

## Zero to Viral: Build TikTok-style Recommendation

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### Al @DevFest





How do you build short videos recommendation when all have you just raw files with no metadata and nothing else?

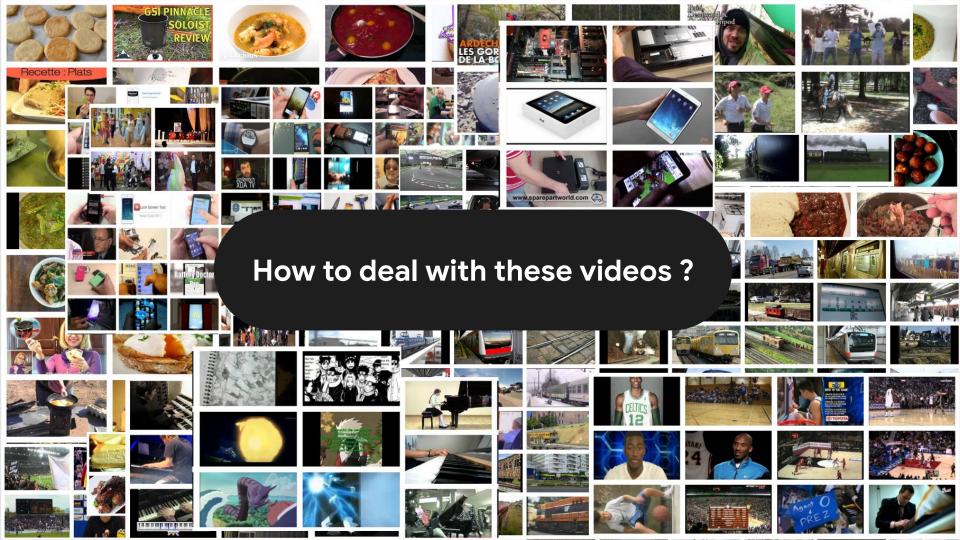


#### **Statistics**



Estimated 1 Billion views per day

34 million TikTok videos are uploaded daily



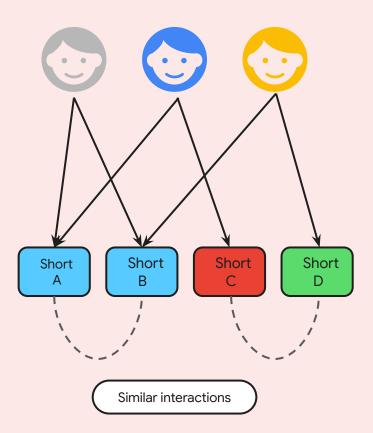
#### What you will get in next 17 minutes

- 1 (Intro: Recommendation System
- (2) (Architecture: Raw Dataset to Final Results
- (3) (Code (Yes, its code!)
- $oldsymbol{4}$  ig) ig( Tips to Scale

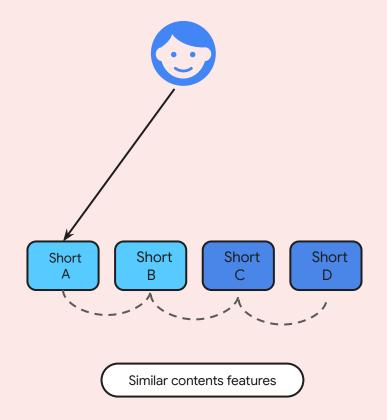
#### A bit about myself

- Build short-videos recommendation at scale in production serving millions of users
- Certified Google Cloud Professional Machine Learning Engineer (PMLE).
- 15 years working experience multi-domains and loves open-source!
- https://github.com/yodiaditya

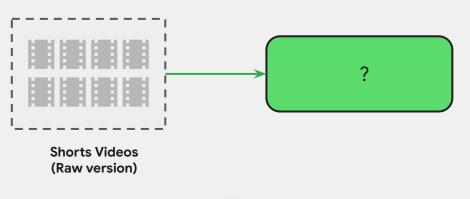
#### Collaborative Filtering



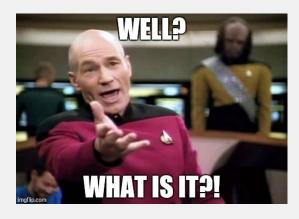
#### Content Based Filtering



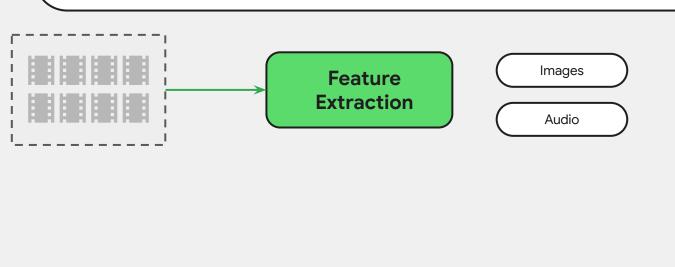
#### Users upload videos to server

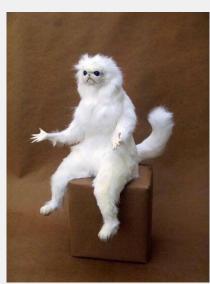






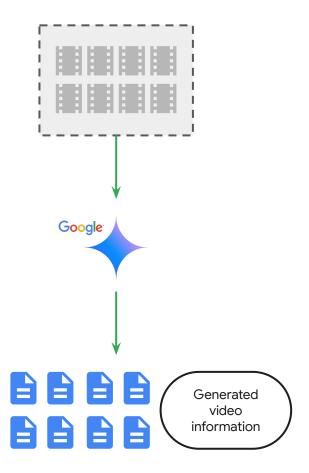
#### First step is to extract the information from the videos





### Gemini: Video Understanding

Model	Video modality details
Gemini 1.5 Flash	Maximum video length:
Go to the Gemini 1.5 Flash model card	<ul> <li>With audio: ~50 minutes</li> <li>Without audio: 60 minutes</li> </ul> Maximum videos per prompt: 10
Gemini 1.5 Pro  Go to the Gemini 1.5 Pro model card	Maximum video length:  • With audio: ~50 minutes  • Without audio: 60 minutes  Maximum videos per prompt: 10
Gemini 1.0 Pro Vision	Maximum video length: 2 minutes
Go to the Gemini 1.0 Pro Vision model card	The maximum videos per prompt: 1  Audio in the video is ignored.



### Next, Embedding Space

#### Supported models 👄

You can get text embeddings by using the following models:

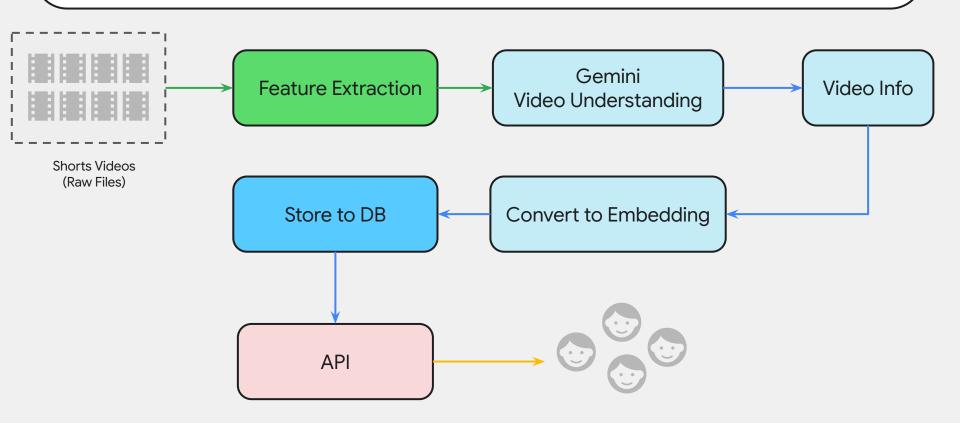
English models	Multilingual models
textembedding-gecko@001	textembedding-gecko-multilingual@001
textembedding-gecko@003	text-multilingual-embedding-002
text-embedding-004	
text-embedding-005	

If you are new to these models, we recommend that you use the latest versions. For English text, use text-embedding-005. For multilingual text, use text-multilingual-embedding-002.





#### Architecture Content-Based Raw Shorts Videos



### Show me the code

```
import vertexai
from vertexai.generative models import GenerativeModel, Part
PROJECT ID = 'YOUR-PROJECT-ID'
vertexai.init(project=PROJECT ID, location="us-central1")
vision model = GenerativeModel("gemini-1.5-flash-002")
# Generate text
response = vision model.generate content(
   Part.from uri(
     "gs://shorts-hdr-dataset/videos/sdr/SDR Animal 45j4.mp4", mime type="video/mp4"
   "Watch each frame in video. Do not make up any information that is not part of the video. Generate title of
video, genres, short description and taggings of content ",
print(response.text)
```



https://media.withyoutube.com/sfv-hdr

Here's a breakdown of the video based on your request:

\*\*Title:\*\* Fluffy Persian Cat

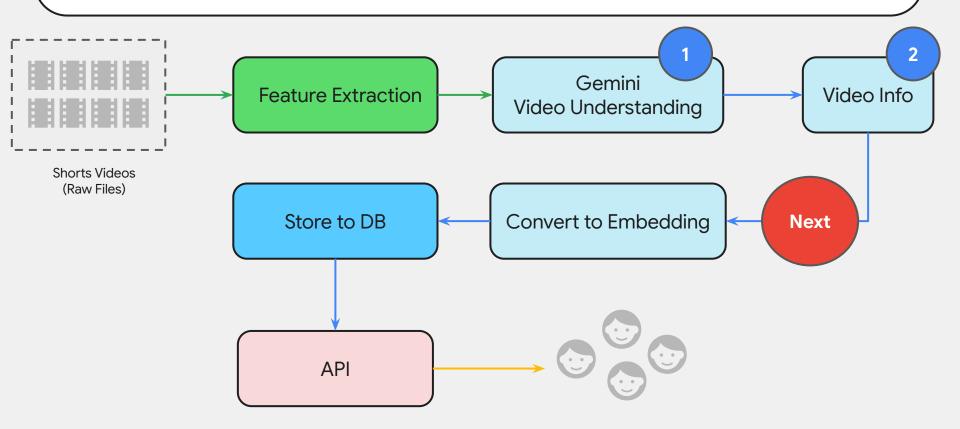
\*\*Genres:\*\* Animals, Cats, Pets

\*\*Short Description:\*\* A beautiful fluffy Persian cat sits on a wooden pallet.

\*\*Taggings:\*\* Persian cat, fluffy cat, cat, kitten, pets, animals, cute, adorable, white cat, long hair cat



#### Architecture Content-Based Raw Shorts Videos



```
import google.generativeai as genai
import os
genai.configure(api_key=os.environ["GEMINI_API_KEY"])
text = f"""
   *Title:** Fluffy Persian Cat **Genres:** Animals, Cats, Pets
   **Short Description:** A beautiful fluffy Persian cat sits on a wooden pallet.
   **Taggings:** Persian cat, fluffy cat, cat, kitten, pets, animals, cute, adorable, white cat, long
hair cat"""
result = genai.embed_content(
   model="models/text-embedding-004",
   content=text)
```

print(str(result['embedding']))

[-0.07675823, 0.027660733, 0.049145754, -0.045824636, 0.061620172, 0.032313757, -0.015517017, -0.03240082, 0.015460015, 0.062071748, -0.03166332, 0.036584534, 0.024837952, 0.009332091, 0.04251558, -0.029856514, 0.05006616, 0.073106274, -0.029902106, -0.025679907, 0.084334895, -0.017889986, 0.059542187, 0.0013102485, -0.04699988,-0.007112105, 0.055407282, 0.018771457, 0.07350461, -0.01813061, 0.04455579, 0.0642487850.00448263, 0.008756887, 0.047846865, -0.025803361, -0.019773666, 0.041618045, 0.051186264, -0.07005315, -0.033514354, -0.026876828, -0.0008525868, 0.08978644, 0.027724743, 0.010746625, -0.061409652, 0.026029717, -0.0034387638, 0.035402797, 0.007589881. -0.023214854. -0.019220363. 0.02619021. -0.026539033. 0.012508128. 0.0040561967. -0.028133305. -0.016629174. 0.020951482. 0.06473735. 0.034282178. 0.051744517. 0.019634075. -0.016341783. 0.021416675. -0.05009744. -0.0076771406. -0.036871992, 0.095764354, -0.02469845, 0.025160229, -0.046339314, 0.024582243,-0.008852839, -0.010245625, 0.011931524, 0.037149847, -0.027476897, 0.021095295. -0.0048002778, 0.012314561, 0.09466355, -0.028999066, -0.010545491, 0.019936696, -0.0075119766, 0.02987802, -0.08925607, 0.0030204963, 0.06398722, -0.005665603,0.055336952. 0.031011213. 0.0803204. -0.046087332. -0.09462911. -0.029559113. 0.017180113. 0.035609562, -0.041688204, 0.016256401, -0.018201571, -0.06340068, 0.05076339, 0.00789683. -0.04663778. 0.003865695. -0.02993782. 0.05526901. -0.059481055. -0.08222241. -0.044425923, 0.03254365, 0.052314047, 0.077546, 0.0024010574, 0.030455906, -0.017964754. -0.03830613. 0.042473085. 0.021389754. 0.031468235. -0.0121501945. -0.010297442. 0.012780487. -0.0086625. -0.019988453. 0.017340286. -0.009815917. 0.068223. 0.02263392. 0.02600014, 0.03895912, -0.01126567, 0.03847954, 0.035594787, -0.08233705, 0.007975588, -0.010537024, -0.042857118, -0.037623975, -0.04778282, 0.007993469, 0.0021377627, -0.046956647, -0.019428788, 0.068178736, -0.002273978, 0.032764167, 0.022261541, -0.050062243. -0.0049963817. -0.0493768. 0.027171357. 0.030386483. 0.035637327. -0.026583953. -0.05276558. -0.051282965. 0.060191035. -0.03913675. 0.009324002. -2.8839599e-05. -0.010929866. -0.0029119896. 0.013345155. -0.013394253. 0.006282062. -0.029159687, -0.023629908, -0.04224124, -0.045065757, -0.023713488, -0.038271822,0.046776716, -0.024885133, -0.05454553, -0.07411608, 0.02653326, 0.01615619, -0.017124163, -0.03665412. -0.07935555. 0.034953605. -0.073878795. -0.035953894. -0.019765098. 0.004043873. 0.024911957. -0.005691729. -0.050333932. -0.000985027. -0.036075074. 0.011225469, -0.016764432, 0.09254729, -0.072244085, -0.037920788, 0.0124474, 0.041128416, -0.018892227, 0.02873513, 0.03394367, 0.005887189, 0.04828047, -0.015023347, -0.010894143,0.003211685, -0.042894274, -0.110083245, 0.013853838, 0.02313884, -0.072523855, 0.024407636, 0.02207754, 0.027719164, -0.01624695, 0.040581483, -0.039446704, 0.062565975. 0.014453987. 0.021910533. -0.019675337. 0.04837492. -0.011042686. -0.02570678. -0.0035492862, -0.012308466, -0.020631982, -0.035574436, 0.042452924, 0.022002371, -0.012221022, -0.039625946, 0.009512582, 0.0071309633, -0.01212419, -0.014330538. 0.06173416, -0.007517297, 0.016374068, 0.08243889, 0.0066845445, 0.053010326, 0.004689015. 0.027499422. 0.05117024. -0.06105194. -0.0407715. -0.006838235. 0.007643424. -0.028924558, 0.031687874, -0.06336678, -0.01165728, -0.003804315, 0.04765609,-0.008533617, 0.006844215, -0.026217818, -0.049941003, -0.010676516, -0.039658975, ...



## Tips to scale

#### Gemini: Video Understanding

#### Limitations

While Gemini multimodal models are powerful in many multimodal use cases, it's important to understand the limitations of the models:

- Content moderation: The models refuse to provide answers on videos that violate our safety policies.
- Non-speech sound recognition: The models that support audio might make mistakes recognizing sound that's not speech.
- **High-speed motion**: The models might make mistakes understanding high-speed motion in video due to the fixed 1 frame per second (fps) sampling rate.
- **Transcription punctuation**: (if using Gemini 1.5 Flash) The models might return transcriptions that don't include punctuation.

```
#!/bin/bash

# Create the output folder if it doesn't exist mkdir -p small files=(*.mp4)

for i in "${!files[@]}"; do file="${files[$i]}" filename=$(basename "$file" .mp4) output="small/${filename}.mp4" gpu=$((i % 2))
```

Sun Nov 24 14:16:48 2024 (Press h for help or a to quit) Driver Version: 550,120 CUDA Driver Version: 12.4 GPU Name Persistence-M Bus-Id Disp.A Volatile Uncorr. ECC Memory-Usage GPU-Util Compute M. Load Average: 3.39 1.96 0.91 AVG GPU MEM: 1.2% CPU: 2.6% MEM: 2.98GiB (2.3%) SWP: 0.00GiB (0.0%) AVG GPU UTL: 10.5% PID USER GPU-MEM %SM %CPU %MEM TIME COMMAND 19387 C dev 549.2MiB 6 263.5 0.2 0:00 ffmpeg -hwaccel cuda -i 10102.mp4 -vf fps=1,scale=1024:1.

CUDA\_VISIBLE\_DEVICES="\$gpu" ffmpeg -hwaccel cuda -i "\$file" -vf
"fps=1,scale=1024:1024:force\_original\_aspect\_ratio=decrease,pad=1024:1024:(ow-iw)/2:(oh-ih)/2" \

-c:v h264\_nvenc -preset slow -cq 30 -c:a aac -b:a 96k -movflags +faststart "\$output" &

echo "Started processing: \$file on GPU \$gpu"

# Wait for processes to complete every two files

if ((i % 2 == 1)); then

wait

fi

done

- Use FFMPEG-GPU NVIDIA
- 2. Reduce size and apply FPS 1

Process 2,6K videos: 31GB to 2.7GB (90% reduction) with FPS 1. 10x faster speed than CPU.

# Use PostgreSQL and PGVector

Enable the extension (do this once in each database where you want to use it)

CREATE EXTENSION vector;

Create a vector column with 3 dimensions

CREATE TABLE items (id bigserial PRIMARY KEY, embedding vector(3));

Insert vectors

INSERT INTO items (embedding) VALUES ('[1,2,3]'), ('[4,5,6]');

Get the nearest neighbors by L2 distance

SELECT \* FROM items ORDER BY embedding <-> '[3,1,2]' LIMIT 5;

Also supports inner product ( <#> ), cosine distance ( <=> ), and L1 distance ( <+> , added in 0.7.0)

Note: <#> returns the negative inner product since Postgres only supports ASC order index scans on

https://github.com/pgvector/pgvector

operators

#### Thank you!

Linkedin
<a href="mailto:linkedin.com/in/yodiaditya">linkedin.com/in/yodiaditya</a>

Github
<a href="https://github.com/yodiaditya/shorts">https://github.com/yodiaditya/shorts</a>

