



Revolutionizing HPCL Operations with Neural Networks

Theme selected: Technological Innovations

Vision: Utilize neural networks to drive operational efficiency, fuel quality, and sustainability in HPCL's value chain. Our project serves as a strategic measure to support HMEL's aim of achieving 38% renewables in the energy mix and a 12.7% reduction in greenhouse gas emissions by 2030

Presented By Team Human



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Core Objectives:

1. Predictive or Assumptive Maintenance:

Proactively address equipment failures with AI.

2. Fuel Quality Assurance and Analysis:

Ensure superior fuel standards using analytics.

3. Logistics Optimization:

Improve fuel distribution speed and accuracy.

4. Process Optimization:

Enhance energy efficiency in refining operations.

5. Sustainability Goals:

Reduce carbon emissions through smarter processes

How Neural Networks Are Used:

1. Predictive Maintenance:

- LSTMs analyze time-series data (e.g., temperature, vibration) to forecast equipment failures.
- Early detection prevents costly downtime.

2. Fuel Quality Monitoring:

- CNNs process spectroscopy images to detect impurities in real-time.
- Autoencoders flag deviations in fuel composition for correction.

3. Logistics Optimization:

- Reinforcement Learning dynamically optimizes delivery routes and schedules.
- Neural networks predict demand, avoiding stockouts and reducing costs.

4. Process Optimization:

- DNNs model chemical reactions for energy-efficient refining.
- Optimized process parameters minimize waste and maximize yield.



Neural Network Integration & Impact

Title: *"Harnessing Neural Networks for Efficiency & Innovation"*

Code Implementation Highlights:

Data Preprocessing:

- Techniques: Normalization, outlier removal, and time-series conversion for sensor data.
- Libraries: Pandas, NumPy, TensorFlow/Keras.

Optimization Algorithms:

- Reinforcement learning (Q-Learning) for logistics optimization.
- Gradient Descent optimizers for refining process modeling.

Neural Network Design:

- Predictive Maintenance: LSTM-based sequential models for time-series predictions.
- Fuel Quality Monitoring: CNNs for image-based impurity detection

Deployment:

- Tools: TensorFlow Serving for API-based integration.
- Platforms: Cloud-based pipelines for scalability.

Impact Metrics:

- **Efficiency:** 40% downtime reduction.
- **Quality:** 25% improvement in fuel consistency.
- **Sustainability:** 30% lower carbon emissions

An implementation of code in google colab:

<https://colab.research.google.com/drive/18gNhfaxKMDMzgF-Xc3mcFLjfSEi3MbWm?usp=sharing>

