

PART 1: THEORETICAL ANALYSIS

Q1. Explain how Edge AI reduces latency and enhances privacy compared to cloud-based AI. Provide a real-world example.

Edge AI refers to running artificial intelligence models directly on local devices (edge devices) such as smartphones, Raspberry Pi boards, drones, smart cameras, or IoT sensors—rather than sending data to the cloud.

How Edge AI Reduces Latency

In cloud-based AI, devices must send data to remote servers, wait for the server to process it, and then receive a response.

This creates delays due to:

- Network traffic
- Internet speed
- Server processing time

Edge AI performs inference **locally**, eliminating the need for round-trip communication.

This results in:

- Real-time responses
- Faster decision-making
- High performance even in low-network environments

How Edge AI Enhances Privacy

Cloud-based systems require transferring raw data (images, audio, personal information) to external servers.

This exposes data to:

- Network attacks

- Unauthorized access
- Compliance issues

Edge AI keeps data **on the device**, meaning:

- Sensitive data never leaves the device
- Reduced risk of breaches
- Better compliance with privacy laws (GDPR, HIPAA)

Real-World Example — Autonomous Drones

Autonomous drones must process:

- Real-time video feed
- Obstacle detection
- Path planning

If a drone relied on the cloud:

- Network delays could cause crashes
- Loss of connectivity would make the drone unusable
- Streaming video creates privacy and bandwidth concerns

By running AI models **onboard**, drones can detect objects, avoid obstacles, and make navigation decisions instantly — with higher safety and privacy.

Q2. Compare Quantum AI and Classical AI in solving optimization problems. What industries benefit most?

Classical AI

Classical AI uses CPUs/GPUs and traditional algorithms to perform:

- Search
- Optimization
- Pattern recognition
- Decision-making

It is limited by classical binary computing (0s and 1s).

Optimization problems with huge search spaces (NP-hard problems) are slow and resource-intensive.

Examples of difficult classical optimization problems:

- Traveling Salesman Problem
- Complex logistics routing
- Portfolio optimization
- Molecular structure prediction

Quantum AI

Quantum AI uses **quantum bits (qubits)**, allowing superposition and entanglement.

This enables quantum computers to evaluate **many possible solutions simultaneously**, offering major advantages in optimization.

Advantages of Quantum AI in Optimization

- Can search extremely large solution spaces much faster
- Better exploration (less stuck in local minima)

- More efficient solving of NP-hard problems
-

Industries That Benefit Most from Quantum AI

1. Finance

- Portfolio optimization
- Risk modeling
- Fraud detection
- High-frequency trading

2. Logistics & Supply Chain

- Route optimization
- Inventory planning
- Scheduling

3. Healthcare & Pharmaceuticals

- Protein folding
- Drug discovery
- Genomics optimization

4. Energy

- Grid optimization
- Renewable resource allocation

- Predictive energy balancing

5. Manufacturing

- Production scheduling
- Quality control optimization

Quantum AI provides exponential speed-ups for complex optimization problems that classical AI struggles with.

TASK 2: AI-DRIVEN IoT AGRICULTURE SYSTEM

1. Sensors Needed

Sensor	Purpose
Soil Moisture Sensor	Detect soil water levels
Temperature Sensor (DHT22)	Plant health and weather monitoring
Humidity Sensor	Measure leaf transpiration
Light Sensor (LDR)	Optimize sunlight exposure
pH Sensor	Soil acidity for fertilizer decisions
Rain Sensor	Predict irrigation needs
Camera Module	Detect crop diseases

2. Proposed AI Model for Crop Yield Prediction

A **Random Forest Regression model** would be ideal because it:

- Handles nonlinear relationships
- Performs well with noisy environmental data

- Offers high accuracy with limited tuning
- Works well for IoT sensor integration

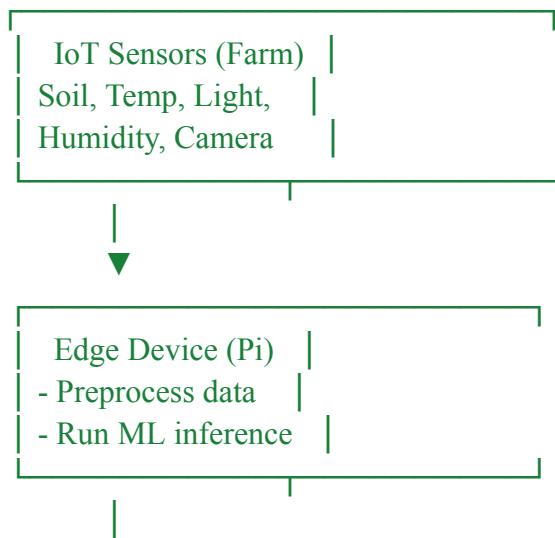
Input features:

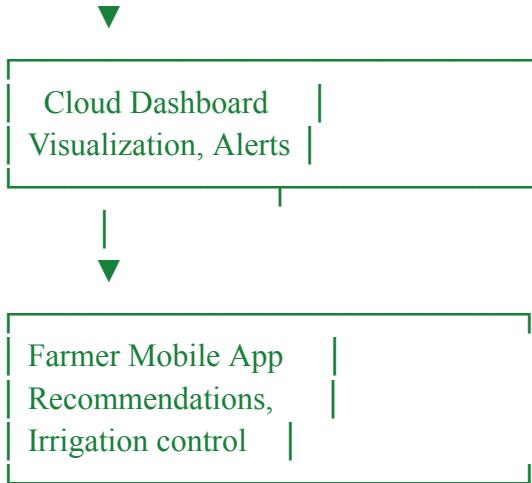
- Soil moisture
- Temperature
- Humidity
- pH
- Rainfall
- Sunlight hours
- NDVI vegetation index

Output:

Predicted crop yield (kg/ha)

3. Data Flow Diagram





How the System Works

1. IoT sensors collect environmental data.
2. Edge device preprocesses and runs ML yield prediction.
3. Cloud receives processed data and updates dashboards.
4. Farmers receive real-time recommendations like:
 - “Irrigate now”
 - “Soil pH too low — add lime”
 - “Disease risk detected”