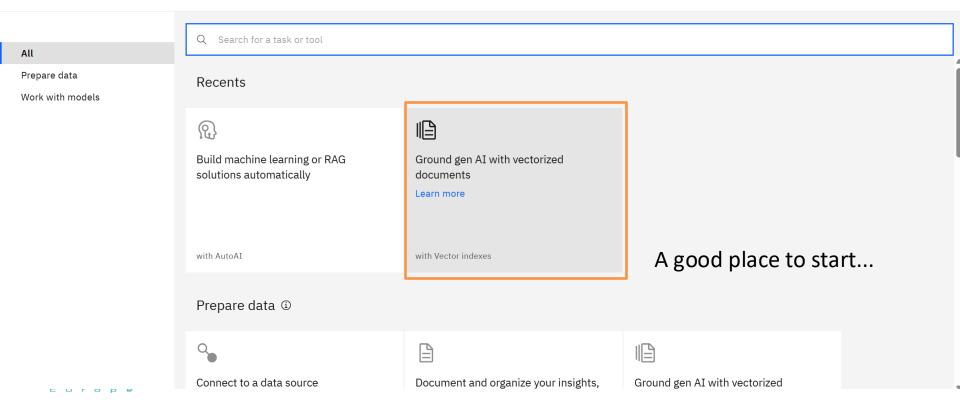
# Working with Wastonx.ai – Prompt Lab

# watsonx.ai<sup>™</sup> Creating and testing a basic RAG model

X

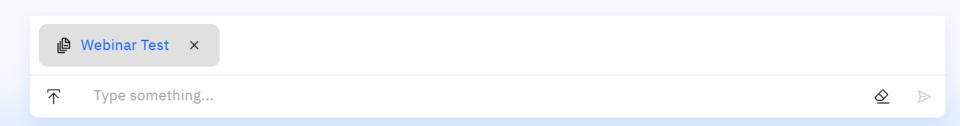
### What do you want to do?

Select a task based on your goal. You'll use a tool to create an asset for that goal.



# watsonx.ai<sup>™</sup> Prompt Lab

- Here we can design, test, and refine prompts against a library of IBM-hosted foundation models or even tuned models you've already deployed.
- We can use it to experiment across models and different applications
  like summarization, Q&A, extraction and classification tasks with ready-made
  examples and adjustable parameters.
- You can try out different prompts, compare outputs, and lock in effective settings without writing code.









# Working with Wastonx.ai – Auto Al RAG

# watsonx.ai™ Auto Al RAG

- Here we can set up experiments that try out lots of parameter settings (embedding model, chunking, LLM etc) and ranks them according to how they performed.
- One way to do this, is to compare how each model scored against a 'perfect' set of example questions and answers as part of an evaluation file.
- Any of the resultant models can then be directly deployed to the outside world for use.

KA	RAG Fattern leaderboard v								
	Rank ↑	Name	Model name	Answer faithfulness (Optimized)	Retrieval method	Hybrid ranker strategy	Number of chunks	Chunk size	Chunk overlap
•	1	Pattern 16	o granite-3-8b-instruct slate-125m-english-rtrvr-v2	0.932	Window Size: 4	N/A	5	2048	0
	2	Pattern 17	ogranite-3-8b-instruct multilingual-e5-large	0.916	Window Size: 4	N/A	5	2048	0
	3	Pattern 13	ogranite-3-8b-instruct multilingual-e5-large	0.906	Window Size: 4	N/A	3	2048	0
	4	Pattern 11	ogranite-3-8b-instruct slate-125m-english-rtrvr-v2	0.895	Window Size: 4	N/A	3	2048	0
	5	Pattern 18	ogranite-3-8b-instruct slate-125m-english-rtrvr-v2	0.894	Window Size: 4	N/A	5	2048	512



RAG Pattern leaderhoard \( \nabla \)

### **AutoAl RAG – Evaluation File**

```
■ RAG Project SPSS SPSS SPSS SYNTA SPSS_RAG_2 SPSS_SYNTA Divide ■
File
    Edit
         View
"question": "When you import data and define a variable-width string field, what is
the maximum width you are allowed to set?",
"correct answer": "A variable-width string field can be as wide as 32,767 bytes.",
"correct answer document ids": ["IBM SPSS Statistics Core System User Guide.pdf"]
"question": "How do you turn on connection pooling for an ODBC query pasted from the
Database Wizard?",
"correct answer": "After the wizard pastes the GET DATA /TYPE=ODBC command, add
Pooling=true at the very end of the quoted CONNECT string",
"correct answer document ids": ["IBM SPSS Statistics Core System User Guide.pdf"]
"question": "What licence role must a user hold to successfully import IBM Cognos BI
packages or reports into SPSS Statistics, and what happens if they dont?",
"correct answer": "The users Cognos Analytics licence must include the Administrator
entitlement and their role must be System Administrator. Without that role, the
Cognos BI data will not fully import.",
"correct_answer_document_ids": ["IBM_SPSS_Statistics_Core_System User Guide.pdf"]
"question": "Which three server-side processes must be copied to an IBM Cognos TM1
server (and the server restarted) hefore SPSS Statistics can exchange data with
Ln 1, Col 1 29.053 characters
                                                            Windows (CRLF)
                              Plain text
                                                                            UTF-8
```



# **Evaluating a RAG model**

- Answer faithfulness: This measures whether the generated answer is consistent with the provided retrieved context. It minimises the likelihood of the response including information that wasn't in the source documents (hallucinations).
- Answer correctness: This evaluates the factual correctness of the information
  presented in the final answer. While faithfulness links the answer to the context,
  correctness verifies the answer against known facts in the documents and the
  evaluation file. Here the focus is on enhanced accuracy and reliability.
- **Context correctness:** This is a *retrieval-quality* metric. In **AutoAI RAG**, Context correctness shows how well the retriever pulled the right chunk(s) of context compared to the benchmark. As such, it evaluates the retriever, not the generated answer.

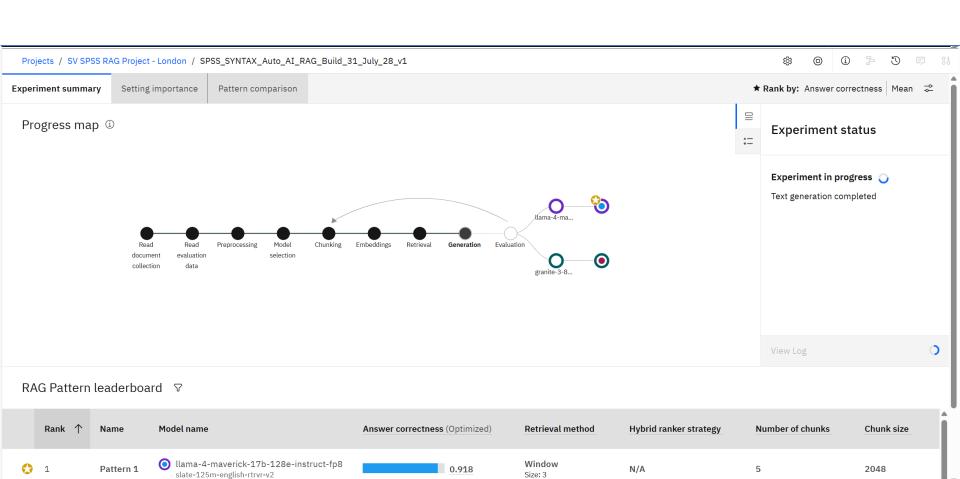


# Evaluating a RAG model: practical advice

- If you want to minimise hallucinations, then **Answer faithfulness** is a good optimum but watch correctness as a secondary metric.
- If you need to match *exact* answers (e.g. as in safety and compliance applications), try to optimise **Answer correctness**.
- You can use Context correctness to debug the retriever itself. Low scores
  here indicate that parameters such as chunking, embedding, or retrieval
  may need tuning.



# Auto Al Rag Experiment progress map







# **Deploying RAG Models**

# **Deployment options - Notebook**

### Save notebook

Save a pattern as a notebook, so you can review, edit, or run the code for the RAG pattern. Or, save as an AI service to create a deployable asset based on the current experiment settings.

# Choose your objective Index building, retrieval, and generation Build an index knowledge base using ChromaDB, retrieve relevant passages from the knowledge base, evaluate the retrieval quality, and build and deploy a function as an endpoint to inference with the RAG Pattern. Name Pattern 3: SPSS\_SYNTAX\_Auto\_AI\_RAG\_Build\_31\_Jul... Deployment space SV\_RAG\_DEPLOYMENT\_LONDON| X | V Why don't I see all of my spaces? ① Runtime environment

Runtime 24.1 on Python 3.11 AutoAI-M (4 vCPU and 16 GB RAM)

# **Deployment options – AI Service**

### Save AI Service

Save a pattern as a notebook, so you can review, edit, or run the code for the RAG pattern. Or, save as an AI service to create a deployable asset based on the current experiment settings.

### Choose your objective Define details Asset type 0 Index building, retrieval, and generation Notebook AI Service Build an index knowledge base using ChromaDB, retrieve relevant passages from the knowledge base, evaluate the retrieval quality, and build and deploy a function as an endpoint to inference with the RAG Pattern. Name Pattern 3: SPSS\_SYNTAX\_Auto\_AI\_RAG\_Build\_31\_Jul... Description (optional) AI Service description Add tags to make assets easier to find. Add a tag

Promote and deploy AI Service to deployment space

Cancel

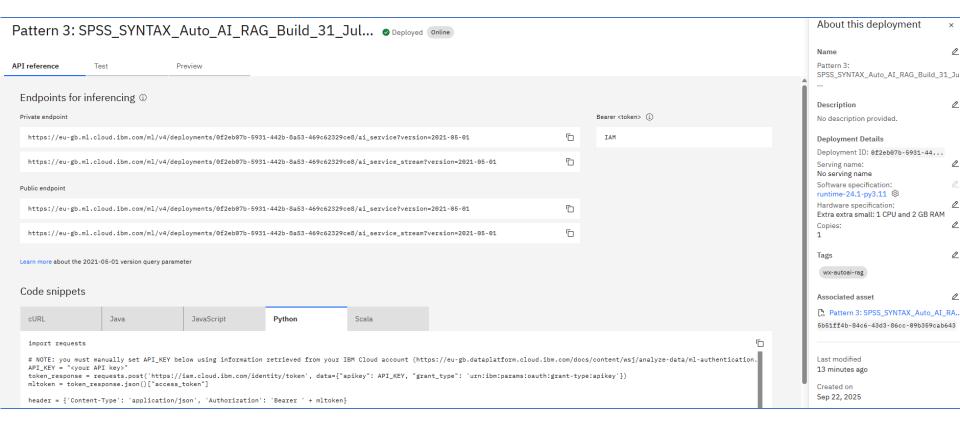
Create and deploy

# Deployment options - Al Service deploying

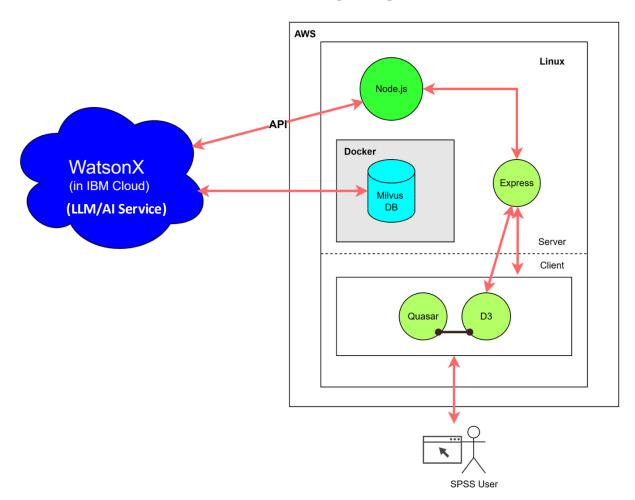
Deployment spaces / SV_RAG_DEPLOYMENT_LONDON / Pattern 3: SPSS_SYNTAX_Auto_AI_RAG_Build_31_Jul					
Deployments	Code	Schema			
∇ Q Search					
Name			Туре	Status	Tags
((p)) Pattern 3: SPSS	_SYNTAX_Auto_Al	_RAG_Build_31_Jul	Online	Initializing	wx-autoai-rag



# API endpoint of deployed AI Service



# chatSPSS deployment on Amazon (AWS)



Open-source components on AWS:

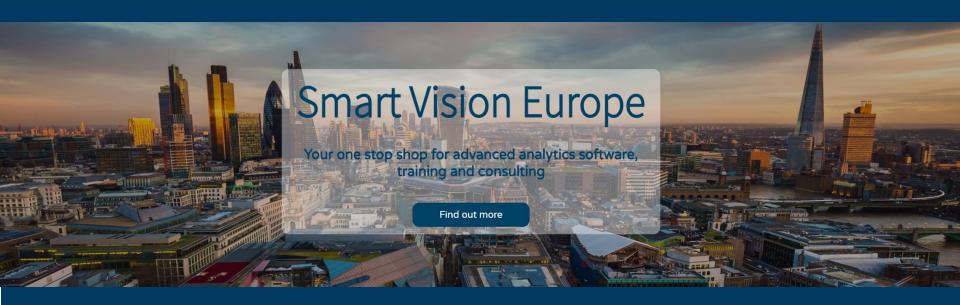
<u>D3</u>

Quasar

**Express** 

Node.js

**Milvus** 



# Working with Smart Vision Europe

# Working with Smart Vision Europe

- Ready to start your Retrieval Augmented Generation AI journey?
- We can assist with every aspect of your RAG project
  - Help with solution design and technical architecture
  - Selection of appropriate tools and technology components
  - Ensuring appropriate governance and security
- As experts in Data Science and AI we will
  - Offer "skills transfer" consulting
  - Collaborate with your internal resources to deliver a complete solution that you will own
  - Offer advice and guidance on engineering (separating code from content), content selection and security relating to RAG projects



# **Smart Vision Europe: Other services**

### Sourcing Software

- You can buy your analytical software from us often with discounts
- Assist with selection, pilot, implementation & support of analytical tools
- <a href="http://www.sv-europe.com/buy-spss-online/">http://www.sv-europe.com/buy-spss-online/</a>

### Training

- Formal classroom/virtual training
- Online self-paced training resources

### Advice and Support

- 'No strings attached' technical and business advice relating to analytics
- Tracked technical support services around the IBM SPSS product line





### Contact us:

+44 (0)207 786 3568

info@sv-europe.com

Twitter: @sveurope

Follow us on Linked In **6** 

Sign up for our Newsletter

# Thank you