

# Turbofan Engine Remaining Useful Life (RUL) Prediction - MLOps Project

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## Overview

This project predicts the Remaining Useful Life (RUL) of turbofan engines using sensor data and advanced deep learning models (CNN-LSTM). It is designed as a full MLOps pipeline, leveraging **MLflow** for experiment tracking, **DVC** for data versioning, and **Docker** for reproducible deployment.

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## Data Description

The dataset is sourced from NASA's CMAPS repository ([behrad3d/nasa-cmaps](#)). It consists of multiple text files:

- **Training Data:** `train_FD00X.txt`
- **Testing Data:** `test_FD00X.txt`
- **RUL Labels:** `RUL_FD00X.txt`

### Data Format

- **Train/Test Files:** Each row represents one engine cycle and contains:
  - Engine ID
  - Cycle number
  - 24 sensor readings (float)
  - Operational settings (float)
  - Example: 1 1 -0.0005 0.0004 100.0 518.67 ... 8145.32
- **RUL Files:** Each row is an integer representing the RUL for a corresponding engine in the test set.

See sample files: - [RUL\\_FD003.txt](#) - [train\\_FD003.txt](#)

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## Data Processing

Data processing is handled in [src/data\\_operation.py](#):

1. **Extraction:** Raw data files are loaded and merged into Pandas DataFrames.
2. **Cleaning:** Unnecessary columns are dropped, missing values handled.
3. **Feature Engineering:**
4. Engine IDs are constructed as `file_name_cycle`.
5. RUL is calculated for each row in train/test sets.
6. Label encoding for categorical columns.
7. **Normalization:** Features are normalized using `MinMaxScaler`.
8. **Windowing:** Data is split into sequences for time-series modeling (e.g., 10 cycles per window).
9. **Saving:** Processed data is saved to [data/processed\\_data/](#).

See [src/data\\_operation.py](#) for implementation details.

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## Model Training

Model training is orchestrated by `model_training_pipeline.py` and `training_pipeline_runner.py`:

- Uses a CNN-LSTM architecture (see notebook: [smarter-maintenance-with-cnn-lstm-94-accurate.ipynb](#)).
  - Hyperparameters (epochs, batch size, validation split) are configurable via CLI.
  - Model metrics (loss, MAE, accuracy) are logged to MLflow.
  - Model is saved and versioned via MLflow.

# MLOps Stack

MLflow

- Tracks experiments, parameters, metrics, and models.
  - See usage in `training_pipeline_runner.py`.

DVC

- Version controls raw and processed data.
  - Use `dvc add data/raw_data/` and `dvc add data/processed_data/` to track data changes.
  - Pipelines can be defined in `dvc.yaml`.

## Docker

- Dockerfile will be provided for reproducible environment.
  - Build and run: `sh docker build -t turbofan-rul .` `docker run -it turbofan-rul`

# Project Structure

```
data/
    raw_data/
    processed_data/
    datasets/
src/
    config.py
    data_operation.py
    pipelines.py
    train_model.py
    test_model.py
    transform_data.py
notebooks/
    smarter-maintenance-with-cnn-lstm-94-accurate.ipynb
training_pipeline_runner.py
model_training_pipeline.py
app.log
dvc.yaml
Dockerfile
README.md
```

## Getting Started

## 1. Clone the repository

```
git clone <repo-url>
cd turbofan_engine_life_prediction
```

## 2. Setup Environment

```
python -m venv venv
source venv/bin/activate # or .\venv\Scripts\activate on Windows
pip install -r requirements.txt
```

## 3. Data Versioning

```
dvc pull
```

## 4. Train Model

```
python training_pipeline_runner.py [epochs] [batch_size] [validation_split]
```

## 5. Track Experiments

- Start MLflow UI: sh mlflow ui
- View results at <localhost:5000>

## 6. Build Docker Image

```
docker build -t turbofan-rul .
docker run -it turbofan-rul
```

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## Contributing

Feel free to open issues or submit pull requests for improvements.

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## License

MIT License

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## References

- NASA CMAPS Dataset: <https://www.nasa.gov/content/prognostics-center-of-excellence-data-set-repository>
- MLflow: <https://mlflow.org/>
- DVC: <https://dvc.org/>

- Docker: