Low-Level Design (LLD) - ALL Classifier

1 Source Tree

2 FastAPI Specification

Path	Method	Request	Response	HTTP Code
/predict	POST	<pre>multipart/form- data {file: image}</pre>	<pre>{ probabili ties: float[4], predicted _class: int }</pre>	200
/feedbac k	POST	<pre>multipart/form- data { correct_label:s tr, predicted_label :str, file:image }</pre>	{ stored:bo ol, feedback_ count:int }	200
/feedbac k/count	GET	_	{ feedback_ count:int }	200
/metrics	GET	_	Prometheus exposition format	200

/ping GET - {status:" 200 ok"}

Exception Handling

```
@app.exception_handler(Exception)
async def fallback_ex(exc:Exception, req:Request):
    logger.exception("unhandled error")
    return JSONResponse({"detail":"server error"}, status_code=500)
```

3 SQLite Schema

```
id INTEGER PRIMARY KEY AUTOINCREMENT,
ts REAL, -- unix timestamp
correct TEXT, -- ground-truth class
predicted TEXT, -- model prediction at time
img BLOB -- orig. JPEG/PNG bytes
);
```

DB file path: \${FEEDBACK_DIR:-/app/data}/feedback.db (volume-mounted in prod stack for extraction).

4 Airflow DAG - all_classification_pipeline

Task-id	Operator	Callable / Cmd	Key output
preprocess _images	PythonOpe rator	resize_images(.)	resized images → /data/processe d
prepare_da ta	PythonOpe rator	<pre>make_manifest()</pre>	CSV manifests
train_clas sifier	PythonOpe rator	train_classifie r()	run_id (XCom)
test_class ifier	PythonOpe rator	<pre>test_classifier (, run_id)</pre>	metrics
register_m odel	PythonOpe rator	<pre>register_model(, run_id)</pre>	MLflow model reg.

Retrain-on-Feedback Sensor

from airflow.sensors.python import PythonSensor

def enough_feedback(**_):
 conn = sqlite3.connect('/app/data/feedback.db')

```
cnt = conn.execute('SELECT COUNT(*) FROM feedback').fetchone()[0]
return cnt >= 100

wait_feedback = PythonSensor(
    task_id = 'wait_feedback',
    python_callable = enough_feedback,
    poke_interval = 3600, # hourly
    timeout = 86400, # 1 day window
)
    wait_feedback >> preprocess_images

(In dev DAG only; omitted from client-facing stack)
```

5 Monitoring Metrics

Metric	Туре	Labels	Purpose
http_requests_total	Count	method,	requests
	er	handler, status	volume
http_request_duration_s	Histo	handler	latency
econds_bucket	gram		P50/P95
process_cpu_seconds_tot	Count	_	CPU
al	er		usage
<pre>node_memory_MemAvailabl e_bytes</pre>	Gaug e	_	free mem

Grafana-ready dashboard JSON

(monitoring/grafana/dashboards/all-api.json) provisioned on container start.

6 Docker Images

Image	Base	Entrypoint
95		
back end	python:3.9-sli m	uvicorn main:apphost 0.0.0.0 port 8000
stre	python:3.9-sli	streamlit run app.py
amli t	m	server.port 8501 server.address 0.0.0.0
mlfl	python:3.9-sli	mlflow serverbackend-store-uri
OW	m	sqlite:///mlruns.db
prom ethe	<pre>prom/promethe us:latest</pre>	<pre>/bin/prometheusconfig.file=/etc/prometheus/pro</pre>
us		metheus.yml

7 Environment Variables Summary

Variable	Default	Used By	Meaning
BACKEN D_URL	http://back end:8000	Streamlit	API base for requests

FEEDBA CK_DIR	/app/data	Backend	where to put feedback.db
FB_THR ESH	100	Airflow sensor	min samples before retrain

(end of LLD)