

REAL VS FAKE IMAGES CLASSIFICATION THROUGH AUTOENCODING

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OBJECTIVE

Use an Autoencoder to classify an image and determine whether it is a real image, or an AI generated image



DATASET INFORMATION

- The source dataset from Kaggle includes 30,000 samples from both real image sets and those generated by Generative AI tools from Midjourney DALL-E, and Stable Diffusion (60,000 total images, 52GB).
- Samples were pre-divided into a 80/20 Train/Test split.
- This project ran on 20% of the dataset in Kaggle ~ **12000 images** (½ real, ½ fake) - 11 GB



PROJECT GOALS

Real



Real



Fake



Fake



Preprocess Images

Modify dataset images to appropriately scale and re-encode color palettes to ensure efficient processing.

Train for High Accuracy

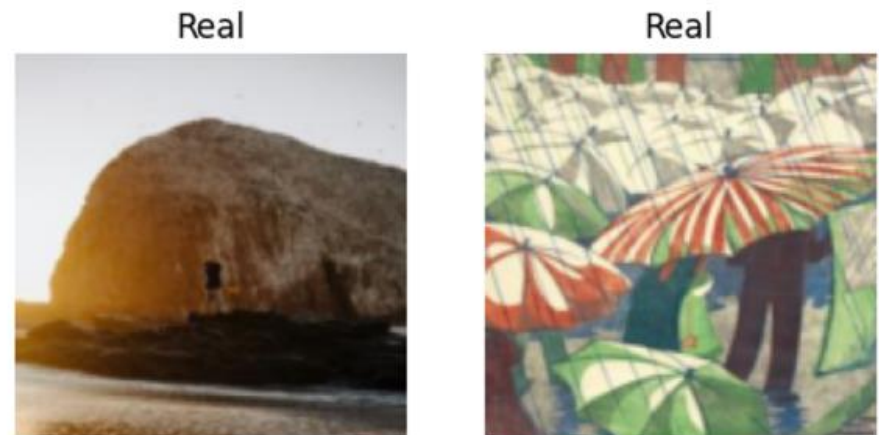
Use techniques discussed in class to achieve a high, and ideally not overfitted, accuracy against a training and test data set.

Investigate Performance on greyscale images

Experiment with generic CNNs, Transfer learning with popular CNN libraries, and GANs.

STANDARD OPERATIONS

- Batch size 16
- 200 Epochs for model training
- Learning rate = 0.0003
- Started with 5% sample and moved to large sample size with initial success
- Randomization of samples seeded for repeatability
- T4 GPU



IMAGES

REAL



FAKE



FAKE



REAL



FAKE



PSEUDOCODE – NETWORKS (RGB & GRAYSCALE)

DATA PROCESSING

- Modify color palette
- Resize to 256 x 256
- Categorical Classification

LOSS FUNCTIONS

- Focal Loss
- Hybrid Loss
- Optuna Training Parameters

TEST

- Visualize Test Results
- Youden's Threshold

BUILD NETWORKS

- Autoencoder Classifier

TRAIN

- Using Finetuned Optuna parameters

RGB - GENERAL MODEL SUMMARY

Autoencoder

- Learn features of input images
- Series of convolutional then transpose layers
- Transfer Encoder
- Batch Normalization in Decoder

Classifier Head

- Predict the class of images using learned features
- Fully Connected Dense Layers
- Global Average Pooling & Dropout

Reconstruction

- Unsupervised task to reconstruct input images
- MSE LOSS
- Hybrid Loss – L1, LPIPS, SSIM

Classification

- Supervised task to classify images
- BCE Loss
- Smoothed BCE Loss
- Focal Loss

(1) CONVOLUTIONAL AE

Model Architecture

- Encoder: 4 Layers of convolutional 2-D layers with ReLu
- Decoder: 4 Layers of transpose 2-D layers with ReLu and Sigmoid
- Classifier: Flatten with 2 fully connected layers with ReLu and Sigmoid

Loss Function

- Reconstruction Loss: MSE Loss
- Classification Loss: BCE Loss
- Total Loss = Reconstruction Loss + Classification Loss

Model Additions

- Input Image size: 128 x 128
- Trained on 1% sample, 200 epochs

(1) CONVOLUTIONAL AE

Misclassified Samples — Original (Top) vs Reconstructed (Bottom)

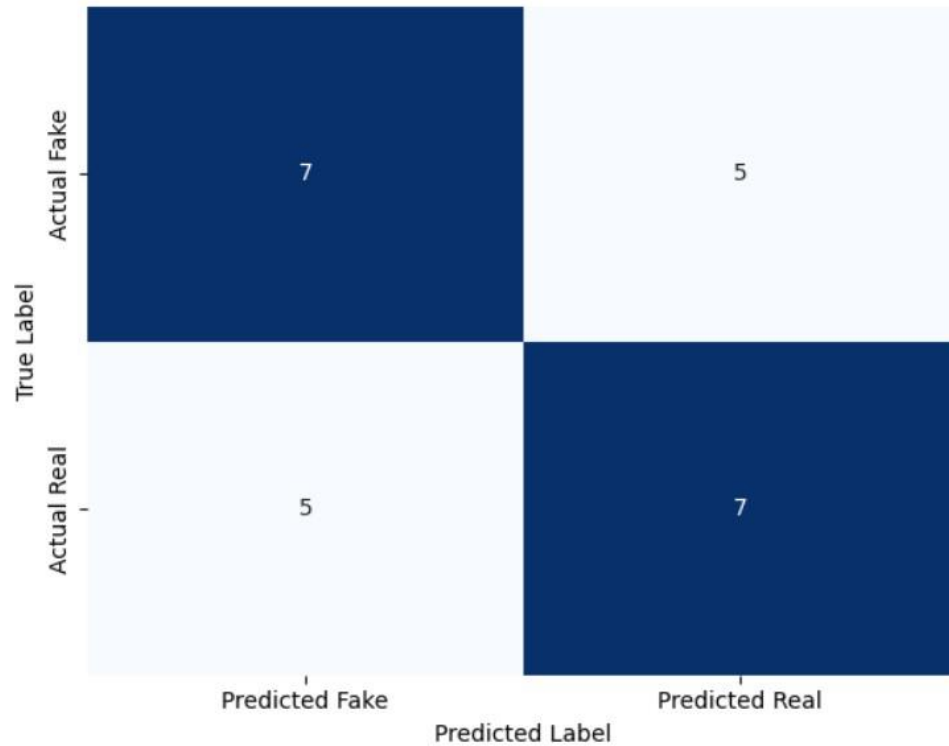


(1) CONVOLUTIONAL AE (15 MIN)

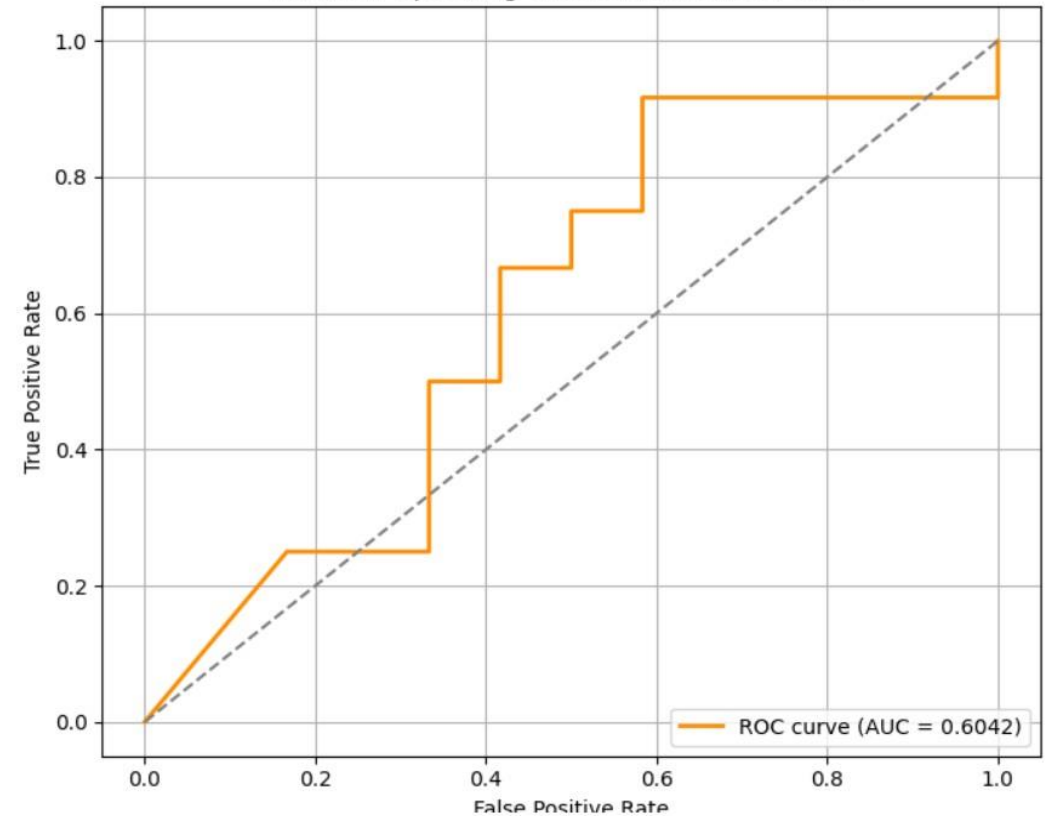
Classification Report on Test Set

Accuracy: 0.5833
Precision: 0.5833
Recall: 0.5833
F1 Score: 0.5833

Confusion Matrix



Receiver Operating Characteristic (ROC) Curve



(2) TRANSFER AE W/ HYBRID LOSS

Model Architecture

- Encoder: **Resnet 18** transferred encoder model
- Layered Decoder
- Fully Connected Classifier: **Global Average Pooling**

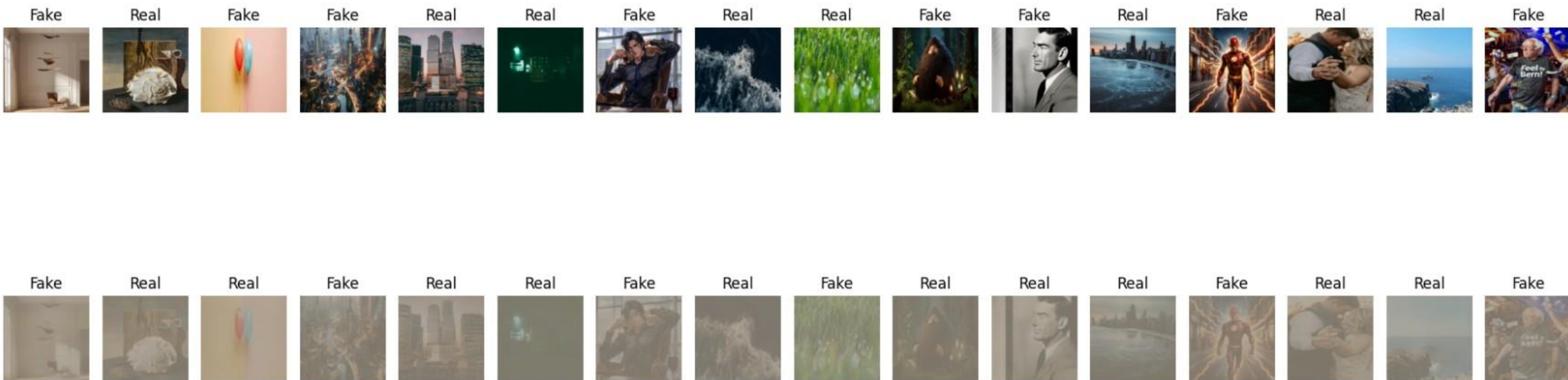
Loss Function

- Reconstruction Loss: **Hybrid Loss**
 - L1
 - SSIM
 - LPIPS
- Classification Loss: **Smoothed BCE Loss**
- Total Loss = recon loss * **r_weight** + class loss * **c_weight**

Model Additions

- Input Image size: **256 x 256**
- **Unfreeze** final 2 layers of transfer encoder during training
- **Freeze** classification loss on first 10% of epochs
- Trained on 5% sample, 200 epochs

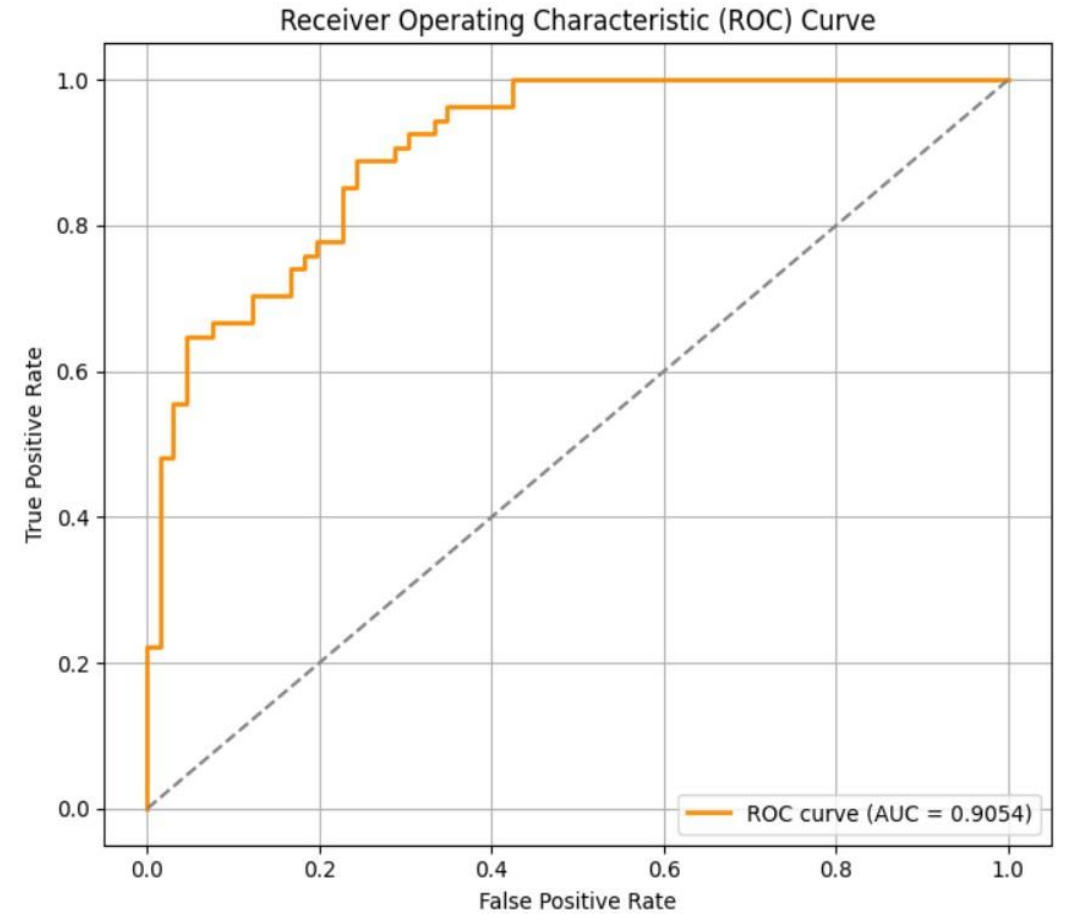
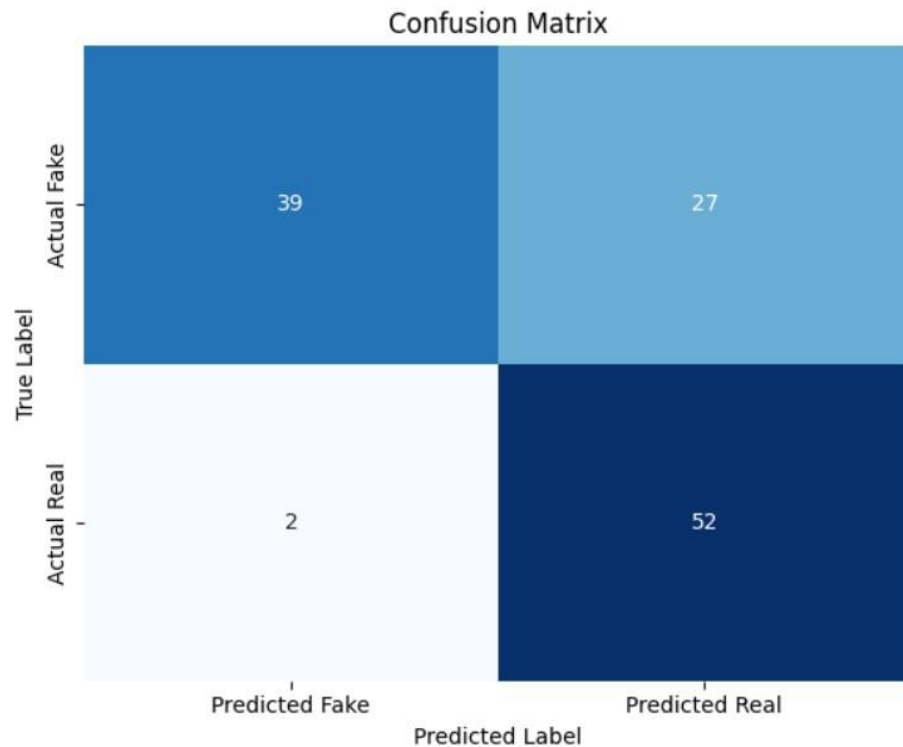
(2) TRANSFER AE W/ HYBRID LOSS



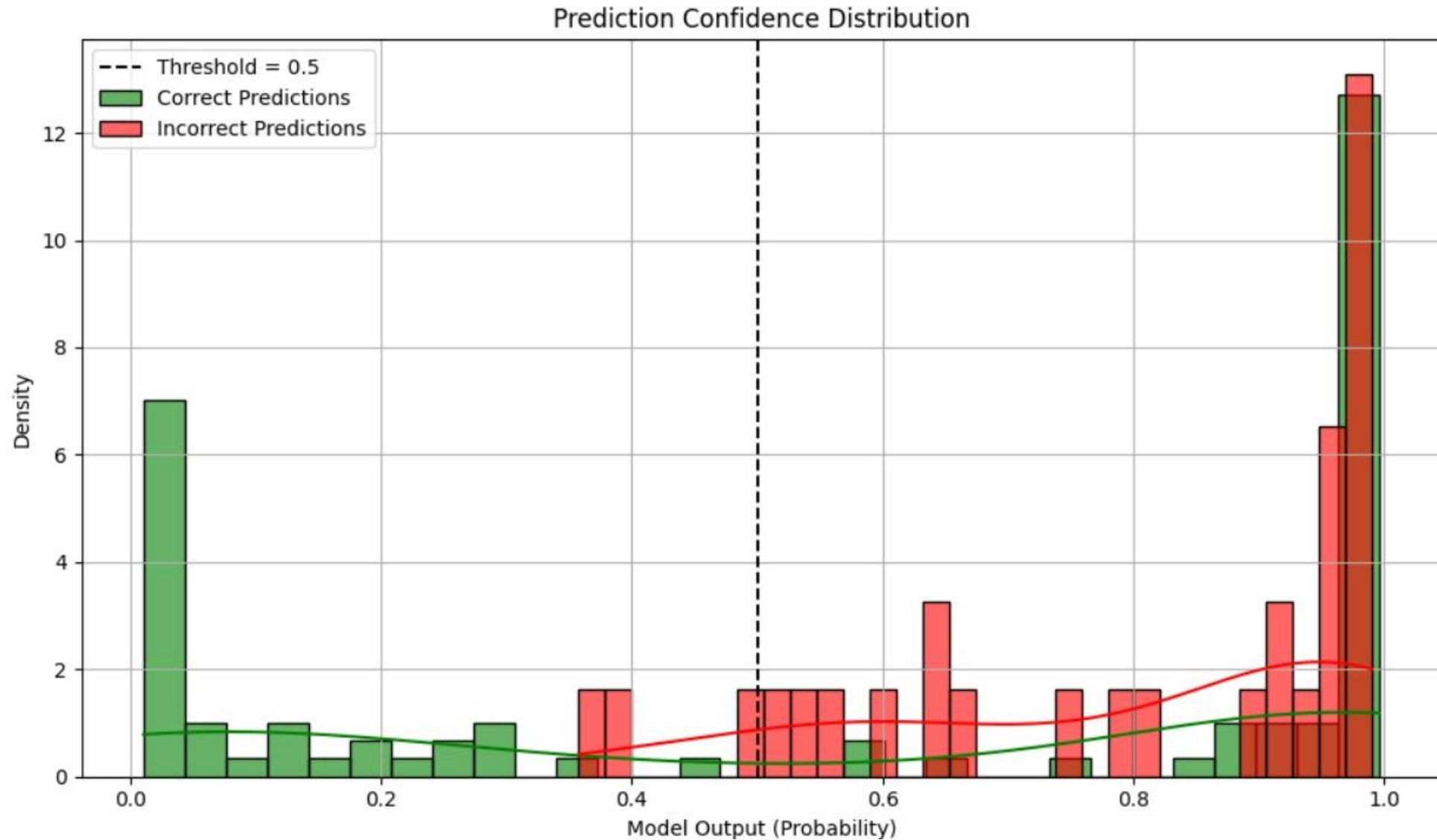
(2) TRANSFER AE W/ HYBRID LOSS (1.5 HOURS)

Classification Report on Test Set

Accuracy: 0.7583
Precision: 0.6582
Recall: 0.9630
F1 Score: 0.7820



(2) TRANSFER AE W/ HYBRID LOSS



(3) FOCAL LOSS W/ OPTUNA

Model Architecture

- Transferred Encoder: Unfrozen layers on training
- Layered Decoder: **Batch Normalization**
- Fully Connected Classifier: **Dropout Layers** & Global Average Pooling

Loss Function

- Reconstruction Loss: Hybrid Loss
- Classification Loss: **Focal** Loss
- Total Loss = recon loss * r_weight + class loss * c_weight (**Progressive weighting** and freeze epochs)

Model Additions


- Unfreeze final **3** layers of transfer encoder during training
- **Optuna Optimization**
- Trained on 5% sample, 200 epochs

(3) FOCAL LOSS W/ OPTUNA

Parameter Comparison:

alpha_r:

↳ Previous: 0.84

↳ New Best: 0.8199844782456824  CHANGED


beta:

↳ Previous: 0.15

↳ New Best: 0.23515562871287  CHANGED

gamma_r:

↳ Previous: 1.0

↳ New Best: 1.4612961670003108  CHANGED


gamma_c:

↳ Previous: 2.0

↳ New Best: 3.1038963555252  CHANGED

alpha_c:

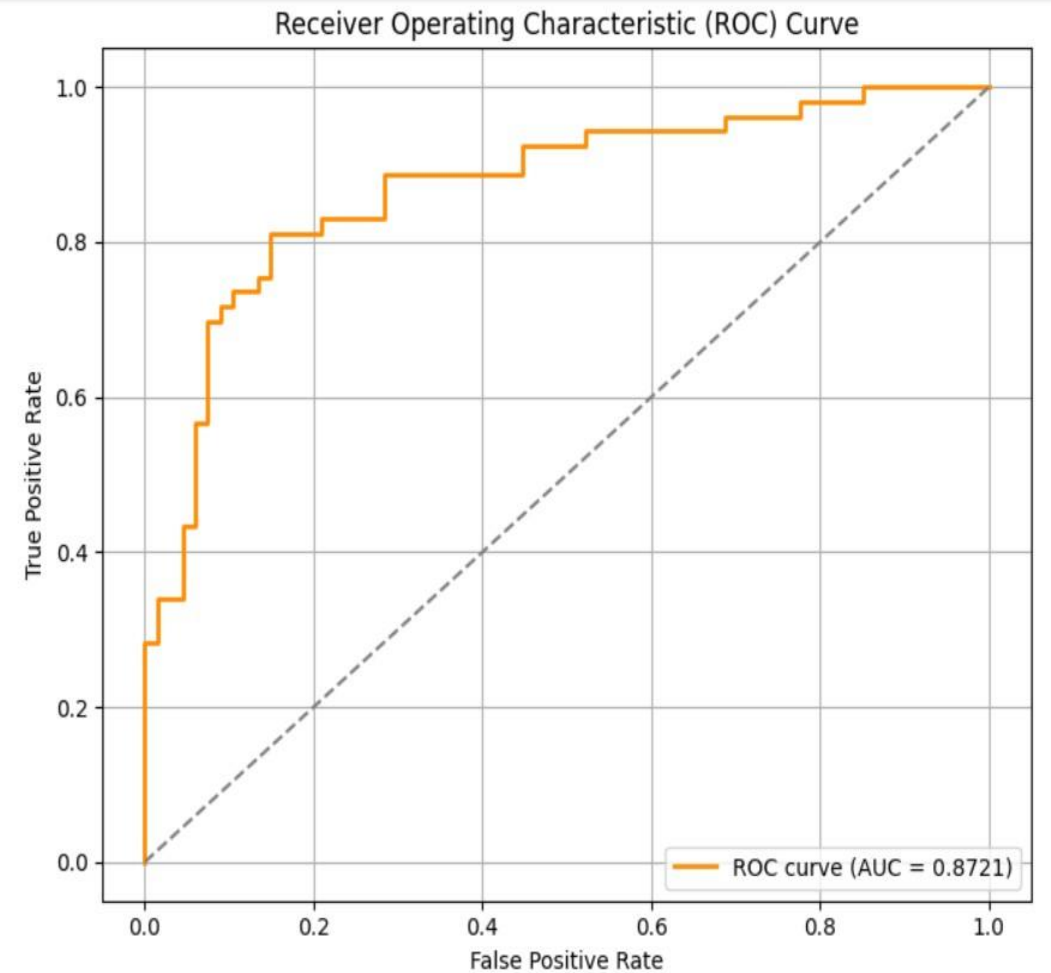
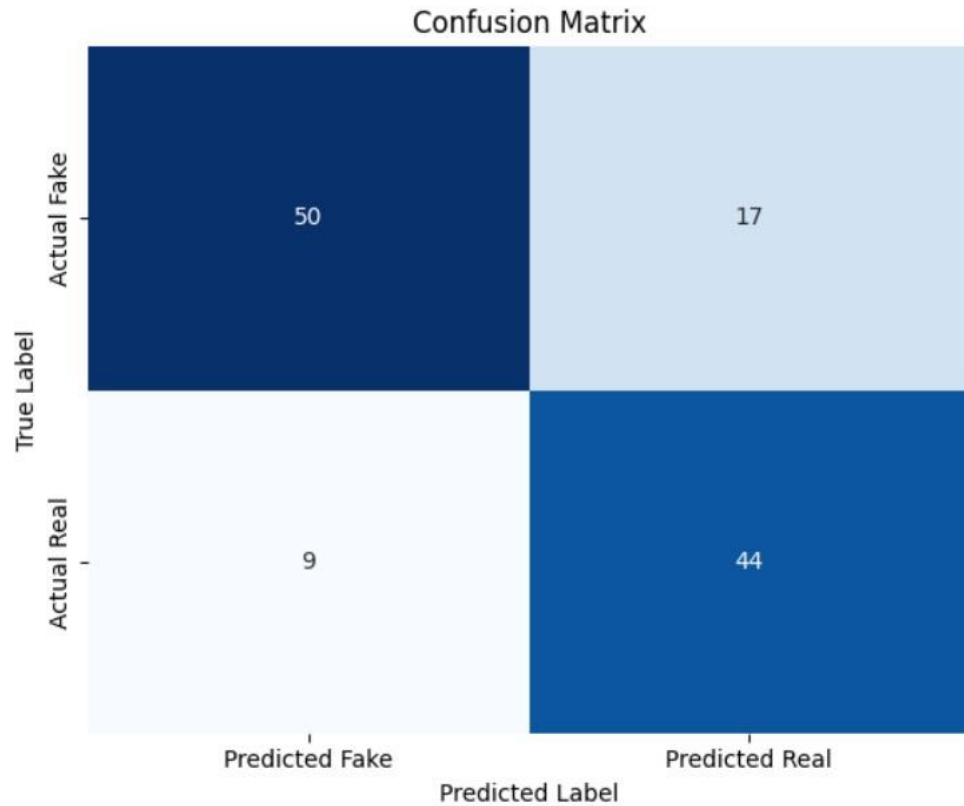
↳ Previous: 0.25

↳ New Best: 0.5518748596679164  CHANGED

* From 5% Sample, 10 epochs, 10 trials

(3) FOCAL LOSS W/ OPTUNA (3 HOURS)

```
warnings.warn(  
Classification Report on Test Set  
Accuracy: 0.7833  
Precision: 0.7213  
Recall: 0.8302  
F1 Score: 0.7719
```



(3) FOCAL LOSS W/ OPTUNA

Misclassified Samples — Original (Top) vs Reconstructed (Bottom)

Actual: Fake



Pred: Real
Conf: 0.85



Actual: Fake



Pred: Real
Conf: 0.51



Actual: Fake



Pred: Real
Conf: 0.55



Actual: Fake



Pred: Real
Conf: 0.50



Actual: Fake



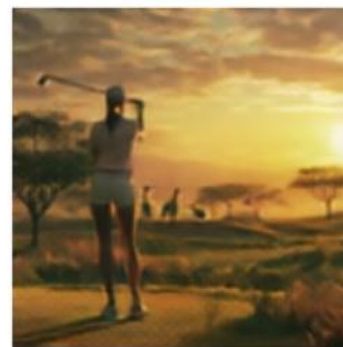
Pred: Real
Conf: 0.83



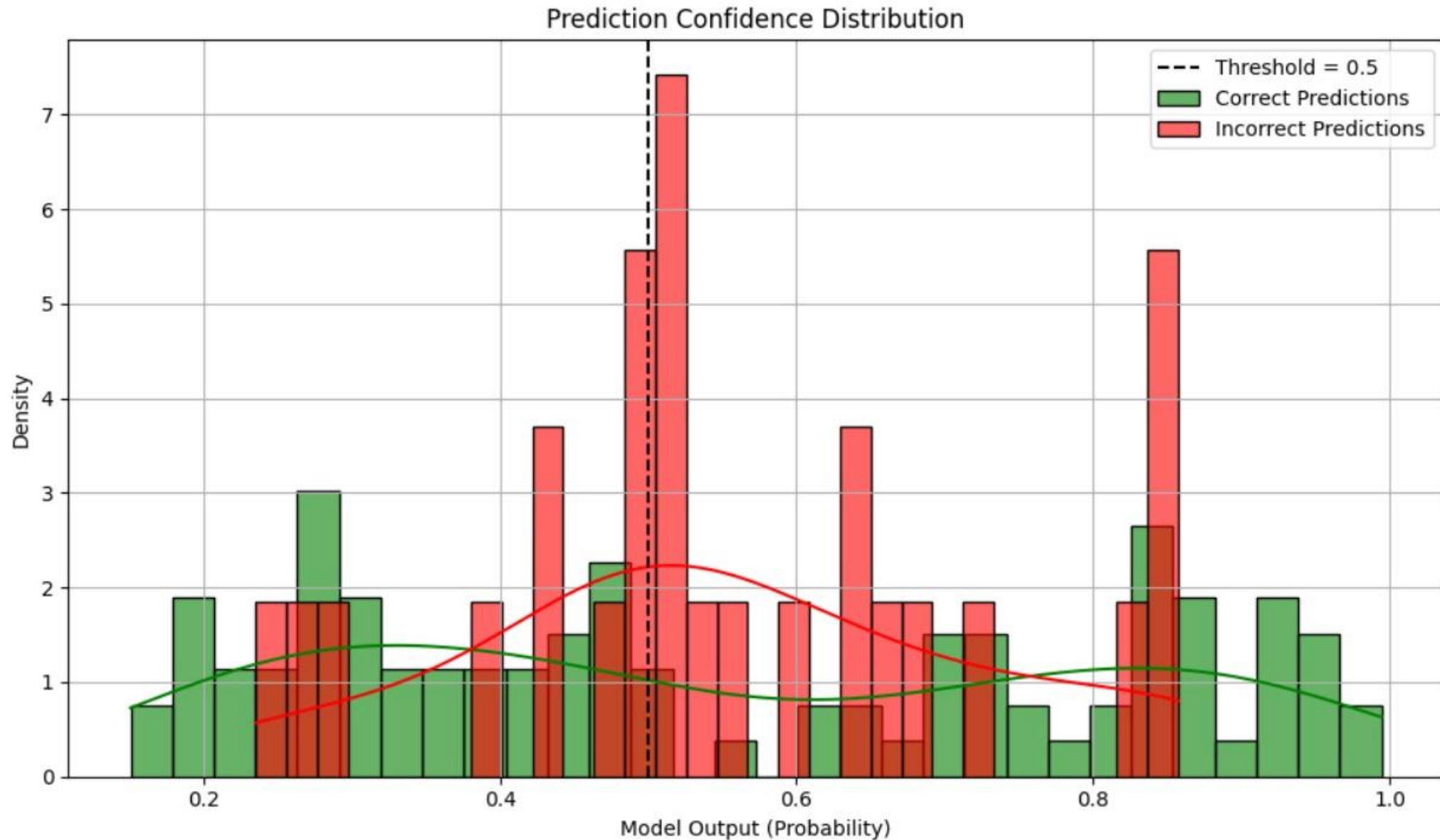
Actual: Fake



Pred: Real
Conf: 0.67



(3) HYBRID/FOCAL LOSS W/ OPTUNA



(4) HYBRID LOSS W/ OPTUNA & YAUDEN

Model Architecture

- Transferred Encoder: Unfrozen layers on training
- Layered Decoder: Batch Normalization
- Fully Connected Classifier: Dropout Layers & Global Average Pooling

Loss Function

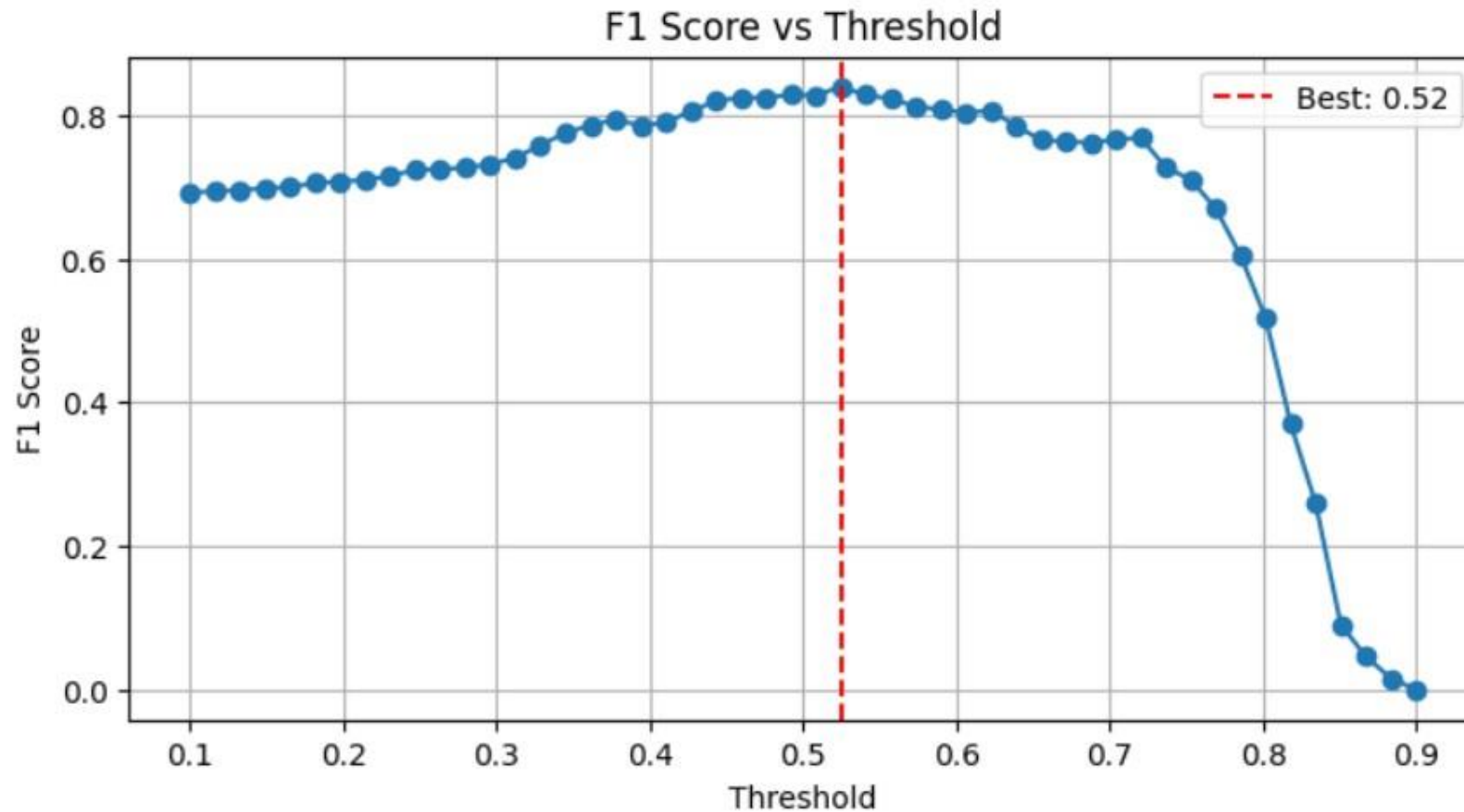
- Reconstruction Loss: Hybrid Loss
- Classification Loss: Focal Loss
- Total Loss = recon loss * r_weight + class loss * c_weight (Progressive weighting and freeze epochs)

Model Additions

- Unfreeze final 3 layers of transfer encoder during training
- **Youden Threshold**
- Trained on 10%, 50% and 100% sample, **100** epochs
- **Save Checkpoints**

(4) HYBRID LOSS W/ OPTUNA & YAUDEN

[↕]



- 🧠 Best F1 threshold: 0.5245
- 📌 Best threshold by Youden's J: 0.5282

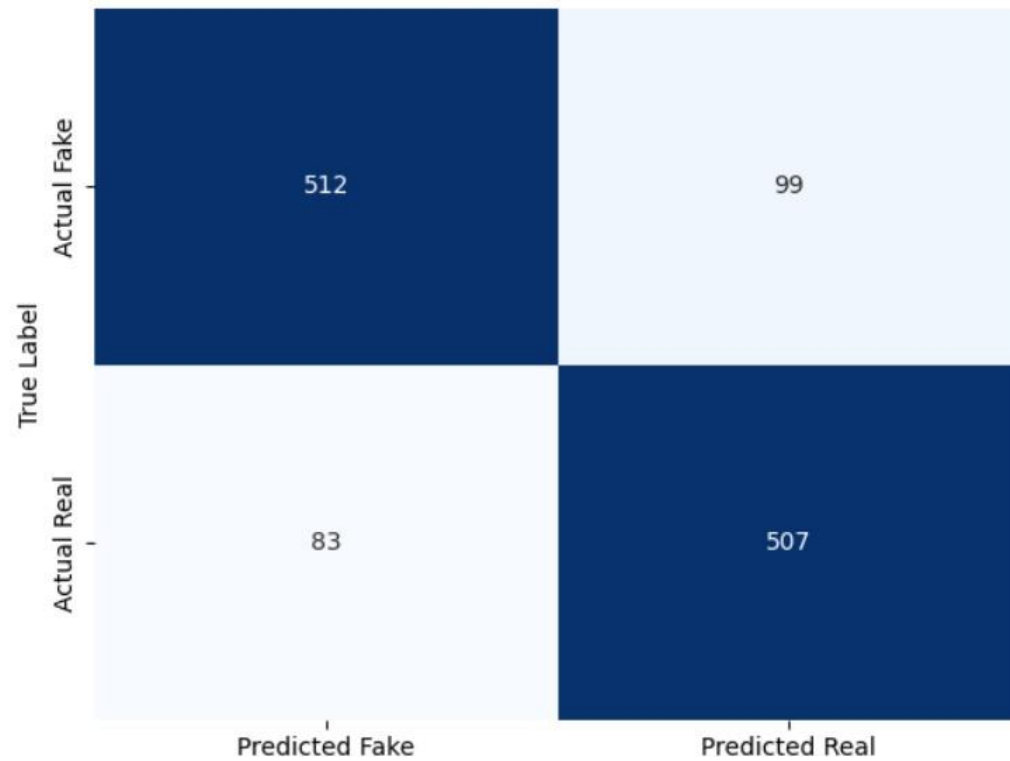
* From 10% Sample

(4) HYBRID LOSS W/ OPTUNA & YAUDEN (12 HOURS)

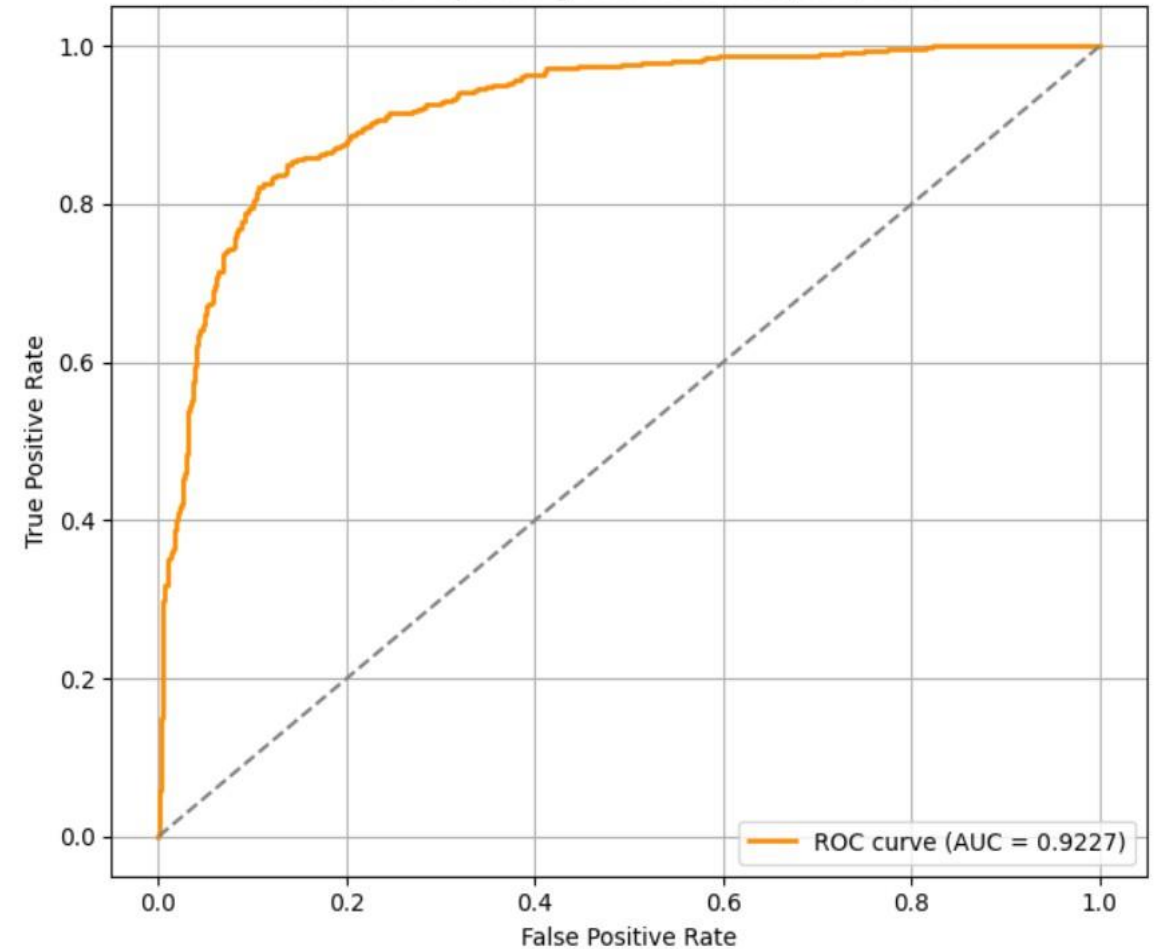
Classification Report on Test Set

Accuracy: 0.8485
Precision: 0.8366
Recall: 0.8593
F1 Score: 0.8478

Confusion Matrix



Receiver Operating Characteristic (ROC) Curve



(4) HYBRID LOSS W/ OPTUNA & YOUNG

Misclassified Samples — Original (Top) vs Reconstructed (Bottom)

Actual: Fake



Pred: Real
Conf: 0.53



Actual: Fake



Pred: Real
Conf: 0.57



Actual: Real



Pred: Fake
Conf: 0.44



Actual: Real



Pred: Fake
Conf: 0.46



Actual: Real



Pred: Fake
Conf: 0.49



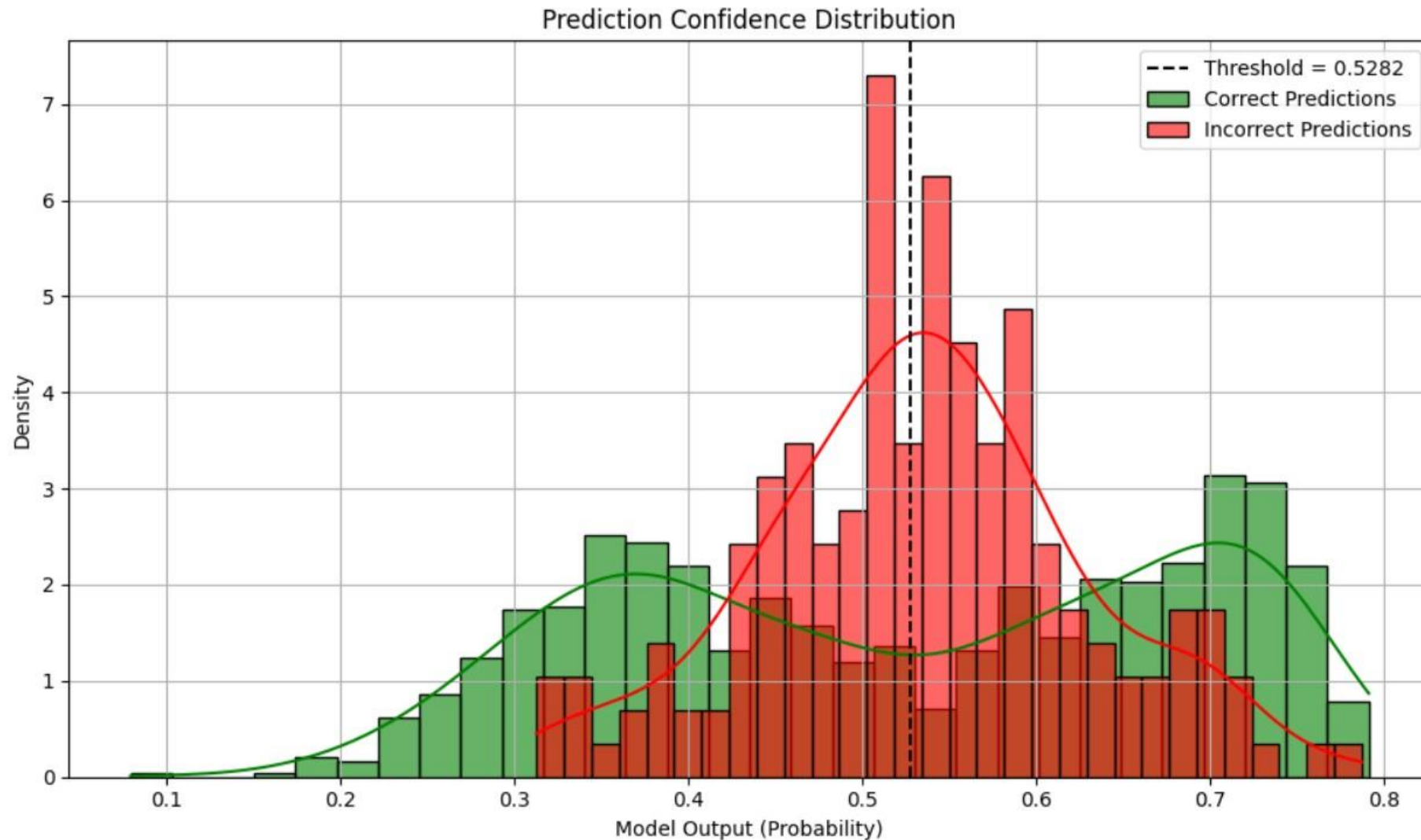
Actual: Real



Pred: Fake
Conf: 0.51



(4) HYBRID LOSS W/ OPTUNA & YAUDEN



(5) BEST MODEL FULL TRAINING (24 HOURS)

Classification Report on Test Set

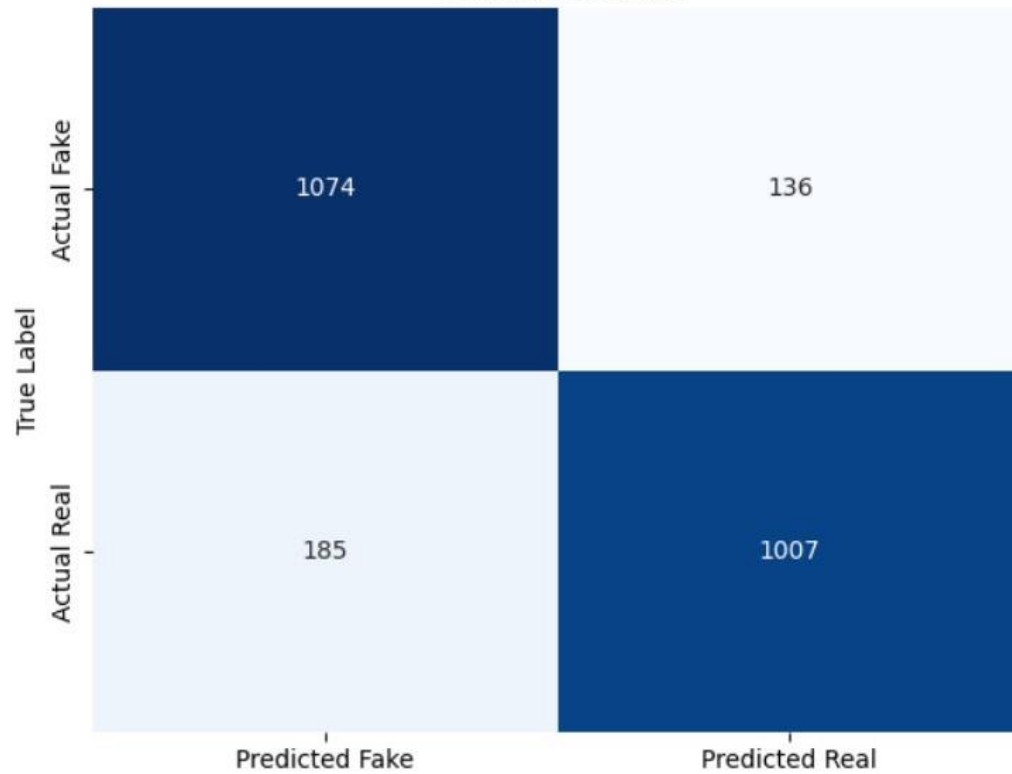
Accuracy: 0.8664

Precision: 0.8810

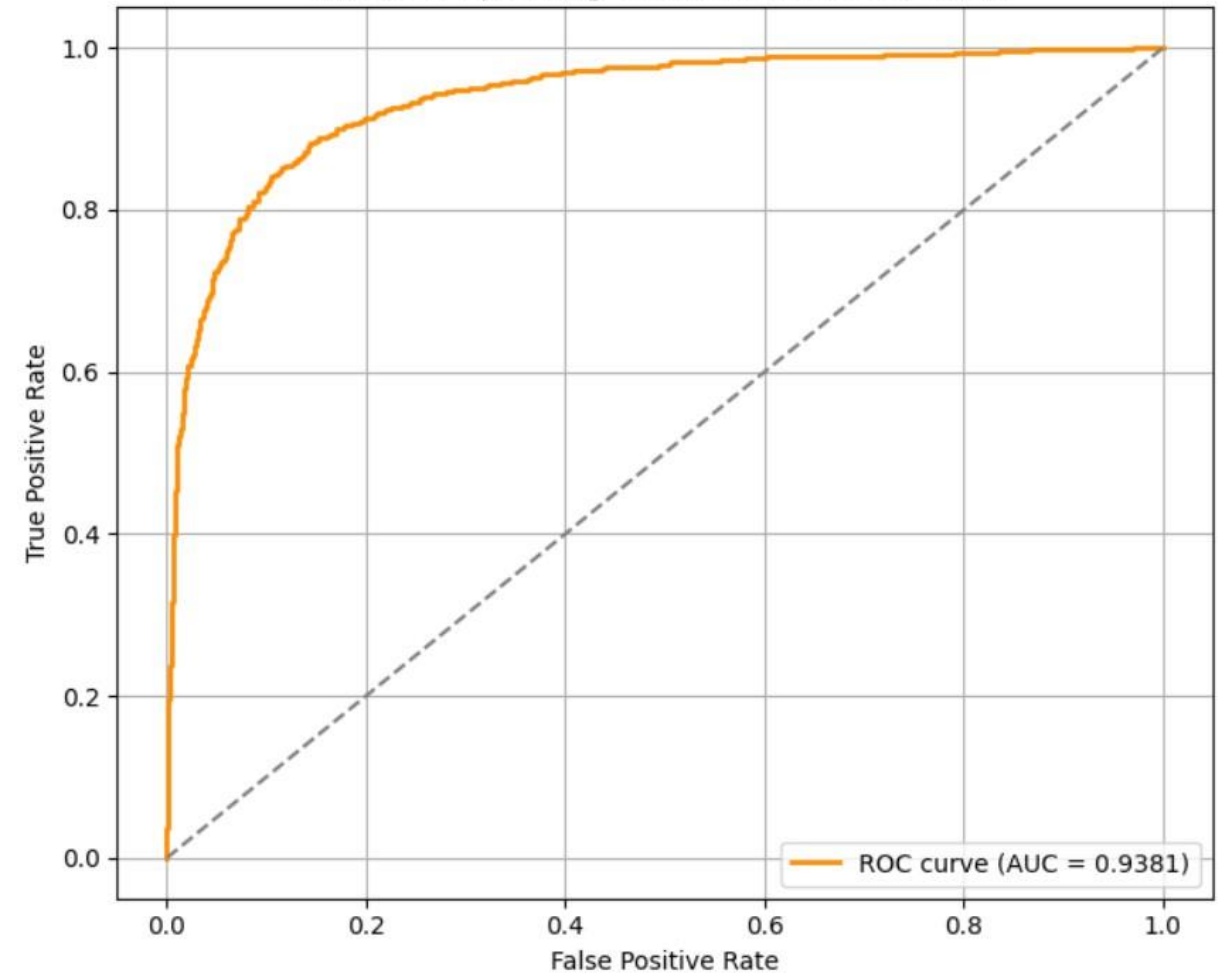
Recall: 0.8448

F1 Score: 0.8625

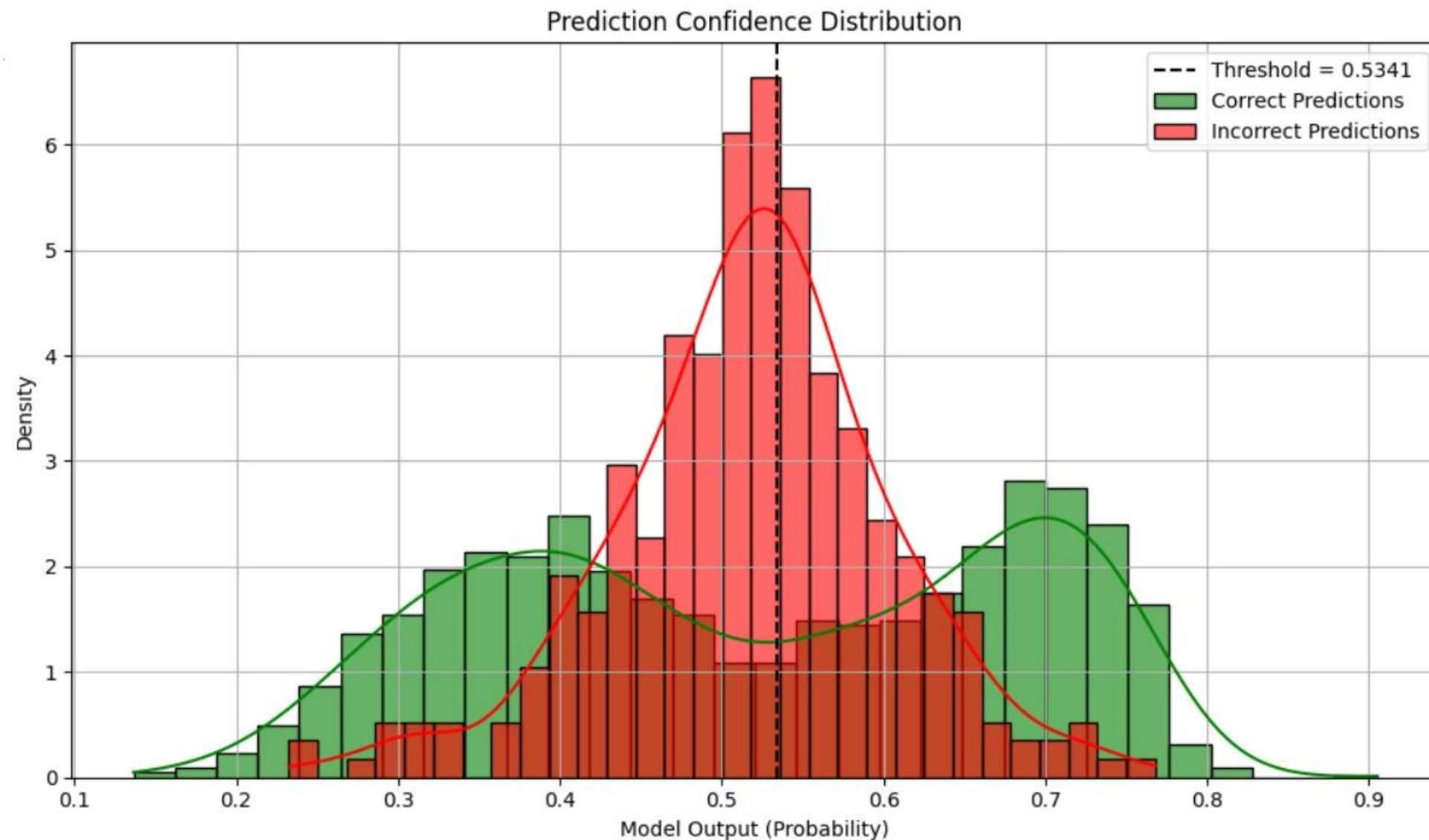
Confusion Matrix



Receiver Operating Characteristic (ROC) Curve



(5) BEST MODEL FULL TRAINING



GREYSCALE GENERAL MODEL SUMMARY

Model Architecture

- Transferred Encoder: Unfrozen layers on training
- Layered Decoder: Batch Normalization
- Fully Connected Classifier: Dropout Layers & Global Average Pooling

Loss Function

- Hybrid Loss w/ Optuna & Youden
- Reconstruction Loss: Hybrid Loss
- Classification Loss: Focal Loss
- Total Loss = recon loss * r_weight + class loss * c_weight (Progressive weighting and freeze epochs)

Model Features

- Grey-Scaled Images
- Unfreeze final 3 layers of transfer encoder during training
- Youden Threshold
- Trained on 5%, 10% and 100% sample, 100 epochs
- Save Checkpoints

OPTUNA

Parameter Comparison:


alpha_r:

↳ Previous: 0.87

↳ New Best: 0.5526043142841599  CHANGED

beta:

↳ Previous: 0.47

↳ New Best: 0.4041598185631064  CHANGED


gamma_r:

↳ Previous: 1.5

↳ New Best: 0.9653969778156798  CHANGED


gamma_c:

↳ Previous: 2.9

↳ New Best: 1.7976319360196373  CHANGED


alpha_c:

↳ Previous: 0.38

↳ New Best: 0.3582681521119193  CHANGED

class_weight:

↳ Previous: 0.26

↳ New Best: 0.2867933630943549  CHANGED

* From 5% Sample, 10
epochs, 10 trials

GREYSCALED (5% - 2 HRS)

Classification Report on Test Set

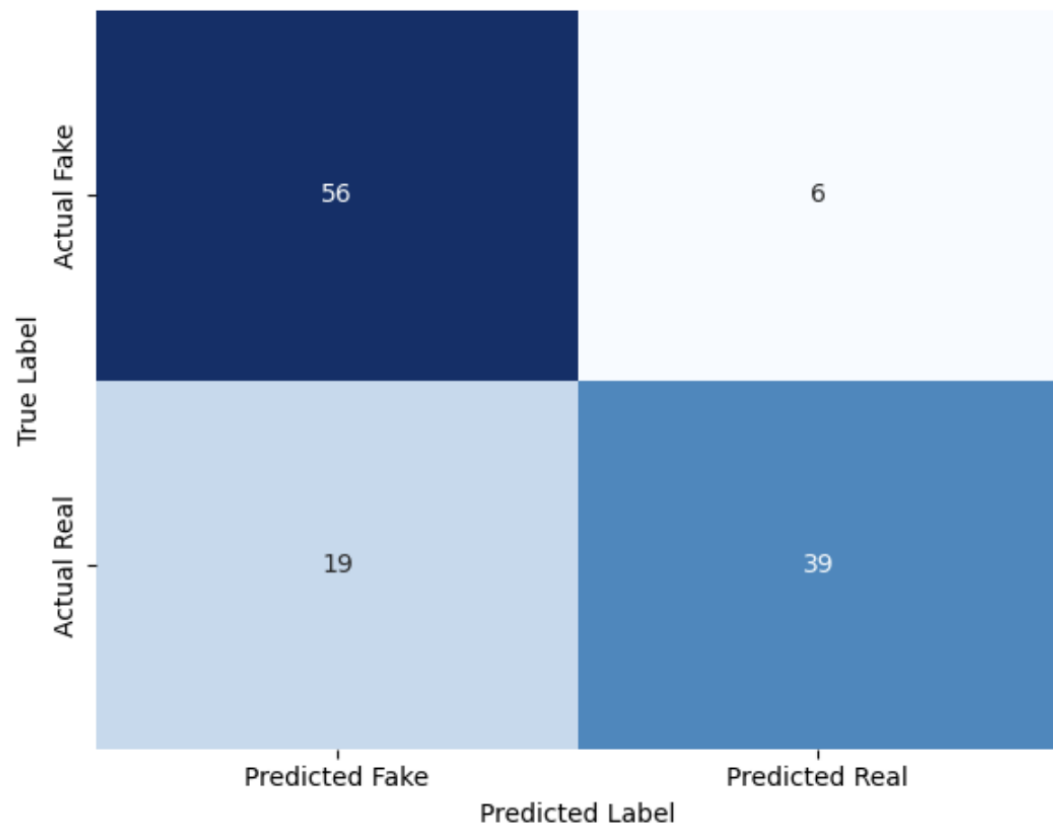
Accuracy: 0.7917

Precision: 0.8667

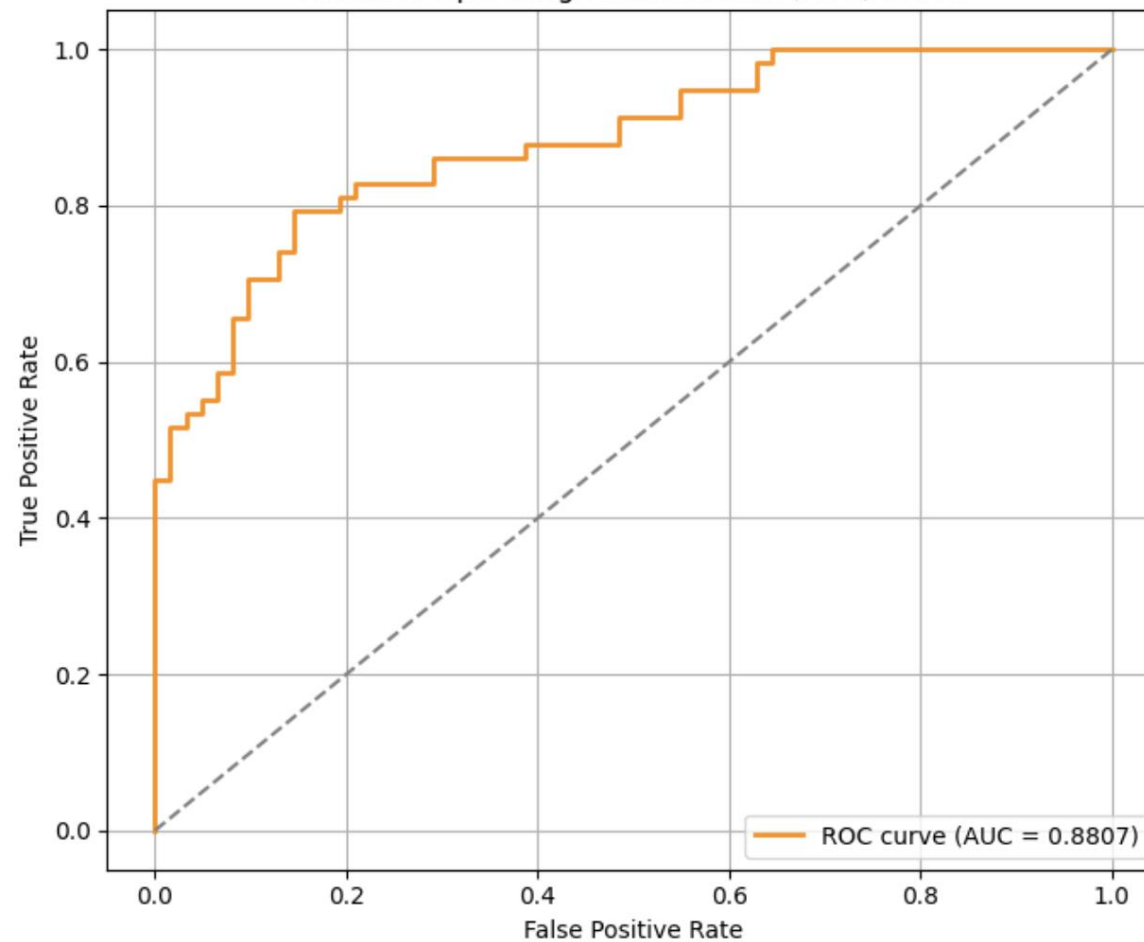
Recall: 0.6724

F1 Score: 0.7573

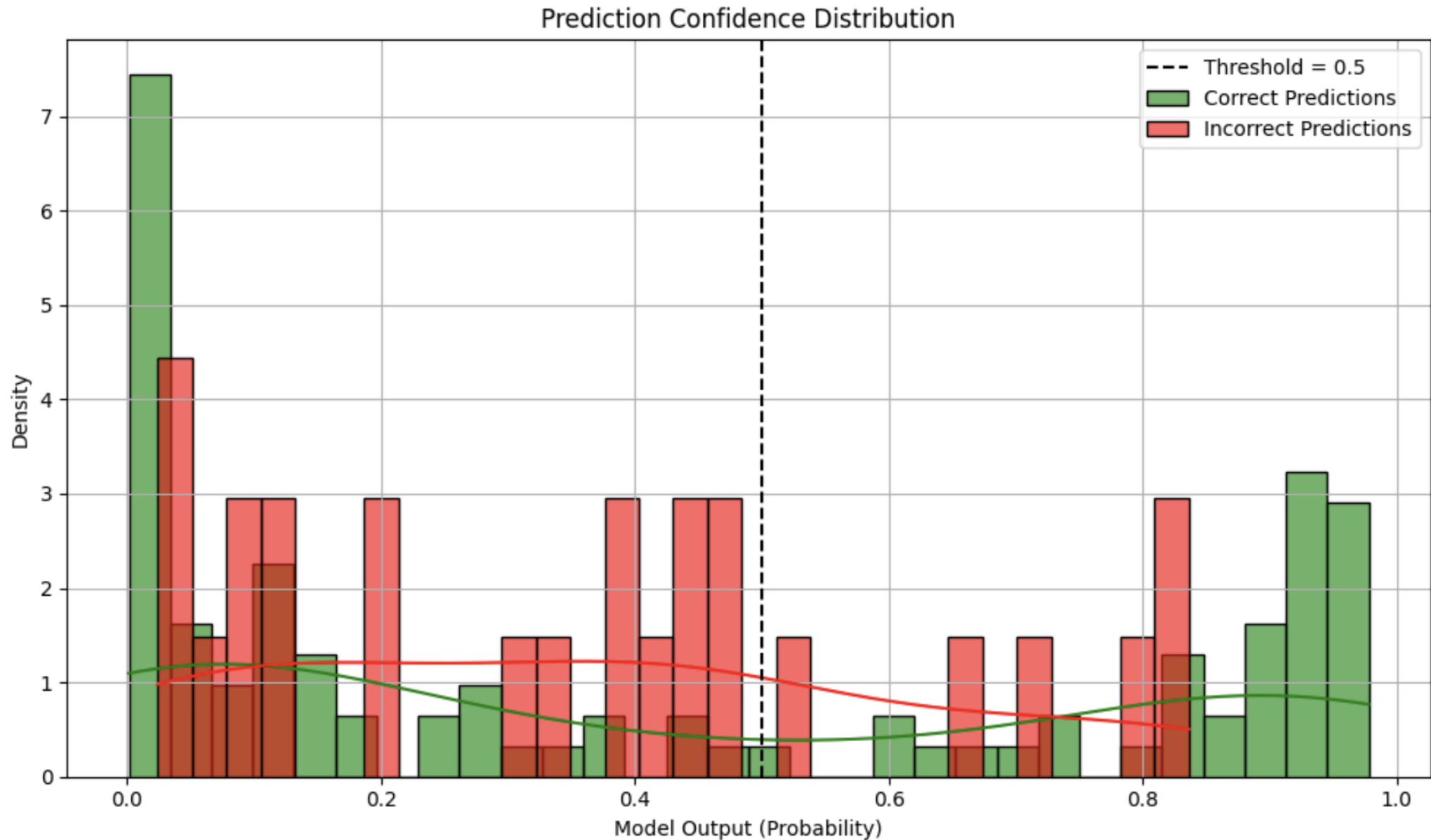
Confusion Matrix



Receiver Operating Characteristic (ROC) Curve



5% TRAINING (2 HOURS)



GREYSCALED (10% - 4 HRS)

Classification Report on Test Set

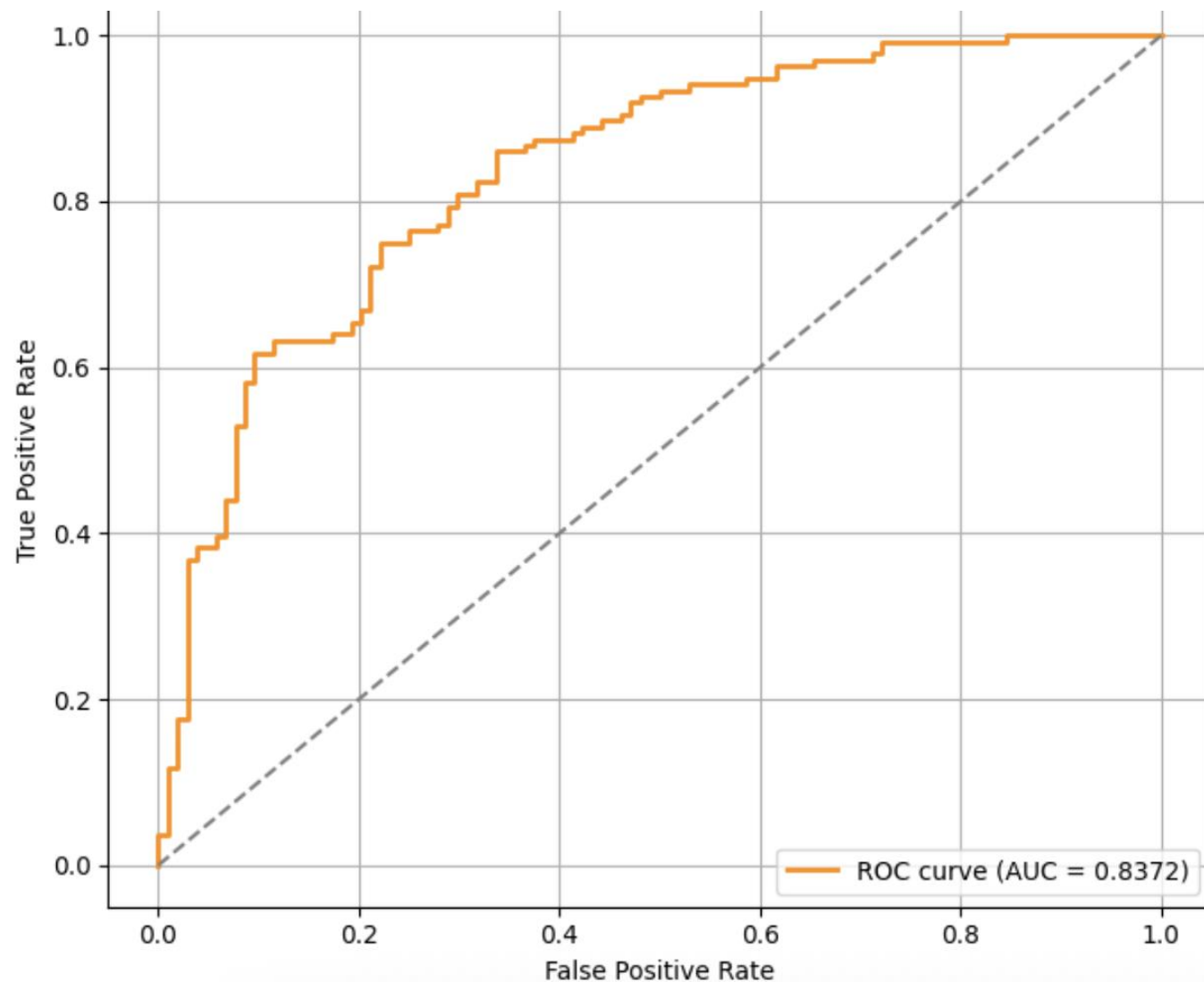
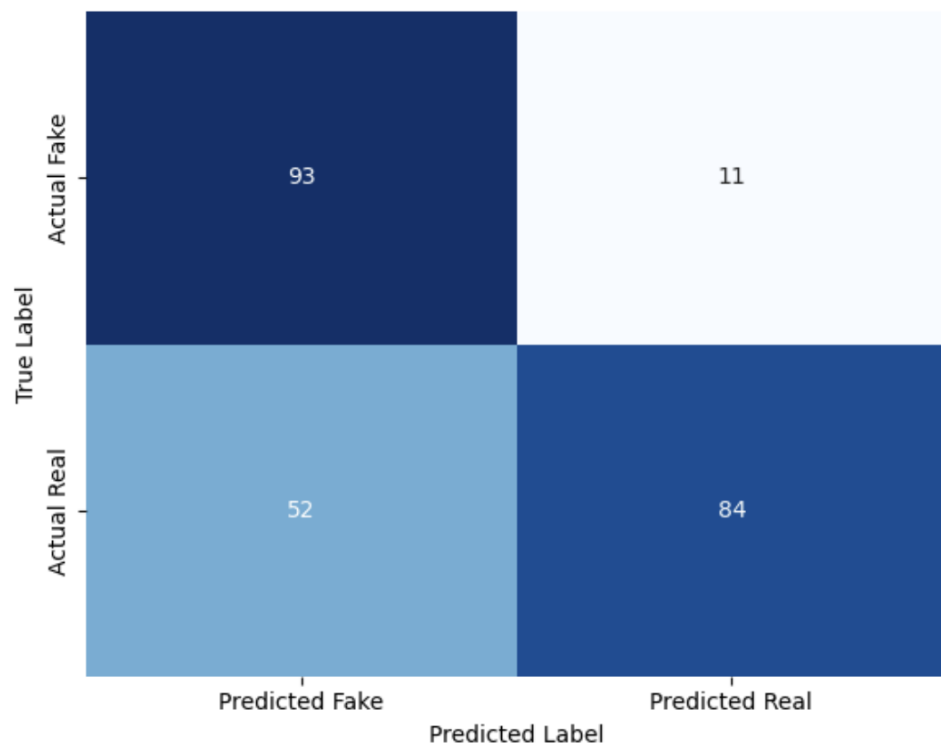
Accuracy: 0.7375

Precision: 0.8842

Recall: 0.6176

F1 Score: 0.7273

Confusion Matrix



GREYSCALED (10% - 4 HRS)

Real



Real



Real



Real



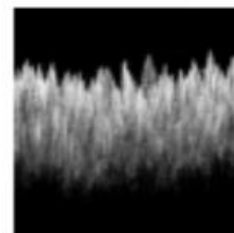
Fake



Real



Real



Real



Real



Real



Real



Real



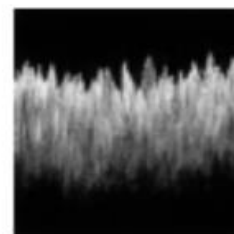
Fake



Fake



Real



Real



Prob: 0.97

Prob: 0.64

Prob: 0.97

Prob: 0.86

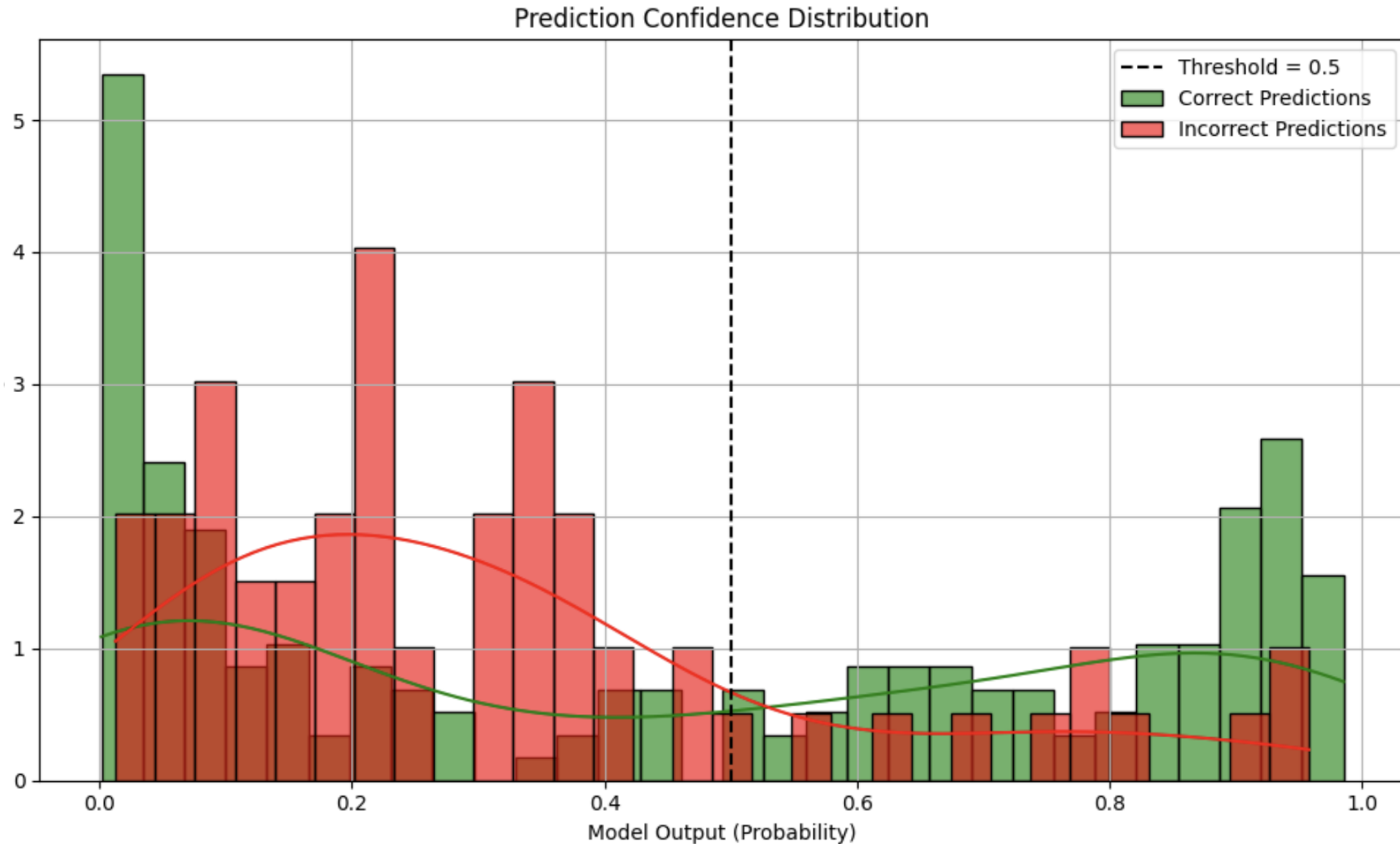
Prob: 0.03

Prob: 0.34

Prob: 0.83

Prob: 0.87

10% TRAINING (4 HOURS)



MODEL RESULTS/RUNTIMES

Sample Size	AE - Model 1	AE - Model 2	AE - Model 3	AE - Model 4	Grey Scaled
5% Sample	58% / 15m (1%)	75% / 1.5 h	78 % / 3 h	x	79% / 2 h
10% Sample	x	x	x	75 % / 5h	74% / 4 h
50% Sample	x	x	x	84.9 % / 12h	x
100% Sample	x	x	x	86.6 % / 24h	19h – accuracy not recorded

*Last semester best results with 100% sample: 83% / 6h (CNN – Pytorch)

OPPORTUNITIES FOR IMPROVEMENT

Image Distribution

Use an unbalanced dataset to prepare for real life applications

Greyscale Images

Test the RGB model with Greyscale images

Use Full Kaggle Set

Train on the entire Kaggle dataset. Our "full" training run is still only 20% of available images

Pre-Classify Images

Classify images into classes like dog, cat, face, etc. Then build one AE for each class. Multiple AEs for each class may work better



CONCLUSIONS

THANK YOU

