# Tool for detecting deepfakes based on biological factors

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# MOTIVATION

- Deepfakes threaten media trust and privacy.
- Biological signals are hard to fake.
- Study the features of biological signals.
- Study breathing patterns for detection deepfakes.

# **IMAGE-BASED**

- Extract three face regions.
- Transform these face regions.
- Extract rPPG using the CHROM method.



# TECH & DATASET

- PhysNet for extract PPG signal
- FaceForensics++ for training models
- Celeb-DF (V2) for testing models
- PyTorch
- MTCNN face detector for sequence-based method
- dlib face detector for image-based method

### **IMPLEMENTATION**

ROI Frame Sequence

- Generate PPG maps per region.
- Classify maps using a CNN.



Figure 2: Facial regions selected for PPG map for image-based approach.

# **SEQUENCE-BASED**

- Extract full face region.
- Model breathing pattern with LSTM/GRU.
- Predict labels based on rPPG signal dynamics.



Figure 3: Facial regions selected for CNN approach.



Figure 1: Breath rate estimation model architecture.

- Extract face regions frame sequences from video.
- Process frame sequence.
- Extrac rPPG signal from facial regions.
- Filter signal to isolate breathing pattern.
- Classify using CNN or RNN models.



Table 1: Cross-dataset evaluation of model performance on FaceForensics++ and Celeb-









Figure 5: Comparison of the power spectral density of a real video (left) and a deepfake video (right).



#### **EXPERIMENT RESULTS**

rancoatener	55.5
Inconsistent head poses	89.0
DeepVision	87.5
DeepLie	72.7
Image-based (CNN)	60.7
Sequence-based (LSTM)	54.7
Sequence-based (GRU)	53.2

Figure 4: Output of the PhysNet model for a sequence of duplicated static frames.

Figure 3: Comparison of the PPG signal of a real video (top) and a deepfake video (bottom)

 

 Table 2: Comparison of different deepfake

detection methods. The names of our models are highlighted in bold.



