VDBMS

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01. Vectors & Vectorisation

What is a vector?

=> An array of numbers:

[0.03712432, 0.030104533, 0.079726376, 0.03848225]

- A vector that represents some data object is called a **vector embedding**.

- **Vectorisation:** The process of converting some data object, e.g. an image, a video, or a JSON document, into a vector embedding. [[0.0618615, -0.534259, 0.545717], [-0.6649279, -0.590766, -0.9606892], [-0.468922, 0.3946947, -0.8946314], [-0.2190834, 0.3915435, 0.5040905], [0.2973618, 0.9794774, 0.7686151], [0.8759577, 0.9395101, 0.7044962], [0.0140189, -0.1538083, -0.5995125], [0.0130325, 0.1164392, -0.2760475], [-0.8062036, -0.0230114, 0.6596168]]

Vectorisation



- In theory: We might want each value in a vector embedding to represent some meaningful attribute of the original object. For instance:



[1920, 1080]



If we were to plot these vectors:





$$d(a,b) = \sqrt{(x_a - x_b)^2 + (y_a - y_b)^2}$$

Vectorisation



- In practice: We use pre-trained models specifically designed to create embeddings. Examples of pre-trained models include:

- + text-embedding-3-small and text-embedding-3-large by OpenAI for text e.g., raw text, code, json, etc.
- + ResNet50 for images
- + ImageBind by Meta AI for images, video, audio, text, depth, thermal and inertial measurement units

These models are either:

- + Open-source and can be run locally as part of a Vector DBMS;
- + Proprietary and only accessible via API.



02. VDBMS

In a Relational DBMS, each record is a row, multiple rows make up a table, and multiple tables make up a database.
In a Vector DBMS, each record is a vector embedding, multiple vector embeddings make up an embedding space.
=> A VDBMS is a DBMS for managing embedding spaces. It supports storing and retrieving vector embeddings, and each record can contain, in addition to the vector embedding, some metadata and the original data, also called a payload.







02. VDBMS

A VDBMS typically consists of:

- A query processor, which supports query types such as:

- + k-nearest neighbor (k-NN) queries
- + Approximate nearest neighbor (ANN) queries
- + Hybrid queries
- + Multi-vector queries
- + Aggregate queries
- A **storage manager**, which **stores** and **indexes vectors** for faster retrieval.
- Optionally, embedding models.

Query Processor							
Interface							
Queries Attribute K-NN ANN Hybrid Data Manip. Range Batch Multi.							
Operators Ins Up Del Sort / Top-K Table Scan Embed Proj Idx Scan Hybrid Scan							
Query Optimizer Query Executor							
Storage Manager							
Search Indexes LSH IVF (k-d) RP (KGraph EFANNA) (NSG) L2H FLANN (ANNOY) FANNG (HNSW) (NGT)							
Vector Storage							



VDBMS Workflow





Popular VDBMSs

	License	First release	Querying with metadata
Pinecone	Proprietary	2021	rich expressions
Chroma	Apache 2.0	2023	rich expressions
Milvus	Apache 2.0	2019	rich expressions
Weaviate	BSD 3-clause/proprietary	2019	supported
Qdrant	Apache 2.0/proprietary	2022	rich expressions
Deep Lake	Apache 2.0/proprietary	2019	rich expressions
	Integration	Querying	Example use-cases
Pinecone	OpenAI, LangChain, others	Java, Python, C#, several others	chatbots, image search
Chroma	LangChain, LlamaIndex	JavaScript, Python, Ruby, others	chatbots
Milvus	OpenAI, LangChain, others	Java, Python, Go, Node.js	chatbots, image/audio/video search
Weaviate	OpenAI, Cohere, PaLM	Java, JavaScript, Python, Go, GraphQL	chatbots, image search
Qdrant	OpenAI, LangChain, others	Python, JavaScript, Go, Rust	chatbots, image search
Deep Lake	LlamaIndex, LangChain	Python, SQL-like TQL	image search



03. Similarity Search

We can perform similarity search on **anything that can be vectorised**:

- Documents
- Images and videos
- Audio

In practice, this means that VDBMSs can be used (and have been used) in:

- Storing and comparing molecular structures
- Recommendation systems
- Facial recognition
- Speech recognition



04. Image Search Demo



- Idea: a meme search app, where the user uploads an image and the app returns similar memes.
- Implementation: We shall use Weaviate to vectorise and store the memes and their embeddings. The embedding model, sentence-transformers/clip-ViT-B-32-multilingual-v1, is open-source and will be run locally.
 The memes will be scraped from knowyourmeme.com,
- Repositories:
- + Backend: https://github.com/ElliotAtHelsinki/memesearch-server
- + Frontend: https://github.com/ElliotAtHelsinki/memesearch-client



Let's scrape the memes

📊 fetch.ts M 🗙



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Let's scrape the memes

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Setting up Weaviate

docker-compose.yml × docker-compose.yml > ... - 0.0.0.0 - '8080' - --scheme - 8080:8080 - 50051:50051 - weaviate data:/var/lib/weaviate restart: on-failure:0 CLIP INFERENCE API: 'http://multi2vec-clip:8080' QUERY DEFAULTS LIMIT: 25 AUTHENTICATION_ANONYMOUS_ACCESS_ENABLED: 'true' ENABLE MODULES: 'multi2vec-clip' CLUSTER HOSTNAME: 'node1' image: cr.weaviate.io/semitechnologies/multi2vec-clip:sentence-transformers-clip-ViT-B-32-multilingual-v1 ENABLE CUDA: '0' weaviate data:

Setting up Weaviate



\$ sudo docker compose up -d
[sudo] password for elliot:
[+] Running 2/0
< Container memesearch-server-weaviate-1 Running
< Container memesearch-server-multi2vec-clip-1 Running</pre>



Vectorising the memes

```
📊 migrate.ts 🗙
src > utils > rs migrate.ts > ...
      const client: WeaviateClient = await weaviate.connectToLocal()
      const collections = await client.collections.listAll()
      if (!collections.find(c => c.name == 'Meme')) {
         await client.collections.create({
          name: 'Meme'.
           vectorizers: vectorizer.multi2VecClip({
            imageFields: ['image'],
            textFields: ['text'],
           properties:
               name: 'image',
              dataType: dataType.BLOB
              name: 'text',
              dataType: dataType.TEXT
      const memeCollection = client.collections.get('Meme')
      const filename = fileURLToPath(import.meta.url)
      const dirname = path.dirname( filename)
      const memesDir = path.join(__dirname, '../../memes')
      const imgFiles = fs.readdirSync(memesDir)
      for (let i = 0; i < imgFiles.length; i++) {</pre>
        const contentsBase64 = await fs.promises.readFile(`${memesDir}/${imgFiles[i]}`, { encoding: 'base64' });
        console.log(await memeCollection.data.insert({ image: contentsBase64, text: imgFiles[i] }))
```

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Querying the memes

meme.ts M 🗙

```
src > resolvers > rs meme.ts > ...
      import { Context } from '@/src/types'
      import { Arg, Ctx, Field, Int, ObjectType, Query, Resolver } from 'type-graphql'
      @ObjectType()
      class Meme {
        @Field(() => String)
        text: string
        @Field(() => String)
        image: string
      @Resolver(Meme)
      export class MemeResolver {
        @Query(() => [Meme])
        async search(
          @Arg('image', () => String, { nullable: true }) image: string,
          @Arg('limit', () => Int, { nullable: true }) limit: number = 1,
          @Ctx() { weaviate }: Context
         ): Promise<Meme[]> {
          const memeCollection = weaviate.collections.get('Meme')
          const searchFileBuffer = Buffer.from(image, 'base64')
          const response = await memeCollection.guery.nearImage(searchFileBuffer, {
            limit,
            returnProperties: ['image', 'text']
          return response.objects.map(o => ({ image: o.properties.image as string, text: o.properties.text as string }))
```

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Querying the memes

<pre>1 query Search(\$limit: Int, \$image: String) { 2 search(limit: \$limit, image: \$image) { 3 image 4 text</pre>							
5 } 6 }							
	5						
Variables Headers Pre-Operation Script Post-Operation Script	≈						
1 []	SON						
<pre>2 "limit": 1, 3 "image": "iVBORw0KGgoAAAANSUhEUgAAAyAAAA00CAYAAABdh1I9AAAABGdBTUEAALGPC/ xhBQAAACBjSFJNAAB6JgAAgIQAAPoAAACA6AAAdTAAAOpgAAA6mAAAF3CculE8AAAABmJLR0QA/wD/AP+gvaeTAACAAELEQVR42uydd7zkVnn +n3Mkzcyte7CXr73ede/2uoDBvWDcjcGhl1BDSwhpJKSQEEIg/AIJhBYIhmBKAFNMsTHYgLvBDTds477eXu/dW2ZGOuf3x9FRG83cmXvnzm3P9/ PRjkajK2mONNr30dsECCGEzHWKQHF/IFjiOHoJIBdrrZcAWAroJVqLxVJiMYCi1ugFtGf+TCWM/ 94F0BfOawC7E9suA3oEEHuEwKBSGBRCDAJqEBDbAbFVCLUxCMQWwNkClDcCGOEpIYSQ +YvgEBBCyJxg0eCucxy9TmuxDtDrhBDrtMY6APvMsPv9LiHwhNbiCUA/IYR4Qgj9h0/7jwB4lqeSEEIoQAghhMyo+3ZxneP467UW67XGcUJgPYClc +T77RECD2itHwTKA1LqB33fvxtprwshhBAKEEIIIVPEEdxztBanKw11guB4wAsmGdjoAE8Cog7AfxaKdwJV08FU0blQQghFCCEEEImR4/ ruicrhX0EwDla4zgAksNSgy8E7tMaPxNC3xIEwS8BDHJYCCGEAoQQQkhjJOCeLCX0FwJnaY0TYZK +SWuUhcCdWuPnSomfANXfAFAcFkIIoQAhhBBiRMcLpMQVAF4KKyR02st2QNwohPphEATXANjFISGEEAoQQgiZb6LjVCn1ywBx0YBVHJKO4WuNX0gpvhUE1e8CZ MYhIYQQChBCCJmjFA+WMngrgFcDWMHxmHYCrXEjxQghhFCAEELIXMJxXffMIMCfCIELea+d6WJEfz4Igu8DqHBICCGEAoQQQmYTq6R0XwuIdwJ6Xw7HrGIXoL</pre>							



Querying the memes



image: '/9i/4AAOSkZJRgABAOEAYABgAAD/40AWRXhpZgAASUkaAAgAAAAAAAAAAAAAAAAAAA/2wBDAAgGBgcGBOgHBwcJCOgKDBONDAsLDBkSEw8UHRofHh0aHBwgJC4nICIsIxwcKDcpLDAxND00Hvc5PTgvPC4zNDL/ tRAAAgEDAwIEAwUFBAQAAAF9AQIDAAQRBRIhMUEGE1FhByJxFDKBkaEII0KxwRVS0FAkM2JyggkKFhcYGRolJicoKSo0NTY30Dk6Q0RFRkdISUpTVFVwV1hZwmNkZwZnaGlqc3R1dnd4eXqDhIWGh4iJipKTlJwwl5iZ CecdcZzwTScktxnEVu+DdT0nR/Fdlf65Y/btNi3+db+Uku/KMF+VyAcMOefStnX9R8E6fos+jeG9Pmv7mbaJNYvflYYN+6THGeVJwDj1g/408ZRXOr6DoVx4R8Kywyz29nJPJpoaZ1LKhYsWwXI5zjr2gW247Acv4y1 PSdY8V3t/odj9h02XZ5Nv5SRbMIob5UJAywJ49awq7P4r2Vpp3xL1e0sbWG1to/J2QwRhEXMKE4UcDkk/jXXeAt18M3Hwc1vU/EN1EyW2oZa4SJfP2qIWEa0eRuPy8EffPTNHMoxTA8eqeys7jUb+2sbSPzLm51WGJMg bnYgAZPA5I617T4CvvDfxEvtT0C58HaRYILUzW81rEBKiggp3P1LZYHIx9K4n4beJv7M17SdK/sPRbv7TqcP+13Vp51xFuZF+R8/LiGRxwSTRzvXTVBY5PWdGv8Aw/qs+16pB5F5Bt8yPerbdyhhyp16EHrVGvZfiV45 /sf4gapYf8It4YvvK8r/AEi+0/zZnzEh+Ztwzi0B7AVh/DPwtaeNvFGratqVnH/Z9mDcyWVohVWdyxWNFH8Aw2BnsBzSU/d5pBY82or3jw3AvifXJNG134WQ6VptyjCG6i014XgIUn55SBn0MAjH0Bg5ry3SNIjsfihY aLdC05jg1q001DoCsoWYIcqc8HB496ane4W0Zor2f4g6p4V8EeKby10rwvpV9qUwjkmF5AGt7UbFARIhgZIBcnjlx16D1b6Tw/q/xqsTosFs+h3Go2aJE1v5cTKfLDjyyBwTuyC0efWiM7q9gscFRXtHj7VPCfgjxZdW 2m+FtLv7+bv5J1uoVNvarsUC00MAAEgFifVh16DP+JfhrTLvfwbe+H90h0+TxFEAIEGxAzeXtvBwP9bg4Haka17abhY8nor3XX9KXwTc2mieH/hxHr0UcaPdahd6c9vZSc7lUhcKenPIGcY4rk/i74Z0/R7rRdX02wbT oNWtTI9iyFDBIoUsCM/KcOoK9ip9aI1E213Cxxmk+GtX1yw1K+06086202Lzrt/MRfLTDHOGIJ4Rume1ZNe9/D7xf/avhLxrd/8ACO+H7P7DYCTyb0y8u04/dzHbKu471+XG00GPrX1+s+0f7Y0gew/4RbwxY+bt/wBI sdP8gZMMD8rbjiOMH2JoiOTbVgOUor01V8A+Nxww8Jaw3Y/NZSH68bOB/sin+Kua8TeBPEHhNt2o2Ra1P3LuA+ZC3T+IdOvfBa1NN2eiA5uiiigEFFFAD4pZIJkmhkaOWNgvOhwVI5BBHO16JafFJdX09NJ8daWmt2K b7/sGSf+jYq898M3sGmeK9Hv7lttvbX0M0rYJwquCTgc9BWVRT5dW+4HtPxF+G/iDxT4ym1/QEttQ07UEiZZo7mMBAI0XJyRkHGcrniqHwn1Sz8MeIvEPhfWrwWc17/oguY5SqpLGXXAcY2k7jhu0Q05FeS0VPI+XlbC 4TakULizto7gg0rbFtmx+leM0VPJt5Bc+gvG0m+J/F08XiL4f+JLgexuI18yzt9SaEo44vFLBV4xkcHIPUmvLfH0jeKtIj03/hKdwe8mmEhigkvmuHt8bd27JIX0R0Jzt9g4+iiMHELng3wbmtLvT/FvhyS7it73V7Hy rXzTgMOkanHaRvBx1wD6Gua1v4YeKfD21Xmp6pZwwWlaVBk+0I3mZcKCoBJ6kd00K46iilaldMD0lfCnhHwaPM8Yap/aeor/AMwiTHvFPpJJxiaDi5Tx/FWb4l+KGs65p/8AZN1HDpGihPLWvtBgFP7nN119hgHPINcR 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text: 'kymdbstyleguide3.JPG'

Moment of truth!



Visit https://memesearch-frontend.elliot-at-helsinki.social/ and upload any image of your choice to test it out! For the image:





Moment of truth!

We get back:



*=

05. Usage with LLMs

a. VecDB as a semantic cache:

- LLM-based chatbots often use external APIs.
- Very often, customers might make the same query

repeatedly, or many customers might ask the same question.

- -> Invoking API calls every time is too costly.
- => Solution: Semantic cache, which stores previous questions and answers.
- => Asked questions would not invoke new API calls.
- => Reduced response time.



VecDB as Semantic Cache



05. Usage with LLMs

b. VecDB to power Retrieval-Augmented Generation (RAG):

Two major problems of LLMs:

- Hallucination
- New influx of information in real-time
- => Solution: VecDB as a dynamic external memory for LLMs.







Q

06. Naïve RAG Demo



- Idea: FoxGPT, which is ChatGPT but equipped with real-time information updates from Fox News

- Implementation: We scrape Fox News articles, vectorise and store them in a **Pinecone** database. This is done daily to ensure that that our GPT has access to the latest articles and news updates. This time, we'll use OpenAI's **text-embedding-3-small** API to vectorise the news articles.

- Repository: https://github.com/elliotathelsinki/foxgpt-backend





Scrape, vectorise, and upsert

```
import { OpenAIEmbeddings } from '@langchain/openai'
import { Pinecone } from '@pinecone-database/pinecone'
import * as cheerio from 'cheerio'
const url = 'https://www.foxnews.com/world/church-england-appears-stop-using-word-church-sound-more-relevant-study'
const scrapedArticleData = await scrape(url)
console.log({ scrapedArticleData })
const openAIClient = new OpenAIEmbeddings({
 apiKey: process.env.OPENAI_API_KEY,
 model: 'text-embedding-3-small',
 dimensions: 1536
1)
const vectorEmbedding = await openAIClient.embedQuery(JSON.stringify(scrapedArticleData))
console.log({ vectorEmbedding })
const pc = new Pinecone({
 apiKey: process.env.PINECONE API KEY
1)
const ns = pc.index(process.env.PINECONE INDEX).namespace(process.env.PINECONE NAMESPACE)
```

await ns.upsert([{ id: url, values: vectorEmbedding, metadata: { payload: JSON.stringify(scrapedArticleData)}}])



Scrape, vectorise, and upsert

—\$ npm run sample

> sample

> tsx --env-file=.env --require reflect-metadata --require tsconfig-paths/register --require dotenv-safe/config src/sample.ts

Fetching https://www.foxnews.com/world/church-england-appears-stop-using-word-church-sound-more-relevant-study (node:35085) [DEP0040] DeprecationWarning: The `punycode` module is deprecated. Please use a userland alternative instead (Use `node --trace-deprecation ...` to show where the warning was created)

scrapedArticleData: {

headline: "Church of England appears to stop using the word 'church' to sound more 'relevant': study",

articleBody: 'The Church of England is apparently dropping the word "church" in the hopes of drawing in more crowds, according to a recent study. The study, called "New Things: A theological investigation into the work of start ches across 11 dioceses in the Church of England," was conducted by the Centre for Church Planting Theology and Research in Durham. The qualitative study inspected language used by 11 Church of England dioceses to describe their ne CATHOLIC BISHOP AND ORTHODOX ARTIST DISCUSS MATERIALISM, SCIENTIFIC ARGUMENTS FOR CHRIST, REUNIFICATION Traditionally, new churches are referred to as "church plants.' In this study, the Centre for church Planting Theology and Research in Durham. The qualitative study inspected language used by 11 dioceses to describe their new churches are referred to as "church plants.' In this study, the Centre for church Planting Theology discover though more than 900 new churches were established by 11 dioceses in the past 10 years, none of them used the word "church plant" on "church.' "Not one diocese used ("fresh expressions' of 'pioneering' in its descriptor," reads the study is conclusion. The study refers to these not-quite-church-plants as "new things," as there was no specificity offerer these "things" are. "That the term 'church' is not used, in favour of other terms (community, congregation etc.) is worthy of thelological reflection. We explore whether the question, 'what is church?' is worth asking.''NY TIMES HYPES to CATHORIA SHOW AND GARG SHOW PARLOR AS ('PLACE TO BE'Six of the 11 dioceses preferred the language of 'worshin' as their main descriptor, and only wor average "community," and only two used "computer ording to the Reverend Dr. Will Foulger. Shows "a misplaced desire to be relevant and modern-sounding". Dr. Foulger admitted in his study that these new language changes are "forcing us to redefine what we think a church is in the Church of E N HERE TO GET THE FOX NEWS APPINE Church of England did not immediately respond to Fox News Digital's request f

datePublished: '2024-08-16T21:55:03-04:00',

dateModified: '2024-08-16T21:55:03-04:00',

description: 'A new study conducted in the UK shows that new "church plants" for the Church of England have avoided the word "church" for the past 10 years or more.'

vectorEmbedding: [

	b			
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0.03315141,	-0.03144648,	0.022137024,	0.021663431,	0.019566096,
0.0054835137,	-0.058725353,	-0.014857244,	0.048847586,	0.027360061,
0.026291097,	-0.014735463,	-0.028226057,	-0.011758602,	0.018050604,
0.0029988494,	0.0005924123,	-0.009201207,	0.031040544,	0.024247888,
-0.021663431,	-0.0040390594,	0.01458662,	0.04692616,	-0.0066133677,
0.0357494,	-0.03677777,	0.01721167,	0.0007687406,	-0.004857696,
0.02382842,	0.0016372737,	-0.026277566,	0.03864507,	-0.051229075,
0.029119115,	-0.0011569165,	0.025249196,	-0.047548592,	0.0047156187,
-0.0033912538,	-0.012638129,	-0.042244367,	0.011582697,	0.005811645,
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-0.04879346,	-0.022001712,	-0.008396102,	-0.014735463,	0.03886157,
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-0.0043198317,	-0.009194442,	0.046303723,	-0.024545575,	-0.0095124245,
0.029092053,	0.04205493,	-0.08470523,	0.024735011,	-0.026155785,
0.028686117,	0.0039206618,	0.00418452,	-0.03523521,	0.03758964,
-0.0029514902,	0.016440393,	-0.044923544,	-0.04170312,	0.029335614,
0.026927063,	0.019972032,	0.03131117,	-0.03896982,	0.026304629,
0.01868657,	0.026805282,	-0.023611922,	-0.029660363,	-0.021000402,
0.0040559736,	-0.032880787,	0.013835639,	0.0596996,	0.024207294,
1436 more :	items			



Scrape, vectorise, and upsert

```
cron.schedule('0 0 * * *', async () => {
    const urls = await fetchNewURLs()
    for (const url of urls) {
        const scrapedData = await scrape(url)
        const embedding = await openAIClient.embedQuery(JSON.stringify(scrapedData))
        await ns.upsert([{ id: url, values: embedding, metadata: { payload: JSON.stringify(scrapedData) } }])
    })
})
```



Setting up a REST endpoint

```
app.post('/query', async (req, res) => {
 let keywords = []
 if (!req.body.keywords || req.body.keywords.length == 0) {
   keywords.push('Latest news')
 else {
   keywords = req.body.keywords
 const embedding = (await embeddings.embedQuery(JSON.stringify(keywords)))
 const result = await ns.query({
   topK: req.body.topK,
   vector: embedding,
   includeValues: true,
   includeMetadata: true
  })
 const payloads = result.matches.map(m => {
   const url = m.id
   const obj = JSON.parse(m.metadata?.payload as string)
   return { ...obj, url }
 res.json(payloads)
})
```



Setting up a REST endpoint

Н	ppscotch • Open source API development ecosystem • Hoppscotch					£) :	_ @ ×
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	POST v https://foxgpt-backend.elliot-at-helsinki.social/query	Send	Save V	0	+ New		() L
	Parameters Body Headers Authorization Pre-request Script Tests		Variables	~			
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	1 * [2 "topK": 5, 3 "keywords": ["ukraine", "middle east", "olympics"] 4]				Impo	rt or create a collection	
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	<pre>1 * [2 * {</pre>	osing ceremony despite banishment over Ukra as neutral individual athletes will be allo not allowed to attend the opening ceremony st Ukraine. Anobsp: 'The IOC Executive Boarc ceremony of the Olympic Games Paris 2024, / because it had invaded Ukraine just four de diered this attack a violation of the Olymp in conflict, starting one week before the AGE ON FOXNEWS.COM The Russian Olympic Com in February.absp: Belarus, which was used t any Belarusian athletes competing as neut from Russia and Belarus combined competed si er, most of them returned home after their two litwipowich and Viuslata Bardrilowekave	when war", weed to attend the y and have not been i has decided that " the IOC said in ays after the end bit truce - a Olympics begin and mittee lost its as a staging trals will also be in Paris - 17 respective events became the first				
П						(2) Help & feedback	<u></u> 4- п п



Defining a GPT Action

src > 💽	
	openapi: 3.1.0
	version: 1.0.0
	the mapping become and the second second
	description: server to query Fox wews articles
	operationId: getArticles
	summary: Get newest news articles from Fox News.
	description: This endpoint retrieves relevant news articles based on keywords to retrieve the top 'k' similar ite
	manual the Andre-
	type: string
	description: The keywords to search for relevant news articles.
	description: The number of top articles to retrieve.
	19997
	200 .
	Bescription: A list of news articles.
	type: object
	description: Title of the news article.
	articleBody
	turns - Atoline
	Spect activity
	description: content of the news article.
	type: string
	description. The time the article was last modified
	description:
	type: string
	description: A short summary of the news article.
	description. The URL at which the article is published
	"ana",
	usscription: invalid request
	15001

Defining a GPT Action



C Some Sector Control Cont	k						⊘Share	Update
د Let your GPT ۱	Edit action: retrieve information or take a Learn more.	S actions outside of Cha	tGPT.			Preview		
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<pre>openapi: 3.1.0 info: title: Fox News API for GPT description: This API allows th database. version: 1.0.0 servers: version: 1.0.0 servers: 1.0.</pre>	e GPT to retrieve the latest hlliot-at-helsinki.social Fox News articles rticles from Fox News. retrieves relevant news arti	: Fox News articles sto	red in a vector to retrieve the Fo	top rrmat		FoxGPT		
Available actions	Mothed Date							
getArticles	POST /que	ery T	est		W Wessage Foxor I			?

Voilà!





Voilà!



=> Visit https://chatgpt.com/g/g-5pzMtoqjN-foxgpt to test it yourself!

AI to the moon!





References



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https://dl.acm.org/doi/pdf/10.1145/3626246.3654691





Thanks!

Do you have any questions? quyanh.nguyen@helsinki.fi



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