

Giraldo Nainggolan

<https://portofoliokreatif.blogspot.com/>

Portfolio



2025

Information Systems



Giraldo Nainggolan

IINNOVATIVE IT PROGRAMMER

Education

- **senior high school**
SMAKN SAMOSIR 2018 - 2021
- **Information Systems Students**
Trunojoyo University 2022 - 2026

During my studies, I have been actively involved in various IT trainings and projects, including software redesign, website and Android application debugging, and user convenience implementation. I am active in the Information Systems Association organization, Google Developer Campus, ITC, and several other additional activities.

Experience

About Me

I am an Information Systems student with an interest in technology and innovation. Experienced in website development, UI/UX, mobile application, data analysis, and project management, I focus on creating digital solutions that are secure, efficient, and impactful.

Skills

- Python
- Javascript
- Java / Kotlin
- Wordpress
- PHP
- Laravel
- SQL

Contact

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PERUM TGR BLOCK I 47/16

Project Management - Surabaya

Reka Village April 2024 - June 2024
Explore my Real Estate Website project built with WordPress CMS. Featuring responsive design, SEO optimization, and advanced property search for seamless browsing.

Product Developmen - Surabaya

UPN East Java June 2024 - Agustus 2024
Discover my FarmBot project: an innovative solution combining automation and precision farming to revolutionize sustainable agriculture and boost efficiency.

Software Staff

Bangkalan Legislative Council Sep 2024 - Nov 2024
Tasked with designing a cross-platform application for the Bangkalan Regional People's Representative Council (DPRD) to facilitate efficient correspondence and communication.

Audit & Corporate Governance

PT Bukit Darmo.Tbk Oct 2024 - Dec 2024
Data entry in the IT governance audit using COBIT 19 involves collecting, verifying, and recording information related to IT Processes, GAMO, and corporate compliance over a three-month period.

Production Planning and Inventory Control

PT Sebastian Jaya Metal Jan 2025 - July 2025
In this role, I am responsible for production planning, inventory control, as well as ensuring a smooth production process in accordance with the demand and availability of raw materials.



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PROJECT

DATA SPECIALIST

Data Science, Analytics, and Computer Vision for Advanced Insights and AI Solutions

Disusun oleh
Giraldo Naiggolan

Fruit Classification using CNN and Image Processing

Dataset properties:

The total number of images: 94110. Only images scaled to 100x100 pixels are counted here. Training set size: 70491 images (one object per image). Test set size: 23619 images (one object per image). The number of classes: 141 (fruits and vegetables). Image size in the largest dataset version: 100x100 pixels.

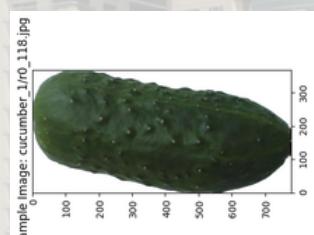
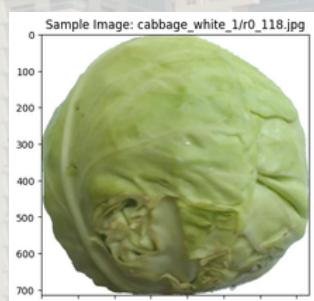
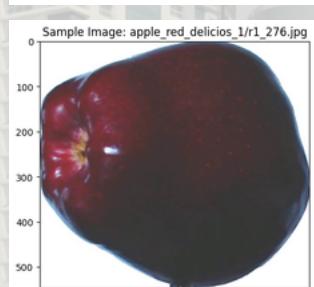
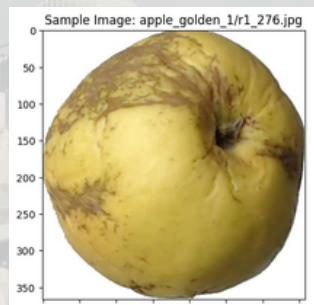
Input (1.01 GB)

Technologies Used:

os, cv2 (OpenCV), matplotlib, seaborn, numpy, pandas, torch (PyTorch), torchvision, sklearn.model_selection (train_test_split), tensorflow.keras, tensorflow.keras.optimizers (Adam), tensorflow.keras.callbacks (EarlyStopping), sklearn.metrics (classification_report, confusion_matrix), mpl_toolkits.mplot3d.Axes3D, sklearn.preprocessing.StandardScaler

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 148, 148, 32)	896
max_pooling2d (MaxPooling2D)	(None, 74, 74, 32)	0
conv2d_1 (Conv2D)	(None, 72, 72, 64)	18,496
max_pooling2d_1 (MaxPooling2D)	(None, 36, 36, 64)	0
conv2d_2 (Conv2D)	(None, 34, 34, 128)	73,856
max_pooling2d_2 (MaxPooling2D)	(None, 17, 17, 128)	0
flatten (Flatten)	(None, 36992)	0
dense (Dense)	(None, 512)	18,940,416
dense_1 (Dense)	(None, 24)	12,312

```
Found 4993 images belonging to 24 classes.  
Found 615 images belonging to 24 classes.  
Total samples in train_generator: 4993  
Total samples in test_generator: 615  
Data batch shape: (32, 224, 224, 3), Label batch shape: (32, 24)  
Steps per epoch (train): 156  
Validation steps (test): 19  
Epoch 1/20  
/opt/conda/lib/python3.10/site-packages/torch/src/trainers/data_adapters/py_dataset_adapter.py:121:  
UserWarning: Your 'PyDataset' class should call 'super().__init__(**kwargs)' in its constructor.  
'**kwargs' can include 'workers', 'use_multiprocessing', 'max_queue_size'. Do not pass these arguments  
to 'fit()', as they will be ignored.  
    self.warn_if_super_not_called()  
156/156 [██████████] 276s/step - accuracy: 0.9418 - loss: 2.6969 - val_accuracy: 0.8783 -  
val_loss: 0.4482  
Epoch 2/20  
156/156 [██████████] 25 877us/step - accuracy: 1.0000 - loss: 0.0192 - val_accuracy: 0.7143 -  
val_loss: 0.7699  
Epoch 3/20  
/opt/conda/lib/python3.10/contextlib.py:153: UserWarning: Your input ran out of data; interrupting  
training. Make sure that your dataset or generator can generate at least 'steps_per_epoch * epochs'  
batches. You may need to use the 'repeat()' function when building your dataset.  
    self.gen.throw(typ, value, traceback)  
156/156 [██████████] 275s/2s/step - accuracy: 0.9845 - loss: 0.0563 - val_accuracy: 0.9243 -  
val_loss: 0.3771  
Epoch 4/20  
156/156 [██████████] 25 1ms/step - accuracy: 1.0000 - loss: 0.0015 - val_accuracy: 0.8571 -  
val_loss: 0.2236  
Epoch 5/20  
156/156 [██████████] 269s/2s/step - accuracy: 1.0000 - loss: 0.0011 - val_accuracy: 0.9194 -  
val_loss: 0.4260  
Epoch 6/20  
156/156 [██████████] 11s 62ms/step - accuracy: 1.0000 - loss: 0.0057 - val_accuracy: 0.8571 -  
val_loss: 0.1305  
Epoch 7/20  
156/156 [██████████] 271s/2s/step - accuracy: 1.0000 - loss: 4.9764e-04 - val_accuracy: 0.9112 -  
val_loss: 0.4374  
Epoch 8/20  
156/156 [██████████] 25 1ms/step - accuracy: 1.0000 - loss: 5.9189e-05 - val_accuracy: 1.0000 -  
val_loss: 0.0384  
Epoch 9/20  
156/156 [██████████] 271s/2s/step - accuracy: 1.0000 - loss: 2.2102e-04 - val_accuracy: 0.9128 -  
val_loss: 0.4623  
Epoch 10/20  
156/156 [██████████] 25 778us/step - accuracy: 1.0000 - loss: 1.8633e-04 - val_accuracy: 0.7143 -  
val_loss: 1.8958  
Epoch 11/20  
156/156 [██████████] 270s/2s/step - accuracy: 1.0000 - loss: 1.5063e-04 - val_accuracy: 0.9112 -  
val_loss: 0.4777  
Epoch 12/20  
156/156 [██████████] 11s 59ms/step - accuracy: 1.0000 - loss: 2.2863e-04 - val_accuracy: 0.8571 -  
val_loss: 0.1162  
Epoch 13/20  
156/156 [██████████] 271s/2s/step - accuracy: 1.0000 - loss: 9.6746e-05 - val_accuracy: 0.9079 -  
val_loss: 0.4944  
Epoch 14/20  
156/156 [██████████] 2s 847us/step - accuracy: 1.0000 - loss: 7.8199e-05 - val_accuracy: 1.0000 -  
val_loss: 0.0273  
Epoch 15/20  
156/156 [██████████] 275s/2s/step - accuracy: 1.0000 - loss: 8.0174e-05 - val_accuracy: 0.9013 -  
val_loss: 0.5085  
Epoch 16/20  
156/156 [██████████] 25 1ms/step - accuracy: 1.0000 - loss: 7.0432e-05 - val_accuracy: 1.0000 -  
val_loss: 0.0030  
Epoch 17/20  
156/156 [██████████] 271s/2s/step - accuracy: 1.0000 - loss: 5.7610e-05 - val_accuracy: 0.9879 -  
val_loss: 0.5048  
Epoch 18/20  
156/156 [██████████] 25 753us/step - accuracy: 1.0000 - loss: 1.8365e-05 - val_accuracy: 0.7143 -  
val_loss: 0.6171  
Epoch 19/20  
156/156 [██████████] 270s/2s/step - accuracy: 1.0000 - loss: 4.7687e-05 - val_accuracy: 0.9038 -  
val_loss: 0.5209  
Epoch 20/20  
156/156 [██████████] 25 711us/step - accuracy: 1.0000 - loss: 5.2908e-05 - val_accuracy: 0.8571 -  
val_loss: 0.2784
```



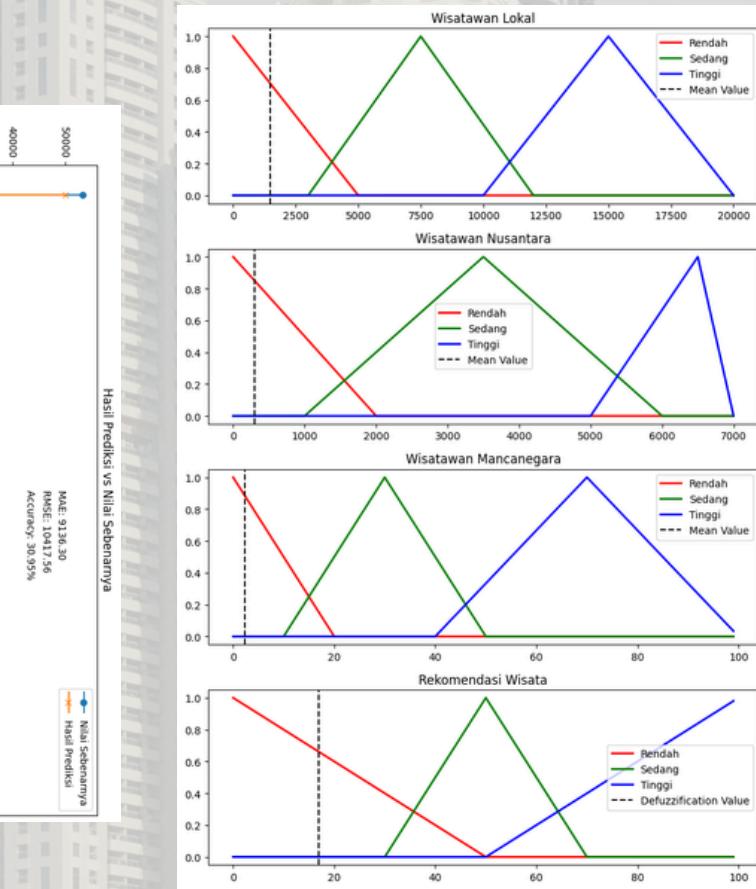
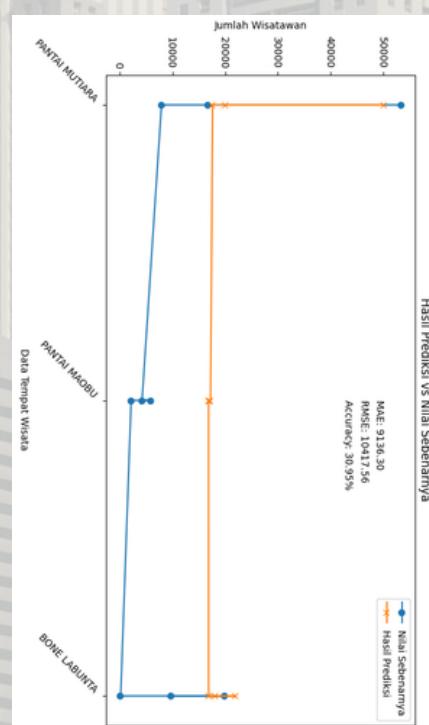
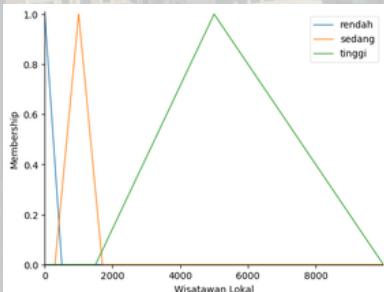
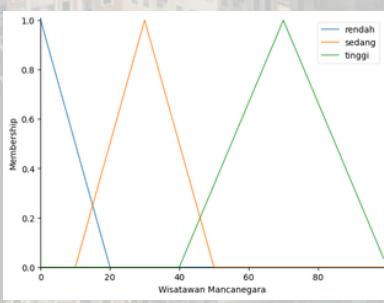
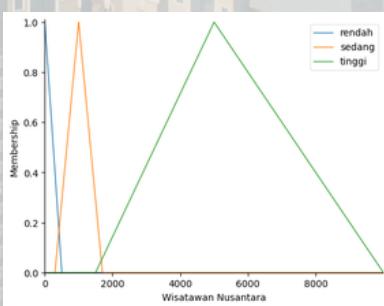
Online shopping-KNN-ROC

Dataset properties:

The dataset provided contains data about several tourist attractions along with the predicted number of tourists who will visit these places. For example, Mutiara Beach is predicted to attract 50,000 tourists, while Maobu Beach has a predicted number of visitors ranging from 16,695 to 17,079 tourists, and Bone Labunta is expected to invite between 16,708 to 21,843 tourists.

Technologies Used:

numpy as np, pandas as pd, scikit-fuzzy, matplotlib.pyplot as plt, skfuzzy as fuzz, sklearn.model_selection, train_test_split, skfuzzy, control as ctrlsklearn.metrics, mean_absolute_error, mean_squared_error, matplotlib.pyplot as plt



U-Net with binary labels

Dataset properties:

This notebook demonstrates the use of the U-Net architecture for image segmentation with binary labels. The goal is to classify each pixel in an image into two categories: the target object and the background.

Technologies Used:

Deep Learning Framework: TensorFlow/Keras, Model Architecture: U-Net, Programming Language: Python, Data Handling Libraries: NumPy, Pandas, Visualization Tools: Matplotlib, Experiment Platform: Kaggle Notebook

The screenshot shows a Kaggle notebook interface. On the left, the sidebar includes 'Create', 'Home', 'Competitions', 'Datasets', 'Models', 'Code', 'Discussions', 'Learn', 'More', 'Your Work', 'U-Net with binary labels', 'FIDE & Google Che...', 'FIDE & Google Effi...', 'notebookika3f575...', and 'View Active Events'. The main area has tabs for 'Notebook', 'Input', 'Output' (which is selected), 'Logs', and 'Comments (0)'. The 'Output' tab displays a grid of 31 small images labeled 'pure_unet'. To the right of the images is a file browser for 'final_scores_unet_watershed.csv' (757 B). The browser shows a hierarchical tree structure with files like raw_unet_1.h5, raw_unet_1.log, raw_unet_30.h5, raw_unet_2.h5, raw_unet_2.log, raw_unet_3.h5, raw_unet_3.log, raw_unet_4.h5, raw_unet_4.log, raw_unet_5.h5, raw_unet_5.log, raw_unet_6.h5, raw_unet_7.h5, raw_unet_7.log, raw_unet_8.h5, raw_unet_9.h5, raw_unet_9.log, AII_unet.npy, AII_unet_watershed.npy, PQ_pure_unet.npy, PQ_pure_unet_watershed.npy, dice_pure_unet.npy, dice_pure_unet_watershed.npy, final_scores_pure_unet.csv, and final_scores_unet_watershed.csv. The bottom of the notebook shows a command line: `_ kaggle kernels output giraldainggolan/u-net-with-binary-labels -p /path/to/dest`. The right side of the interface has buttons for 'Output', 'images', 'validation', 'pure_unet', 'watershed_unet', 'models', 'AII_pure_unet.npy', 'AII_unet_watershed.npy', 'PQ_pure_unet.npy', 'dice_pure_unet.npy', 'dice_pure_unet_watershed.npy', 'final_scores_pure_unet.csv', and 'final_scores_unet_watershed.csv'.

<https://www.kaggle.com/code/giraldainggolan/u-net-with-binary-labels>

FIDE & Google Chess AI

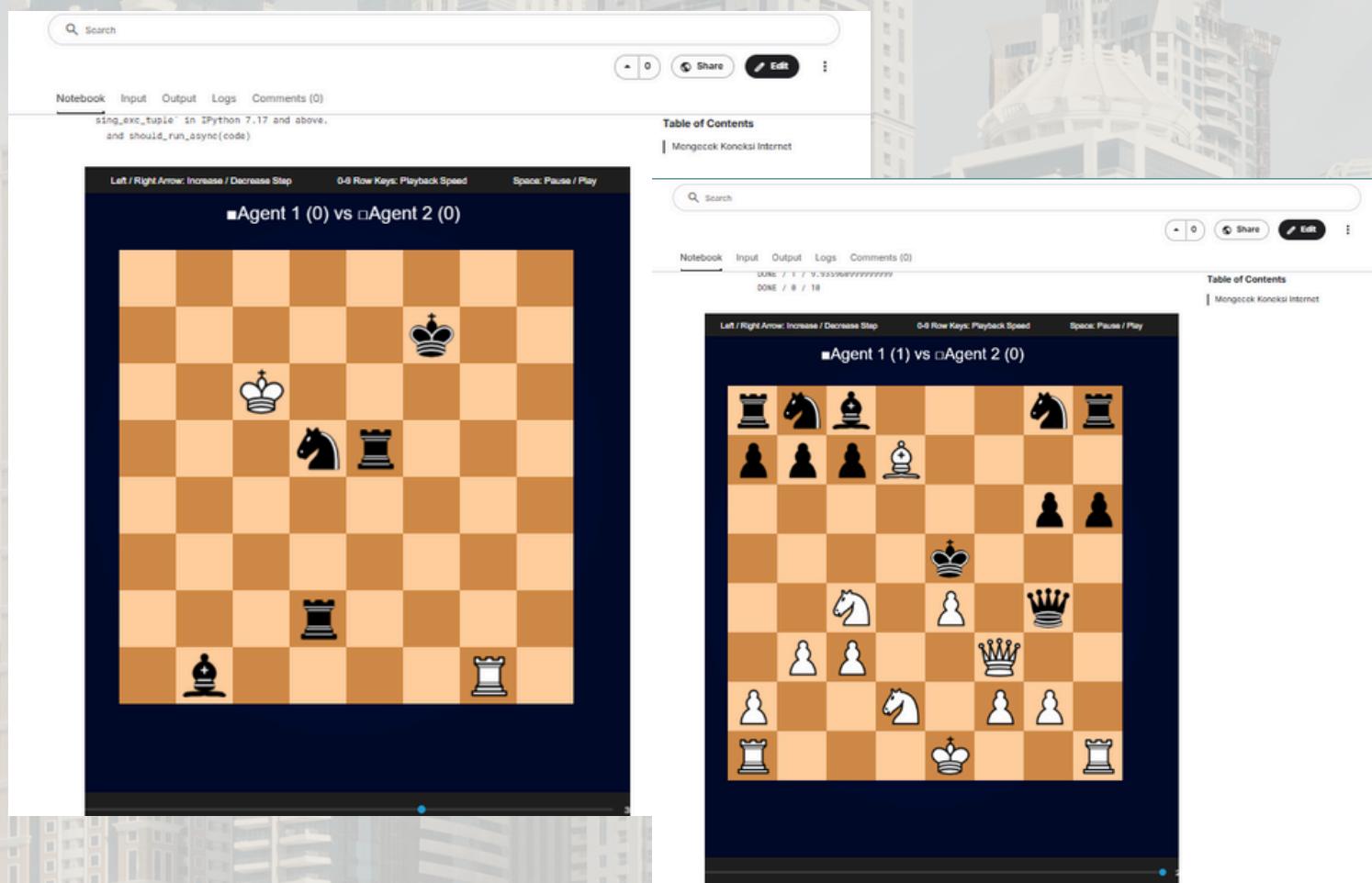
Dataset properties:

The "FIDE & Google Chess AI: Thinking Smarter, Not Hard" project explores AI technology in chess by developing a more strategic and efficient system beyond brute-force calculations. By integrating optimized search algorithms like Monte Carlo Tree Search (MCTS) and Alpha-Beta Pruning with Google's AI technology, the project enhances move analysis and game strategy.

Leveraging FIDE's chess database, the AI adapts to Grandmaster play patterns in real-time, making its decision-making process more intelligent and dynamic.

Technologies Used:

Python, TensorFlow & PyTorch, Stockfish & AlphaZero, Google Cloud AI, FIDE Chess Database, Monte Carlo Tree Search (MCTS), Alpha-Beta Pruning, Jupyter Notebook.



Machine Learning Regression

Dataset properties:

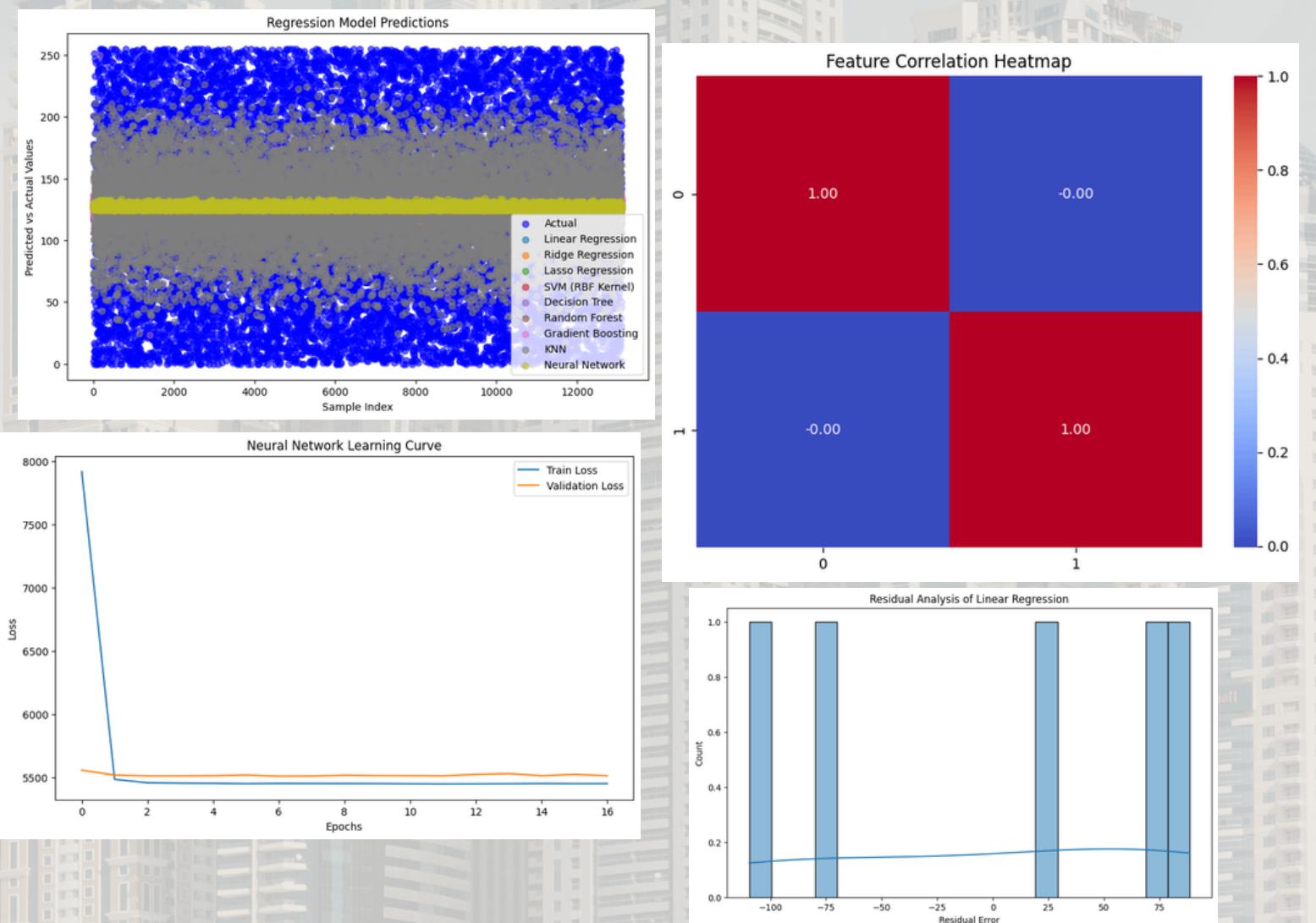
This dataset is designed to explore various regression methods in machine learning, including Linear Regression, Polynomial Regression, Ridge Regression, Lasso Regression, and Decision Tree Regression. It is suitable for training and experimenting with different regression models.

Variables in the dataset:

- *input1* → The first numerical variable used for prediction.
- *input2* → The second numerical variable used for prediction.
- *output* → The target variable to be predicted.

Technologies Used:

Python, Pandas , NumPy , Exel, Matplotlib & Seaborn, Scikit-Learn, Jupyter Notebook



Data-Driven Rainfall Prediction

Dataset properties:

It includes Date & Time (YYYY-MM-DD HH:MM:SS) to track temporal patterns, Temperature ($^{\circ}\text{C}$) representing the daily average temperature, and Humidity (%), which indicates moisture levels in the air. Additionally, Air Pressure (hPa) is included to analyze atmospheric conditions, along with Wind Speed (km/h) and Wind Direction ($^{\circ}$) to assess air movement patterns. The dataset also records Rainfall (mm) as the primary target variable, measuring the amount of precipitation per day. Lastly, Weather Conditions provide categorical data on the general state of the weather (e.g., Clear, Cloudy, Rainy), further enhancing the prediction model's accuracy.

Technologies Used:

Data Preprocessing & Analysis, Machine Learning Model, Deployment & Tools, Output / Goals

