

Industrial Project 2024 Part 02

Comprehensive Guide for Project Execution and Usage











Group A:

Sonain Jamil, Kasem Amnuayrotchanachinda, Muhammad Turab

✉ sonainjamil@ieee.org

November 21, 2024

Contents

1	Introduction 	3
2	Project Features 	3
3	Installation and Setup 	3
3.1	Requirements 	3
3.2	Setup Instructions 	4
4	Configuration 	4
5	Usage 	4
5.1	Stitching	5
5.2	PCA Processing	5
6	Project Structure 	5
7	Key Modules 	5
8	Testing and Results 	6
9	Example Results 	6
10	References 	6

1 Introduction

The **Industrial Project 2024 Part 02** focuses on processing spectral image data through image stitching and Principal Component Analysis (PCA). This project enables the assembly of multiple spectral images into a single, high-resolution stitched image and performs dimensionality reduction on the stitched data using PCA.

2 Project Features

- **Image Stitching:** Combines multiple hyperspectral images into a single, high-resolution stitched image.
- **Transformation Options:** Applies transformations to the stitched image before PCA, including:
 - No Transformation
 - Standard Normal Transformation
 - First Derivative (with contrast enhancement)
 - Second Derivative (with contrast enhancement)
- **PCA on Spectral Data:** Reduces the dimensionality of the stitched hyperspectral image, saving each principal component as an individual grayscale PNG image.
- **Configurable Parameters:** Utilizes a `config.yaml` file to set paths, transformation options, and other processing parameters.

3 Installation and Setup

3.1 Requirements

- Python 3.8 or higher.
- Required Python libraries:
 - `numpy`
 - `matplotlib`
 - `PyYAML`
 - `opencv-python`
 - `scikit-learn`

3.2 Setup Instructions

1. Clone the repository to your local machine:

```
git clone https://github.com/sonainjameel/  
Industrial_Project_2024_Part2.git  
cd Industrial_Project_2024_Part2
```

2. Install the required dependencies:

```
pip install -r requirements.txt
```

3. Ensure you have the correct input data in the directories specified in `config.yaml`.

4 Configuration

The project relies on a `config.yaml` file for setting input paths and processing parameters. Below is an example configuration:

```
paths:  
  input_dir: "data/input/images" # Directory containing input  
    hyperspectral images  
  stitched_output: "data/stitched/stitched_image.hdr" # Path to save  
    stitched hyperspectral image  
  pca_output_dir: "data/pca_output" # Directory for saving PCA  
    component images  
parameters:  
  stitching:  
    num_columns: 8 # Number of images per row in the stitched output  
  pca:  
    n_components: 20 # Number of PCA components to keep  
  transformation: "first_derivative" # Options: "none", "  
    standard_normal", "first_derivative", "second_derivative"  
  lower_percentile: 2 # Percentile for contrast enhancement in  
    derivatives  
  upper_percentile: 98
```

Listing 1: Example config.yaml File

5 Usage

To run each part of the pipeline, specify the configuration file and task (`stitch` or `pca`) in the command line.

5.1 ✂ Stitching

Run the stitching process to create a stitched hyperspectral image from the input directory:

```
python main.py config.yaml stitch
```

Listing 2: Run the Stitching Process

The stitched image will be saved at the location specified in the `stitched_output` path of the `config.yaml` file.

5.2 ✂ PCA Processing

Perform PCA on the stitched image to reduce dimensionality and generate principal component images:

```
python main.py config.yaml pca
```

Listing 3: Run the PCA Process

The PCA component images will be saved in the directory specified in the `pca_output_dir` field of the `config.yaml` file.

6 📁 Project Structure 🏗

Industrial_Project_2024_Part2/

-- config.yaml	# Configuration file for paths and parameters
-- main.py	# Main script to run the pipeline
-- requirements.txt	# List of dependencies
-- data/	# Directory for input and output data
-- input/	# Folder containing input hyperspectral images
-- stitched/	# Folder to save stitched output
-- pca_output/	# Folder to save PCA component images
-- modules/	# Core project modules
-- __init__.py	# Initialize module
-- stitching.py	# Functions for image stitching
-- pca_processing.py	# Functions for PCA
-- transformations.py	# Functions for spectral transformations

7 📁 Key Modules ⚙

- `stitching.py`: Handles the assembly of input spectral images into a single stitched image.

- `pca_processing.py`: Performs dimensionality reduction using PCA, saving each principal component as a grayscale image.
- `transformations.py`: Implements transformations such as standard normalization, first derivative, and second derivative with optional contrast enhancement.

8 Testing and Results

The following test cases were conducted:

- Verified stitching functionality with different layouts and resolutions.
- Tested PCA component generation with varying numbers of components.
- Ensured accuracy and clarity of transformed spectral data using different transformations.

9 Example Results

Below are some example outputs from the project:



Figure 1: Principal Component 1 output from PCA.

10 References

- Python Documentation: <https://www.python.org/doc/>
- Libraries: NumPy, OpenCV, Matplotlib, PyYAML, Scikit-Learn.