#### Unit I — Introduction to IoT: Smart Healthcare

#### Problem overview

Design a scalable, secure IoT ecosystem for remote patient monitoring, asset tracking, and clinical decision support.

# **Key challenges**

- **Data confidentiality & privacy:** HIPAA-like requirements, encryption at-rest and in-transit, anonymization for analytics.
- Latency and reliability: Critical alerts (fall detection, arrhythmia) require low-latency paths and fault-tolerant delivery.
- **Heterogeneity:** Multiple device types (wearables, bedside monitors, imaging systems) and protocols (BLE, Wi-Fi, LoRaWAN).
- **Scalability:** Tens of thousands of endpoints and high-frequency telemetry (e.g., ECG streams).
- Interoperability & standards: FHIR for clinical data, HL7 messaging for legacy systems.
- Regulatory & auditability: Secure logging, device attestations, firmware provenance.

## **Architectural requirements**

- 1. **Device layer:** Constrained devices and gateways. Support secure boot, TPM/secure elements.
- 2. **Edge/fog layer:** Local preprocessing, real-time analytics, anomaly detection, and transient data caching to reduce cloud cost and latency.
- 3. **Communication layer:** Transport security (TLS/DTLS), MQTT for telemetry, CoAP for constrained devices, HTTP/REST for management.
- 4. **Cloud platform layer:** Device management, identity & access management (IAM), long-term storage, analytics, and ML model hosting.
- 5. **Application layer:** Clinical dashboards, alerting, EHR integration (FHIR), role-based access for clinicians.

# **Enabling technologies & APIs**

- MQTT/AMQP for pub/sub telemetry; MQTT TLS + client certs for secure mutual auth.
- CoAP/DTLS for constrained devices.
- REST APIs for backend services and EHR integration (use FHIR RESTful endpoints).
- OAuth2 / OpenID Connect for user & app authentication.
- Edge toolkits: TensorFlow Lite for on-device inference; Apache NiFi/StreamSets for data flow.

# **Domain-specific concepts**

• Patient identity mapping (link device IDs to patient IDs securely, tokenization).

- Clinical-grade SLAs (guarantees for alert delivery times).
- Quality of Service (QoS) levels in MQTT to prioritize critical messages.



### Unit II — Infrastructure & Service Discovery: LPWAN in Smart Waste Management

#### Scenario

City-wide network to monitor fill-level sensors on waste bins, route optimization, and scheduled collection.

### Why LPWAN?

- Long range & low power: Devices operate years on battery and cover wide urban/rural areas.
- Cost-effective for low-bandwidth telemetry (fill-level, battery status, tamper alerts).

### Comparison: LoRaWAN vs 6LoWPAN

- **LoRaWAN**: Star-of-stars topology with gateways; adaptive data rate; good for sparse uplink telemetry; network servers handle deduplication, ADR; not IP-native.
- **6LoWPAN**: IPv6 over low-power radios (e.g., IEEE 802.15.4); supports mesh; IP-native allowing REST/CoAP directly to endpoints; better where mesh is needed.

# Layered protocol architecture (example with LoRaWAN)

- Application layer: JSON/CBOR payloads, application server processes telemetry.
- Transport/session: LoRaWAN MAC and network server (handles join, ADR, frame counters).
- **Network:** Gateways forward packets over TCP/TLS to network server.
- Physical/MAC: LoRa PHY parameters (SF, bandwidth).

### **Device & service discovery**

- **Device provisioning:** OTAA (Over The Air Activation) or ABP with secure keys managed in network server.
- Service discovery: For 6LoWPAN, use CoAP resource discovery (/.well-known/core) and mDNS in local domains; for LoRaWAN, discovery occurs via device registry mapping DevEUI

  Application.

# Real-time requirements & strategies

• Use edge gateways to perform local aggregation and burst-tolerant uplinks.

- **Duty cycle & downlink limits** of LoRaWAN mean avoid frequent downlinks schedule bulk configuration windows.
- QoS for critical events (fire, overflow) routed via redundant paths (cellular fallback on gateway).

# Diagram — network layout

# Unit III — Python & Raspberry Pi: Home Automation Prototype

# System design

Raspberry Pi acts as central controller (edge hub). Sensors (temperature, motion, light) and actuators (relay-controlled lights, motorized curtains) are interfaced via GPIO, I<sup>2</sup>C, SPI, and UART.

# Hardware interfacing

- GPIO: Digital inputs (PIR motion), outputs to drive relays (via transistor/optocoupler).
- I<sup>2</sup>C: Environmental sensors (BME280) and displays (OLED) share I<sup>2</sup>C bus (pull-ups needed).
- **SPI:** High-speed sensors (e.g., ADC like MCP3008) or external radio modules.
- UART/Serial: Communication with microcontrollers (Arduino) or Zigbee coordinators.

# **Sample Python components**

- Use RPi.GPIO or gpiozero for GPIO control.
- smbus2 for I<sup>2</sup>C, spidev for SPI, pyserial for UART.
- Lightweight local broker: mosquitto for MQTT on Pi; publish sensor telemetry to ThingSpeak or cloud via HTTPS.

# ThingSpeak integration

- Send periodic HTTP POST or MQTT messages to ThingSpeak channels (API keys used as write keys).
- Use field formatting and timestamps. For alerts, push to IFTTT or SMS gateway.

# Data flow & control loop

- Local automation rules run on Pi (e.g., motion + night -> turn on light) to avoid cloud dependence for critical controls.
- Cloud used for long-term logging, visualization, remote overrides.

Diagram — prototype

Unit IV — Cloud for IoT: Smart Traffic with Hybrid Cloud & Fog

### **Problem**

Real-time traffic signal optimization, incident detection, and historical analytics for planning.

# **Architectural components**

- 1. **Edge/Fog nodes:** Roadside units (RSUs) and signal controllers run low-latency inference (vehicle count, emergency vehicle preemption).
- 2. **Message bus:** Kafka clusters for durable ingestion of high-throughput telemetry (video metadata, sensor counts).
- 3. **Stream processing:** Apache Spark Structured Streaming for windowed aggregations, feature extraction, and feeding ML models.
- 4. **Long-term storage:** Time-series DB (InfluxDB / Prometheus for metrics), object storage for video (S3), and data warehouse for planning (Snowflake/Redshift).
- 5. Control plane: Microservices handle route suggestions, signal timing pushes, and public APIs.

### Hybrid cloud + fog rationale

- **Fog** handles low-latency control loops (sub-100ms) and de-identification of video at the edge to preserve privacy.
- Cloud handles heavy ML training, global coordination, long term analytics.

## IoT Data Analytics Pipeline (Kafka + Spark)

- 1. Devices → Edge gateway → Kafka producers → Kafka topics (ingestion)
- 2. Spark Structured Streaming consumes Kafka topics, performs sliding-window counts and anomaly detection
- 3. Results → Control topic (Kafka) → Edge consumer to adjust signals; also stored in time-series DB and fed to dashboards

# Diagram — pipeline & placement

# Unit V — AWS IoT: Industrial Deployment

### Use case

Industrial plant with thousands of sensors for predictive maintenance, process control, and telemetry visualizations.

### **AWS IoT features used**

- AWS IoT Core: secure device connectivity (MQTT over TLS) and rules engine for routing messages to AWS services.
- **Device Management:** fleet provisioning, jobs, OTA firmware updates.
- Device Shadow: virtual device state to allow apps to read/set desired states even when
  device offline.
- Resource tagging: organize resources (by plant, line, criticality) for cost allocation and access control.

# Data storage & retrieval

- Route telemetry via IoT Rules to **Kinesis** or **Kafka** (MSK) and then to **S3** for raw data; use **Timestream** or **InfluxDB** for time-series analytics.
- Use **Lambda** for lightweight transforms and **AWS Glue** for ETL into data warehouse (Redshift).

# Scalability & best practices

- Thing registry with hierarchical naming and provisioning templates.
- Use **X.509 certs** and AWS IoT jobs for secure OTA updates.
- Sharding topics and using partition keys for downstream consumers to scale ingestion.
- Implement backpressure handling (IoT Core → Kinesis limits) and auto-scaling consumer groups.

# **Example workflow**

- 1. Device publishes telemetry (MQTT TLS) to plant/line1/machineA/telemetry.
- 2. IoT Rule matches and sends to Kinesis  $\rightarrow$  Lambda for enrichment  $\rightarrow$  Timestream & S3.
- 3. Dashboard queries Timestream for near-real-time KPIs and S3 for batch ML training.
- 4. To update setpoint, operator writes desired state to Device Shadow; device syncs on next connect.

# Diagram — AWS IoT deployment

# Common design recommendations (cross-unit)

- **Security-first:** hardware root-of-trust, mutual TLS, principle of least privilege, encrypted storage.
- Edge processing: move time-sensitive logic to edge to reduce latency and cloud costs.
- **Standards & interoperability:** favor IP-native stacks where possible, and adopt healthcare/industrial standards as appropriate.
- Observability: centralized logging, distributed tracing, device health telemetry, and SLA monitoring.
- **Testing & simulation:** digital twins and test harnesses to validate scale, failover, and update workflows before production deployment.