

## DeepFakesON-Phys: DeepFakes Detection based on Heart Rate Estimation



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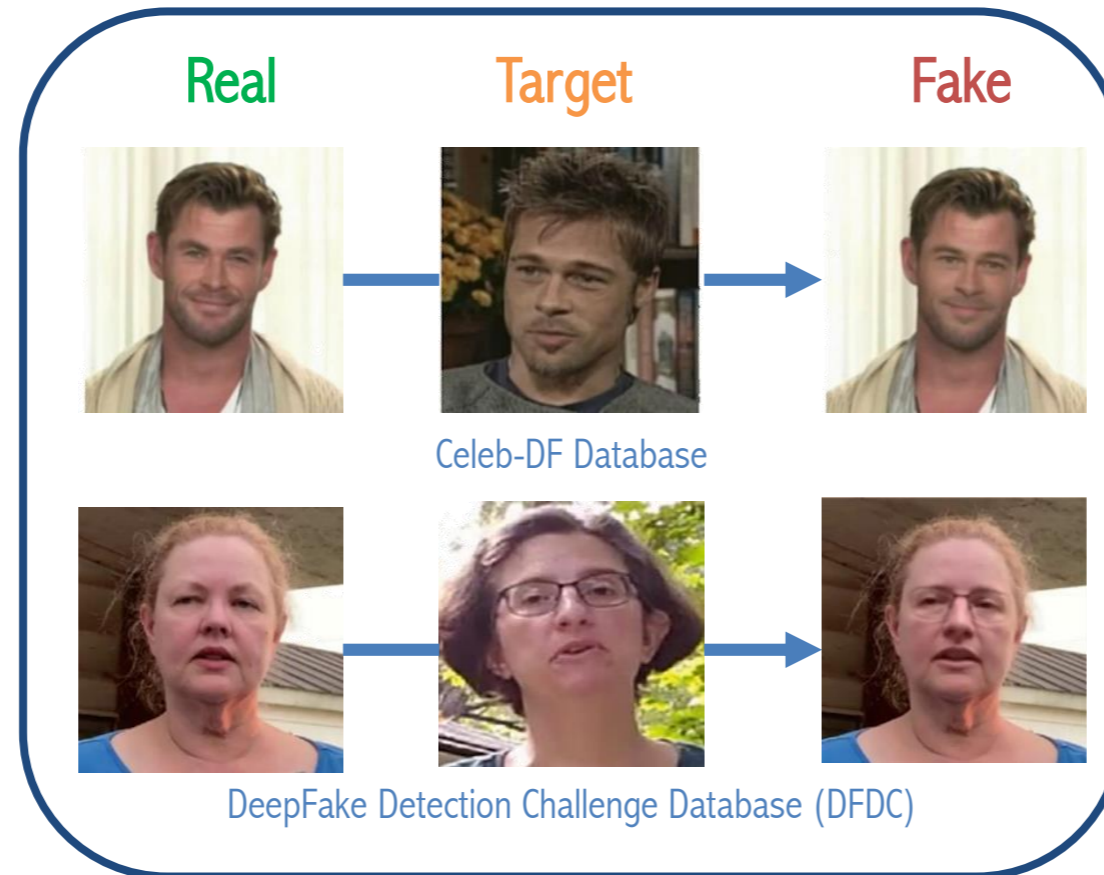
Prof. Julian  
FIERREZ



Prof. Aythami  
MORALES

# Introduction

- **DeepFake (Identity Swap)** is referred to a deep learning based technique able to create fake videos by **swapping** the face of a person by the face of another person [1].



[1] Tolosana, R.; Vera-Rodriguez, R.; Fierrez, J.; Morales, A.; and Ortega-Garcia, J. 2020. “DeepFakes and Beyond: A Survey of Face Manipulation and Fake Detection”. *Information Fusion* 64: 131–148.

# Introduction

- **Face manipulation techniques:** mostly based on AutoEncoders (AE) [2] and Generative Adversarial Networks (GAN) [3].
- **Very realistic visual results:** specially in the latest generation of public DeepFakes [4].

**Real** Video  
(Robert de Niro)



**DeepFake** Video  
(Al Pacino)

[2] Kingma, D. P.; and Welling, M. 2013. “Auto-Encoding Variational Bayes”. In *Proc. Int. Conf. on Learning Represent.*

[3] Goodfellow, I.; Pouget-Abadie, J.; Mirza, M.; Xu, B.; Warde-Farley, D.; Ozair, S.; Courville, A.; and Bengio, Y. 2014. “Generative Adversarial Nets”. In *Proc. Advances in Neural Information Processing Systems.*

[4] Tolosana, R.; Romero-Tapiador, S.; Fierrez, J.; and Vera-Rodriguez, R. 2020. “DeepFakes Evolution: Analysis of Facial Regions and Fake Detection Performance”. In *Proc. International Conference on Pattern Recognition Workshops.*

# Introduction

- **Face Recognition Presentation Attack:** using photographs, videos, and masks [5].



- **3D Masks** : somehow similar to DeepFake digital manipulations.
  - Physical vs digital mask over the real face.
- **Texture and shape**-based techniques **not efficient** against hyperrealistic 3D Masks [6].
  - Same with realistic DeepFake methods.
  - Other approaches are necessary → Physiology.

[5] Hernandez-Ortega, J.; Fierrez, J.; Morales, A.; and Galbally, J. 2019. "Introduction to Face Presentation Attack Detection". In *Handbook of Biometric Anti-Spoofing*, 187–206. Springer.

[6] Erdogmus, N.; and Marcel, S. 2014. "Spoofing Face Recognition with 3D Masks". *IEEE Transactions on Information Forensics and Security* 9(7): 1084–1097.

# Introduction

- **3D Masks do not emulate the physiology of human beings** [6], i.e. HR, blood oxygen, breath rate.
  - **Estimating them** is a powerful tool for 3D Masks detection.
- **Do DeepFake manipulations consider the physiological aspects in the synthesis process?**
- Detection based on pulse detection → **Remote Photoplethysmography** [7], used in:
  - E-learning [Hernandez-Ortega *et al.* 2020].
  - Health Care [Mc-Duff *et al.* 2015].
  - Human-Computer Interaction [Tan and Nijholt 2010].
  - Security [Marcel *et al.* 2019].



[7] Hernandez-Ortega, J.; Fierrez, J.; Morales, A.; and Tome, P. 2018. "Time Analysis of Pulse-Based Face Anti-Spoofing in Visible and NIR". In *Proc. IEEE Conf. on Comp. Vision and Pattern Recognition Workshops*.

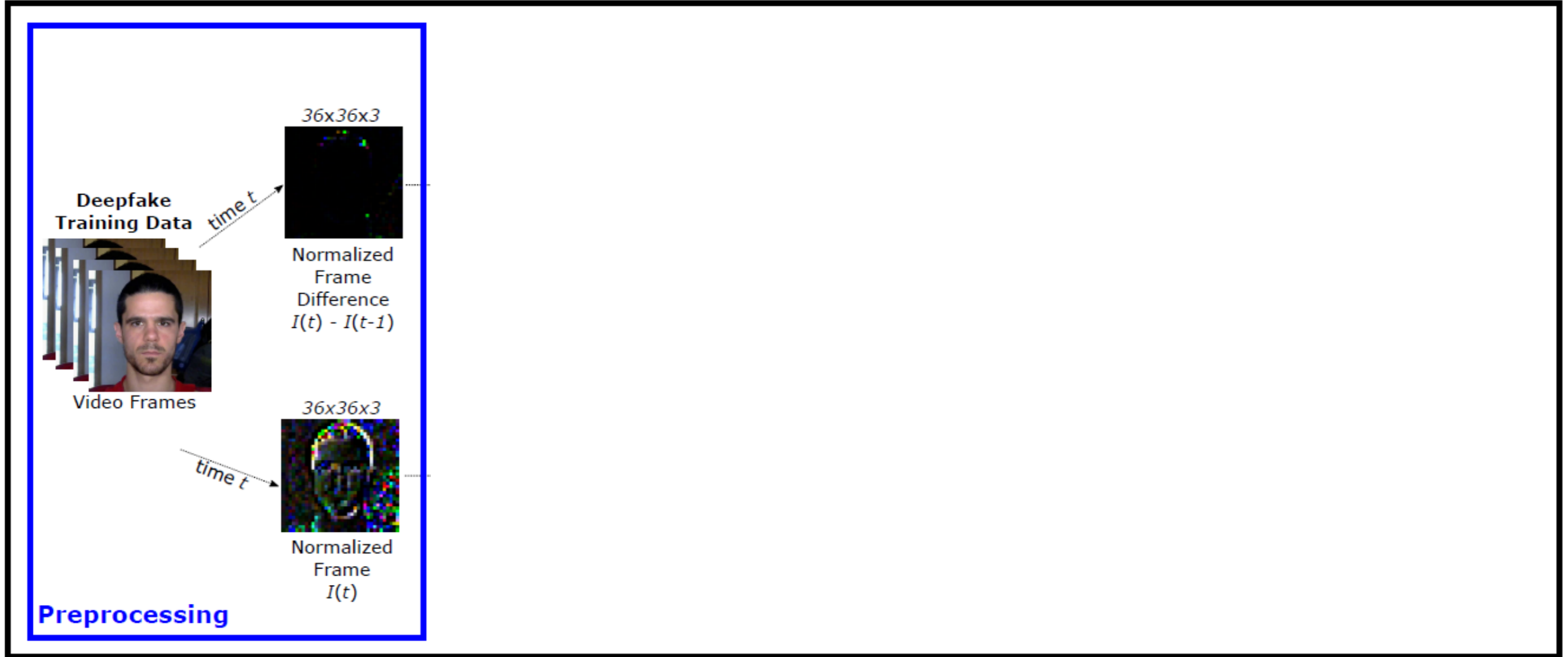
# Contributions

- **DeepFake detector based on physiological measurement: DeepFakesON-Phys.**
  - Based on Deep Learning.
  - rPPG features pretrained for heart rate estimation.
  - Adapted using knowledge transfer.
  - Information related to the heart rate → **Real** or **Fake**.
- Trained and tested with 2<sup>nd</sup> generation DeepFake DBs:
  - **DFDC Preview.**
  - **Celeb-DF v2.**

**DeepFakesON-Phys** → solution to the weaknesses of detectors based on the visual artifacts and fingerprints inserted during the synthesis process.

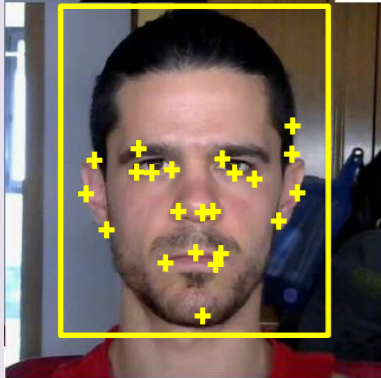
# Proposed Framework

DeepFakesON-Phys



# Preprocessing

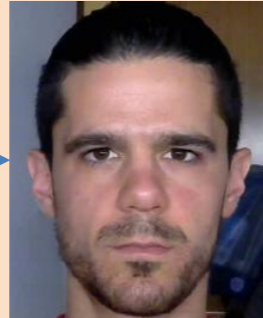
## 1. Face Detection & Tracking



MTCNN Face Detector  
&  
KLT Feature Tracker

## 2.1 Normalized Frame

Face Frame  
 $F(t)$



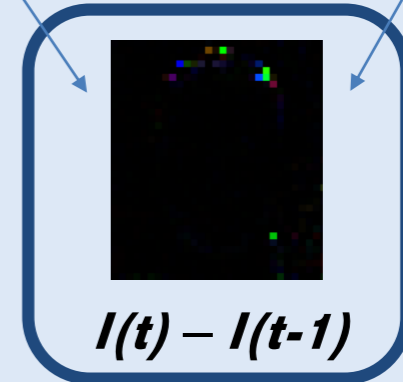
Normalized Frame  
 $I(t)$ :



$$I(t) = (F(t+1) - F(t)) / (F(t+1) + F(t))$$

## 2.2 Normalized Frame Difference

Normalized Frames

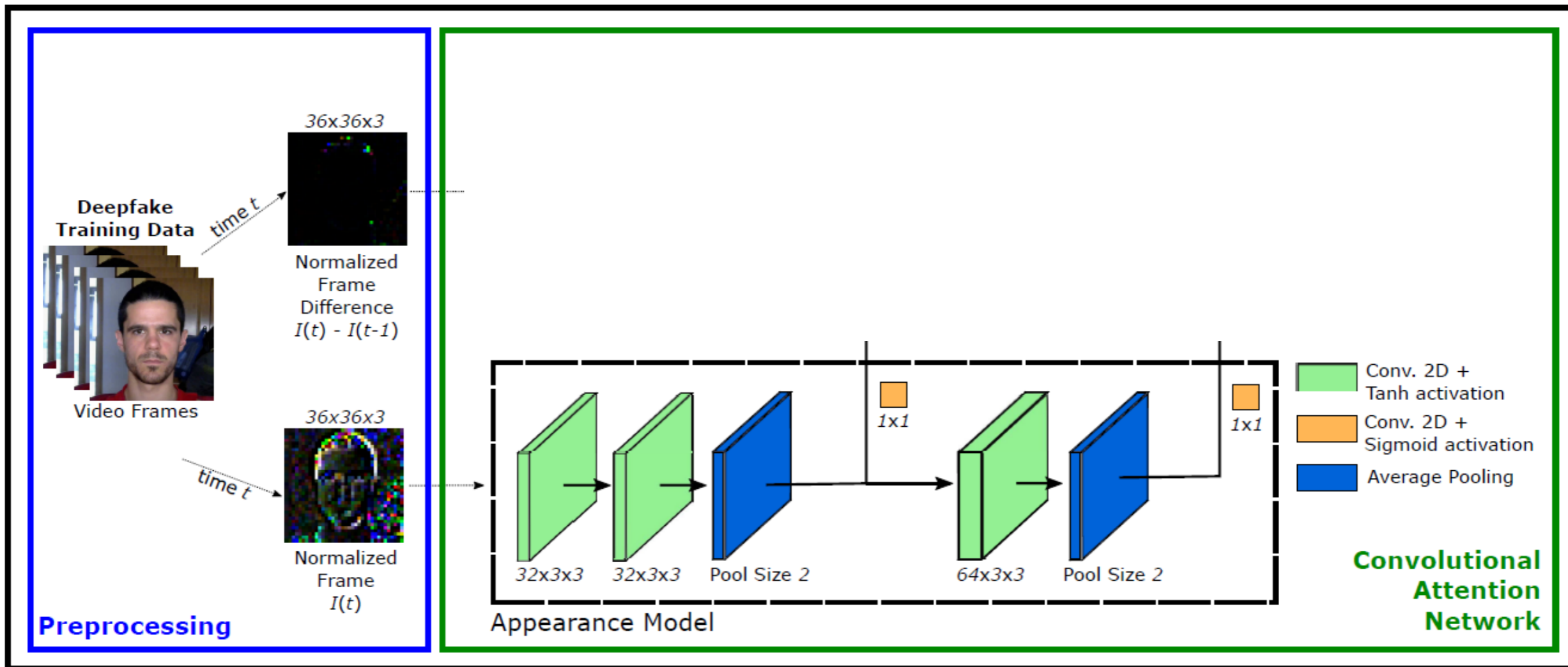


Normalized Frame Difference



# Proposed Framework

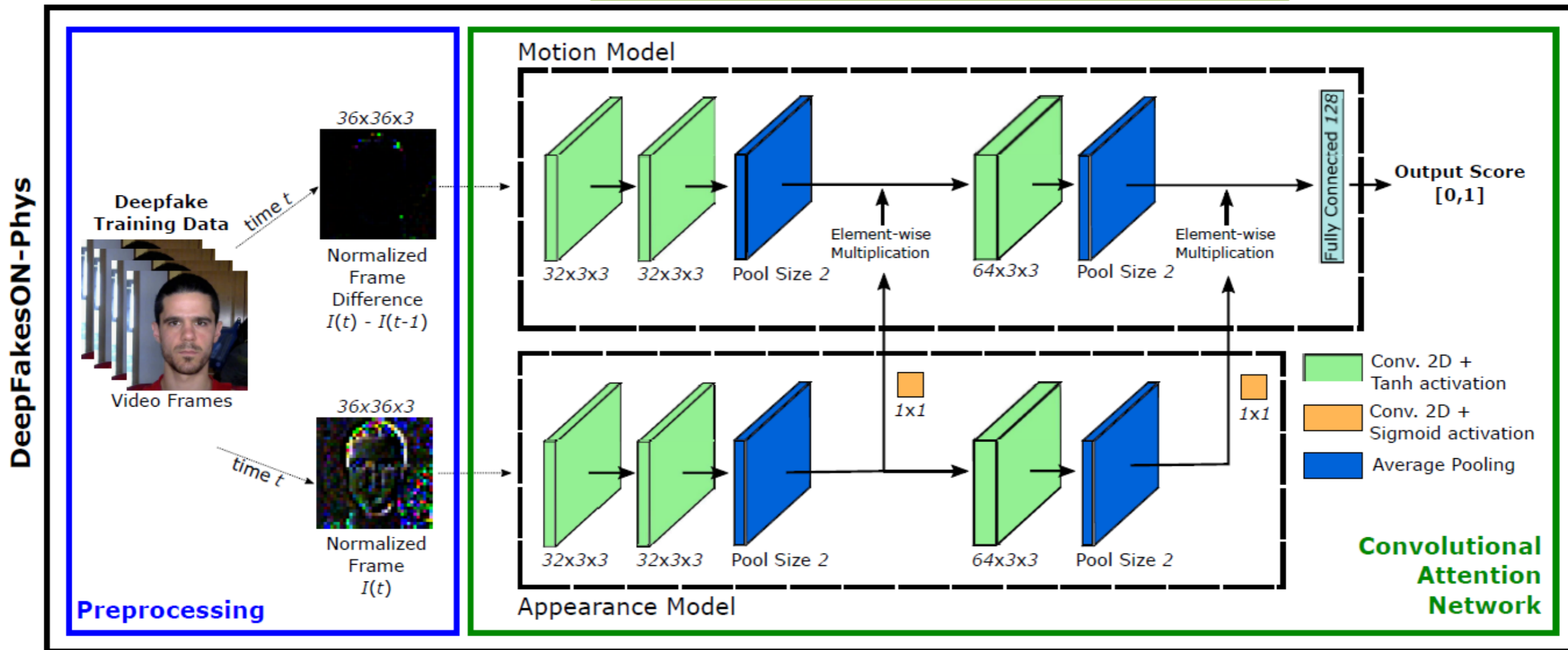
DeepFakesON-Phys



Appearance model: static information  $\rightarrow$  Attention

# Proposed Framework

Motion model: temporal information + attention



Appearance model: static information  $\rightarrow$  Attention

# Databases — 2nd Generation

## Celeb-DF v2 [9]

- 590 real (Youtube)
- 5,639 fake videos (Deep Learning)



## DFDC Preview [10]

- 1,131 real (Actors)
- 4,139 fake videos (Various)



[9] Y. Li, X. Yang, P. Sun, H. Qi, and S. Lyu. 2020. “Celeb-DF: A LargeScale Challenging Dataset for DeepFake Forensics”. In *Proc. IEEE/CVF Conf. on Computer Vision and Pattern Recognition (CVPR)*.

[10] B. Dolhansky, R. Howes, B. Pflaum, N. Baram, and C. C. Ferrer. 2019. “The Deepfake Detection Challenge (DFDC) Preview Dataset”. *arXiv preprint.:1910.08854*.

# DeepFakesON-Phys: Development

1) Model **based on DeepPhys** [11] (Heart rate from facial video) → **Not public.**

[11] W. Chen, and D. McDuff. 2018. “Deepphys: Video-based Physiological Measurement using Convolutional Attention Networks”. In *Procs. of the European Conf. on Computer Vision (ECCV)*.

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[12] J. Hernandez-Ortega, *et al.* 2020. “A Comparative Evaluation of Heart Rate Estimation Methods using Face Videos”. In *Procs. of the Computers, Software, and Applications Conf. (COMPSAC)*.

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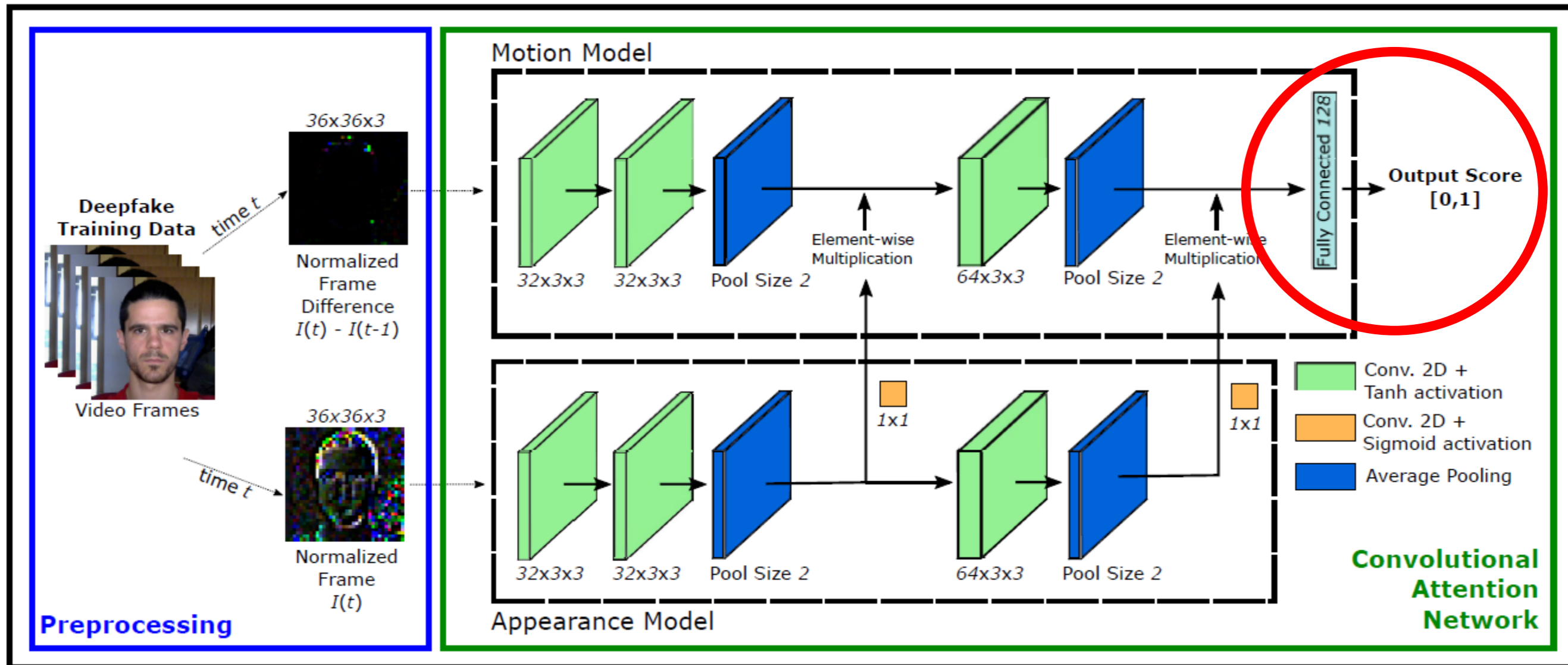
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# DeepFakesON-Phys: Development and Evaluation

DeepFakesON-Phys





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- 4) Changed the **last FC** and the output layers of the former model (two classes, real or fake).
- 5) **Fixed** all **weights** up to the final fully-connected layer.
- 6) **Trained the network** for 100 more epochs and choose the best performing model based on validation accuracy.
  - One model per training database.

[11] W. Chen, and D. McDuff. 2018. “Deepphys: Video-based Physiological Measurement using Convolutional Attention Networks”. In *Procs. of the European Conf. on Computer Vision (ECCV)*.

[12] J. Hernandez-Ortega, *et al.* 2020. “A Comparative Evaluation of Heart Rate Estimation Methods using Face Videos”. In *Procs. of the Computers, Software, and Applications Conf. (COMPSAC)*.

# Experimental Results

Evaluation Metrics → Area Under the Curve (AUC) and Accuracy (Frame level).

## Celeb-DF v2

Study	Method	Classifier	AUC (%)
Yang, Li, and Lyu 2019	Head Pose	SVM	54.6
Li <i>et al.</i> 2020	Face Warping	CNN	64.6
Afchar <i>et al.</i> 2018	Mesoscopic	CNN	54.8
Dang <i>et al.</i> 2020	Deep Learning	CNN + Attention	71.2
Tolosana <i>et al.</i> 2020a	Deep Learning	CNN	83.6
Qi <i>et al.</i> 2020	Physiological	CNN + Attention	-
Ciftci, Demir, and Yin 2020	Physiological	SVM/CNN	Acc. = 91.5
DeepFakesON-Phys [Ours]	Physiological	CNN + Attention	99.9 Acc. = 98.7

# Experimental Results

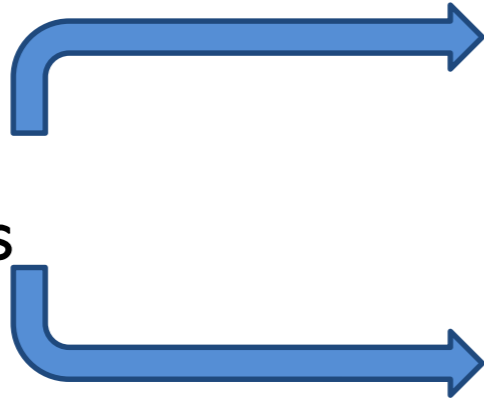
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## DFDC Preview

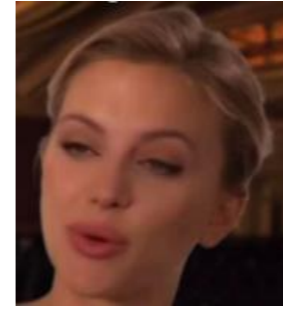
Study	Method	Classifier	AUC (%)
Yang, Li, and Lyu 2019	Head Pose	SVM	55.9
Li <i>et al.</i> 2020	Face Warping	CNN	75.5
Afchar <i>et al.</i> 2018	Mesoscopic	CNN	75.3
Dang <i>et al.</i> 2020	Deep Learning	CNN + Attention	-
Tolosana <i>et al.</i> 2020a	Deep Learning	CNN	91.1
Qi <i>et al.</i> 2020	Physiological	CNN + Attention	Acc. = 64.1
Ciftci, Demir, and Yin 2020	Physiological	SVM/CNN	-
DeepFakesON-Phys [Ours]	Physiological	CNN + Attention	98.2 Acc. = 94.4

# Conclusions

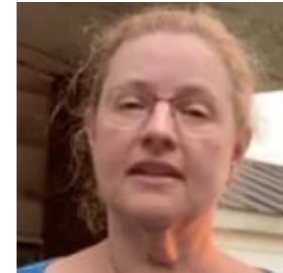
Two 2nd generation  
DeepFake databases



Celeb-DF v2



DFDC Preview



Two of the latest and most  
challenging DeepFake video  
databases.

## DeepFakesON-Phys:

Outperformed other **state-of-the-art fake detectors** based on face warping and pure deep learning features, among others.

Revealed that **current DeepFake techniques do not pay attention to the heart-rate-related or blood-related physiological information.**

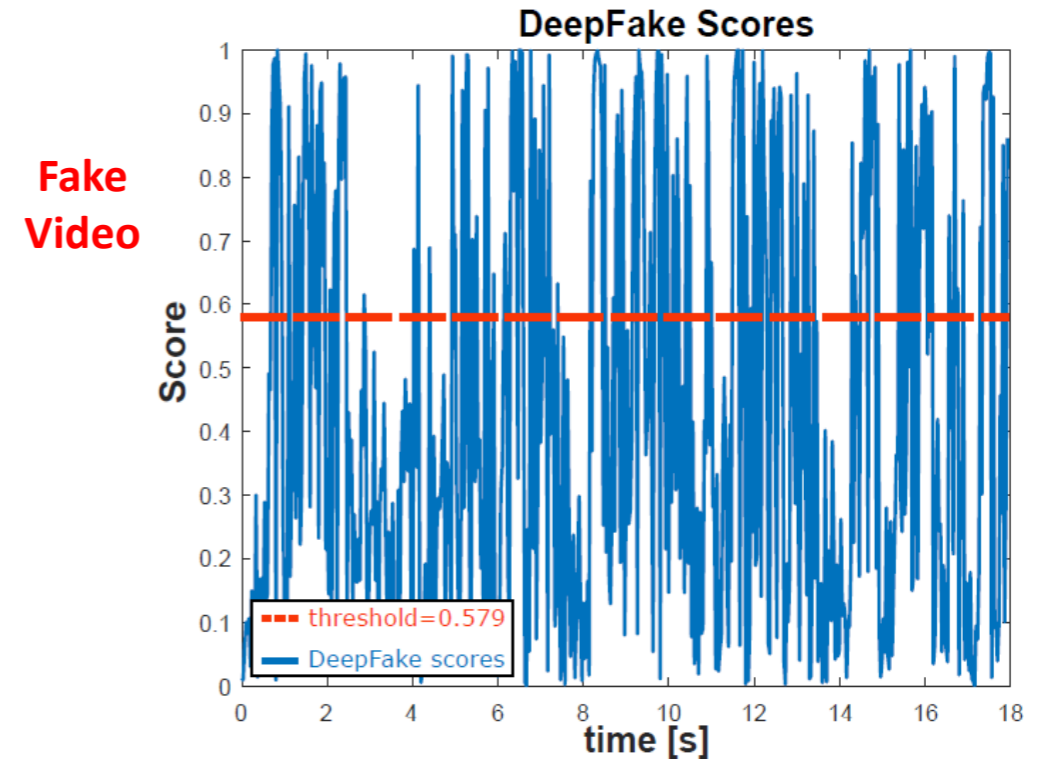
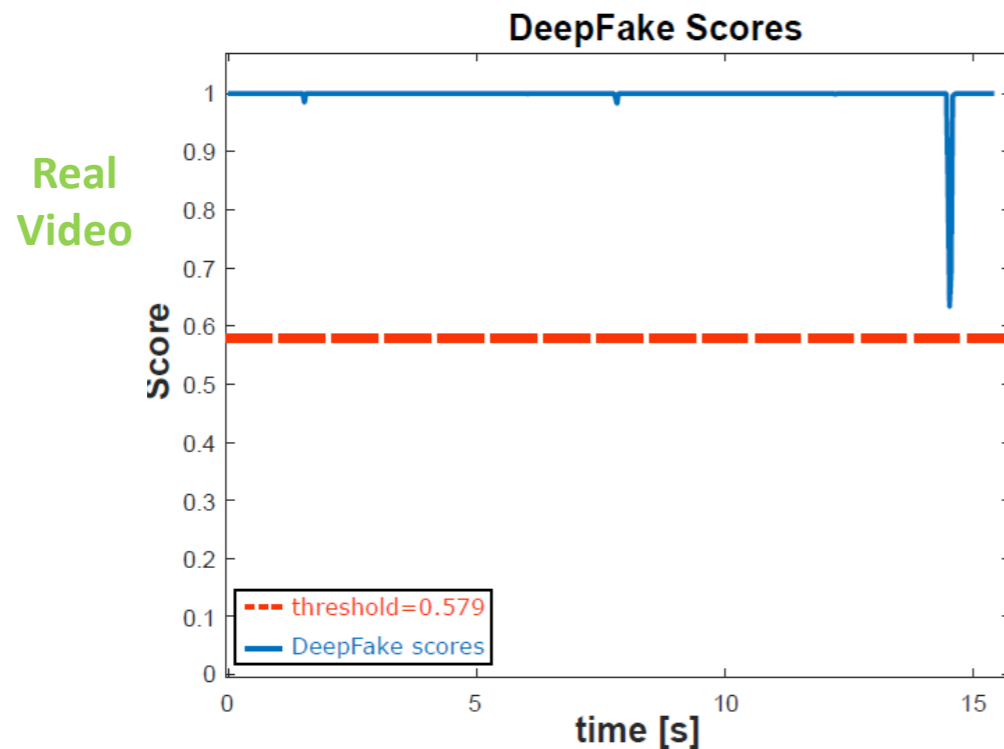
# Future Work

1. Analysis of the **robustness against unseen face manipulations** [13].

[13] Tolosana, R., Vera-Rodriguez, R., Fierrez, J., Morales, A., and Ortega-Garcia, J. 2020. “DeepFakes and Beyond: A Survey of Face Manipulation and Fake Detection”. *Information Fusion* 64: 131–148.

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1. Analysis of the **robustness against unseen face manipulations** [13].
2. Applying **temporal integration** to frame data [14].

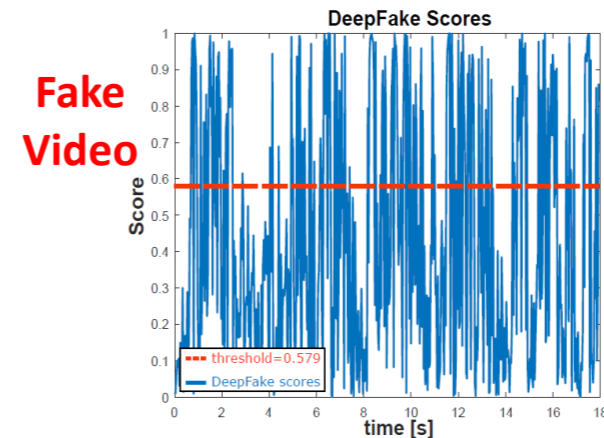
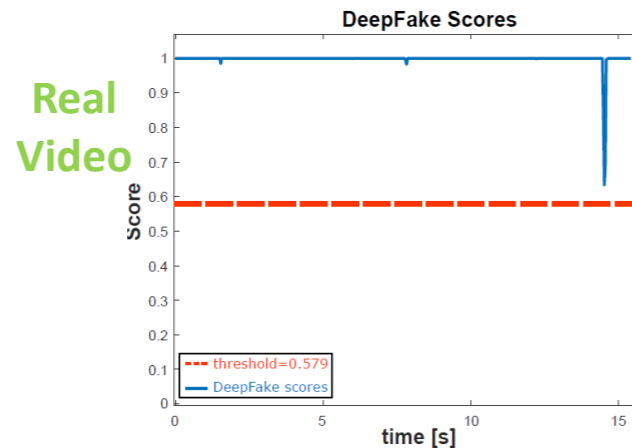


[13] Tolosana, R., Vera-Rodriguez, R., Fierrez, J., Morales, A., and Ortega-Garcia, J. 2020. “DeepFakes and Beyond: A Survey of Face Manipulation and Fake Detection”. *Information Fusion* 64: 131–148.

[14] Hernandez-Ortega, J., Fierrez, J., Morales, A., and Tome, P. 2018. “Time Analysis of Pulse-Based Face Anti-Spoofing in Visible and NIR”. In *Proc. IEEE Conf. on Comp. Vision and Pattern Recognition Workshops (CVPRw)*.

# Future Work

1. Analysis of the **robustness against unseen face manipulations** [13].
2. Applying **temporal integration** to frame data [14].



3. Studying **other face manipulation techniques**, e.g. face morphing [15].

[13] Tolosana, R., Vera-Rodriguez, R., Fierrez, J., Morales, A., and Ortega-Garcia, J. 2020. “DeepFakes and Beyond: A Survey of Face Manipulation and Fake Detection”. *Information Fusion* 64: 131–148.

[14] Hernandez-Ortega, J., Fierrez, J., Morales, A., and Tome, P. 2018. “Time Analysis of Pulse-Based Face Anti-Spoofing in Visible and NIR”. In *Proc. IEEE Conf. on Comp. Vision and Pattern Recognition Workshops (CVPRw)*.

[15] Raja, K., et al. 2020. “Morphing Attack Detection - Database, Evaluation Platform and Benchmarking”. *IEEE Transactions on Information Forensics and Security*.



## Know More:

R. Tolosana, R. Vera-Rodriguez, J. Fierrez, A. Morales and J. Ortega-Garcia, "**DeepFakes and Beyond: A Survey of Face Manipulation and Fake Detection**", *Information Fusion*, 2020.

J. C. Neves, R. Tolosana, R. Vera-Rodriguez, V. Lopes, H. Proenca and J. Fierrez, "**GANprintR: Improved Fakes and Evaluation of the State of the Art in Face Manipulation Detection**", *IEEE Journal of Selected Topics in Signal Processing*, 2020.

J. Hernandez-Ortega, *et al.* "**Time Analysis of Pulse-based Face Anti-spoofing in Visible and NIR**". In *Proceedings of the IEEE Conference on Computer Vision and Pattern Recognition Workshops*. 2018.

J. Hernandez-Ortega, *et al.* "**Introduction to Face Presentation Attack Detection.**" *Handbook of Biometric Anti-Spoofing*. Springer. 2019.

<http://biometrics.eps.uam.es>

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