

Image Forensics Project: Tampering Localization & Source Camera Identification

This project integrates two essential computer vision techniques in digital forensics:

1. Tampering Localization: Detecting and segmenting manipulated regions in an image using a ConvNeXt-based segmentation model.
2. Source Camera Identification (SCI): Identifying the source camera of an image using PRNU (Photo-Response Non-Uniformity) patterns extracted via denoising (FFDNet) and classified using a ResNet model.

Project Structure

Project Structure:

```
project-root/
  configs/          # Config files for MMSegmentation (ConvNeXt)
  data/
    CASIA2/         # Dataset for tampering localization
    custom/         # SCI dataset (D01D35 folders)
  mmseg/           # MMSegmentation framework
  ffdnet_tf/       # FFDNet implementation in TensorFlow
  prnu_utils/      # PRNU extraction and camera ID helpers
  train.py         # Training script for ConvNeXt segmentation
  test.py          # Evaluation script for segmentation
  sci_train.py     # Training script for SCI (ResNet + PRNU)
  sci_test.py      # Testing script for SCI
  interface.py     # Optional GUI (Tkinter-based)
  README.md
```

1. Environment Setup

1. Environment Setup

Python Environment:

```
$ conda create -n image-forensics python=3.8 -y
$ conda activate image-forensics
```

Dependencies:

```
$ pip install torch torchvision torchaudio --index-url
https://download.pytorch.org/whl/cu121
$ pip install mmcv-full==1.6.2 -f
https://download.openmmlab.com/mmcv/dist/cu121/torch2.0/index.html
$ pip install mmssegmentation==0.29.0
```

```
$ pip install tensorflow==2.10.0
$ pip install scikit-image opencv-python matplotlib tqdm pandas seaborn
```

2. Tampering Localization

2. Tampering Localization (ConvNeXt + MMSeg)

Dataset Structure:

```
data/CASIA2/
  images/
  masks/
  ...
```

To Train:

```
$ python train.py
```

To Test:

```
$ python test.py
```

3. Source Camera Identification

3. Source Camera Identification (PRNU + FFDNet + ResNet)

Dataset:

Organized in 'data/custom/' with D01D35 folders for each device.

Preprocessing:

- Denoising with FFDNet (TensorFlow)
- PRNU Extraction ($\text{Residual} = \text{Original} - \text{Denoised}$)
- Feature extraction with ResNet (Keras or PyTorch)

To Train:

```
$ python sci_train.py
```

To Test:

```
$ python sci_test.py
```