Overview of Azure Al Services and Their Integration with RAG Applications

By RDP Team









Data & Al Azure

Agenda

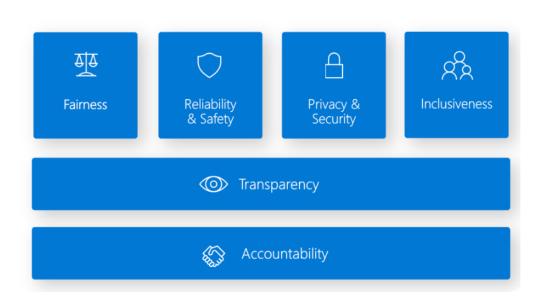
- Introduction to Azure Al Services
- Using Azure Portal with various Azure Al Services
- •Code based demo for working with Azure AI Services
- LLM/RAG introduction and benefits of integrating with Azure AI services



Introduction to Azure Al Services

- •Prebuilt AI models allow for fast development without the need to train your own models.
- •Easily integrate advanced capabilities like natural language processing, computer vision, and speech recognition.
- •Quickly deploy scalable and reliable AI-powered applications without custom model training.
- •Ensures high standards of security, compliance, and performance with minimal setup.



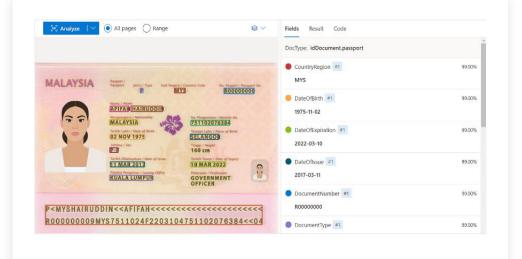


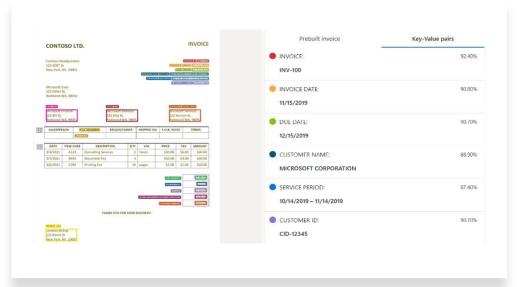
Responsible Al

- •Fairness: Al systems should treat all individuals equitably, avoiding bias or discrimination.
- •Reliability and Safety: Al systems must function consistently and safely across various scenarios.
- •Transparency: Al decisions should be clear and explainable to users and stakeholders.
- Privacy and Security: All systems must protect personal data and comply with privacy regulations.
- •Inclusiveness: Al systems should consider diverse users and ensure accessibility for all.
- •Accountability: Developers and organizations must be responsible for AI outcomes and maintain oversight throughout the system's lifecycle.

Computer Vision: OCR for Document Digitization

- OCR extracts text from printed and handwritten documents, converting them into digital formats.
- Supports multiple languages, making it ideal for global document digitization.
- Useful for digitizing invoices, receipts, and business cards, improving document management.
- Uses deep learning models to ensure high accuracy and scalability for large projects.





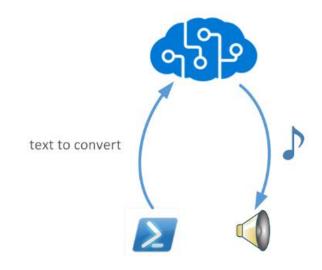
Speech Services: Speechto-Text and Text-to-Speech

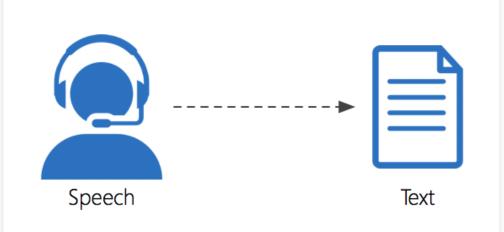
•Speech-to-Text:

•Converts spoken language to text using advanced recognition technology. Supports multiple languages and accents, ideal for transcription and voice commands.

•Text-to-Speech:

•Turns text into natural-sounding speech with prebuilt and custom neural voices. Uses deep neural networks for high-quality synthesis and supports SSML for customization.

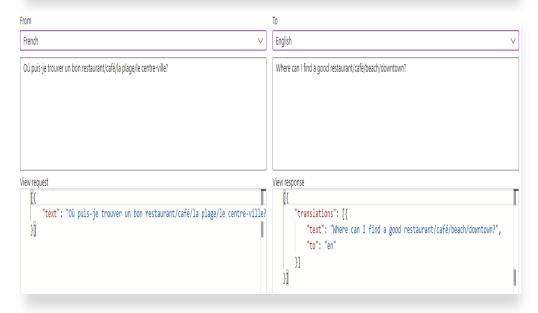






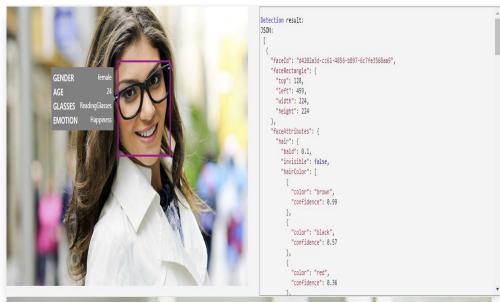
- •Azure Translator provides real-time translation across multiple languages, making it ideal for global communication and multilingual applications.
- •Easily integrates into apps, websites, and services to offer seamless language translation with high accuracy.





Face API

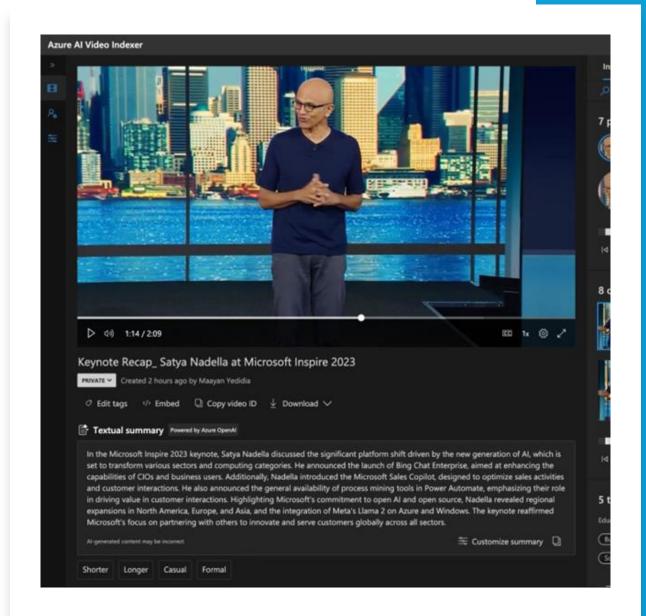
- •Azure Face API detects and analyzes human faces in images, providing features like face recognition, emotion detection, and facial attribute extraction.
- •It enables applications to perform tasks such as identity verification, emotion analysis, and facial authentication with high accuracy.





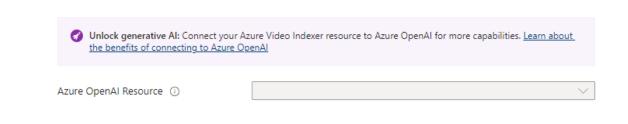
Video Indexer

- •Transcribes speech and translates it into multiple languages with subtitles.
- Detects objects, faces, and emotions within video content.
- •Extracts metadata like keywords, scene segmentation, and sentiment analysis.
- •Performs OCR to recognize and extract text appearing in videos.



Video Indexer

- By enabling AI features with an Azure OpenAI resource, you unlock additional capabilities such as:
- Automated Video Summarization: Generate concise summaries of video content.
- **Enhanced Accuracy**: Benefit from more coherent and human-like summarization and metadata extraction.
- **Customization**: Tailor the tone and length of video summaries to your needs.
- **Broader Use Cases**: Suitable for education, media, and corporate applications.



Text Analytics: Key Phrase Extraction and Entity Recognition

•Key Phrase Extraction:

•Automatically identifies and extracts the main concepts and phrases from unstructured text, helping to summarize and categorize content efficiently.

•Named Entity Recognition (NER):

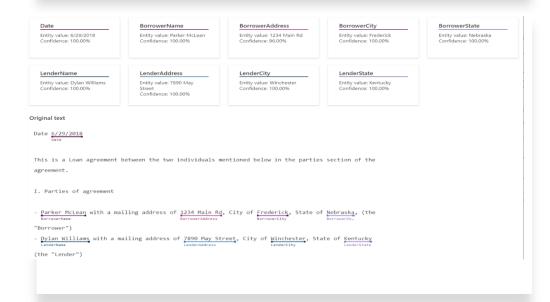
•Detects and categorizes entities such as people, places, and dates within text, aiding in the organization and understanding of large volumes of information.

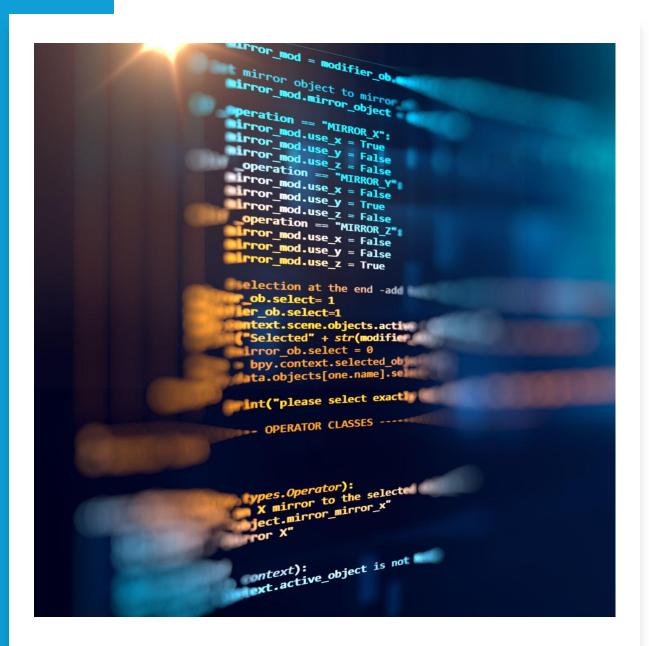
Key phrases

Pike place market, favorite Seattle attraction

Original text

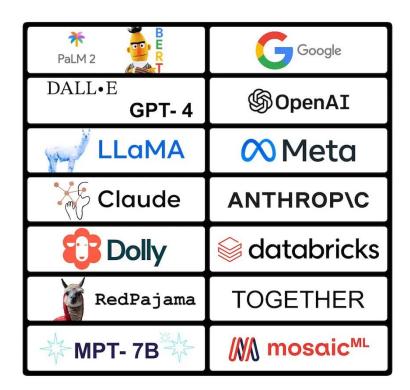
Pike place market is my favorite Seattle attraction.
Key phrase

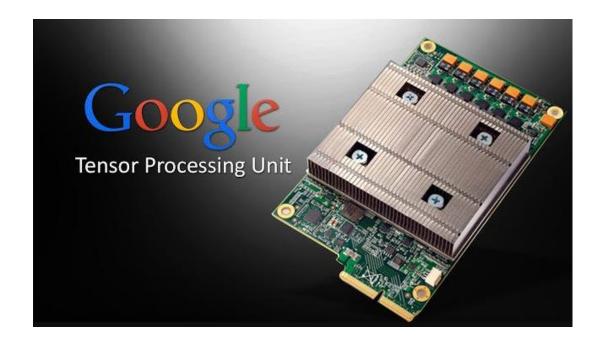




Text Analytics API – Python Code Based Demo







Foundation Models

- It is highly impractical, if not impossible to train your own LLM due to AI expertise needed and the amount of computing power needed.
- It is more practical to use pre-trained models, called Foundation Models from the big tech companies

✓ Response



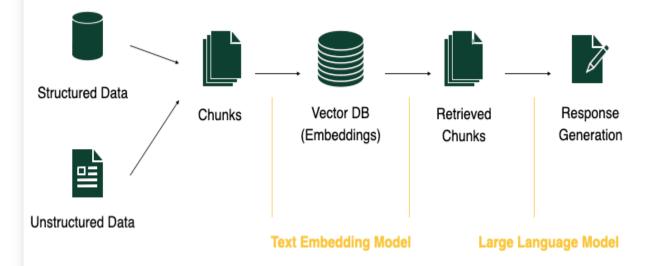
According to the New York Times, Joe Biden won the 2024 Presidential election. Biden received 306 electoral votes, while his opponent, Donald Trump, received 232 electoral votes. Biden also won the popular vote, receiving 74.2 million votes to Trump's 70.5 million votes.

Who was the sole survivor of Titanic?

The sole survivor of the Titanic disaster was Charles Joughin. He was a British baker and the chief baker aboard the Titanic. Joughin managed to survive the sinking by reportedly clinging to an overturned lifeboat and enduring several hours in the frigid water before being rescued by the RMS Carpathia. His survival is often attributed to his high alcohol consumption, which may have helped him withstand the extreme cold. Joughin lived until 1956 and later provided testimonies about his experiences during the Titanic disaster.

LLM Problems - Hallucinations

- LLMs often generate confident answers on topics they weren't trained on, leading to incorrect information (hallucinations).
- RAG helps reduce hallucinations by providing additional, real-time context to ground the LLM's responses.

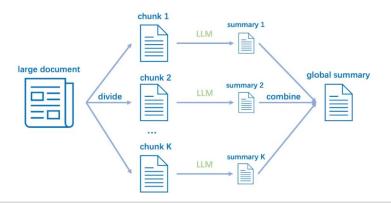


RAG (retrieval-augmented generation)

- Convert Text to Numbers: The input text is turned into a list of numbers (a vector) that represents its meaning.
- **Find Similar Text**: This vector is compared to other vectors in a database using cosine similarity to find similar texts.
- Retrieve Similar Text: The most similar text from the database is retrieved.
- **Use in LLM**: The retrieved text is added to the large language model's context, improving the accuracy of its response.

Chunking

- Chunking breaks long texts into manageable sections
- A common strategy is to split text by a number of words or sentences
- Chunking strategies are actively being researched



Input Text:

- •"Data is becoming a critical asset for businesses. Companies need efficient ways to manage and analyze it. Cloud platforms provide scalable solutions for handling large datasets. Azure offers various services for data storage, processing, and analysis. Machine learning models can extract valuable insights from this data. These insights help drive better decision-making across industries."
- •Chunk 1: "Data is becoming a critical asset for businesses. Companies need efficient ways to manage and analyze it."
- •Chunk 2: "Cloud platforms provide scalable solutions for handling large datasets. Azure offers various services for data storage, processing, and analysis."
- •Chunk 3: "Machine learning models can extract valuable insights from this data. These insights help drive better decision-making across industries."

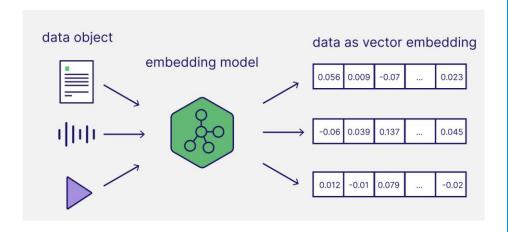
Method	Pros	Cons
TF-IDF	Simple, interpretable	Doesn't capture context, large vectors
Bag of Words	Simple, fast	Ignores grammar and word order
Word2Vec	Captures semantic meaning	Requires large corpus to train well
GloVe	Captures both local and global context	Requires memory for co-occurrence matrix
FastText	Works well with rare words	Slightly more complex than Word2Vec
Doc2Vec	Useful for document-level tasks	More complex to train and fine-tune
LSA	Reduces dimensionality, finds topics	Needs tuning of SVD, may lose some interpretability
BERT	Context-aware embeddings	Computationally intensive

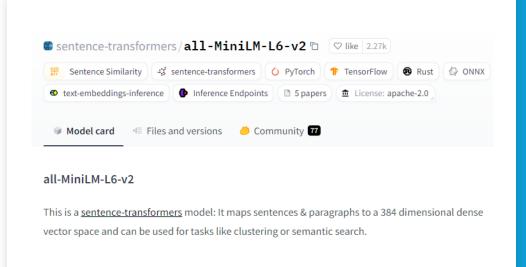
Vector Representation

- You can turn text into numbers (called vectors) using different methods. In LLMs (Large Language Models), an embedding model converts the text into a vector.
- Example:
- Text: "I love cake"
- Vector: [0.01, 0.02, 0.062, ...]
- Text: "Cats are cute"
- Vector: [0.05, 0.08, 0.09, ...]
- Text: "Let's go hiking"
- [0.03, 0.07, 0.11, ...]

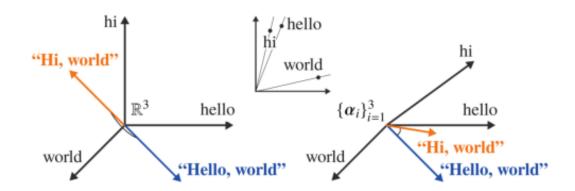
Embedding Model

- •Preferred Approach for LLMs: Embedding models are a common method to convert text into vectors for LLMs.
- •Open Source Availability: You can find open-source embedding models on platforms like Hugging Face.
- •Versatile Use Cases: Embedding models are crucial for tasks like semantic search, clustering, and similarity analysis by transforming text into dense vector spaces.





Cosine Similarity



Cosine Similarity shows how similar two sentences are in meaning by turning them into numbers (vectors).

•Example:

- "I like cats" and "I like dogs" would have a high similarity because they mean almost the same thing.
- "Johnny Bravo" would have low similarity with both because it's unrelated.

The score goes from **0** (completely different) to **1** (exactly the same).

$$\cos(heta) = rac{\mathbf{A} \cdot \mathbf{B}}{\|\mathbf{A}\| \|\mathbf{B}\|} = rac{\sum\limits_{i=1}^n A_i B_i}{\sqrt{\sum\limits_{i=1}^n A_i^2} \sqrt{\sum\limits_{i=1}^n B_i^2}}$$

Vector A = [2,3,4] Vector B = [5,6,7]

Step 1: Dot Product of A and B

$$ec{A} \cdot ec{B} = (2 imes 5) + (3 imes 6) + (4 imes 7) = 10 + 18 + 28 = 56$$

Step 2: Magnitude of A

$$||\vec{A}|| = \sqrt{(2^2 + 3^2 + 4^2)} = \sqrt{4 + 9 + 16} = \sqrt{29} \approx 5.385$$

Step 3: Magnitude of B

$$||\vec{B}|| = \sqrt{(5^2 + 6^2 + 7^2)} = \sqrt{25 + 36 + 49} = \sqrt{110} \approx 10.488$$

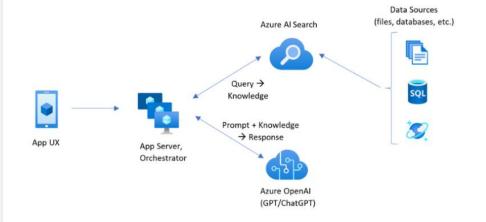
Step 4: Cosine Similarity Calculation

Cosine Similarity =
$$\frac{56}{5.385 \times 10.488} = \frac{56}{56.478} \approx 0.9915$$



Benefits:

- State of the Art Models: Microsoft's partnership with OpenAI ensures access to cutting-edge models.
- All-in-One Platform: Azure provides LLMs via Azure OpenAl, vector databases with Azure Al Search, and storage with Azure Blob.
- **Seamless Integration**: Use other Azure AI services like Document Intelligence to extract and feed data into your vector databases.
- **Scalable & Secure**: Easily scale RAG systems with integrated, enterprise-ready AI services.
- **Production Ready**: Gain other benefits from the Azure ecosystem such as access control, networking, security, and data encryption.



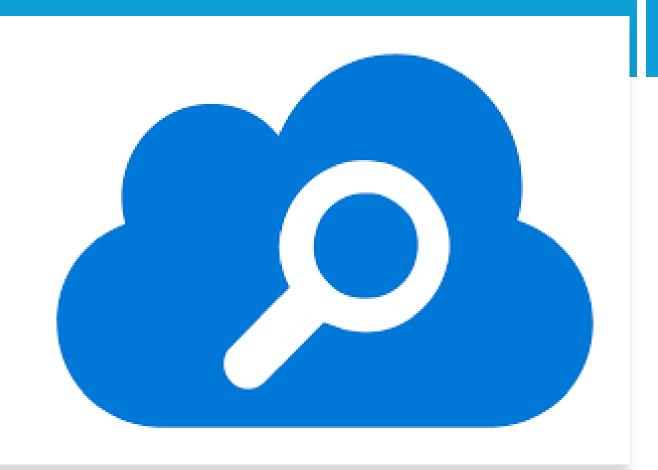
Azure Al Search

Azure Al Search is a vector store for semantic search.

Supports full-text search and vector embeddings.

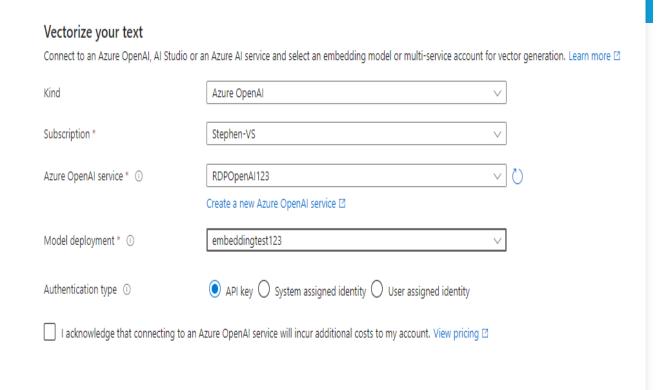
Enables k-nearest neighbor (KNN) similarity search.

Integrates with AI-powered features like OCR and entity recognition.



Azure Al Search

- •Built-in Embedding: Azure Al Search offers a built-in embedding model for vectorizing your data.
- •Time-saving: Eliminates the need for managing your own embedding models, reducing overhead.
- •Seamless Integration: Easily connect to Azure services like Blob Storage, Data Lake, and OneLake for data import and processing.



Search explorer

rdpaisearch

Index vector-1726067000869

🔎 specialists best practices to prepare for the behavioral shift towards increased reliance on mobile Web Development Intern Wisconsin Public Televisio... 💢 Results "@odata.context": "https://rdpaisearch.search.windows.net/indexes('vector-1726067000869')/\$metadata#docs(* "@search.answers": [], "value": ["@search.score": 0.03306011110544205, "@search.rerankerScore": 3.193474292755127, "@search.captions": [10 "text": "specialists best practices to prepare for the behavioral shift towards increased reliance "highlights": "specialists best practices to prepare for the behavioral shift towards increase 14 "chunk_id": "0ee117f73a61_aHR0cHM6Ly9yZHBzdG9yYWdlMTIzNDU2LmJsb2IuY29yZS53aW5kb3dzLm5ldC9yZXN1bWVzLzY1 "parent id": "aHR0cHM6Ly9yZHBzdG9yYWdlMTIzNDU2LmJsb2IuY29yZS53aW5kb3dzLm5ldC9yZXN1bWVzLzY1Zjk4NTZmYTNh 15 "chunk": "specialists best practices to prepare for the behavioral shift towards increased reliance on 16 "title": "65f9856fa3aa7b7a8214fa6b simple-1.png", 18 "text vector": [19 0.0028651152, -0.018100582, 20 21 0.03197816, -0.032144856, 23 -0.006497734, 24 0.019142443, 25 -0.017586598, 26 -0.016308583, 27 -0.0064873155, 28 -0.02186517, 29 0.02725506, 30 0.0030474407, 0.0047057346,

Query options API version 2024-05-01-preview Vector search ① On On Hide vector values in search results Off Semantic ranker ① On On Semantic configuration ① vector-1726067000869-semantic-configuration Create nev Spell correction (Preview) ① Off Query language ① American English (en-... 🗸

Close

Azure Al Search

- You can integrate AI Search with AI Services using AI Search using Skillsets
- This lets you leverage the AI services directly before loading your data into AI Search
- Example: There are documents that contained unstructured data and you need to load the data into Azure AI Search for your LLM/AI application.
 Instead of writing the code for the data extraction, leverage Azure AI Search Skillsets to automatically extract the relevant data and load it into the data store.

Import data

Add enrichments					
Run cognitive skills over a source data field to create additional searchable fields. Learn about additional skills and extensibility here.					
Skillset name * ①					
azureblob-skillset					
✓ Enable OCR and merge all text into merged_content field ①					
Source data field *					
merged_content					
Enrichment granularity level ①					
Source field (default)					
☐ Enable incremental enrichment ①					
Checked items below require a field name.					
Text Cognitive Skills	Parameter		Field name		
Extract people names			people		
Extract organization names			organizations		
Extract location names			locations		
Extract key phrases			keyphrases		
✓ Detect language			language		
Translate text	Target Language English	~	translated_text		
Extract personally identifiable information			pii_entities		
✓ Image Cognitive Skills		Field name			
Generate tags from images		imageTags			
Generate captions from images		imageCaption			

college

Search

Results

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Thank You