Electrolysis-Agent-Report

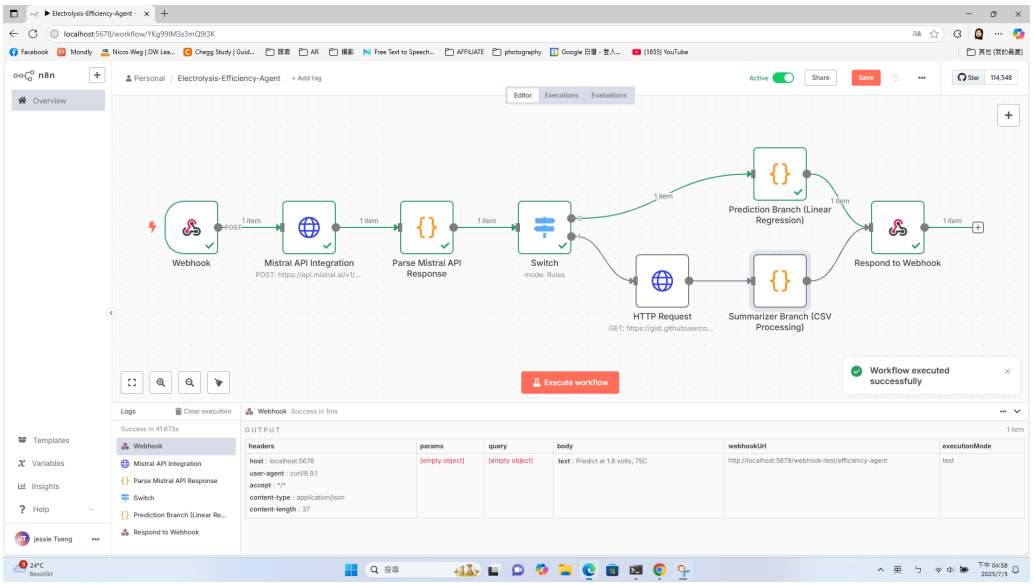
System Architecture

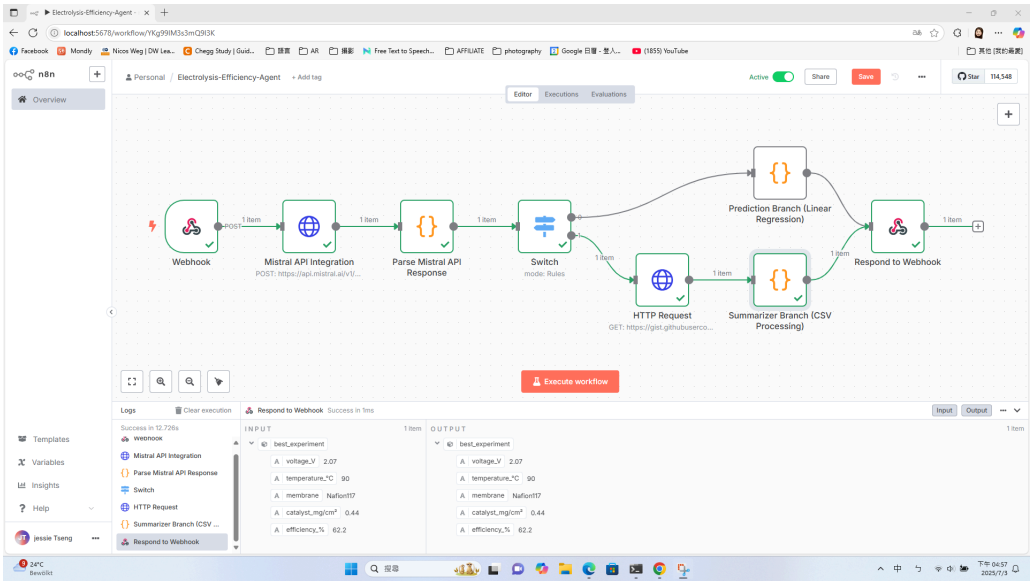
The system is built using **n8n** and consists of the following components:

1. **Webhook**– Accepts free-text requests from users.
2. **Mistral API Integration** – Sends the user's input to a hosted **LLM model** (e.g., Mistral) that interprets the request.
3. **Parse Mistral API Response** – Extracts structured information from the LLM's response, including:
   - `intent` (either `"predict_run"` or `"summarize_best"`)
   - relevant numeric parameters such as `voltage` and `temperature`
4. **Switch** – Directs the workflow based on the extracted `intent.`
5. **1.1 Prediction Branch** – Implements a pre-trained linear regression model with fixed coefficients.

   2.1 HTTP Requestre – trieves a public CSV file of 50 experimental runs.

   2.2 Summarizer Branch – parses the CSV content by splitting the string into rows and columns, converting each row into a JSON object, identifying the row with the **highest efficiency_%.**
6. **Respond to Webhook** – Returns JSON output to the user.





Linear Regression Model

📌 Data Preprocessing Steps

1. **Source**: Dataset imported from external CSV URL.
2. **Column Standardization**: Renamed the following columns for consistency:
   - `voltage_V` → `voltage`
   - `temperature_°C` → `temperature`
   - `efficiency_%` → `efficiency`
   - `catalyst_mg/cm²` → `catalyst`
3. **Missing Data Handling**: Dropped all rows with missing values using `dropna()`.
4. **Feature Selection**:
   - Features used: `voltage`, `temperature`
   - Target variable: `efficiency`
5. **Train-Test Split**:
   - 80% training data, 20% test data
   - `random_state=42` for reproducibility
6. **Standardization**:

- Input features were standardized using `StandardScaler` to ensure zero mean and unit variance.

7. **Regression Model**:
   - Applied `LinearRegression` from `scikit-learn` wrapped in a pipeline.

📈 Regression Model Performance

- **Model**: StandardScaler + LinearRegression
- **Test Set RMSE**:

  ✅ **{rmse:.2f}**

  (Root Mean Squared Error on hold-out test set)
- **Cross-Validation RMSE (5-Fold)**:

  ✅ **{cv_rmse.mean():.2f}**

  (Mean RMSE across 5 shuffled splits using K-Fold CV)
- Model 1 (There is another model I try)

```
Test RMSE: 1.04
5-Fold CV RMSE: 0.87
```

- **Coefficient Summary (Real Scale)**:

Term	Coefficient
Intercept	22.1581
Voltage	10.294
Temperature	0.1918

Example Inputs & Outputs

Prompt	Response
`Predict run at 1.85V and 80C`	`{"prediction": "58.53%", "parameters": {"voltage": 1.85, "temperature": 80}}`
`Summarize the best efficiency run`	`{"best_experiment": {"voltage_V": 2.07, "efficiency_%": 62.2, ...}}`
What is the expected efficiency at 1.7V and	[{"prediction": "51.17%","parameters": {"voltage": 1.7,"temperature": 60},"model_info": "Linear Regression

Prompt	Response
60°C?	(voltage & temp)"}]