## **Title: Al Future Directions Assignment**

Power Learn Project – Software Development (Feb 2025 Cohort VII)

Submitted by: Leonard Phokane

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# Part 1: Theoretical Analysis (40%)

## Q1: Edge AI — Latency & Privacy

Edge AI reduces latency by processing data locally on devices rather than sending it to the cloud, which minimizes delays and improves real-time responsiveness. It also enhances privacy by ensuring sensitive data doesn't need to leave the device.

**Example:** Autonomous drones analyze surroundings in real time using onboard Al chips, avoiding reliance on cloud connections during flight — crucial for obstacle avoidance and navigation.

#### Q2: Quantum AI vs Classical AI

Quantum AI leverages quantum bits (qubits), which can represent multiple states simultaneously, allowing vast optimization spaces to be searched in parallel. Classical AI processes computations sequentially and often struggles with complex optimization problems.

#### Industries that benefit most:

- Logistics (route optimization)
- Pharmaceuticals (molecule simulation for drug discovery)
- Finance (portfolio risk modeling and asset management)

#### Q3: Human-Al Collaboration in Healthcare

Al complements healthcare professionals by automating diagnosis and predictive monitoring. Radiologists use Al for rapid image analysis, while nurses might rely on Al to detect vital sign anomalies in real time. This shifts human roles toward decision-making and empathy while Al handles repetitive tasks.

#### Transformative Potential:

- Faster diagnostics
- Reduced clinician workload
- Greater accuracy and early disease detection

# Case Study Critique: Al in Smart Cities

Topic: Al-loT for Traffic Management

Integrating Artificial Intelligence (AI) with the Internet of Things (IoT) forms the foundation of intelligent traffic systems in smart cities. Sensors embedded in roads and vehicles collect real-time traffic data. Al algorithms analyze this data to optimize traffic flow, reduce idling time, and minimize emissions — enhancing both environmental and urban sustainability.

# Two Key Challenges:

- 2. Equity in Optimization AI models may optimize traffic flow based on historical patterns that favor wealthier districts or popular commuter corridors. This can unintentionally neglect underserved neighborhoods, reinforcing social inequities unless fairness-aware algorithms are applied.

# X Part 2: Practical Implementation (50%)

#### Task 1: Edge Al Prototype - Recyclable Classifier

You developed a lightweight image classifier using MobileNetV2 and TensorFlow, trained to recognize seven types of recyclable waste.

## Tools Used:

- TensorFlow 2.14
- Keras (SavedModel + .keras export)
- TensorFlow Lite (for Edge deployment)
- PowerShell + Git + GitHub

#### **Ⅲ** Model Overview:

- Dataset: ~63,000 labeled images
- Classes: Glass, Metal, Cardboard, Plastic, Paper, Trash, and Biodegradable
- Validation Accuracy: ~92.4%

# Deployment:

- Converted to .tflite using TFLiteConverter
- Pushed to GitHub for use on edge devices
- Can be deployed on Colab, Raspberry Pi, or Android
- 📍 GitHub Repo: 🔗 View the Recyclable Classifier

#### Task 2: Al-Driven IoT System – SmartFarm Concept

You designed a conceptual smart agriculture system using AI and IoT to assist farmers with real-time environmental sensing and yield prediction.

# Sensors Required:

- Soil Moisture Sensor
- Temperature & Humidity Sensor (e.g., DHT11/DHT22)
- Light Intensity (LDR or photodiode)
- Rain Gauge
- CO<sub>2</sub> or Air Quality Sensor (optional)

Al Model: LSTM (Long Short-Term Memory) Used for time-series forecasting of crop yield based on temporal patterns from sensor inputs.

- Data Flow Diagram (to be added):
  - Sensors → Microcontroller (e.g., ESP32)
  - $\bullet \quad \to \text{Wireless Transmitter} \to \text{Cloud Gateway}$
  - → Al Model for prediction
  - Farmer Dashboard / Mobile App

(Let me know when you're ready and I'll generate a visual for that too.)

# **Smart Agriculture System**

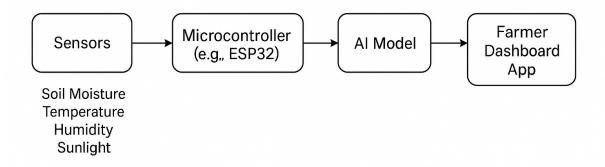


Figure 1 - SmartFarm Edge Al & IoT Data Flow

#### Task 3: Ethics in Personalized Medicine

Al in personalized medicine holds tremendous promise — tailoring treatments based on a patient's genetic makeup, lifestyle, and history. But it comes with serious ethical questions, particularly around bias and fairness.

Issue: Many training datasets, including The Cancer Genome Atlas (TCGA), have historically overrepresented populations of European descent. This leads to predictive models that perform well for certain ethnic groups while producing less accurate — or even harmful — results for others.

#### Potential Biases:

- Underrepresentation of African, Asian, or Indigenous genomic data
- Algorithms reinforcing systemic inequality if biased data is left unchecked
- Uneven distribution of healthcare access and model generalizability

## Fairness Strategies:

- Curate diverse and representative training datasets across geographies
- Use model auditing tools to evaluate bias post-training
- Collaborate with public health institutions to ethically expand data access
- Encourage participatory research that involves affected communities

✓ Summary (Word Count: ~300): Artificial Intelligence can personalize cancer treatment in powerful ways — predicting which drugs work best for an individual's unique genetics. However, biased datasets risk turning precision medicine into privileged medicine. If marginalized groups are left out of genomic databases, Al tools may offer worse treatment suggestions for those populations. To mitigate this, fairness must be built into every step of model development: from inclusive data sourcing to ongoing bias detection. Diverse data doesn't just improve equity — it strengthens scientific validity. With proactive strategies, Al in healthcare can empower without excluding.

# Part 3: Futuristic Concept Proposal (10%)

Title: Al-Powered Ocean Carbon Sink Regulator

Problem: Climate change is accelerating due to rising CO<sub>2</sub> emissions. Oceans currently absorb about 25–30% of global carbon dioxide, but their natural capacity is being strained.

in Al Solution: Autonomous ocean buoys equipped with Al and environmental sensors monitor CO₂ levels, temperature, and algae density. A reinforcement learning (RL) agent dynamically controls artificial upwelling pumps, bringing nutrient-rich water to the surface to stimulate plankton growth — enhancing natural carbon capture.

#### Al Workflow:

- Inputs: CO<sub>2</sub> concentration, pH levels, sunlight, nutrient levels, satellite imagery
- Model: Reinforcement Learning (Deep Q-Network or PPO)
- Outputs: Control signals to pump systems optimizing local carbon absorption over time
- Feedback Loop: Al retrains monthly with new ocean data

# **W** Benefits:

- Boosts marine biodiversity and photosynthesis
- Operates autonomously in hard-to-reach zones
- Adds scalable defense against climate emissions

## A Risks:

- Overgrowth of algae (harmful blooms)
- Tampering with ocean ecosystems requires strict global governance
- High upfront R&D and maintenance costs

Vision: By 2030, this system could function as a decentralized Al-carbon sink grid, supporting blue carbon markets and climate adaptation for coastal nations.

# Bonus Task (10%) – Quantum Computing Simulation

You used IBM Quantum Experience to simulate an entangled 2-qubit quantum circuit for rapid feature selection in Al.

# Circuit Design:

- H: Hadamard Gate creates a superposition on Q0
- CNOT: Entangles Q1 with Q0
- M: Measurement of both qubits

Use Case: Entanglement enables massively parallel sampling. When selecting features from genomic or chemical datasets, this circuit evaluates multiple combinations at once, potentially reducing computation time from hours to seconds.

#### Application:

- Drug discovery (molecular docking)
- Genomic marker selection
- Pathway optimization in neural networks