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## Capstone Project:

# Smart Manufacturing: Fuel Efficiency and Turbine Health Analytics on Azure Cloud

## Business Scenario

Modern manufacturing environments—especially in sectors like shipping, energy, and aerospace—are equipped with industrial machines like gas turbines that continuously generate high-frequency telemetry data from embedded sensors. These sensors track vital parameters such as torque, pressure, fuel flow, and decay coefficients, which are critical to monitoring machine health and optimizing fuel efficiency.

However, this raw sensor data is often siloed in edge devices or local storage. There is a growing need to centralize this data in the cloud to build scalable, analytics-ready platforms that support:

- **Remote monitoring of turbine health**
- **Anomaly detection and efficiency benchmarking**
- **Automated data ingestion and transformation pipelines**
- **Real-time API-based access to machine KPIs**
- **Cloud-based storage and deployment for reliability and scale**

This capstone simulates the journey of building a **cloud-native monitoring platform** using **Azure services (ADLS, ADF, Databricks)** and **FastAPI**, integrating software engineering best practices and analytics to generate operational insights.

## Project Objectives

1. Ingest and transform turbine sensor data using Python and SQL.
2. Design SRS, HLD, and UML diagrams for a cloud-based turbine monitoring system.
3. Build modular ETL pipelines that clean, transform, and store telemetry data.
4. Develop REST APIs using FastAPI to access turbine health KPIs.
5. Perform exploratory data analysis (EDA) and visualize trends.
6. Deploy data pipelines and dashboards on Azure (ADLS, ADF, Power BI).
7. Build unit tests and deployment readiness documentation.

## Project Dataset

### Sensor Telemetry Dataset (CSV format)

| Field Name | Description                                |
|------------|--|
| Lp         | Lever position                             |
| V          | Ship speed (knots)                         |
| GTT        | Gas Turbine shaft torque (kN·m)            |
| GTn        | Gas Turbine revolutions per minute (rpm)   |
| GGn        | Gas Generator revolutions per minute (rpm) |
| Ts         | Starboard Propeller Torque (kN)            |

|                     |  |
|---------------------|--|
| Tp                  | Port Propeller Torque (kN)             |
| T48                 | HP Turbine exit temperature (°C)       |
| T1                  | Compressor inlet air temperature (°C)  |
| T2                  | Compressor outlet air temperature (°C) |
| P48                 | HP Turbine exit pressure (bar)         |
| P1                  | Compressor inlet pressure (bar)        |
| P2                  | Compressor outlet pressure (bar)       |
| Pexh                | Exhaust gas pressure (bar)             |
| TIC                 | Turbine Injection Control (%)          |
| mf                  | Fuel flow rate (kg/s)                  |
| decay_coeff_comp    | Compressor decay coefficient           |
| decay_coeff_turbine | Turbine decay coefficient              |

## Capstone Phase with learner Tasks & Deliverables

### Phase 1: Documentation & Architecture (SDLC, SRS, HLD, UML)

#### Tasks:

- Draft **Software Requirements Specification (SRS)** for the turbine health platform.
- Define **functional and non-functional requirements**.
- Create **High-Level Design (HLD)** to represent modules (Ingestion, ETL, Monitoring, API).
- Generate **UML diagrams**:
  - **Use Case Diagram** – turbine data ingestion, anomaly detection, API access.
  - **Deployment Diagram** – Azure VMs, ADLS, ADF, FastAPI integration.
  - **Component Diagram** – ingestion, transformation, visualization.

### Phase 2: Data Ingestion & Transformation (Python, Pandas, NumPy)

#### Tasks:

- Write Python code to **ingest CSV sensor logs**.
- Perform **data cleaning** (nulls, outliers, smoothing).
- Compute **derived features** (e.g., temperature-pressure ratios, torque differentials).
- Use **Pandas and NumPy** for initial EDA.

### Phase 3: SQL & Exploratory Analysis

### Tasks:

- Create SQL tables: sensor\_readings, alerts, turbine\_metadata.
- Use **SQL for transformations**: joins, filters, aggregations.
- Analyze:
  - Fuel usage patterns over time
  - Turbine efficiency comparisons
- Visualize results using **Matplotlib / Seaborn**.

## Phase 4: ETL and Azure Integration (ADF, ADLS)

### Tasks:

- Simulate **ETL pipeline** using Python scripts and Azure Data Factory:
  - Extract: Load CSVs from local storage to **Azure Data Lake Gen2**.
  - Transform: Cleaned sensor logs processed into **silver layer**.
  - Load: Store curated data into structured Azure storage (simulated with CSV or SQL).
- Integrate sensor data into **hierarchical ADLS folders**.

## Phase 5: REST APIs using FastAPI

### Tasks:

- Build **FastAPI-based endpoints**:
  - GET /health-summary – aggregate fuel usage & decay status.
  - GET /sensor-metrics – recent telemetry for a turbine.
  - POST /anomaly-alerts – log new anomaly detection.
- Use **Pydantic** for schema validation.
- Expose Swagger documentation and enable file uploads.

## Phase 6: Testing & Release Readiness

### Tasks:

- Write **unit tests** using pytest for:
  - Data ingestion
  - API endpoints
  - KPI calculations
- Prepare **Test Summary Report**:
  - Pass/Fail table
  - Defect logs
- Define **Acceptance Criteria**:
  - Data load time thresholds
  - API uptime
  - Alert trigger latency