

Configuration Manual

MSc Research Project
MSc Cybersecurity (Part-Time)

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MSc Project Submission Sheet



School of Computing

Student Name: Alana Griffiths Kelly
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Programme: MSc Cybersecurity (Part Time) **Year:** 2025
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MSc (Research) Practicum/Internshio
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Module: Supervisor – Michael Pantridge
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Lecturer:
Submission Due Date: 11/12/2025
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Configuration Manual
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Project Title:
223
.....
Word Count: **Page Count:** 5

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Configuration Manual

Alana Griffiths Kelly

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1. System Requirements and Environment Setup

1.1. Software Requirements

- Python Version: Python 3.9 or later
- Package manager: pip
- Required Libraries:
 - o PyTorch
 - o torchvision
 - o facenet-pytorch
 - o MTCNN
 - o PIL
 - o glob

1.2. Setup Steps

Step 1: Install Python

Step 2: Install required packages

“pip install facenet-pytorch torchvision torch”

Step 3: Install FFmpeg

Download from <https://ffmpeg.org/download.html>

“pip install ffmpeg-python”

2. System Configuration and Pipeline Setup

2.1. Preprocessing Pipeline Configuration

2.1.1. Video Frame Extraction (FFmpeg)

Extract frames at 5 FPS

```
# Extract Frames from Video
ffmpeg

```

2.1.2. Face Detection Using MTCNN

MTCNN is used to detect and crop face regions from each frame. The cropped faces are automatically resized to 224x224.

```
mtcnn = MTCNN(image_size=224) # Resize face crops to 224x224 pixels
```

2.1.3. *Image Normalisation*

```
# Prepare Images for PyTorch Model
transform = transforms.Compose([
    transforms.Resize((224,224)),
    transforms.ToTensor()
])
```

2.2. *Model Architecture Configuration*

```
class CNN_ViT_Fusion(nn.Module):
    def __init__(self, num_classes=2):
        super().__init__()

        # 1. CNN Backbone (EfficientNet-B0)
        self.cnn = models.efficientnet_b0(weights="IMAGENET1K_V1")
        cnn_out = self.cnn.classifier[1].in_features
        # Remove final classification layer
        self.cnn.classifier[1] = nn.Identity()

        # 2. Vision Transformer Backbone (ViT-B/16)
        self.vit = models.vit_b_16(weights="IMAGENET1K_V1")
        vit_out = self.vit.heads.head.in_features
        self.vit.heads.head = nn.Identity()

        # 3. Fusion Layer
        fusion_dim = cnn_out + vit_out
        self.classifier = nn.Linear(fusion_dim, num_classes)

    def forward(self, x):
        # CNN Feature Vector
        cnn_feats = self.cnn(x)

        # ViT Feature Vector
        vit_feats = self.vit(x)

        # Concatenate Features
        fused = torch.cat([cnn_feats, vit_feats], dim=1)

        # Final Classifier
        return self.classifier(fused)
```

2.2.1. *EfficientNet-B0 (CNN Backbone)*

```
# 1. CNN Backbone (EfficientNet-B0)
self.cnn = models.efficientnet_b0(weights="IMAGENET1K_V1")
cnn_out = self.cnn.classifier[1].in_features
# Remove final classification layer
self.cnn.classifier[1] = nn.Identity()

# CNN Feature Vector
cnn_feats = self.cnn(x)
```

2.2.2. ViT-B/16 (Transformer Backbone)

```
# 2. Vision Transformer Backbone (ViT-B/16)
self.vit = models.vit_b_16(weights="IMAGENET1K_V1")
vit_out = self.vit.heads.head.in_features
self.vit.heads.head = nn.Identity()

# ViT Feature Vector
vit_feats = self.vit(x)
```

2.2.3. Feature Fusion

```
# 3. Fusion Layer
fusion_dim = cnn_out + vit_out
self.classifier = nn.Linear(fusion_dim, num_classes)

# Concatenate Features
fused = torch.cat([cnn_feats, vit_feats], dim=1)
```

2.2.4. Classification head

```
# Final Classifier
return self.classifier(fused)
```

2.2.5. Softmax Interference

```
#Run Detection on Cropped Faces
for face_file in glob.glob("faces/*.jpg"):
    face_img = Image.open(face_file)
    tensor = transform(face_img).unsqueeze(0) # Unsqueeze adds batch dimension as expects 4D input
    with torch.no_grad(): # Disables gradient tracking to speed up inference
        output = model(tensor) # Runs through DF model
        prob_fake = F.softmax(output, dim=1)[0,1].item() # Gets probability of "fake"
    scores.append(prob_fake)
```

2.3. Video-Level Classification

The final probability that a video is fake is computed by averaging frame predictions. A threshold of 0.5 classifies a video as manipulated.

```
scores = []
```

```
#Run Detection on Cropped Faces
for face_file in glob.glob("faces/*.jpg"):
    face_img = Image.open(face_file)
    tensor = transform(face_img).unsqueeze(0) # Unsqueeze adds batch dimension as expects 4D input
    with torch.no_grad(): # Disables gradient tracking to speed up inference
        output = model(tensor) # Runs through DF model
        prob_fake = F.softmax(output, dim=1)[0,1].item() # Gets probability of "fake"
        scores.append(prob_fake)
```

```
video_score = sum(scores) / len(scores) # Avg fake probability across frames
```

3. Running the System and Reproducing Results

3.1. Running the Full Pipeline

Run preprocessing:

Extract frames

```
os.makedirs("frames", exist_ok=True) # Create Folder 'Frames'
(
    # Extract Frames from Video
    ffmpeg
    .input(video_path) # Load Video
    .output('frames/frame_%05d.jpg', r=5) # Extract 5FPS to separate images named 'frame_00001.jpg' etc.
    .run() # Execute's FFmpeg Command
)
```

Detect and crop faces

```
os.makedirs("faces", exist_ok=True)
# Detect and Crop Faces from Frames
for frame_file in glob.glob("frames/*.jpg"): # Gets all frame filenames from folder 'Frames'
    frame_img = Image.open(frame_file) # Open Frame as Image
    face = mtcnn(frame_img) # Detect face & return cropped tensor
    if face is not None: # Ensure face found in frame
        face_img = to_pil_image(face) # Convert Tensor to PIL
        face_img.save(f"faces/{os.path.basename(frame_file)}") # Save crops to new folder
```

Run model

```
# Load the Hybrid Model
model = CNN_ViT_Fusion()
model.eval()
```

Get probability

```
video_score = sum(scores) / len(scores) # Avg fake probability across frames
print(f"\nDeepfake Probability: {video_score:.3f}")
```

3.2. Output Files

/frames/ – extracted frames

/faces/ – cropped face images