# Voltage Park — Single Node Text-to-Video API

## Overview

This project implements a **Kubernetes-deployed asynchronous text-to-video API** using **FastAPI**, **Celery**, **Redis**, and GPU-bound **Rust workers** orchestrated by Python.

It satisfies the Voltage Park take-home assignment for a single node with 8x H100 GPUs.

#### Core idea:

- FastAPI serves as the HTTP API gateway.
- Celery manages job orchestration and retries.
- Redis acts as the broker/result backend and status cache.
- Rust GPU workers run the genmo/mochi-1-preview model.
- PVC provides fast local scratch space for video generation.
- MinIO stores completed videos for retrieval via presigned URLs.

## High-Level Architecture

![arch\_diagram](Screenshot 2025-08-08 at 4.11.39 AM LINK)

## Components

- FastAPI API Gateway
  - Async, high-performance Python web framework.
  - Handles:
    - POST /jobs → Create a job, return job\_id
    - GET /jobs/{id} → Status (pending, processing, completed, failed)
    - GET /jobs → Paginated job list
    - GET /jobs/{id}/result → Presigned download URL (MinIO) or direct MP4

#### Scaling:

- Stateless → scale horizontally (HPA) by CPU/memory or RPS.
- No GPU allocation.

### 2. Celery — Job Orchestration

- Mature Python task queue framework.
- Handles retries, routing, and concurrency limits.
- Queue separation:

- short-low quick jobs
- long-high heavy/long-running jobs

#### Scaling:

- HPA based on queue\_depth.
- Workers pinned to GPU resources via Kubernetes.

#### 3. Redis — Broker + Cache

- Fast, in-memory datastore.
- Stores:
  - Celery task queue
  - Task results
  - Job status cache

### Scaling:

- Single instance with AOF persistence for MVP.
- Optional Redis Cluster for high concurrency.

#### 4. Rust GPU Workers

- Long-running processes in GPU pods.
- Host the mochi-1 model for text-to-video.
- Python task (Celery) calls Rust via gRPC (preferred) or PyO3 FFI.

#### **GPU Binding:**

```
resources:
limits:
nvidia.com/gpu: 2
```

• Each worker pod processes tasks sequentially per GPU to avoid memory contention.

## 5. PVC — Persistent Scratch Storage

- Local NVMe-backed (or fast network) persistent volume for:
  - Temporary frames
  - o Partial encodes
- Prevents data loss on pod restarts.

## 6. MinIO — Artifact Storage

- **S3-compatible** object store deployed in the cluster.
- Stores final videos.
- API returns presigned URLs so frontend can download directly without routing through the API pods.

## gRPC API Specification (Python ↔ Rust Workers)

The Rust workers expose a **gRPC service** consumed by the Python Celery tasks. This keeps orchestration in Python while GPU-heavy work stays in Rust.

Proto file: video\_generator.proto

```
syntax = "proto3";
package video;
service VideoGenerator {
  // Submit a text-to-video generation request
  rpc Generate (GenerateRequest) returns (GenerateResponse);
  // Check job status (optional if Celery handles status)
  rpc GetStatus (StatusRequest) returns (StatusResponse);
}
message GenerateRequest {
  string job_id = 1;  // UUID from Celery
string prompt = 2;  // User's text prompt
  int32 resolution_width = 3; // e.g., 1920
  int32 resolution_height = 4; // e.g., 1080
  int32 duration_seconds = 5; // Clip length
                                  // "low", "medium", "high"
 string quality = 6;
}
message GenerateResponse {
  bool accepted = 1;  // Whether the worker accepted the job
string message = 2;  // Info or reason for rejection
message StatusRequest {
 string job_id = 1;
message StatusResponse {
 string job_id = 1;
                               // "pending", "processing",
"completed", "failed"
  string state = 2;
 string output_path = 3;  // Path on PVC or S3 URL
string error_message = 4;  // If failed
}
# gRPC API & Kubernetes Deployment Specification (Python ↔ Rust Workers)
**Goal:** Keep orchestration in Python (FastAPI + Celery) while
offloading GPU-heavy work to Rust workers exposed via **gRPC**. Workers
run in the same pod as the GPU process (or alone) and write
intermediates to a PVC, then upload final artifacts to MinIO.
```

```
## Architecture Overview

```text
Clients —(HTTPS) — FastAPI (video-api)

Redis (task queue / status cache)

Celery worker (Python orchestrator) —(gRPC) —

Rust worker(s)

NVIDIA GPU

PVC scratch (RWO/RWX)

MinIO (S3-compatible)

API serves status & presigned URL
```

#### Flow:

- 1. FastAPI enqueues a Celery job with job\_id and request payload.
- 2. Celery task dials the Rust worker **gRPC** service (Generate) inside the cluster (same namespace).
- 3. Rust worker loads the model (on first call), generates video to a **PVC scratch** path.
- 4. Worker uploads final video to **MinIO** and records job state (Redis update or callback to API).
- 5. API exposes status + presigned URL to clients.

#### **Advantages:**

- Language isolation: Rust can evolve independently of orchestration.
- Clear contracts: gRPC schema is the source of truth.
- Scales cleanly: Add workers; API/Celery remain unchanged.

# gRPC API

proto (tonic-compatible)

```
syntax = "proto3";
package video;

service Generator {
  rpc Generate(GenerateRequest) returns (stream GenerateEvent);
  rpc GetJob(GetJobRequest) returns (GetJobResponse);
}
```

```
message GenerateRequest {
 map<string, string> options = 5; // width, height, fps, seed,
duration, etc.
message GenerateEvent {
 oneof event {
   Progress progress = 1;
   Result result = 2;
   Error error = 3;
 }
}
message Progress {
 string job_id = 1;
 int32 step = 2;
 int32 total_steps = 3;
 string message = 4;
 double gpu_util = 5;  // Optional, for live UX/metrics
message Result {
 string job_id = 1;
 string artifact_path = 2;  // Final local path before upload
string s3_url = 3;  // s3://bucket/key
 string etag = 4;
 int64 bytes = 5;
}
message Error {
string job_id = 1;
 int32 code = 2;
                                  // Application error codes
 string message = 3;
message GetJobRequest { string job_id = 1; }
message GetJobResponse {
  enum Status { UNKNOWN = 0; QUEUED = 1; RUNNING = 2; SUCCEEDED = 3;
FAILED = 4; }
  string job id = 1;
  Status status = 2;
 string s3_url = 3; // filled if SUCCEEDED string error = 4; // filled if FAILED int32 progress = 5; // 0..100
 string error = 4;
}
```

```
use tonic::{Request, Response, Status};
use tonic::codegen::futures_core::Stream;
use tokio_stream::wrappers::ReceiverStream;
pub struct GeneratorSvc { /* state: model cache, redis, minio, etc. */ }
#[tonic::async_trait]
impl video::generator_server::Generator for GeneratorSvc {
    type GenerateStream = ReceiverStream<Result<video::GenerateEvent,</pre>
Status>>;
    async fn Generate(
        &self.
        req: Request<video::GenerateRequest>,
    ) -> Result<Response<Self::GenerateStream>, Status> {
        // load or get cached model
        // spawn task producing Progress/Result/Error events
        // stream back via mpsc channel
        todo!()
    }
    async fn GetJob(
        &self,
        req: Request<video::GetJobRequest>,
    ) -> Result<Response<video::GetJobResponse>, Status> {
        // read from Redis/state store
        todo!()
    }
}
```

## Python client (Celery task)

```
import grpc
from video_pb2 import GenerateRequest
from video_pb2_grpc import GeneratorStub

@celery.task(bind=True, acks_late=True)
def generate_task(self, job_id: str, prompt: str, options: dict):
    chan = grpc.insecure_channel("video-worker:50051") # or mTLS
    stub = GeneratorStub(chan)
    req = GenerateRequest(job_id=job_id, prompt=prompt, model="mochi-1-
preview", scratch_dir="/scratch", options=options)

for event in stub.Generate(req):
    if event.HasField("progress"):
        redis.hset(f"job:{job_id}", mapping={"status": "RUNNING",
"progress": pct(event.progress)})
```

#### **API semantics:**

- **Idempotency:** job\_id is unique; repeated Generate(job\_id) should resume or no-op.
- Deadlines/timeouts: client sets gRPC deadline; server respects and continues in background if desired.
- Backpressure: streaming progress avoids long-polling and supports cancellation.
- **Observability:** include step counts and GPU util in events for metrics.

## **Kubernetes**

The examples assume namespace video and GPU scheduling via the NVIDIA Device Plugin/Operator.

#### Namespace & RBAC

```
aPiVersion: v1
kind: Namespace
metadata:
  name: video
apiVersion: v1
kind: ServiceAccount
metadata:
  name: runtime
  namespace: video
apiVersion: rbac.authorization.k8s.io/v1
kind: Role
metadata:
  name: runtime-role
  namespace: video
rules:
  - apiGroups: [""]
    resources: ["pods", "pods/log", "secrets", "configmaps"]
    verbs: ["get", "list", "watch"]
apiVersion: rbac.authorization.k8s.io/v1
kind: RoleBinding
metadata:
  name: runtime-rb
```

```
namespace: video
subjects:
    - kind: ServiceAccount
    name: runtime
    namespace: video
roleRef:
    apiGroup: rbac.authorization.k8s.io
    kind: Role
    name: runtime-role
```

## Config & Secrets

```
apiVersion: v1
kind: Secret
metadata:
  name: s3-credentials
  namespace: video
stringData:
  AWS_ACCESS_KEY_ID: "minio"
  AWS_SECRET_ACCESS_KEY: "<redacted>"
  AWS_S3_ENDPOINT: "http://minio.video.svc.cluster.local:9000"
 AWS_S3_BUCKET: "videos"
apiVersion: v1
kind: ConfigMap
metadata:
  name: app-config
  namespace: video
  REDIS_URL: "redis://redis-master.video.svc.cluster.local:6379/0"
  SCRATCH_DIR: "/scratch"
  MODEL_NAME: "mochi-1-preview"
```

## PVC (scratch)

```
apiVersion: v1
kind: PersistentVolumeClaim
metadata:
   name: video-scratch
   namespace: video
spec:
   accessModes: ["ReadWriteOnce"]  # Use RWX if multiple pods read
same path
   resources:
    requests:
        storage: 500Gi
   storageClassName: fast-nvme
```

```
apiVersion: apps/v1
kind: Deployment
metadata:
  name: video-api
  namespace: video
spec:
  replicas: 2
  selector:
    matchLabels:
      app: video-api
  template:
    metadata:
      labels:
        app: video-api
      annotations:
        prometheus.io/scrape: "true"
        prometheus.io/port: "8000"
    spec:
      serviceAccountName: runtime
      containers:
        - name: api
          image: your-dockerhub/video-api:latest
            - containerPort: 8000
            - configMapRef: { name: app-config }
            - secretRef: { name: s3-credentials }
          readinessProbe:
            httpGet: { path: /healthz, port: 8000 }
            periodSeconds: 5
          livenessProbe:
            httpGet: { path: /livez, port: 8000 }
            periodSeconds: 10
apiVersion: v1
kind: Service
metadata:
  name: video-api
  namespace: video
spec:
  selector:
    app: video-api
  ports:
    - port: 80
      targetPort: 8000
      protocol: TCP
apiVersion: autoscaling/v2
kind: HorizontalPodAutoscaler
```

```
metadata:
  name: video-api
  namespace: video
spec:
  scaleTargetRef:
    apiVersion: apps/v1
    kind: Deployment
    name: video-api
  minReplicas: 2
  maxReplicas: 10
  metrics:
    - type: Resource
      resource:
        name: cpu
        target:
          type: Utilization
          averageUtilization: 70
```

## Rust Worker Deployment + Service + PDB

```
apiVersion: apps/v1
kind: Deployment
metadata:
  name: video-worker
  namespace: video
spec:
  replicas: 4
  selector:
    matchLabels:
      app: video-worker
  template:
    metadata:
      labels:
        app: video-worker
      annotations:
        prometheus.io/scrape: "true"
        prometheus.io/port: "9464" # if exposing metrics
    spec:
      serviceAccountName: runtime
      nodeSelector:
        nvidia.com/gpu.present: "true"
      tolerations:
        - key: nvidia.com/gpu
          operator: Exists
          effect: NoSchedule
      volumes:
        - name: scratch
          persistentVolumeClaim:
            claimName: video-scratch
      containers:
```

```
- name: worker
          image: your-dockerhub/video-worker:latest
            - containerPort: 50051 # gRPC
          resources:
            limits:
              nvidia.com/gpu: 2
            requests:
              cpu: "2"
              memory: "8Gi"
          volumeMounts:
            - name: scratch
              mountPath: /scratch
          envFrom:
            - configMapRef: { name: app-config }
            - secretRef: { name: s3-credentials }
          readinessProbe:
            tcpSocket: { port: 50051 }
            periodSeconds: 5
          livenessProbe:
            tcpSocket: { port: 50051 }
            periodSeconds: 10
apiVersion: v1
kind: Service
metadata:
  name: video-worker
  namespace: video
spec:
  selector:
    app: video-worker
  ports:
    - port: 50051
      targetPort: 50051
      protocol: TCP
apiVersion: policy/v1
kind: PodDisruptionBudget
metadata:
  name: video-worker-pdb
  namespace: video
spec:
  minAvailable: 75%
  selector:
    matchLabels:
      app: video-worker
```

Network Policies (lock down traffic)

```
apiVersion: networking.k8s.io/v1
kind: NetworkPolicy
metadata:
  name: api-allow
  namespace: video
spec:
  podSelector:
    matchLabels: { app: video-api }
  ingress:
    - from:
                                     # adjust for ingress controller
        - podSelector: {}
      ports:
        - protocol: TCP
          port: 80
apiVersion: networking.k8s.io/v1
kind: NetworkPolicy
metadata:
  name: worker-allow
  namespace: video
spec:
  podSelector:
    matchLabels: { app: video-worker }
  ingress:
    - from:
        - podSelector:
            matchLabels: { app: video-api }
        - podSelector:
            matchLabels: { app: celery } # if separate Celery
deployment
      ports:
        - protocol: TCP
          port: 50051
```

# Deployment Plan (cluster-level)

### 1. Verify cluster access

```
kubectl ——kubeconfig kubeconfig.yml get nodes
```

### 2. Install NVIDIA GPU Operator + device plugin

```
helm repo add nvidia https://helm.ngc.nvidia.com/nvidia
helm repo update
helm install gpu-operator nvidia/gpu-operator -n gpu-operator --
create-namespace
```

#### 3. Install Redis (Bitnami)

```
helm repo add bitnami https://charts.bitnami.com/bitnami
helm install redis bitnami/redis -n video --create-namespace \
--set architecture=replication --set auth.enabled=false
```

### 4. Deploy API

```
kubectl apply -f k8s/api.yaml
```

## 5. Deploy Workers

```
kubectl apply -f k8s/worker.yaml
```

#### 6. Create PVC

```
kubectl apply -f k8s/pvc.yaml
```

## 7. Deploy MinIO

```
helm install minio bitnami/minio -n video \
    --set resources.requests.memory=1Gi \
    --set mode=standalone \
    --set auth.rootUser=minio,auth.rootPassword=<redacted>
```

# **Scaling Guidelines**

Scenario	API replicas	Worker pods	GPUs/pod	Storage Strategy
Many small jobs	4	4	2	PVC → MinIO
Heavy long jobs	2	4	2 (1 task/GPU)	PVC → MinIO
Mixed	HPA on RPS	Separate queues	1–2	PVC per worker
Very large outputs	2-3	4	2	Direct stream to MinIO

## Notes:

• Consider **one task per GPU** to avoid context thrash.

- Use separate Celery queues (e.g., short, long) mapped to different workers.
- For multi-reader workflows, prefer RWX PVCs (e.g., NFS/CSI) or remove PVC and stream directly to MinIO.

## API (FastAPI) – status & presigned URL

```
from fastapi import FastAPI, HTTPException
import redis, boto3, os
r = redis.Redis.from_url(os.environ["REDIS_URL"])
app = FastAPI()
@app.get("/jobs/{job_id}")
def get_job(job_id: str):
    data = r.hgetall(f"job:{job_id}") or {}
    if not data:
        raise HTTPException(404, "unknown job")
    return {k.decode(): v.decode() for k,v in data.items()}
@app.get("/download/{job_id}")
def presign(job_id: str):
    s3 = boto3.client("s3", endpoint_url=os.environ["AWS_S3_ENDPOINT"],
aws_access_key_id=os.environ["AWS_ACCESS_KEY_ID"],
aws secret access key=os.environ["AWS SECRET ACCESS KEY"])
    key = r.hget(f"job:{job_id}", "s3_url")
    if not key:
        raise HTTPException(404, "no artifact")
    bucket = os.environ["AWS_S3_BUCKET"]
    url = s3.generate_presigned_url("get_object", Params={"Bucket":
bucket, "Key": key.decode()}, ExpiresIn=3600)
    return {"url": url}
```

# Security

- mTLS for gRPC between Celery and Rust workers (SPIRE/certs or cert-manager Issuer).
- NetworkPolicy to restrict gRPC to API/Celery only.
- ServiceAccount with minimal RBAC; avoid node-wide permissions.
- Secrets: use Kubernetes Secrets + optional KMS envelope encryption.
- Pod Security: run as non-root, drop CAP\_SYS\_ADMIN, read-only root FS where possible.

```
securityContext:
   runAsNonRoot: true
   runAsUser: 1000
   fsGroup: 1000
   allowPrivilegeEscalation: false
   readOnlyRootFilesystem: true
```

## Observability

- **Metrics** (Prometheus): GPU utilization (DCGM exporter), queue depth, job throughput, success/failure rates, p95 latency, upload time.
- Logs: structured JSON logs from API/worker; forward via Fluent Bit or OpenTelemetry Collector.
- **Tracing**: OpenTelemetry SDK in API and Rust worker (OTLP → collector).
- Alerts: worker crashloop, PVC > 80% full, queue depth > threshold, high GPU idle.

Example Prometheus annotations are shown on Deployments above.

# **Development Workflow**

- 1. Stand up local FastAPI + Celery + Redis.
- 2. Integrate \*\*HuggingFace \*\*`` (guard model weights with .gitignore; use HF\_TOKEN).
- 3. Test generation locally with GPU (Docker + --gpus all).
- 4. Build Docker images for API & worker and push to registry.
- 5. Apply manifests (kubectl apply -f k8s/), then kubectl logs -n video -l app=video-worker.
- 6. Submit a job; verify progress and final artifact in MinIO.

## Optional Frontend (MVP)

- Minimal SPA or HTML with:
  - Text prompt input
  - Submit button → POST /jobs
  - Status polling → GET /jobs/{id}
  - Download with presigned URL → GET /download/{id}

# Hardening Checklist

•

# **Quick Commands**

```
# Port-forward API
kubectl -n video port-forward svc/video-api 8080:80

# Tail worker logs
kubectl -n video logs -f deploy/video-worker -c worker

# Watch GPU pods
kubectl -n video get pods -l app=video-worker -w
```