# **Summary**

## Code

https://github.com/ankushb92/Voltage-Park-Assessment

## **Services**

- API Service (api-service)
- Redis (redis-service)
- Inference worker (inference-worker)
- Inference engine (mochi-service)
- Storage Service (minio-service)
- Web App (web-app-service)

#### **API Service**

Exposes APIs: Submit Job, Enhance Job, Get Job Status, List Jobs, Get Generated Output File

Enhance Job: The default result of submit job is optimized for faster results (by running with a low inference count). If the user likes the result, they can call the Enhance Job API for that job and then we give them a higher quality output of the same video (by generating with a high inference count while keeping the seed and the prompt the same).

## **Redis**

Used as an in-memory queue and also keeps track of the status of jobs.

We have the following queues (using redis lists):

- PENDING\_JOBS
- PROCESSING\_JOBS

We also maintain Job details (using redis hash) as follows:

```
key: <job_id>
mapping: {
    "submitted_at": <timestamp>,
    "prompt": <str>,
    "enhanced": <bool>,
    "seed": <int>,
    "status": CREATED | PENDING | PROCESSING | COMPLETED | FAILED
    "started_processing_at": null | <timestamp>,
    "completed_processing_at": null | <timestamp>,
}
```

(It would be ideal to use a relational DB like Postgres for maintaining Job details, especially for optimally querying the List Jobs API. I went with this approach in the interest of time).

#### Inference worker

Picks up jobs from the Redis queue and runs inference on those jobs.

Using async IO, we spawn multiple async tasks (which we call 'workers') that pick up jobs from the PENDING\_JOBS queue, only if the length of PROCESSING\_JOBS queue is less than a threshold (set to 4 for now).

## Inference engine

I used this docker image I found for the Mochi model: https://replicate.com/genmoai/mochi-1?input=docker

Deployed 2 replicas of the engine with 4 GPUs each.

(I wanted to use the official mochi repo directly and build my own service using it. However, I went with this approach because I was running out of time)

## **Storage Service**

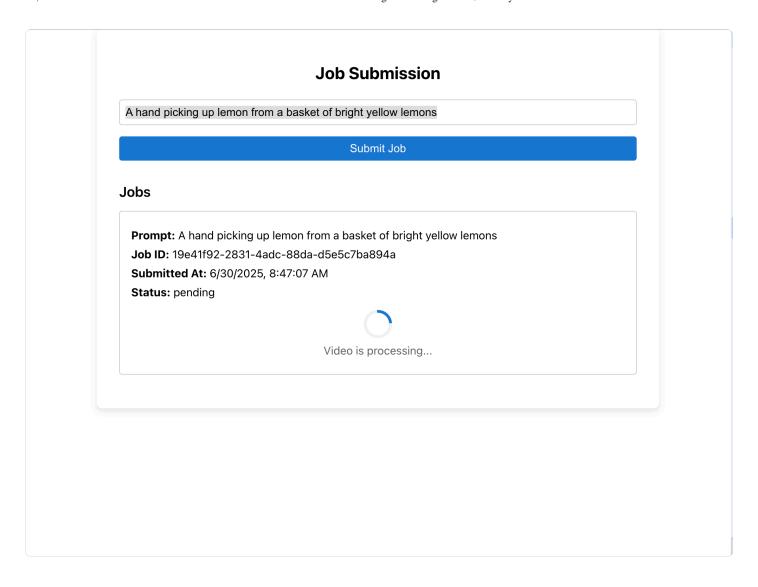
Used minio for object storage. Chose minio because it is S3 API compatible, which I'm familiar with.

Built a PV (persistent volume) of 200 Gi and attached a PVC to the minio deployment. The inference worker saves the model using the minio API and minio can generate links to the file that can be then served to the UI

### Web App

A simple react app that allows user to enter a prompt and calls the submit\_job API.

Once a job is submitted, it polls the <a href="get\_job\_status">get\_job\_status</a> API every 5 seconds until it gets a <a href="COMPLETED">COMPLETED</a> status, and then displays the video. Any completed video can be enhanced with a higher inference count with the enhance button which calls the <a href="enhance">enhance</a> job API.



# Deployment

**Dockerfiles**: deploy/images/<service\_name>/Dockerfile (for each service)

k8s configuration: deploy/k8s/<service\_name>.yaml has configuration for each service

**Build script**: deploy/build\_push.sh : builds docker image and pushes to docker hub

## Considerations/tradeoffs

• Performed benchmarking with a specific prompt (5 second video with inference count set to 100):

8 GPUs: ~2.5 minutes to load and init model. ~2.5 minutes for inference

4 GPUs: ~3 minutes to load and init model. ~4.75 minutes for inference

2 GPUs: Took 4+ minutes to load and init model. ~9 minutes for inference

- Considered creating inference engine replicas with 2 GPU + 6 GPU combination, rather than 4 + 4. This would have led to greater GPU utilization during periods of low traffic.
- Should we reduce the number of inference steps during high load (when PROCESSING\_JOBS queue is long)? This is not a good idea for enhance\_job requests though.
- · Content Safety

• GPU time-sharing: https://docs.nvidia.com/datacenter/cloud-native/gpu-operator/latest/gpu-sharing.html