Real-time Lead Scoring: Production-Ready Serving Infrastructure (Salesforce Take-Home)

This repo delivers a **production-quality** reference implementation for a real-time lead scoring system using **FastAPI**, **AWS ECS Fargate** (behind an **ALB**), **Amazon SageMaker** (optional endpoint integration), **Snowflake** connectivity patterns, and **Terraform** for IaC. It is designed to handle ~300 RPS with p99 < 1s under typical payload sizes, and demonstrates best practices for **security**, **observability**, **CI/CD**, **testing**, **and MLOps**.

✓ This project is runnable locally via Docker or uvicorn with a light-weight mock model. AWS components are provisioned with Terraform (scoped to a cost-conscious subset by default).

Quick Start (Local)

```
# 1) Create and activate a virtual env (optional)
python3 -m venv .venv && source .venv/bin/activate

# 2) Install dependencies
pip install -r requirements.txt

# 3) Run unit tests
pytest -q

# 4) Start the API
uvicorn app.main:app --host 0.0.0.0 --port 8080 --proxy-headers --
forwarded-allow-ips='*'
# or with Docker:
docker build -t lead-scoring-api:local
docker run -p 8080:8080 lead-scoring-api:local
```

Open: http://localhost:8080/docs for Swagger.

Load Testing with Locust

We have included a load testing setup using Locust.

This allows you to simulate concurrent requests and validate system performance under load.

Setup

```
cd load_tests
pip install -r requirements.txt
locust -f load_tests/locustfile.py --host http://localhost:8080
```

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High-Level Architecture

- Ingress: AWS ALB + AWS WAF (rate limiting + managed rules)
- Compute: ECS Fargate service running a FastAPI app (async) with structured JSON logging
- Model Inference:
 - Option A (default demo): In-process mock XGBoost-compatible scorer
 - Option B (prod): Call a SageMaker real-time endpoint over private VPC endpoint (recommended for heavy models)
- **Data Lake Writes**: Results are pushed asynchronously to **Kinesis Firehose -> S3** (parquet) with partitioning
- Features: 50 features accepted; schema validated with Pydantic
- · Observability:
 - Cloudwatch metrics exposed (/metrics) + CloudWatch Container Insights
 - AWS X-Ray tracing (enabled via SDK and sidecar/daemon)
 - SageMaker Model Monitor (template) for data/quality drift
- Security:
 - Cognito (JWT) or mTLS in private mesh; Secrets in AWS Secrets Manager
 - Least-privilege IAM; ECR image scan; CI security checks (Bandit, Trivy, tfsec, Checkov)

See docs/architecture.md for the diagram and detailed choices.

Endpoints

- POST /score returns a score 1–5 and latency metadata
- GET /metrics Cloudwatch metrics

Sample request:

```
{
  "lead_id": "123",
  "features": {
    "f1": 0.12, "f2": 1.0, "f3": 0.0, "...": 0.3
  }
}
```

CI/CD

- GitHub Actions
 - Lint (ruff), type-check (mypy), unit tests (pytest, coverage)
 - Security: bandit (py), trivy (image), tfsec + checkov (Terraform)
 - Docker build & push to **ECR** (OIDC to AWS)
 - Terraform plan/apply to **staging** on main branch; manual approval for **prod**
- Blue/Green deployment on ECS via new task definition & target group weight shift

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Snowflake Integration (Pattern)

This repo shows a lightweight pattern for obtaining features or writing scored outputs to Snowflake using **Snowflake Connector for Python** with **external authentication** (IAM role + Secrets Manager). In local mode the calls are stubbed.

Scope & Cost Decisions

- **Implemented**: Local runnable service, ECS task/service Terraform, ECR, ALB, WAF (basic), CW logs/alarms, Kinesis Firehose to S3 (delivery stream), CI security gates.
- **Documented templates**: SageMaker endpoint + Model Monitor, Cognito integration.
- **Deferred** (explain in docs/scope_decisions.md): PrivateLink to Snowflake, end-to-end Lakehouse table creation, full Grafana stack (CloudWatch metrics suffice for demo).

Future Improvements

- Canary deployments with AWS App Mesh
- Online feature store (Feast / SageMaker Feature Store)

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• Advanced drift monitors and bias metrics with alerts to Slack (SNS)