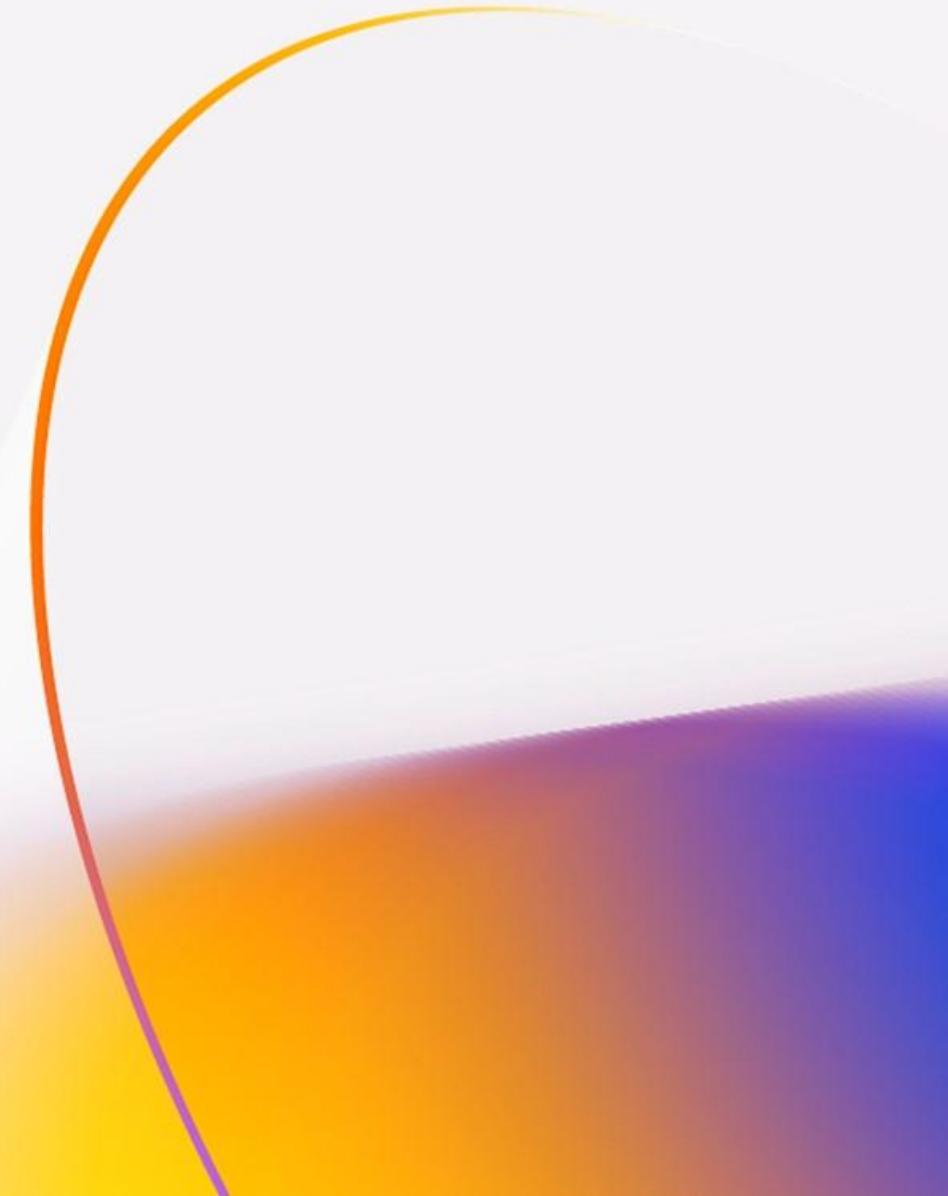




# AI Tech Accelerator Community for Swiss Startups

Be inspired. Be productive. Be Yourself.

Dominique Broeglin  
AI Global Black Belt  
@dbroeglin



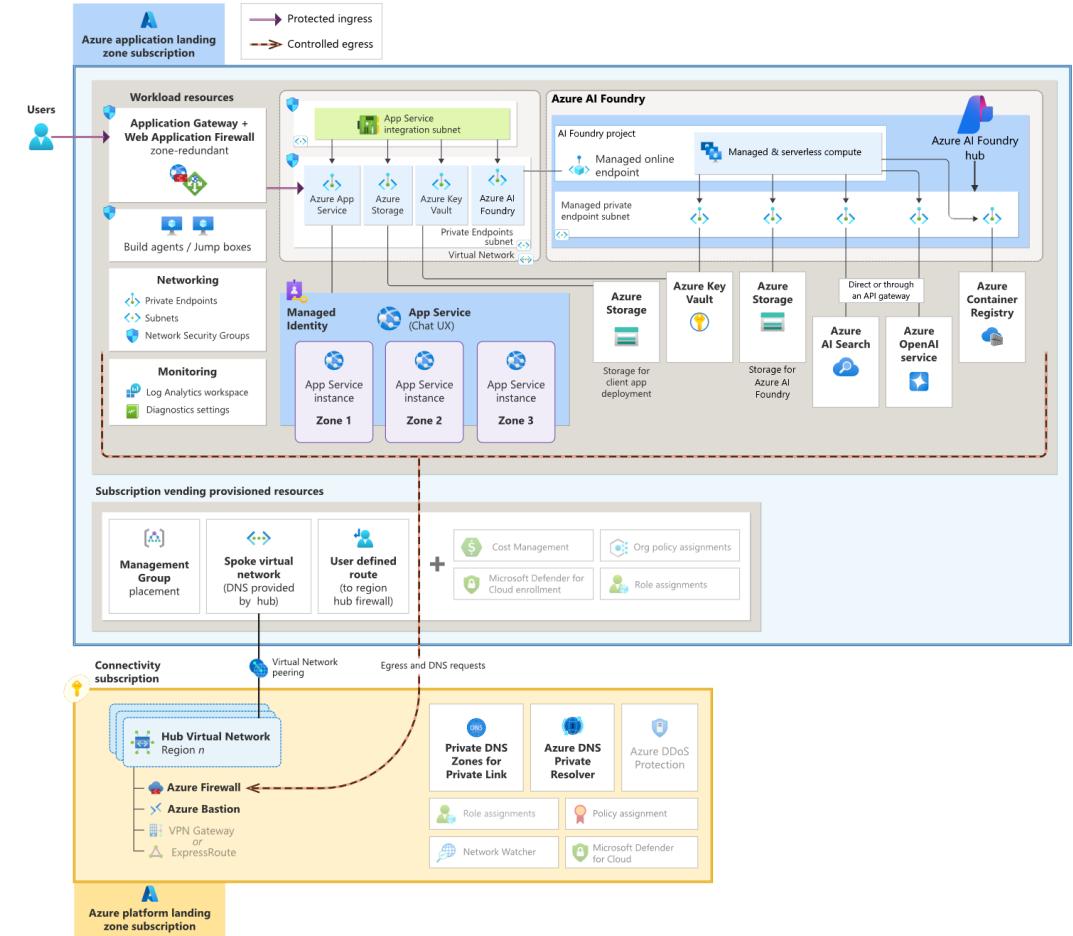


Azure Architecture ~~(90°)~~ 5'

# Azure Landing Zones & Reference Architectures

## Resources:

- Cloud Adoption Framework
- Azure Well Architected\$
- Platform Engineering



<https://learn.microsoft.com/en-us/azure/architecture/ai-ml/architecture/azure-openai-baseline-landing-zone>

# Get started with Azure AI Agents through Azure AI Foundry SDK

**Get started with Agent Service on Microsoft Learn Docs**

[https://aka.ms/MicrosoftLearn\\_Agents](https://aka.ms/MicrosoftLearn_Agents)

Watch the Microsoft Ignite sessions on-demand

- [Streamlining customer service with AI-powered agents](#)
- [Building intelligent multi-agent systems with Azure AI](#)

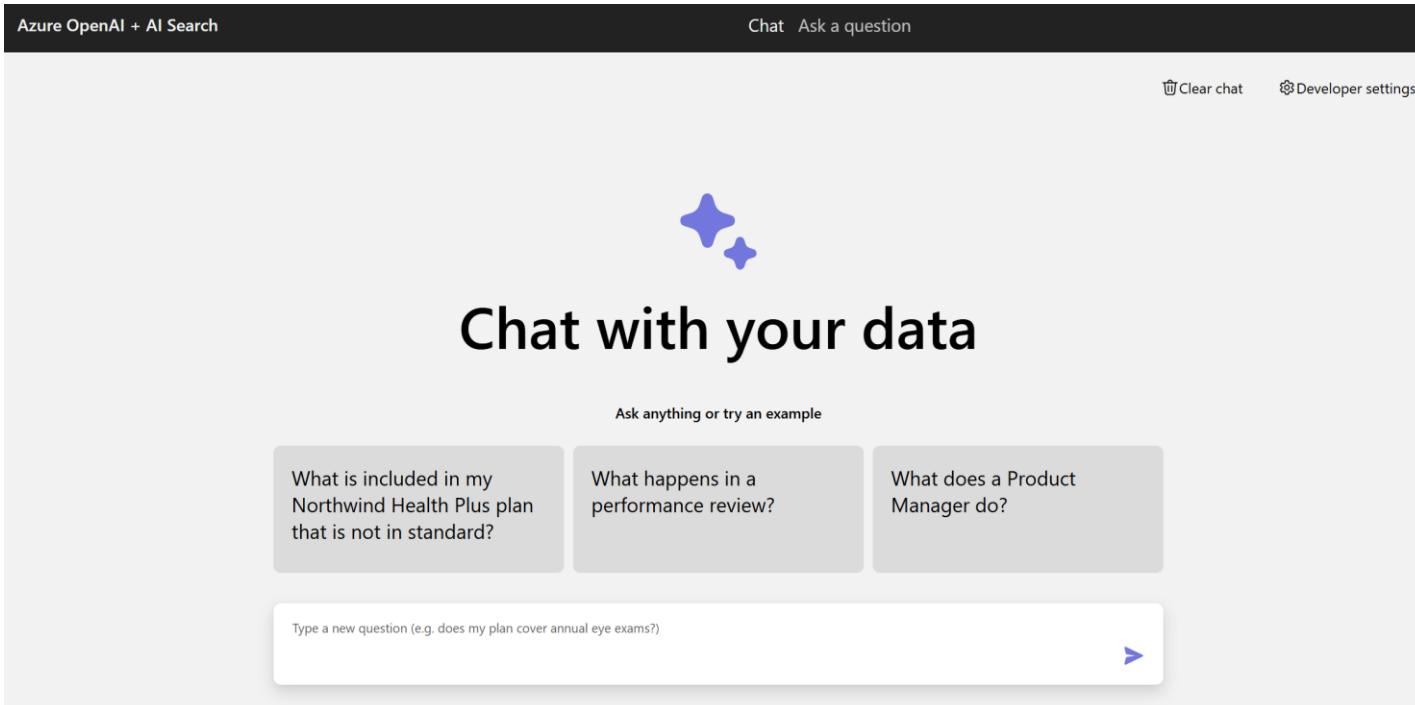
Read the [Techcommunity announcement](#)



# Azure AI Accelerators Catalogue

# Azure + OpenAI RAG

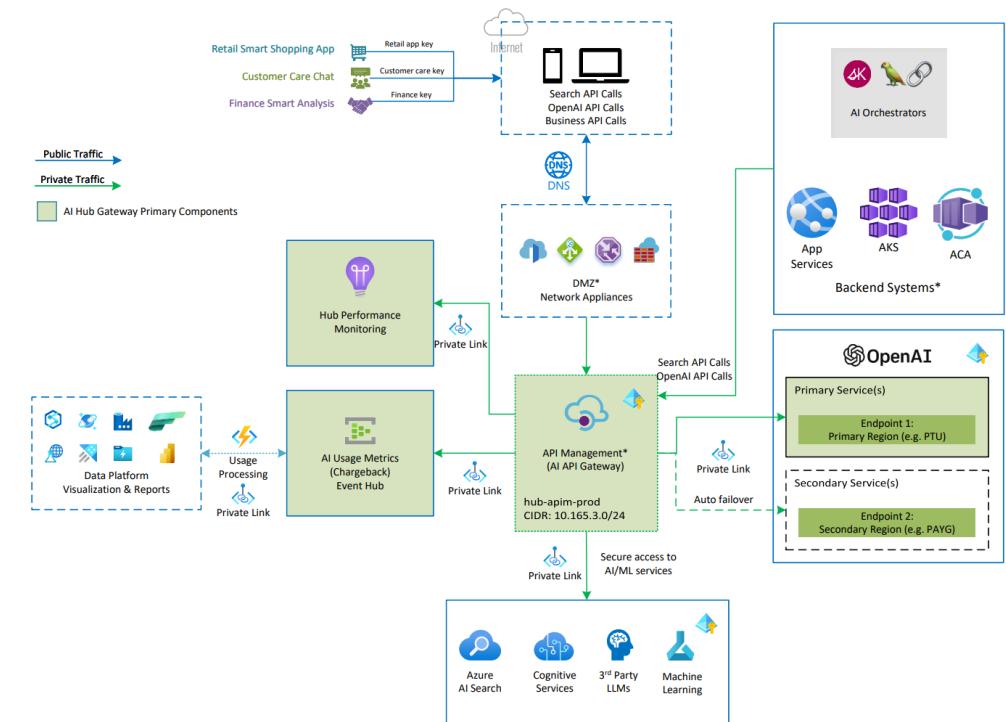
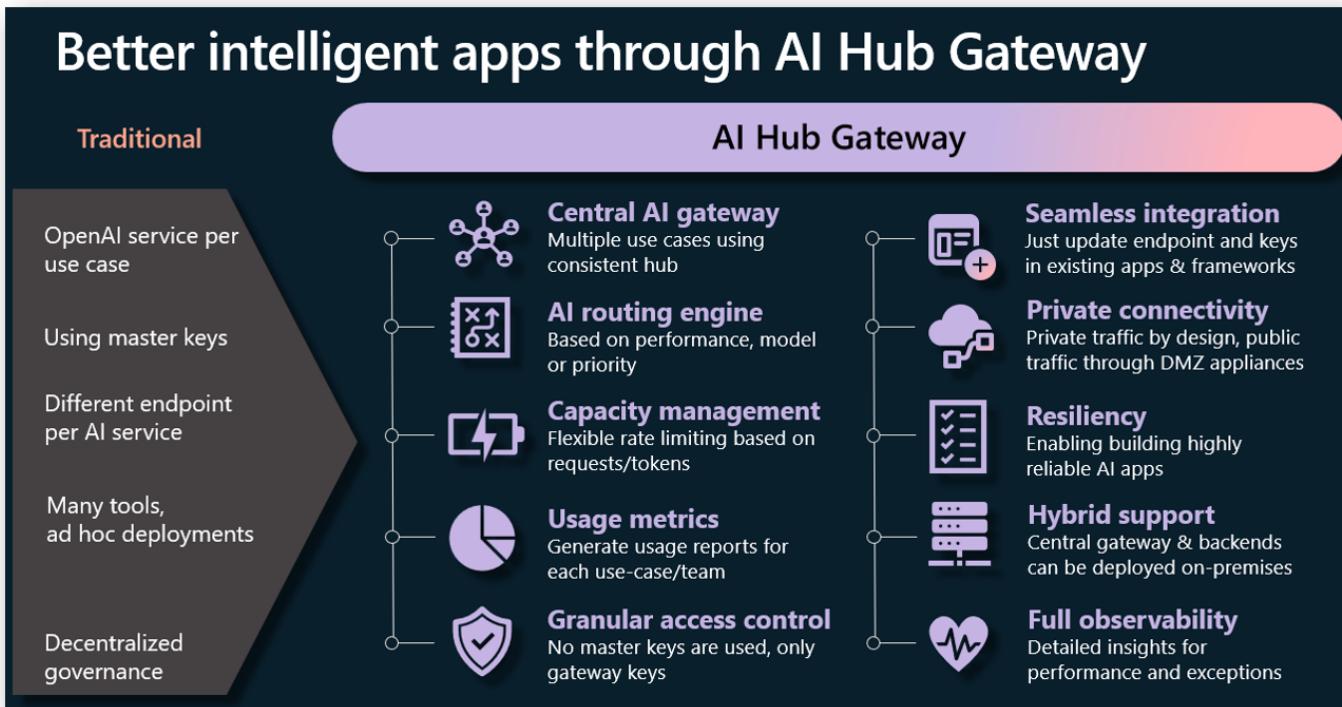
[github.com/Azure-Samples/azure-search-openai-demo](https://github.com/Azure-Samples/azure-search-openai-demo)



# AI Hub Gateway

## Landing Zone accelerator

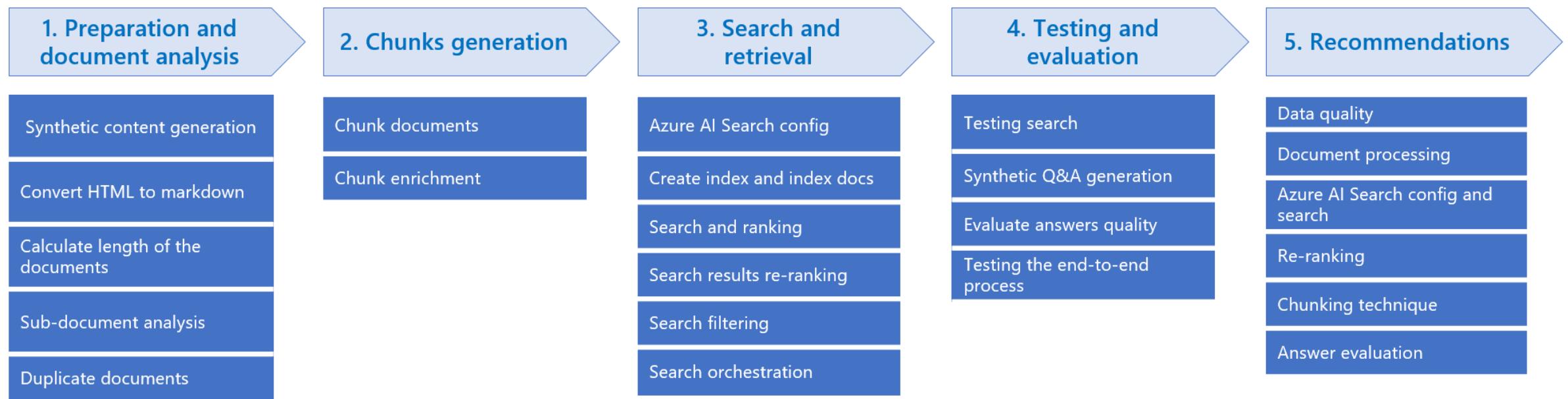
[github.com/Azure-Samples/ai-hub-gateway-solution-accelerator](https://github.com/Azure-Samples/ai-hub-gateway-solution-accelerator)



# Design-and-evaluation- of-RAG-solutions

Programmatic approach to better RAG

# Reference Architecture: Design and evaluation of RAG solutions



<https://github.com/Azure-Samples/Design-and-evaluation-of-RAG-solutions>

## Description

Even though **RAG is one of the most recurring implementations** around GenAI with multiple sample references, there are **no comprehensive end-to-end recommendations for design and testing** that can be easily shared with customers & partners. This repo makes it easy to:

1. Reflect the methodology to provide recommendations to build a RAG Solution following the best practices, along with tools and techniques for document analysis, processing, testing and evaluation.
2. Provide a set of reusable code snippets (12) that you can use as a sample for your projects or for learning purposes (Workshops, Hackathons, PoCs, MVPs, etc.)

# ARGUS

(Automated Retrieval and GPT Understanding System)

# Complex Document Extraction, Classification & Reasoning made Simple&Reliable.



*Argus [*a:gəs*] :  
"All-seeing Argus"  
*is a many-eyed giant in Greek  
mythology.**

FSI



## Insurance Claims / Policy Changes

- ✓ Typically, **60k/yr** policy changes to process
  - ✓ Typically, **2-8m/yr** claims to process
  - ✓ Argus provides **5-20m\$/yr** savings

## Manufacturing



## Delivery Notice / Bill of Delivery

- ✓ Typically, **100k/yr** delivery notices to process
  - ✓ Argus provides **5-10m\$/yr** savings

Pharmaceuticals / Life Science



## Life Science Extraction/Classification

- ✓ Typically, **10-30k/yr** Lab and Research documents
  - ✓ Argus provides **10-20m\$/yr** savings

Currently Expanding in Manufacturing and Energy & Resources

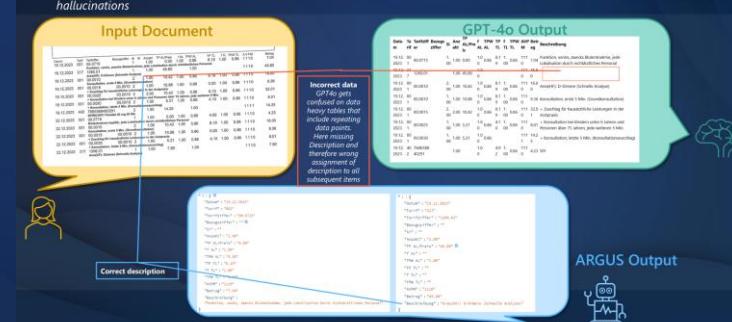
## Azure Document Intelligence vs. ARGUS

*ARGUS understands the context of the document before processing it (i.e., invoice, medical report, shopping list etc.)*

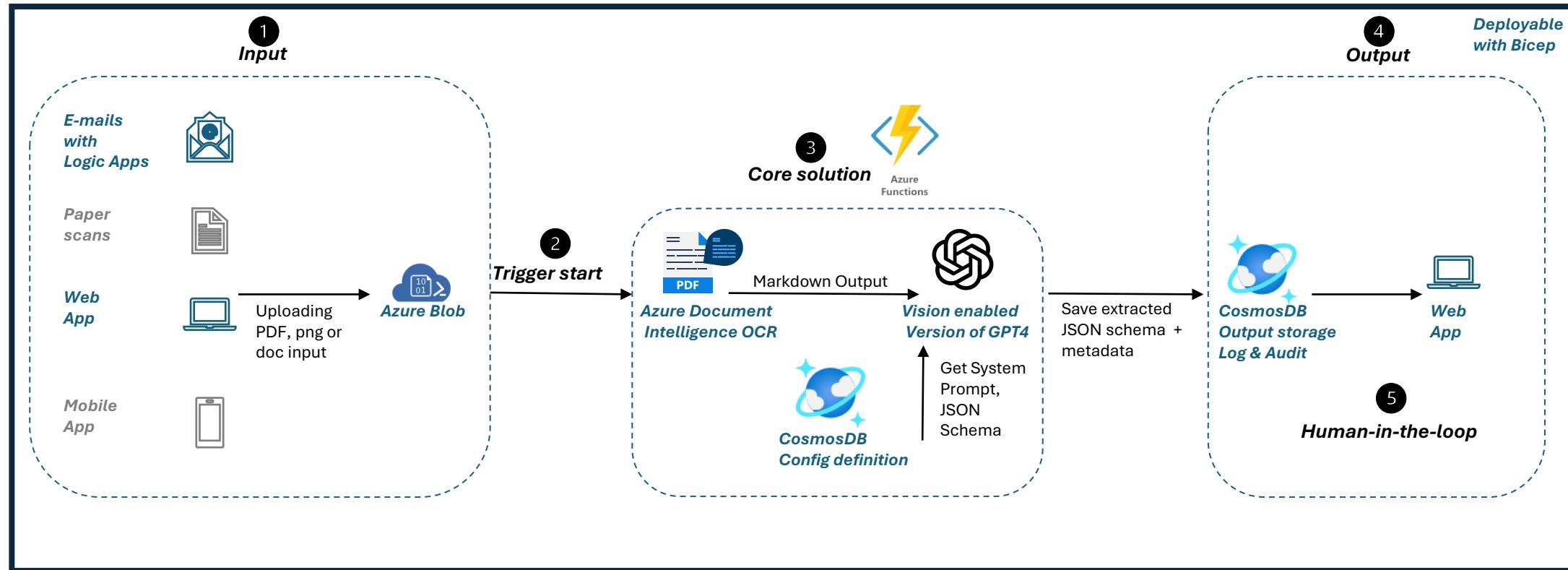


GPT 4 (omni) vs. ARGUS

*ARGUS handles data heavy documents (i.e., large tables, excel files etc.), improves on robustness and avoids laziness and hallucinations*



# Solution Overview



# ARGUS on Azure Github

README Code of conduct MIT license MIT license

## ARGUS: Automated Retrieval and GPT Understanding System

Argus Panoptes, in ancient Greek mythology, was a giant with a hundred eyes and a servant of the goddess Hera. His many eyes made him an excellent watchman, as some of his eyes would always remain open while the others slept, allowing him to be ever-vigilant.

### This solution demonstrates Azure Document Intelligence + GPT4 Vision

Classic OCR (Object Character Recognition) models lack reasoning ability based on context when extracting information from documents. In this project we demonstrate how to use a hybrid approach with OCR and LLM (multimodal Large Language Model) to get better results without any pre-training.

This solution uses Azure Document Intelligence combined with GPT4-Vision. Each of the tools have their strong points and the hybrid approach is better than any of them alone.

Notes:

- The document-intelligence resource needs to use the markdown preview feature (limited regions: West EU and East US at the moment).
- The Azure OpenAI model needs to be vision capable i.e. GPT-4T-0125, 0409 or Omni

### Solution Overview

- Frontend:** A Streamlit Python web-app for user interaction. UNDER CONSTRUCTION
- Backend:** An Azure Function for core logic, Cosmos DB for auditing, logging, and storing output schemas, Azure Document Intelligence, GPT-4 Vision and a Logic App for integrating with Outlook Inbox.
- Demo:** Sample documents, system prompts, and output schemas.

[Azure-Samples/ARGUS \(github.com\)](#)

Deployable with One-click deployment

### Deployment

#### One-Click Deployment

Click the button to directly deploy to Azure:

 Deploy to Azure

[Deploy to Azure](#) offers a one click deployment without cloning the code. Alternatively, follow the instructions below.

#### Deployment with `azd up`

##### 1. Prerequisites:

- Install [Azure Developer CLI](#).
- Ensure you have access to an Azure subscription.
- Create an OpenAI resource and deploy a vision-capable model.
- Ensure Docker is running

##### 2. Deployment Steps:

- Run the following command to deploy all resources:

`azd up`



# SOP Creator



# Automating SOP Document Creation

# A genAI-powered video transcription, standardization, and search solution.

## Issues & Challenges

- Manual Transcription & Standardization
  - Searchability & Retrieval
  - Maintenance & Updates



## Solution

- The SOP Genie turns videos into SOP documents, by leveraging genAI.
  - Achieves unparalleled accuracy in SOP creation, transcription and standardization.
  - **Saves Customers >\$200,000 per factory!**



## Outcomes

- Time saved across various business processes
  - Enhancement of processes & knowledge transfer, and onboarding procedures.
  - Scrap Avoidance through improved accuracy



## UX Overview - What it looks like

# UX Overview - What it looks like

### SOP Document Generator

Download following video file:  
<https://youtu.be/0E3DUU86Geo>

Save video file for:

[Download video file](#)



3:22 / 7:40

---

#### Video File Information

File: ./data/How Nescafe Coffee Is Made in the factory Coffee Bean Harvesting Process.mp4  
Size: 27.68 MB  
Last Modified: 2024-08-08 13:09:09  
Duration: 460 seconds  
Length of video: 00:07:40  
Number of frames: 13805

### Video Frames



#### Audio file information

Extracting audio from video file  
Completed in 0:00:03.339075  
File: ./results/How Nescafe Coffee Is Made in the factory Coffee Bean Harvesting Process.wav  
Size: 77.41 MB  
Last Modified: 2024-08-08 13:09:15

#### Waveplot of audio file



### Extract text from audio file

Transcribing audio file  
Completed in 0:03:55.696477  
Transcript text lenght: 7810  
Transcript: Every day, around 2.4 billion cups of coffee is consumed worldwide. To meet this high demand millions of people harvest tons of coffee beans. But how are 175,000 jars of coffee produced every day? Let's visit the Nescafe factory to discover how one of the most popular beverages in the world is made. Coffee is the second most consumed beverage in the world after water. It is believed to have been discovered by a shepherd who noticed that his goats became more active after eating certain beans. Coffee has been cultivated for over 1000 years. And is now produced in more than sixty countries around the world. The history of Nescafe dates back to the early 20th century when Swiss chemist Max Morgenthaler began working on a way to produce soluble coffee. In 1930, Nestle acquired the rights to Morgenthaler's patent and began producing and marketing soluble coffee under the Nescafe brand. The company invested heavily in advertising and promotion, and Nescafe quickly became a leading brand in the coffee market. During World War Two, Nescafe became a staple for the U.S. troops and the Allied forces as it was easy to transport and prepare in the battlefield. After the war, Nescafe became a popular product worldwide and Nestle continued to innovate in the production and packaging of soluble coffee. The flavor of coffee depends on the region where it is cultivated and the roasting and blending methods used by the producers. Brazil is the largest coffee producer in the world, accounting for over 35% of coffee production. Coffee grows in the form of flowers and nine months later, the buds become grape-sized berries. Coffee is harvested from the coffee plants once it is fully ripe, which can be identified by its intense red color. From flowering to harvest, it usually takes between 6:00 and 9:00 months. Harvesters collect the ripe berries while leaving the green ones to ripen further. They return to collect the berries from the same plant as they mature. Harvesters try to ensure that there are no more than two green berries on each branch to prevent overripening.

#### Text transcript with confidence level

	Word	Offset	Duration	Confidence	Offset_in_secs	Duration_in_secs
0	every	1,100,000	2,400,000	0.8623	0.11	0.21
1	day	3,500,000	4,400,000	0.6895	0.35	0.4
2	around	7,900,000	4,400,000	0.5825	0.79	0.4
3	two	12,300,000	2,000,000	0.8789	1.23	0.2

# SOP Creator

Automating Standard Operating Procedure (SOP) Document Generation:  
A genAI-powered video transcription, standardization,  
and search solution.

## UX Overview - What it looks like

**SOP Document Generator**

Download following video file:  
<https://youtu.be/0E3BUU8IGeo>

Save video file to:  
/data

Download video file

**PROCESS**

Video File Information  
File: /data/How Nescafe Coffee Is Made in the factory Coffee Bean Harvesting Process.mp4  
Size: 27.68 MB  
Last Modified: 2024-08-08 13:09:15  
Duration: 460 seconds  
Length of video: 00:07:40  
Number of frames: 13805

**Video Frames**  
Frame 1200 Frame 2700 Frame 4200 Frame 5700  
Frame 6200 Frame 6600 Frame 8100 Frame 12400

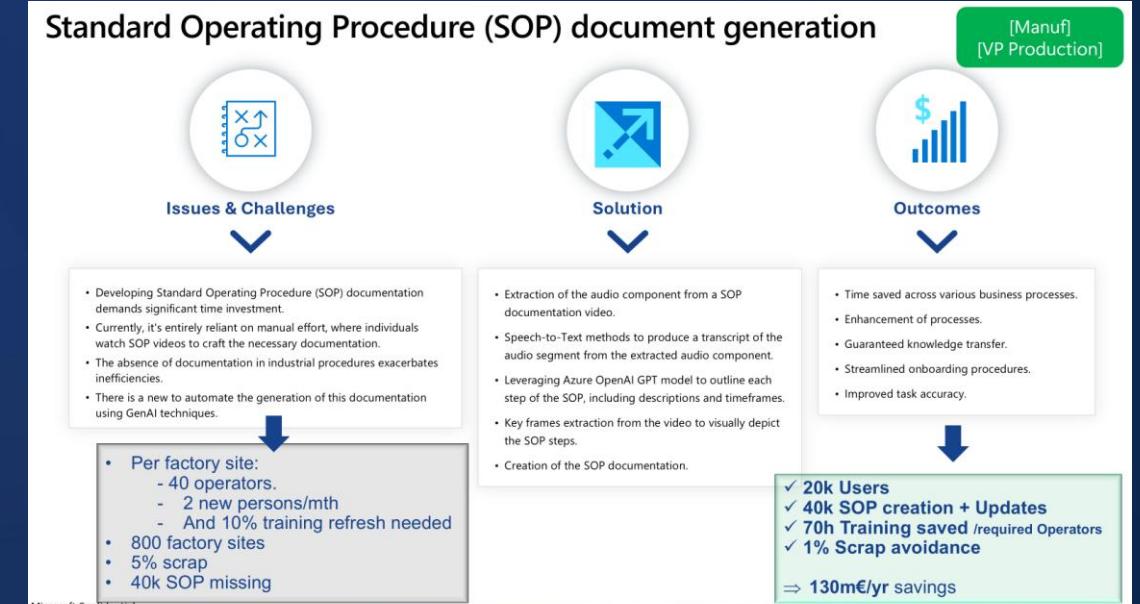
**Extract text from audio file**  
Transcribing audio file  
Completed in 00:03:55.696477  
Transcript text length: 7810  
Transcript: Every day, around 2.4 billion cups of coffee is consumed worldwide. To meet this high demand, millions of people harvest tons of coffee beans. But how are 175,000 jars of coffee produced every day? Let's visit the Nescafe factory to discover how one of the most popular beverages in the world is made. Coffee is the second most consumed beverage in the world after water. It is believed to have been discovered by a shepherd who noticed that his goats became more active after eating certain beans. Coffee has been cultivated for over 1000 years. And is now produced in more than sixty countries around the world. The history of Nescafe dates back to the early 20th century when Swiss chemist Max Morgenthaler began working on a way to produce soluble coffee. In 1930, Nestle acquired the rights to Morgenthaler's patent and began producing and marketing soluble coffee under the Nescafe brand. The company invested heavily in advertising and promotion, and Nescafe quickly became a leading brand in the coffee market. During World War Two, Nescafe became a staple for the U.S. troops and the Allied forces as it was easy to transport and prepare in the battlefield. After the war, Nescafe became a popular product worldwide and Nestle continued to innovate in the production and packaging of soluble coffee. The flavor of coffee depends on the region where it is cultivated and the roasting and blending methods used by the producers. Brazil is the largest coffee producer in the world, accounting for over 35% of coffee production. Coffee grows in the form of flowers and nine months later, the buds become grape-sized berries. Coffee is harvested from the coffee plants once it is fully ripe, which can be identified by its intense red color. From flowering to harvest, it usually takes between 6-8 and 9-10 months. Harvesters collect the ripe berries while leaving the green ones to ripen further. They return to collect the berries from the same plant as they mature. Harvesters try to ensure that there are no more than two green berries per plant.

**Audio file information**  
Extracting audio from video file  
Completed in 00:00:03.339075  
File: ./results/How Nescafe Coffee Is Made in the factory Coffee Bean Harvesting Process.wav  
Size: 77.41 MB  
Last Modified: 2024-08-08 13:09:15

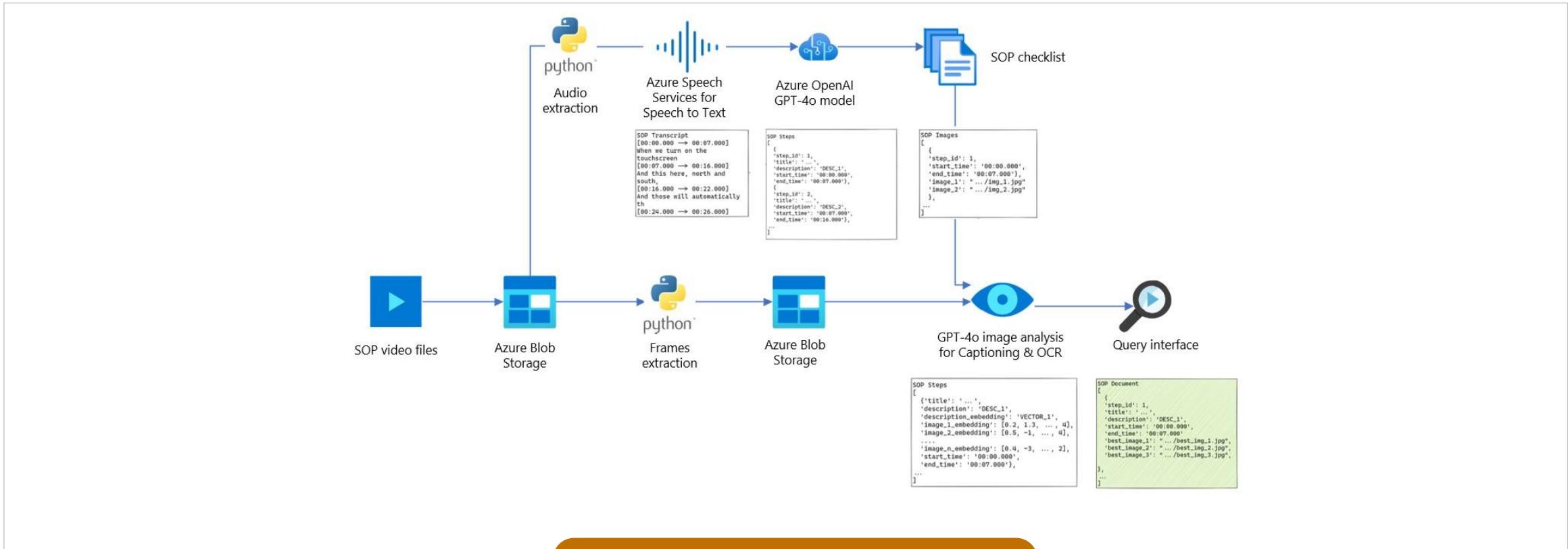
**Waveplot of audio file**  
Waveplot of ./results/How Nescafe Coffee Is Made in the factory\_Coffee Bean Harvesting Process.wav

**Text transcript with confidence level**

Word	Offset	Duration	Confidence	Offset_in_secs	Duration_in_secs
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2	around	7,000,000	4,400,000	0.9823	0.79
3	two	12,300,000	2,000,000	0.8789	1.23



# Reference Architecture: SOP Genie



<https://github.com/sjuratov/sop-doc-creation>

## Description

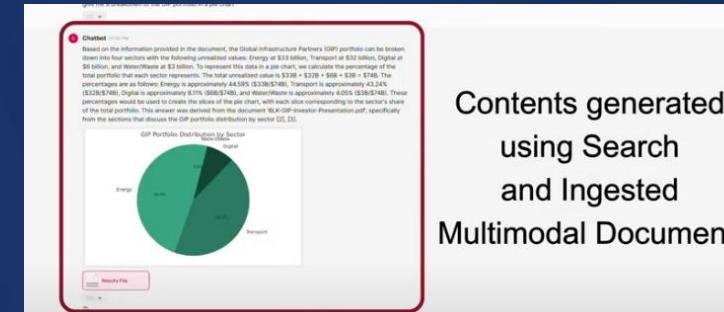
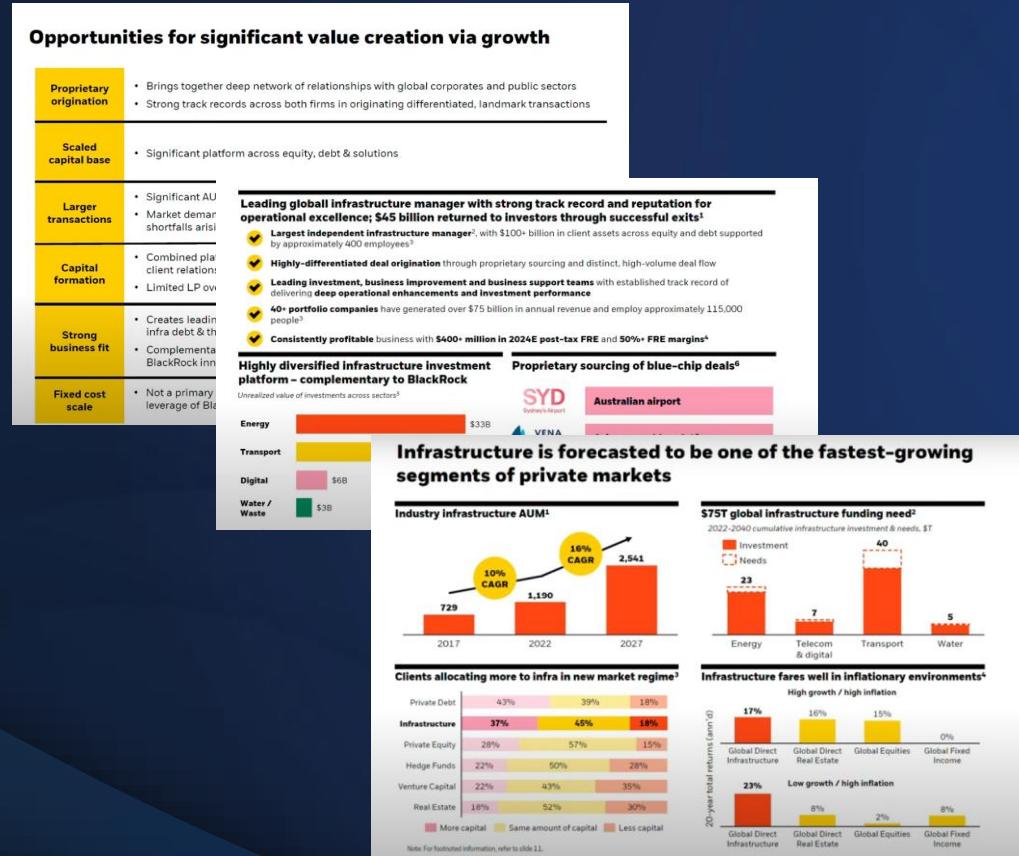
This reference architecture is using azd to automatically deploy relevant Azure services. Azure “one-click-deployment-button” will be an alternative way to deploy. Once deployed, end users are using Web App URL to access Streamlit application that provides UX. After user selects input video, processing pipeline does following:

- Downloads video
- Displays video metadata
- Extracts audio from video file
- Transcribes audio using Azure Speech
- Using transcription as an input, calls Azure OpenAI to create SOP outline/steps
- Using SOP outline, it creates SOP documents that contains following for each step:
  - title, summary, keywords, offset in video, displays video screenshot and explanation of it at the given offset, image OCR

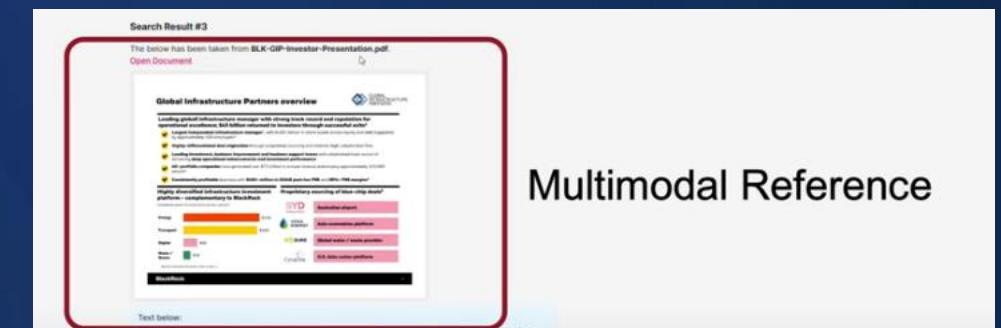
# Multimodal Document Analysis & Creation



[Research CoPilot: Multimodal RAG with Code Execution - YouTube](#)



Contents generated using Search and Ingested Multimodal Documents

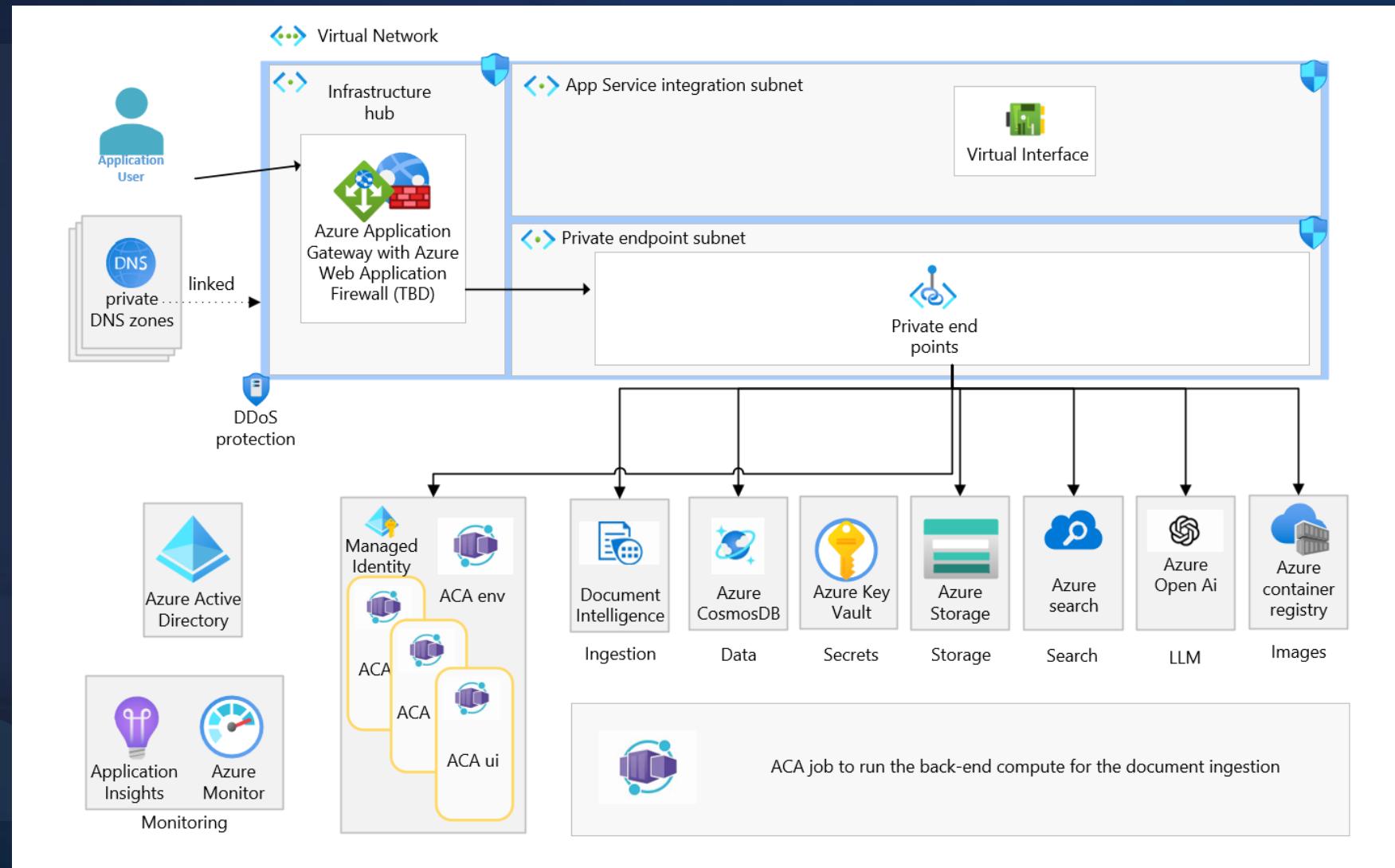


Multimodal Reference



# Multimodal **RAG** **WITH CODE EXECUTION**

# Solution Overview



# Research Copilot - Github

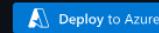
<https://github.com/Azure-Samples/multimodal-rag-code-execution>

Deployable with One-click deployment

The screenshot shows the GitHub README page for the 'multimodal-rag-code-execution' repository. At the top, there are links for 'README', 'Code of conduct', 'MIT license', and 'MIT license'. Below these, a note says 'Note: Start with the Tutorial Notebooks in the Tutorials folder [here](#)'. The main title 'Research CoPilot: Multimodal RAG with Code Execution' is displayed in large, bold, blue text. A detailed description follows, mentioning 'Multimodal Document Analysis with RAG and Code Execution: using Text, Images and Data Tables with GPT4-V, TaskWeaver, and Assistants API:'. Three numbered bullet points describe the implementation: 1. "Chat-With-Your-Multimodal-Data": Implemented a GenAI solution to automatically ingest and analyze multimodal documents, including texts, tables, and images, and produce searchable semantic contents. 2. Content Generation: The solution can be used to output most of the contents needed for standardized formats, such as memos or PowerPoint presentations that highlights essential facts about the ingested data. 3. Analytical Queries with OpenAI Assistants API and Taskweaver: Develop an advanced feature allowing analysts to interact with the multimodal data using a chat interface integrated with a Code Interpreter. This tool will support complex analytical queries, including calculations and on-the-fly graph and file generation.

## Deployment steps (Azure Portal)

1. Click the "Deploy to Azure" button



2. Fill in parameters

There are no special parameters for this deployment. Optional parameters are available to customize the deployment (see below). Typically, only `openAIName` and `openAIRName` are used to reuse an existing Azure OpenAI resource.

Average deployment time is 10 minutes when no existing container registry is set.

## Customization

For additional deployment customization, follow the detailed instructions available in the [Deployment README](#) guide.

# Guided Image Generation

## Customer Use Cases



### Brand Engagement

Coca Cola and Red Bull campaigns for consumers to create images based on given themes (e.g., Christmas, sports)

### Monetization by Product Customization

Post service for creating and ordering personalized postcards, such as for St. Patrick's Day



### Personalized Content Embedded in Products

Mercedes working on enabling passengers to create custom infotainment background images



Microsoft confidential

### Business Content Creation Tools

Image creation capabilities enhance ideation, design, communications and marketing functions



**Create Your Perfect Scene**

Prompt ok: **Hate: safe, Self-harm: not detected, Self-harm: safe, Sexual: safe, Violence: safe**

Photorealistic-style image of a woman's face with an explosion of vibrant, colorful powder radiating from the center on a black background. The woman has striking, expressive eyes and a serene yet powerful expression. The colors of the powder range from bright yellows and reds to deep blues and purples, creating a stark contrast between the vivid colors and the dark background highlights the intensity of the explosion. The image captures the fine details of the powder particles, creating a sense of motion and energy. The lighting is dramatic, emphasizing the contours of her face and the texture of the powder.

**Image Generation**

- Video Analysis
- Gallery
- Organizer
- Dialog

Protect brands/products:  Off  Invertable  Replace

Image generation model: DALL-E-3

HD Vivid

Image size: 1024x1024

Visual style: Photorealistic

Brand detection models: GPT-4o

Image moderation thresholds:

Hate	Low	SoftHarm	One
Sexual	One	Violence	One

Save image  Image cache

Download image Generate video clip

GPT-4o: No brands found (1.1s)

Image content safety: Hate: safe, Self-harm: safe, Sexual: safe, Violence: safe

Woman face with centered explosion of colorful powder on a black background

### 1. Validate and Revise Prompt

AOAI GPT-4o with Content Filter

- Identify jailbreak attempts
- Check for NSFW instructions: hate, self-harm, sexual, violence
- Enhance user scene description, add style and theme.
- Consider use case requirements (e.g., neutralize competitor brands)

### 2. Generate Image

DALL E-3

Stable Diffusion XL on Azure ML

Stable Diffusion 3, FLUX.1 [pro]

- Select model
- Select image properties
- Convert image to video clip

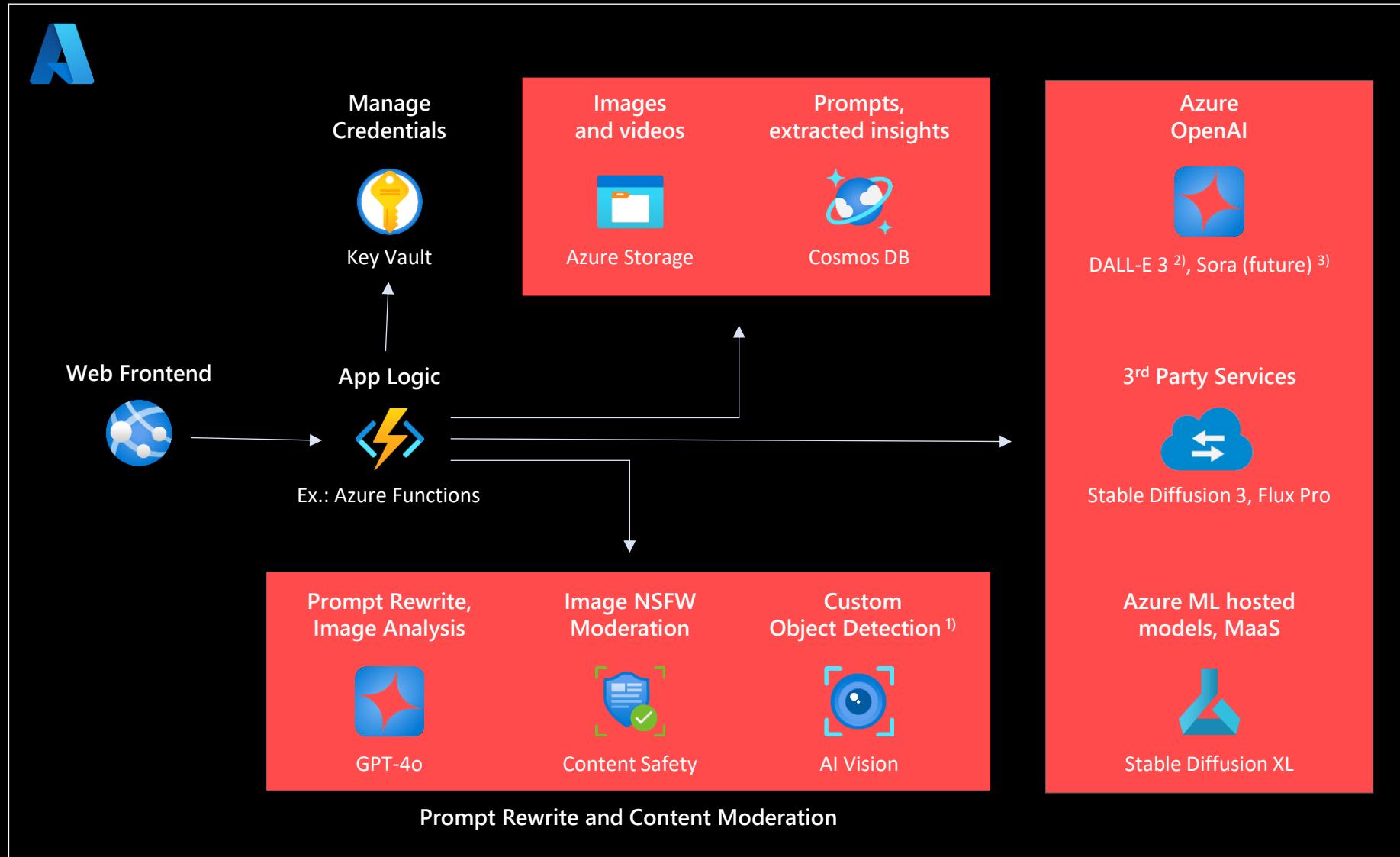
### 3. Validate Image

Azure Content Safety (NSFW)

AOAI GPT-4o

- Check for NSFW content
- Check for domain-specific objects (e.g., competitor visuals)

# Reference Architecture



- 1) Optionally train a custom model if GPT-vision model does not provide sufficient quality for your use case.
- 2) Built-in DALL-E prompt rewrite should be disabled for best results. Reach out to AI GBB team for more details.
- 3) Sora not yet available. Sample app includes basic image to video capabilities for demonstration.

# Guided Content Generation

Enable content creation for customers and employees, adhering to responsible AI standards and organizational policies

01

Validate and revise user prompt within guardrails (style, scene, brand awareness)

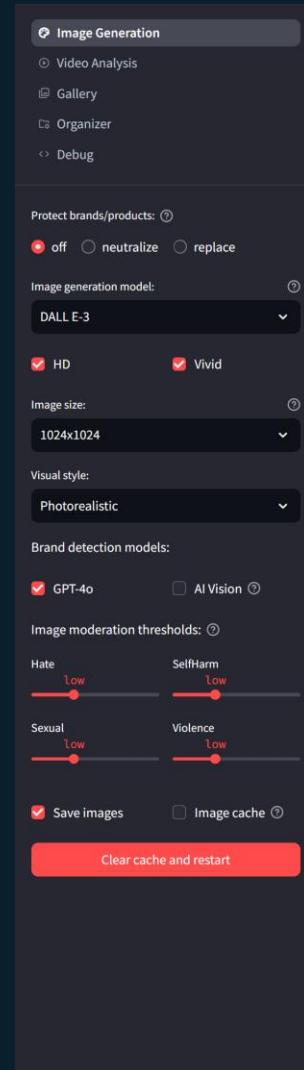
02

Generate image (DALL-E 3, FLUX, Stable Diffusion). Convert result to video clip.

03

Validate visual result: NSFW, domain-specific objects (e.g., competitor visuals)

<https://github.com/Azure/gen-cv/tree/main/guided-content-generation>



## Create Your Perfect Scene

Prompt ok: hate: safe, jailbreak: not detected, self\_harm: safe, sexual: safe, violence: safe

Photorealistic-style image of a woman's face with an explosion of vibrant, colorful powder radiating from the center on a black background. The woman has striking, expressive eyes and a serene yet powerful expression, with the powder creating a dynamic burst of reds, blues, yellows, and greens. The contrast between the vivid colors and the dark background highlights the intensity of the explosion. The image captures the fine details of the powder particles, creating a sense of motion and energy. The lighting is dramatic, emphasizing the contours of her face and the texture of the powder.



Download image

Generate video clip

GPT-4o: No brands found (3.1 s)

Image content safety: Hate: safe, SelfHarm: safe, Sexual: safe, Violence: safe

Woman face with centered explosion of colorful powder on a black background

Image Generation Video Analysis GalleryProtect brands/products: [?](#) off  neutralize  replaceImage generation model: [?](#)

FLUX.1 [pro]

Aspect Ratio [?](#) Steps [?](#)1:1 [?](#) 25 [-](#) [+](#)Guidance [?](#) Interval [?](#)3 [-](#) [+](#) 2 [-](#) [+](#)

Visual style:

Comic [?](#)

Brand detection models:

 GPT-4o  AI Vision [?](#)Image moderation thresholds: [?](#)Hate SelfHarm   
low lowSexual Violence   
low low

# Create Your Perfect Scene

Describe your image

Image Generation

Video Analysis

Gallery

Organizer

Debug

# Image Gallery

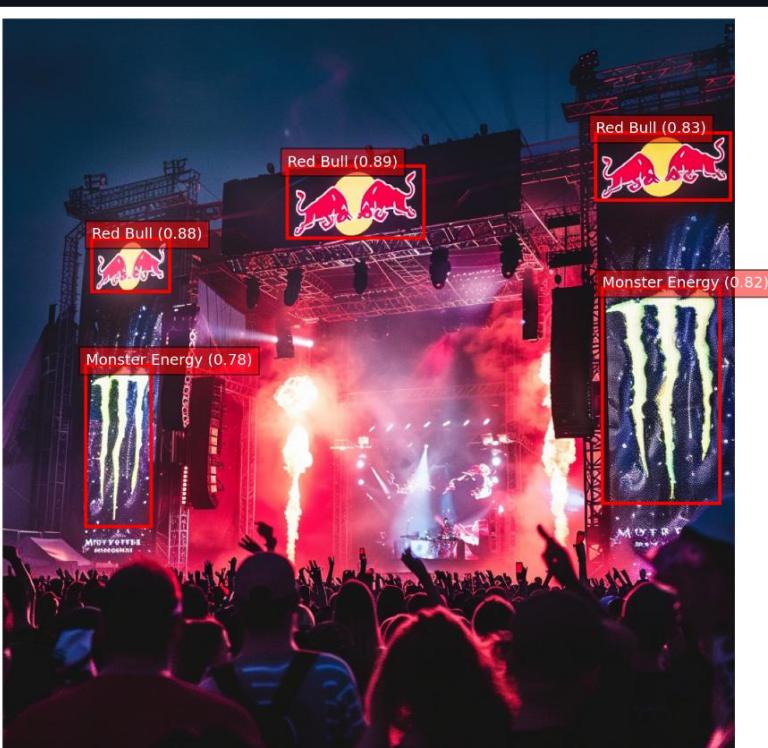
Select an image for analysis



Analyze



Analyze



AI Vision Analysis: Found Red Bull, Monster Energy (3.6 s)

GPT-4o: Found Red Bull, Monster Energy (3.5 s)

Image content safety: Hate: safe, SelfHarm: safe, Sexual: safe, Violence: safe

Custom object detection or  
GPT-4o for specific content  
analysis



Analyze



Analyze



Customer example. Do not share.

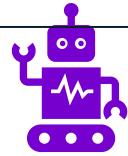
# AI for Call Centers



## Virtual & Assistants

Fully Automated  
Assist agents with real-time insights based on user's usage behavior  
Surface up-sell / cross-sell offers  
Recommend the next best action

During the call



## Call transcription

Speech-to-text service  
Speaker separation and diarization  
Transcription translation in one language



## Post-call analytics

PII detection / redaction  
Sentiment analysis  
Azure OpenAI-powered entity recognition, language understanding



## Call summarization

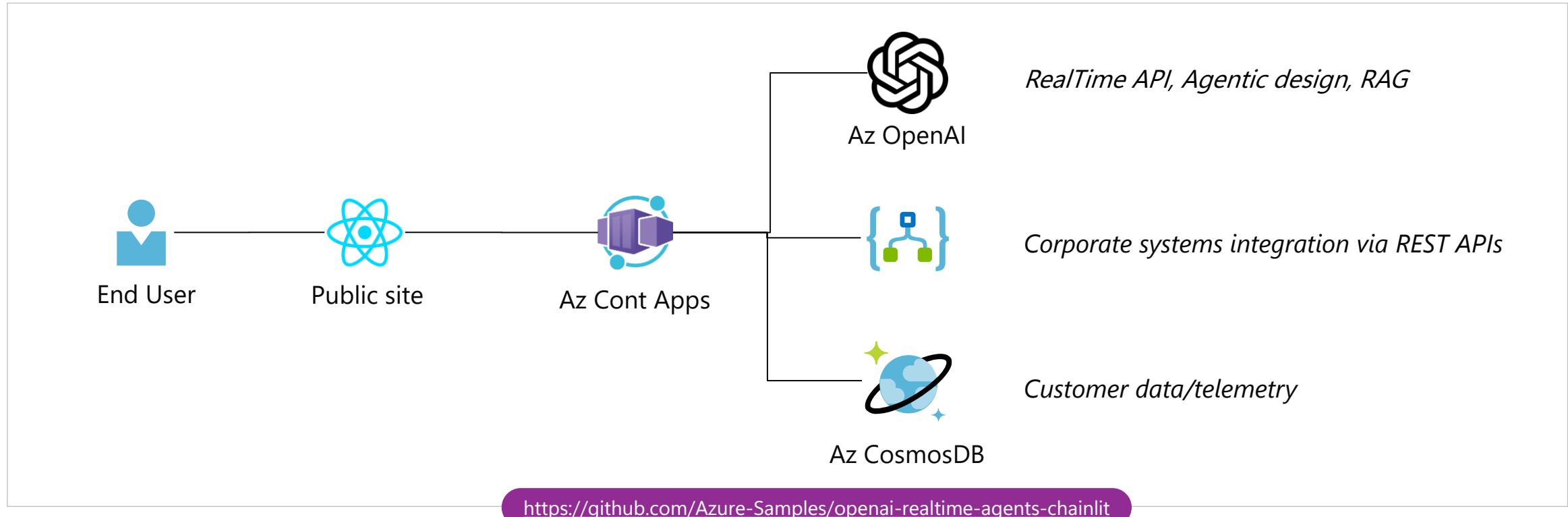
Extract key takeaways from the transcript  
Derive insights for similar issues in the future  
Propose ticket status

After the call ends



<https://www.youtube.com/watch?v=8dJx2oD4pR4>

# Reference Architecture: RealTime Voice Contact Center

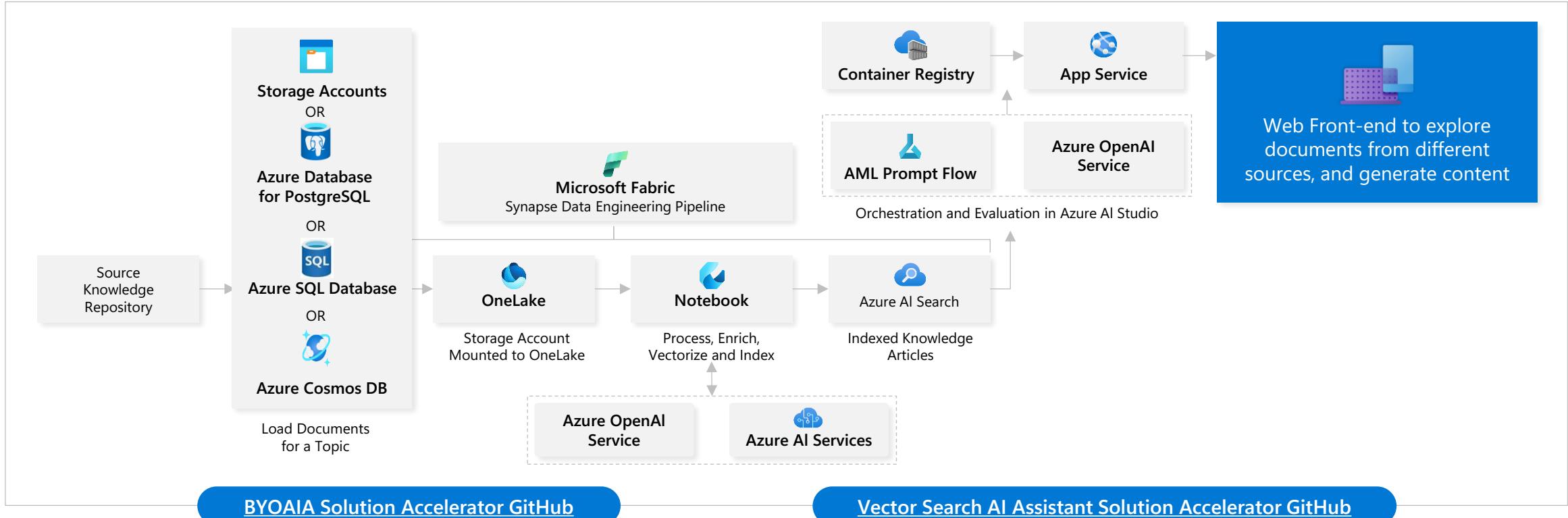


## Description

- Web frontend (either Chainlit or React/Vanilla)
- Integration with GPT-4o RealTime Voice API for best latency and handling of user interruptions
- Leverages agentic design to help assistant address customer asks, routing them to the most appropriate and capable agent
- Uses agents to integrate corporate systems and perform actual operations with LogicApps and/or REST APIs

# Contact Center #1

## Custom copilot



### Custom copilot

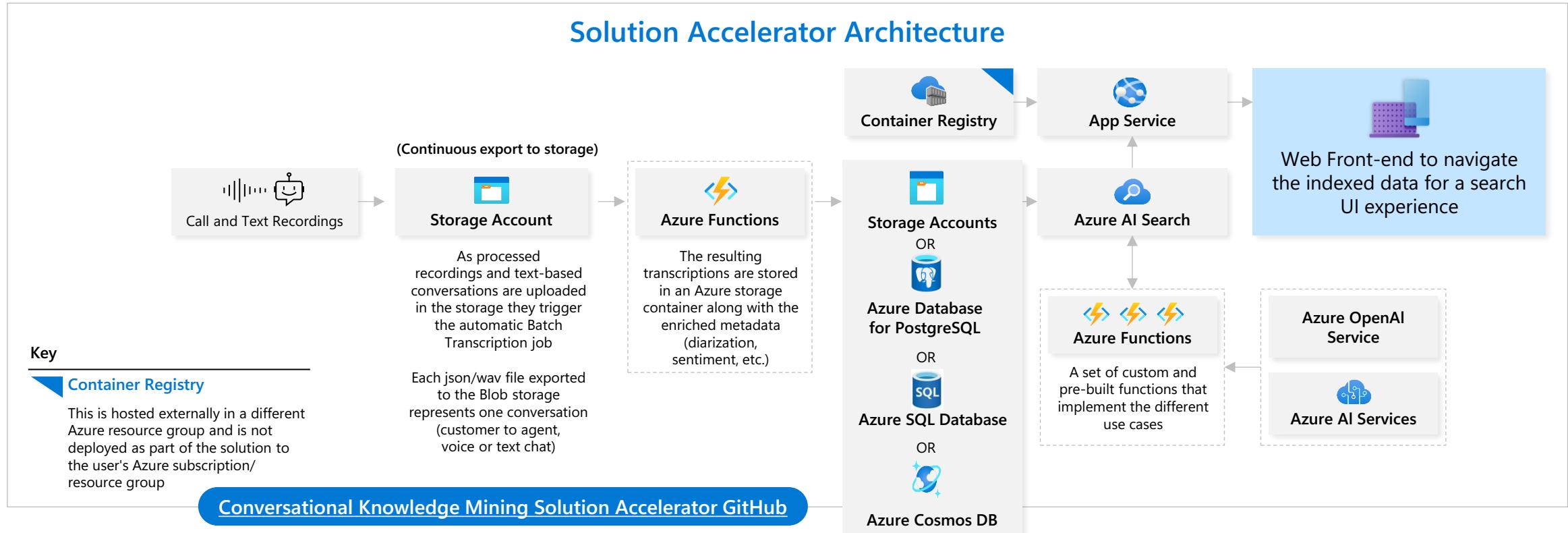
A custom copilot for enterprise applications performs two things:

1. Search/summarize data through a chat UI and/or
2. Task completion within the enterprise app using private or public data.

Data may be read from a variety of data sources, including Storage Account, CosmosDB, Azure Database for PostgreSQL, or Azure SQL. An additional reference architecture can be found in the appendix.

# Contact Center #2

## Conversational Knowledge Mining



### Knowledge Mining

This technical pattern enables customers to conduct automated knowledge mining of human conversations (call center, clinician, meetings) to gain valuable post-interaction insights and patterns.

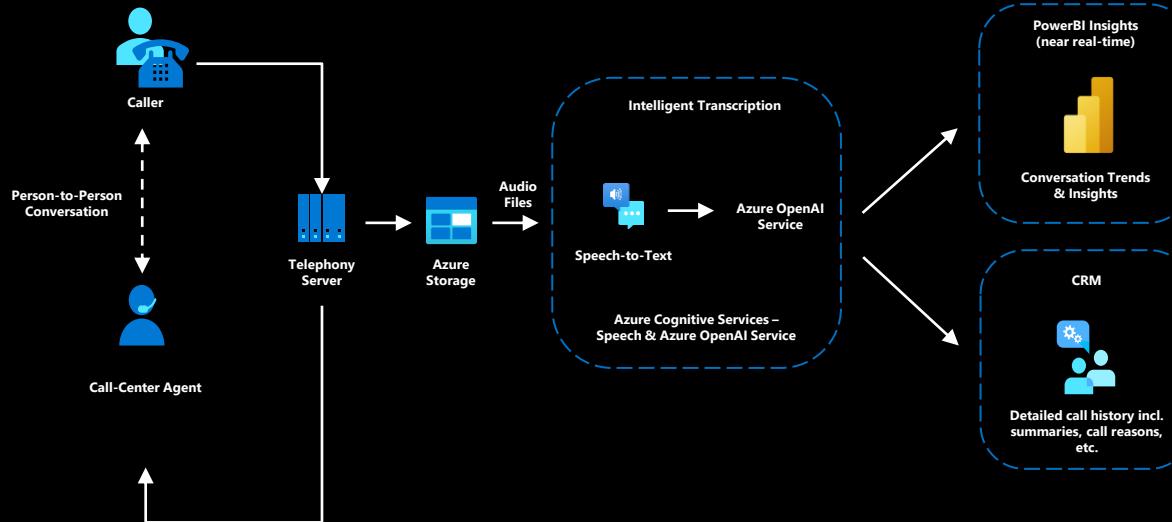
Data may be read from a variety of data sources, including Storage Account, Azure Cosmos DB, Azure Database for PostgreSQL, or Azure SQL



# Customer service and support insights

(e.g., 24/7 customer service)

Handler support (e.g., automated actions and follow-ups)



## Business Impact

- **Improved Customer Experience:** Real-time transcription and analysis of customer interactions allow for quicker and more informed responses to customer needs.
- **Enhanced Efficiency:** Automated transcription and AI-driven insights reduce manual monitoring and increase the efficiency of call handling and follow-ups.
- **Increased Agent Performance:** Insights into conversation trends and specific call issues help in training and coaching agents, leading to improved performance.

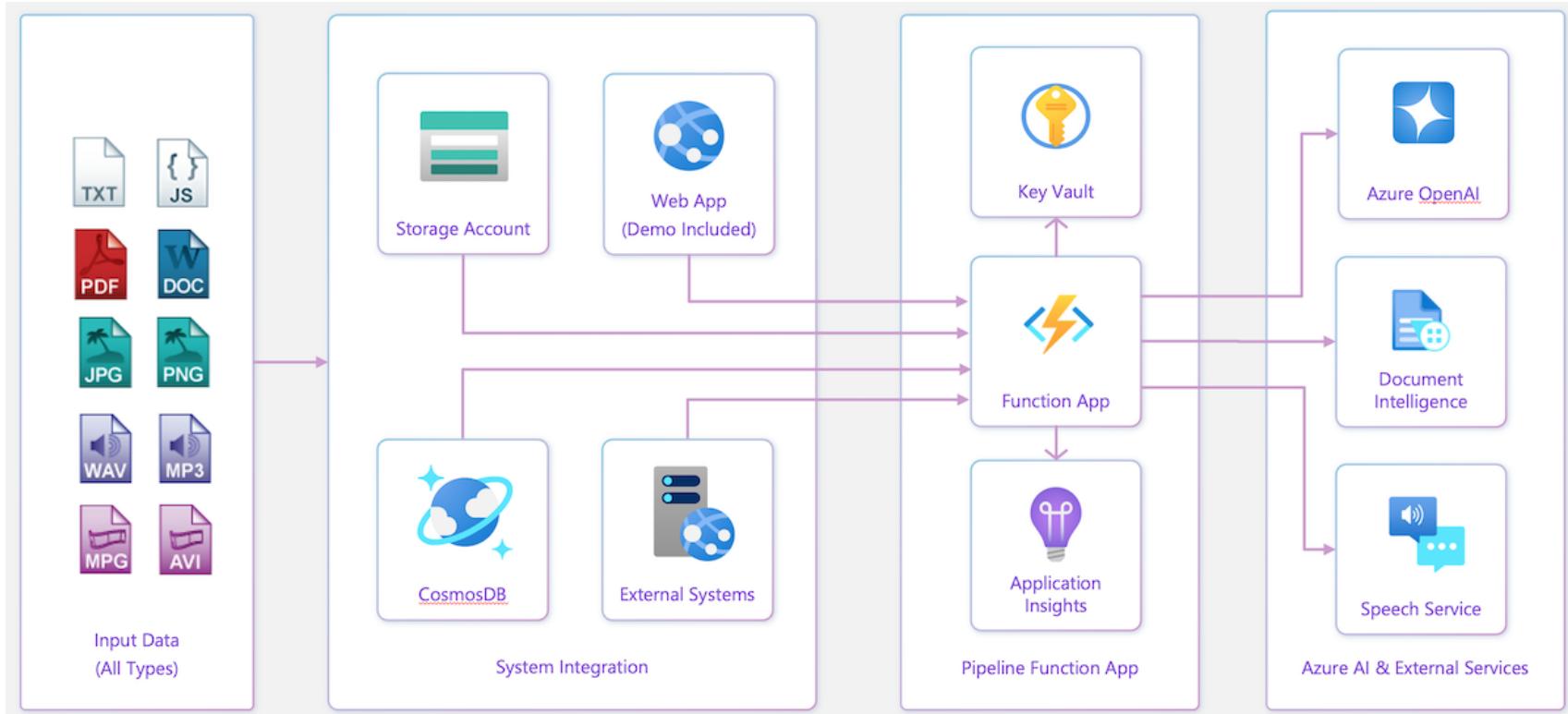
## Business case

- 2 minutes avg saved per call for 100 calls.
- company can **save 41.6% of an FTE's time** annually
- Estimated savings = €12,480 per year per 100 calls/day (with 30k annual salary per FTE)

$$\frac{12k * 1000 \text{ FTE agents}}{12M \text{ EUR saved/year}}$$

**Targeted Persona:** Customer service employees / Customer service managers / Quality assurance

# Azure Multimodal AI + LLM Processing Accelerator



<https://github.com/Azure/multimodal-ai-lm-processing-accelerator>

## Description

This accelerator is a **customizable code template for building and deploying production-grade data processing pipelines** that incorporate Azure AI services and Azure OpenAI/AI Studio LLM models. It uses a variety of data pre-processing and enrichment components to make it easy to build complex, reliable and accurate pipelines that solve real-world use cases. If you'd like to use AI to summarize, classify, extract or enrich your data with structured and reliable outputs, this is the code repository for you.

# TimeGen-1: Generative forecasting

[https://aka.ms/Nixtla\\_TimeGen1](https://aka.ms/Nixtla_TimeGen1)

# Nixtla TimeGEN-1

## Model Value Proposition

1. Deploy TimeGEN-1 with ease through Models-as-a-Service in Azure AI
2. Make forecasting predictions in a matter of seconds using Azure's distributed capabilities
3. Generate accurate results without training using TimeGEN-1's foundational model

Industry Names	North America	West EU	Asia
Manufacturing	Inventory management Production planning	Inventory management Production planning	Inventory management Production planning
Finance	Stock prices Currency exchange rates	Stock prices Macroeconomic development variables Sustainability score	Stock prices GDP growth Currency exchange rates
Retail	Sales forecast Product demand	Sales forecast Product demand	Sales forecast Product demand

Available in US E, France Central.

## Quick links:

- [Tech Community Blog](#)
- [TimeGEN-1 guides and how-tos](#)
- [TimeGEN-1 examples](#)

## Differentiators:

1. Easy to use and deploy, eliminating the need for a dedicated team of machine learning engineers.
2. Fast, make forecasting predictions in seconds, significantly reducing time and computing costs.
3. Accurate forecasts for new time series, no training required.
4. Flexible, fine tune with your own data, or detect anomalies.

Thank you