Kyra Works Internship Onboarding Task

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Build predictive analytics for theft prevention

Task: Define Basic Log Aggregation Strategy Problem: Outline a strategy for collecting and centralizing logs from different system components. Expected Outcome: Document describing the log aggregation approach. Tools: Text editor. Research: Log management. Consider: Sources, format, storage.

Objective

To collect, centralize, and standardize logs from multiple system components (e.g., applications, servers, databases, IoT devices, and security systems) to enable predictive analytics for theft prevention.

1. Sources of Logs

Identify where important information comes from:

- Application logs: User behavior, failed login attempts, accesses of sensitive areas.
- System logs: OS-level events, authentication logins, process runs.
- Database logs: Execution of queries, data access patterns.
- IoT/sensor logs: Access control, motion sensors, camera operations.
- Network logs: Firewall, VPN, router traffic, anomaly events.

2. Log Format Standardization

- Use **structured formats** like JSON or CSV to ensure logs are machine-readable.
- Include **essential fields** for predictive analysis:
 - o Timestamp (with timezone)
 - o Source/component
 - Event type (e.g., login, motion detected)
 - o User or device ID
 - o Metadata (IP address, location, access point)

3. Log Collection & Transmission

- Use **lightweight agents** installed on source systems to forward logs.
- Transport logs securely (TLS/SSL) to central storage.
- Examples:
 - o Fluentd/Logstash: For transforming and shipping logs.
 - o **Filebeat**: Lightweight shipper for system/application logs.

4. Centralized Storage

- Central log server or cloud storage:
 - Elasticsearch, Splunk, or cloud storage options (AWS S3 + Athena, GCP Logging).
- Make sure indexing and searchability to query historical patterns in a timely manner.
- Use retention policies for historical logs (e.g., 90 days, as per regulations).

5. Preprocessing & Normalization

- Parse logs into a **common schema**.
- Remove redundant or irrelevant fields.
- Anonymize sensitive data (for compliance) if needed.

6. Analytics & Alerting

- Feed centralized logs to predictive models to detect anomalies (e.g., unusual access patterns, repeated failed logins, suspicious sensor events).
- Set up real-time alerting on suspicious events.

7. Security & Compliance

- Ensure logs are **tamper-proof** (write-once storage or checksum verification).
- Restrict access to logs to authorized personnel only.
- Comply with local privacy/data regulations.

8. Implementation Tools Overview

Task Tool/Method

Log collection Filebeat, Fluentd, Logstash

Storage Elasticsearch, Splunk, AWS S3, GCP Logging

Processing & Normalization Logstash, custom Python scripts

Analytics Python (Pandas, Scikit-learn), MLflow Alerting Kibana alerts, Grafana, custom scripts

Conclusion

A basic log aggregation method is to collect logs from multiple sources, normalize formats, store securely centrally, and preprocess for analysis. The setup accommodates predictive analytics for theft detection where the system is able to detect anomalies and send notifications in advance. Strong indexing, security, and compliance are essential to preserve integrity and use of logs.

Build model update/deployment automation

Task: Store Feedback Data Problem: Design and implement storage for the alert feedback data. Expected Outcome: Database schema/table for feedback records. Tools: Database. Research: Data modeling.

Consider: Linking to original alert, user info.

Database Schema Design

We need to capture:

- **Original alert** details (linking back to alerts generated by the system).
- User feedback (confirmation, false positive, severity, comments).
- User info (who gave the feedback).
- **Timestamps** (when feedback was provided).

Tables

1. alerts

```
Holds all system-generated alerts.
CREATE TABLE alerts (
  alert id SERIAL PRIMARY KEY,
  alert type VARCHAR(100),
  alert message TEXT,
  created at TIMESTAMP DEFAULT CURRENT TIMESTAMP);
```

2. users

```
Tracks users who can provide feedback.
CREATE TABLE users (
  user id SERIAL PRIMARY KEY,
 username VARCHAR(100) UNIQUE NOT NULL,
  email VARCHAR(150),
 role VARCHAR(50),
  created at TIMESTAMP DEFAULT CURRENT TIMESTAMP);
```

3. alert feedback

```
Stores feedback on alerts.
CREATE TABLE alert feedback (
  feedback id SERIAL PRIMARY KEY,
  alert id INT NOT NULL,
  user id INT NOT NULL,
  feedback status VARCHAR(50) CHECK (feedback status IN
('true positive', 'false positive', 'false negative', 'uncertain')),
  severity level INT CHECK (severity level BETWEEN 1 AND 5),
  comments TEXT,
  created at TIMESTAMP DEFAULT CURRENT TIMESTAMP,
  FOREIGN KEY (alert id) REFERENCES alerts(alert id) ON DELETE CASCADE,
  FOREIGN KEY (user id) REFERENCES users (user id) ON DELETE SET NULL);
```

Example Workflow

- 1. System raises an alert \rightarrow stored in alerts.
- 2. User reviews alert → stores feedback in alert feedback.
- 3. Later, feedback data can be used for:
 - o Measuring false positive/negative rates.
 - o Training improved alert classification models.
 - o **Prioritizing retraining** based on severity/uncertainty.

Example Insert

-- Insert sample alert

INSERT INTO alerts (alert_type, alert_message)

VALUES ('Intrusion', 'Suspicious login from unknown IP');

-- Insert sample user

INSERT INTO users (username, email, role)

VALUES ('alice', 'alice@example.com', 'analyst');

-- Insert feedback

INSERT INTO alert_feedback (alert_id, user_id, feedback_status, severity_level, comments) VALUES (1, 1, 'false positive', 2, 'Login was from a trusted VPN');

Outcome

This schema ensures **feedback data is properly linked** to alerts and users. It supports auditability, enables tracking model performance drift, and provides structured data for retraining models in deployment automation.

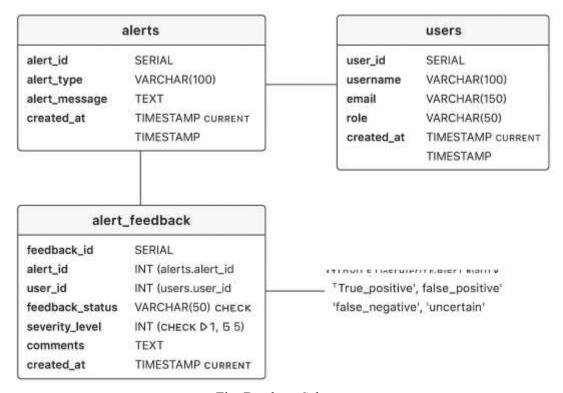


Fig: Database Schema