

Facet

A Specification for Organizing Local Photos by Detected Faces

(Includes Full Source Code Appendices)

Ashutosh Gupta

Version 1.0

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1 Introduction

1.1 Motivation

Digital photo collections often grow uncontrollably, making it difficult to find pictures of specific individuals. While cloud services offer face recognition, privacy concerns arise from uploading personal photos to remote servers. There is a need for a tool that can organize photos based on faces locally, keeping user data private.

1.2 Project Goal

The **Facet** project aims to develop a command-line tool that automatically scans a user-specified directory of photos, detects faces within those photos, groups photos containing the same individuals using facial recognition and clustering, and organizes copies of these photos into separate folders, all running entirely on the user's local machine.

1.3 Scope

This blueprint outlines the design for Facet Version 1.0, focusing on:

- Scanning a local folder (and subfolders) for common image file types.
- Detecting faces in images using established computer vision libraries.
- Generating unique numerical embeddings (face fingerprints) for each detected face.
- Clustering similar face embeddings together using unsupervised machine learning (DB-SCAN) to identify unique individuals without prior knowledge.
- Creating output folders (e.g., **Person_1**, **Person_2**) representing each identified cluster (person).
- Copying the original photo files containing faces belonging to a cluster into the respective person's output folder.
- Providing command-line arguments for specifying input and output directories and tuning clustering parameters.

Features outside the initial scope include: a graphical user interface (GUI), real-time processing, identifying known people based on reference images, merging clusters, tagging photo metadata instead of copying files, and advanced error handling for corrupted images.

2 System Architecture

Facet operates as a command-line script that processes images sequentially through a pipeline involving file system scanning, face detection, face embedding generation, clustering, and file organization.

Workflow Description:

1. **Initialization:** The script (`sort_faces.py`) parses command-line arguments for input and output directories and optional clustering parameters.
2. **File Discovery:** The script scans the specified input directory recursively to find all supported image files.
3. **Face Processing Loop:** For each image found:
 - The image is loaded into memory.

Architecture Diagram Placeholder

(A diagram showing: User Input (CLI Arguments: Input Path, Output Path) → Python Script (`sort_faces.py`) → Scan Files (`os.walk`) → For Each Image: Load Image (`face_recognition/Pillow`) → Detect Faces (`face_recognition`) → Generate Embeddings (`face_recognition`) → Collect All Embeddings → Cluster Embeddings (`sklearn.DBSCAN`) → Map Images to Cluster IDs → Create Output Folders (`os.makedirs`) → Copy Image Files (`shutil.copy2`) → Output Folders (`Person_1, Person_2...`)).

Figure 1: High-Level System Architecture Diagram

- The `face_recognition` library detects the locations (bounding boxes) of all faces within the image.
 - For each detected face, a unique 128-dimension numerical embedding (vector) is generated.
 - Each embedding is stored along with the path to the original image file.
4. **Clustering:** After processing all images, all collected face embeddings are passed to the DBSCAN clustering algorithm from `scikit-learn`. DBSCAN groups embeddings that are close to each other in the 128-dimensional space, effectively grouping faces of the same person. Each face embedding is assigned a cluster label (integer). Noise points (faces that don't fit well into any cluster) are assigned label -1.
 5. **Image Grouping:** The script creates a mapping between cluster labels and the set of original image paths associated with faces belonging to that cluster.
 6. **File Organization:**
 - The specified output directory is created if it doesn't exist.
 - For each cluster label (excluding noise label -1):
 - A new subdirectory is created within the output directory (e.g., `Person_1`, `Person_2`, ...).
 - The script iterates through the unique image paths associated with the current cluster.
 - A copy of each unique image file is placed into the corresponding person's subdirectory using `shutil.copy2` (preserving metadata).
 7. **Completion:** The script finishes after processing all clusters and copying files, logging summary information.

3 Core Components

3.1 Main Script (`sort_faces.py`)

- **Technology:** Python 3

- **Role:** Orchestrates the entire workflow. Handles argument parsing, file system scanning (`os`), image loading, calls face detection/encoding functions, initiates clustering, maps results, and performs file copying (`shutil`). Includes logging for progress feedback.

3.2 Face Detection & Recognition Library

- **Technology:** `face_recognition` (Python library, based on `dlib`)
- **Role:** Provides high-level functions to:
 - Load image files.
 - Detect the locations of faces within images (using HOG or CNN models).
 - Generate 128-dimension face embeddings (face fingerprints) from detected faces.

3.3 Clustering Library

- **Technology:** `scikit-learn` (Python library)
- **Role:** Provides the DBSCAN (Density-Based Spatial Clustering of Applications with Noise) algorithm used to group similar face embeddings together without needing to know the number of people beforehand.

3.4 Supporting Libraries

- **Technology:** `numpy`, `Pillow`, `opencv-python`
- **Role:**
 - `numpy`: Used for efficient numerical operations, particularly handling the face embedding arrays required by `scikit-learn`.
 - `Pillow`: An imaging library, often used as a backend or dependency by `face_recognition` for image loading and manipulation.
 - `opencv-python`: While `face_recognition` can use its own methods, OpenCV is often installed as it provides underlying image processing capabilities or alternative face detectors if needed. The script might implicitly benefit from its presence.

4 Technology Stack

- **Programming Language:** Python 3
- **Face Recognition:** `face_recognition` library (using `dlib` backend)
- **Clustering:** `scikit-learn` library (DBSCAN algorithm)
- **Numerical Computation:** `numpy`
- **Image Handling:** `Pillow`, `opencv-python` (often dependencies)
- **Command-Line Interface:** `argparse` (Python standard library)
- **File System Operations:** `os`, `shutil` (Python standard library)
- **Version Control:** Git

5 Setup and Installation

This section provides the detailed steps required to set up the Facet project environment locally.

5.1 Prerequisites

- **Python:** Version 3.x. Verify with `python -version` or `python3 -version`. ([Download](#))
- **pip:** Python package installer (usually included with Python).
- **Build Tools (Potentially Required for dlib):** The `face_recognition` library depends on `dlib`. Installing `dlib` often requires:
 - **CMake:** ([Download](#))
 - **C++ Compiler:**
 - * **Linux (Debian/Ubuntu):** `sudo apt update && sudo apt install build-essential cmake`
 - * **macOS:** Install Xcode Command Line Tools: `xcode-select -install` (includes Clang compiler and CMake might need separate install via [Homebrew](#): `brew install cmake`).
 - * **Windows:** Install C++ build tools for Visual Studio (select "Desktop development with C++" workload during VS Installer). Make sure CMake is installed and added to PATH.

Refer to the official [face_recognition installation guide](#) for OS-specific details if you encounter issues.

5.2 Detailed Setup Steps

1. Obtain Project Files:

- **Manual:** Download or create the project files (`sort_faces.py`, `requirements.txt`, etc.) in a directory named **Facet**. Navigate into this directory using your terminal.

2. Create Python Virtual Environment (Recommended):

Isolates project dependencies. Execute within the **Facet** directory:

```
python -m venv venv
# Activate (Linux/macOS): source venv/bin/activate
# Activate (Windows CMD): venv\Scripts\activate.bat
# Activate (Windows PS): .\venv\Scripts\Activate.ps1
```

Your terminal prompt should indicate the active environment (e.g., `(venv)`).

3. Install Python Dependencies:

Install the required packages using `pip`. **Ensure prerequisites (CMake, C++ compiler) are installed before this step.**

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

(See Appendix [C](#) for contents. This might take some time, especially compiling `dlib`.)

Setup is complete. Proceed to Section [6](#) for running the tool.

6 Usage

1. Activate the Python virtual environment (if not already active):
`source venv/bin/activate` (or equivalent for your OS).
2. Navigate to the project's root directory (Facet).
3. Run the main script from the command line, providing the required input and output paths:

```
python sort_faces.py --input_folder "/path/to/your/photos" --output_folder "/path/to/sorted_output"
```

- Replace `"/path/to/your/photos"` with the full path to the directory containing the images you want to scan.
- Replace `"/path/to/sorted_output"` with the full path where the organized folders (Person_1, Person_2, etc.) should be created. This directory will be created if it doesn't exist.

4. (Optional) Tune clustering parameters:

```
python sort_faces.py -i "/photos" -o "/sorted" --eps 0.5 --min_samples 3
```

- Use `-eps` (default: 0.55) to control similarity tolerance (lower is stricter).
 - Use `-min_samples` (default: 2) to set the minimum number of times a face must appear to form a distinct cluster.
5. The script will log its progress to the console. Processing can take a significant amount of time depending on the number and size of photos.
 6. Once finished, check the specified output folder for subdirectories named `Person_X` containing copies of the photos organized by detected faces.

7 Project Structure

The directory structure is as follows (items marked `(*)` are typically managed locally and ignored by Git):

```
Facet/
    .gitignore          # Specifies files/folders for Git to ignore
    LICENSE             # Project license file
    README.md           # Project documentation file
    sort_faces.py        # Main Python script for face sorting
    requirements.txt     # Python package dependencies

    venv/ (*)           # Python virtual environment (local)

# Potential additional files/folders created locally (*)
# __pycache__/ (*)     # Python bytecode cache (local)
# Output folder specified by user (e.g., sorted_output/) (*)
#     Person_1/ (*)
#     Person_2/ (*)
#     ...
```

Listing 1: Project Directory Layout

Note: Files/folders marked with `(*)` are generally created locally during setup or runtime and are typically excluded from the Git repository via the `.gitignore` file (Appendix D).

8 Future Work

Potential enhancements for Facet:

- **GUI Interface:** Develop a graphical user interface (e.g., using Tkinter, PyQt, Kivy) for easier folder selection, progress visualization, and viewing results.
- **Naming/Merging Clusters:** Allow users to review the generated `Person_X` folders and assign actual names or merge folders representing the same person clustered separately.
- **Known Face Recognition:** Implement functionality to recognize pre-defined individuals based on reference photos.
- **Metadata Tagging:** Add an option to write face information (identified person name/ID) to image metadata (e.g., EXIF/XMP keywords) instead of, or in addition to, copying files.
- **Performance Optimization:** Explore batch processing of images, GPU acceleration (if dlib is compiled with CUDA support), or alternative, faster face detection models (e.g., from OpenCV Zoo).
- **Incremental Updates:** Add capability to scan a folder again and only process newly added photos, integrating them into existing clusters or forming new ones.
- **Improved Clustering:** Experiment with other clustering algorithms or post-processing steps to refine cluster quality.
- **Configuration File:** Use a configuration file (e.g., YAML, JSON) for settings like `eps`, `min_samples`, image extensions, etc., instead of only command-line arguments.

9 License (Summary)

This project is assumed to be distributed under the MIT License, based on common practice. The full license text placeholder is provided in [Appendix A](#).

A License (MIT)

The following is a placeholder for the MIT License. Replace with the actual license if available, or use this standard text.

MIT License

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SOFTWARE.

Listing 2: MIT License Text Placeholder ('LICENSE')

B Main Script (sort_faces.py)

```
1 import face_recognition
2 import os
3 import shutil
4 import argparse
5 import numpy as np
6 from sklearn.cluster import DBSCAN
7 from collections import defaultdict
8 import logging
9
10 # Configure logging
11 logging.basicConfig(level=logging.INFO, format='%(asctime)s - %(levelname)s - %(message)s')
12
13 # --- Configuration ---
14 # Supported image file extensions
15 IMAGE_EXTENSIONS = ('.jpg', '.jpeg', '.png', '.gif', '.bmp', '.tiff')
16
17 # DBSCAN parameters (tune these based on your results)
18 # eps: Max distance between samples for one to be considered as in the neighborhood of the
19 #       other.
20 #       Lower values mean faces need to be more similar to be grouped. Start around 0.5-0.6.
21 DBSCAN_EPS = 0.55
22 # min_samples: Number of samples (faces) in a neighborhood for a point to be considered as a
23 #       core point.
24 #       Essentially, how many times does a face need to appear to be considered a
25 #       distinct person?
26 DBSCAN_MIN_SAMPLES = 2 # Increase if you only want to group people appearing more often
27
28 # --- Helper Functions ---
29
30 def find_image_files(folder_path):
31     """Recursively finds all image files in the given folder."""
32     image_files = []
33     logging.info(f"Scanning for images in: {folder_path}")
34     for root, _, files in os.walk(folder_path):
35         for filename in files:
36             if filename.lower().endswith(IMAGE_EXTENSIONS):
37                 image_files.append(os.path.join(root, filename))
38     logging.info(f"Found {len(image_files)} image files.")
39     return image_files
40
41 def extract_face_data(image_paths):
42     """Extracts face encodings and their corresponding paths from images."""
43     face_data = [] # List of dictionaries: {'path': path, 'encoding': encoding_vector}
44     total_images = len(image_paths)
45     logging.info(f"Starting face detection and encoding for {total_images} images...")
46
47     for i, image_path in enumerate(image_paths):
48         logging.info(f"Processing image {i + 1}/{total_images}: {os.path.basename(image_path)}")
49         try:
50             # Load image
51             image = face_recognition.load_image_file(image_path)
52             # Find face locations (using CNN model is more accurate but slower, default is
53             # HOG)
54             # face_locations = face_recognition.face_locations(image, model="cnn")
55             face_locations = face_recognition.face_locations(image)
56
57             if not face_locations:
58                 logging.debug(f"No faces found in {os.path.basename(image_path)}")
59                 continue
60
61             # Get face encodings
```

```

58         # Providing known locations speeds up encoding
59         face_encodings = face_recognition.face_encodings(image, known_face_locations=
face_locations)
60
61         logging.debug(f"Found {len(face_encodings)} face(s) in {os.path.basename(
image_path)}")
62
63         # Store each encoding with its path
64         for encoding in face_encodings:
65             face_data.append({'path': image_path, 'encoding': encoding})
66
67         except Exception as e:
68             logging.warning(f"Could not process image {image_path}: {e}")
69
70     logging.info(f"Finished encoding. Found {len(face_data)} total face instances.")
71     return face_data
72
73 def cluster_faces(face_data):
74     """Clusters face encodings using DBSCAN."""
75     if not face_data:
76         logging.warning("No face data to cluster.")
77         return None, {} # Return None for labels, empty dict for data
78
79     logging.info("Starting face clustering...")
80     encodings = np.array([data['encoding'] for data in face_data])
81
82     # Create and fit the DBSCAN model
83     # n_jobs=-1 uses all available CPU cores
84     clusterer = DBSCAN(eps=DBSCAN_EPS, min_samples=DBSCAN_MIN_SAMPLES, metric='euclidean',
n_jobs=-1)
85     clusterer.fit(encodings)
86
87     labels = clusterer.labels_ # Get cluster labels for each face encoding
88     num_clusters = len(set(labels)) - (1 if -1 in labels else 0) # -1 label is for noise/
outliers
89     logging.info(f"Clustering complete. Found {num_clusters} distinct clusters (people)
excluding noise.")
90
91     return labels, face_data
92
93 def organize_photos_by_cluster(labels, face_data, output_dir):
94     """Copies photos into folders based on cluster labels."""
95     if labels is None:
96         logging.error("Cannot organize photos, clustering failed or produced no results.")
97         return
98
99     logging.info(f"Organizing photos into: {output_dir}")
100     os.makedirs(output_dir, exist_ok=True)
101
102     # Group image paths by cluster label
103     images_by_cluster = defaultdict(set) # Use set to store unique image paths per cluster
104     for label, data in zip(labels, face_data):
105         if label != -1: # Ignore noise points
106             images_by_cluster[label].add(data['path'])
107
108     if not images_by_cluster:
109         logging.warning("No valid clusters found to organize photos.")
110         return
111
112     # Copy files for each cluster
113     cluster_count = 0
114     for label, image_paths in images_by_cluster.items():
115         cluster_count += 1

```

```

116     person_folder_name = f"Person_{cluster_count}" # Assign sequential names
117     person_output_path = os.path.join(output_dir, person_folder_name)
118     os.makedirs(person_output_path, exist_ok=True)
119     logging.info(f"Copying {len(image_paths)} unique images for {person_folder_name} (
Cluster Label {label})...")
120
121     for image_path in image_paths:
122         try:
123             dest_path = os.path.join(person_output_path, os.path.basename(image_path))
124             # Avoid copying if somehow the exact same file is listed twice for the
cluster
125             if not os.path.exists(dest_path):
126                 # copy2 preserves metadata like creation/modification time
127                 shutil.copy2(image_path, dest_path)
128             else:
129                 logging.debug(f"Skipping already copied file: {dest_path}")
130         except Exception as e:
131             logging.error(f"Failed to copy {os.path.basename(image_path)} to {
person_folder_name}: {e}")
132
133     logging.info("Finished organizing photos.")
134
135
136 # --- Main Execution ---
137 if __name__ == "__main__":
138     parser = argparse.ArgumentParser(description="Sort photos locally by detected faces
using clustering.")
139     parser.add_argument("-i", "--input_folder", required=True, help="Path to the folder
containing photos to scan.")
140     parser.add_argument("-o", "--output_folder", required=True, help="Path to the folder
where sorted photos (Person_1, Person_2, ...) will be copied.")
141     # Optional arguments for tuning
142     parser.add_argument("--eps", type=float, default=DBSCAN_EPS, help=f"DBSCAN epsilon (max
distance). Default: {DBSCAN_EPS}")
143     parser.add_argument("--min_samples", type=int, default=DBSCAN_MIN_SAMPLES, help=f"DBSCAN
min samples per cluster. Default: {DBSCAN_MIN_SAMPLES}")
144
145
146     args = parser.parse_args()
147
148     # Validate input path
149     if not os.path.isdir(args.input_folder):
150         logging.error(f"Input folder not found or is not a directory: {args.input_folder}")
151         exit(1)
152
153     # Use provided tuning parameters if given
154     DBSCAN_EPS = args.eps
155     DBSCAN_MIN_SAMPLES = args.min_samples
156     logging.info(f"Using DBSCAN settings: eps={DBSCAN_EPS}, min_samples={DBSCAN_MIN_SAMPLES}
")
157
158
159     # 1. Find images
160     image_paths = find_image_files(args.input_folder)
161     if not image_paths:
162         logging.info("No image files found in the specified folder.")
163         exit(0)
164
165     # 2. Extract face data (encodings)
166     face_data = extract_face_data(image_paths)
167     if not face_data:
168         logging.info("No faces detected in any of the images.")
169         exit(0)

```

```
170
171     # 3. Cluster faces
172     cluster_labels, clustered_face_data = cluster_faces(face_data)
173
174     # 4. Organize photos based on clusters
175     organize_photos_by_cluster(cluster_labels, clustered_face_data, args.output_folder)
176
177     logging.info("Face sorting process finished.")
```

Listing 3: Face Sorting Script (*sort_faces.py*)

C Python Dependencies (requirements.txt)

```
face_recognition>=1.3.0
scikit-learn>=1.0.0
numpy>=1.19.0
opencv-python>=4.5.0
Pillow>=8.0.0
```

Listing 4: Python Dependencies ('requirements.txt')

Note: Specific versions might vary. Use 'pip freeze > requirements.txt' after installation for exact versions.

D Git Ignore Rules (.gitignore)

A standard Python '.gitignore' is recommended.

```
# Byte-compiled / optimized / DLL files
__pycache__/
*.py[cod]
*$py.class

# Distribution / packaging / Build artifacts
build/
dist/
*.egg-info/
*.egg

# Environments
.env
.venv
env/
venv/
ENV/
env.bak/
venv.bak/

# IDE / Editor specific
.idea/
.vscode/
*.sublime-*

# OS specific
.DS_Store
Thumbs.db
ehthumbs.db

# Test / Coverage reports
htmlcov/
.coverage
.pytest_cache/
nosetests.xml
coverage.xml

# Logs and databases
*.log
*.sqlite
*.sqlite3

# User-generated output (if not intended to be committed)
# sorted_output/ # Example output directory name
```

```
# LaTeX temporary files (if compiling blueprint in project dir)
*.aux
*.log
*.out
*.toc
*.synctex.gz
```

Listing 5: Example Git Ignore Rules (‘.gitignore’)