

t-SNE for temporal feature

2022.06.27

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1. Problems

Problems

- Deepfake Detection (Input)
 - Video → temporal inconsistency
 - Frame → spatial inconsistency
- Model Generalization
 - Cross-manipulation
 - Cross-dataset

Model Generalization

■ Cross Manipulation

FTCN (Video based)[1]

Method	Train on remaining three				Avg
	DF	FS	F2F	NT	
Xception [43]	93.9	51.2	86.8	79.7	77.9
CNN-aug [54]	87.5	56.3	80.1	67.8	72.9
PatchForensics[11]	94.0	60.5	87.3	84.8	81.7
CNN-GRU [45]	97.6	47.6	85.8	86.6	79.4
Face X-ray[32]	99.5	93.2	94.5	92.5	94.9
LipForensics-Scratch[22]	93.0	56.7	98.8	98.3	86.7
LipForensics[22]	99.7	90.1	99.7	99.1	97.1
ours	99.9	99.9	99.7	99.2	99.7

SBI (Frame based)[2]

Method	Test Set AUC (%)				
	DF	F2F	FS	NT	FF++
Face X-ray + BI [40]	99.17	98.57	98.21	98.13	98.52
PCL + I2G [66]	100	98.97	99.86	97.63	99.11
EFNB4 + SBIs (Ours)	99.99	99.88	99.91	98.79	99.64

Model Generalization

■ Cross Dataset

FTCN (Video based)[1]

Method	CDF	DFDC	FSh	DFo	Avg
Xception [43]	73.7	70.9	72.0	84.5	75.3
CNN-aug [54]	75.6	72.1	65.7	74.4	72.0
PatchForensics [11]	69.6	65.6	57.8	81.8	68.7
CNN-GRU [45]	69.8	68.9	80.8	74.1	73.4
Multi-task [39]	75.7	68.1	66.0	77.7	71.9
FWA [34]	69.5	67.3	65.5	50.2	63.1
Two-branch [36]	76.7	—	—	—	—
Face X-ray [32]	79.5	65.5	92.8	86.8	81.2
LipForensics [22]	82.4	73.5	97.1	97.6	87.7
ours	86.9	74.0	98.8	98.8	89.6

SBI(Frame based)[2]

Method	Input Type	Training Set		Test Set AUC (%)				
		Real	Fake	CDF	DFD	DFDC	DFDCP	FFIW
DSP-FWA [42]	Frame	✓	✓	69.30	-	-	-	-
Face X-ray + BI [40]	Frame	✓		-	93.47	-	71.15	-
Face X-ray + BI [40]	Frame	✓	✓	-	95.40	-	80.92	-
LRL [14]	Frame	✓	✓	78.26	89.24	-	76.53	-
FRDM [45]	Frame	✓	✓	79.4	91.9	-	79.7	-
PCL + I2G [66]	Frame	✓		90.03	99.07	67.52	74.37	-
Two-branch [48]	Video	✓	✓	76.65	-	-	-	-
DAM [68]	Video	✓	✓	75.3	-	-	72.8	-
LipForensics [28]	Video	✓	✓	82.4	-	-	-	-
FTCN [67]	Video	✓	✓	86.9	94.40*	<u>71.00*</u>	74.0	<u>74.47*</u>
EFNB4 + SBIs (Ours)	Frame	✓		93.18	<u>97.56</u>	72.42	86.15	84.83

2. t-SNE

T-SNE

■ FTCN[1]

(a) 3D R50



• Real

(b) 2D R50



• NeuralTexture

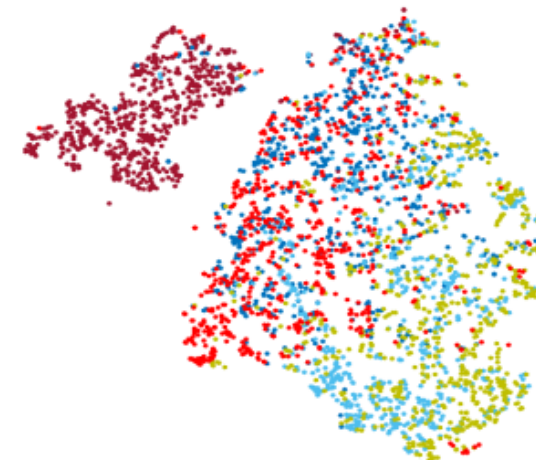
(c) 3D R50-FTCN



• Deepfake

• FaceSwap

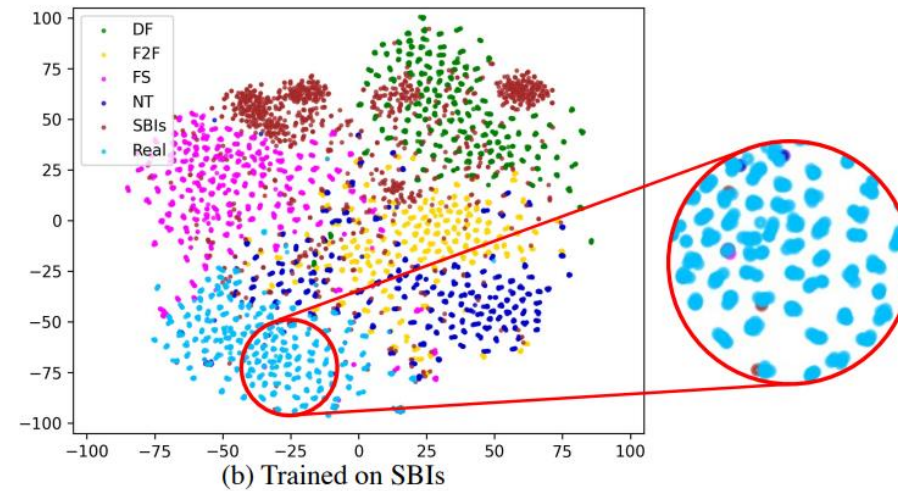
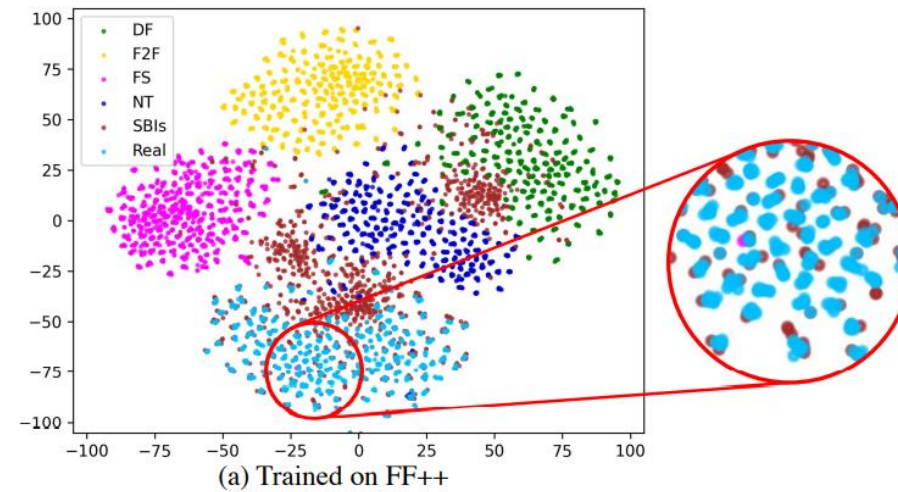
(d) 3D R50-FTCN+TT



• Face2Face

T-SNE

■ SBI[2]



3. Experiment

Experiment

■ FTCN

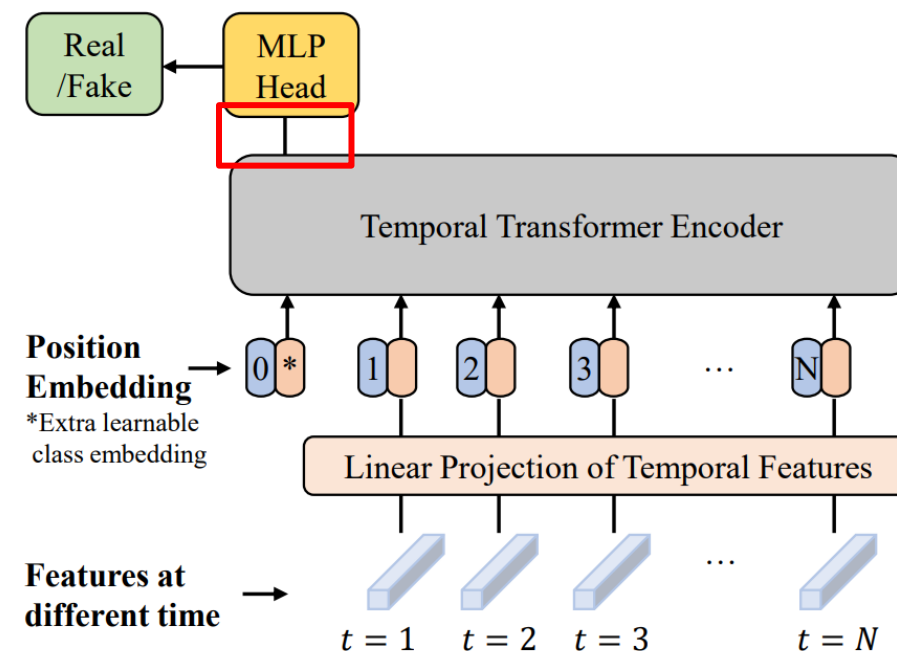
■ Video feature

	layer	output size
conv ₁	$5 \times 1 \times 1, 64, \text{stride } 1, 1, 1$	$64 \times 32 \times 224 \times 224$
pool ₁	$1 \times 5 \times 5 \text{ max, stride } 1, 4, 4$	$256 \times 32 \times 56 \times 56$
res ₂	$\begin{bmatrix} 1 \times 1 \times 1, 64 \\ 3 \times 1 \times 1, 64 \\ 1 \times 1 \times 1, 256 \end{bmatrix} \times 3$	$256 \times 32 \times 56 \times 56$
pool ₂	$2 \times 1 \times 1 \text{ max, stride } 2, 1, 1$	$256 \times 16 \times 56 \times 56$
res ₃	$\begin{bmatrix} 1 \times 1 \times 1, 128 \\ 3 \times 1 \times 1, 128 \\ 1 \times 1 \times 1, 512 \end{bmatrix} \times 4$	$512 \times 16 \times 28 \times 28$
res ₄	$\begin{bmatrix} 1 \times 1 \times 1, 256 \\ 3 \times 1 \times 1, 256 \\ 1 \times 1 \times 1, 1024 \end{bmatrix} \times 6$	$1024 \times 16 \times 14 \times 14$
res ₅	$\begin{bmatrix} 1 \times 1 \times 1, 512 \\ 3 \times 1 \times 1, 512 \\ 1 \times 1 \times 1, 2048 \end{bmatrix} \times 3$	$2048 \times 16 \times 7 \times 7$
	spatial-related average pool	$2048 \times 16 \times 1 \times 1$

■ Check

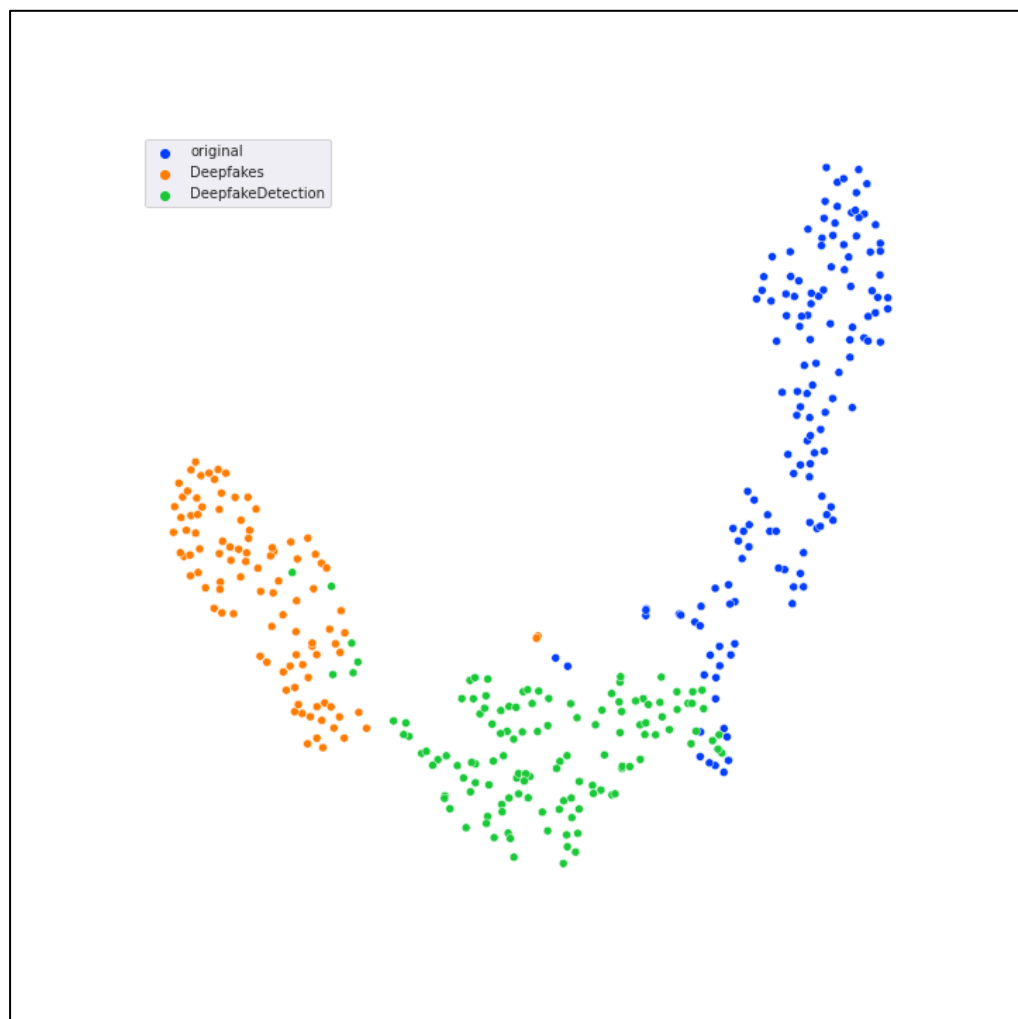
■ Dataset (1st, 2nd)

■ Temporal Consistency (back, random)

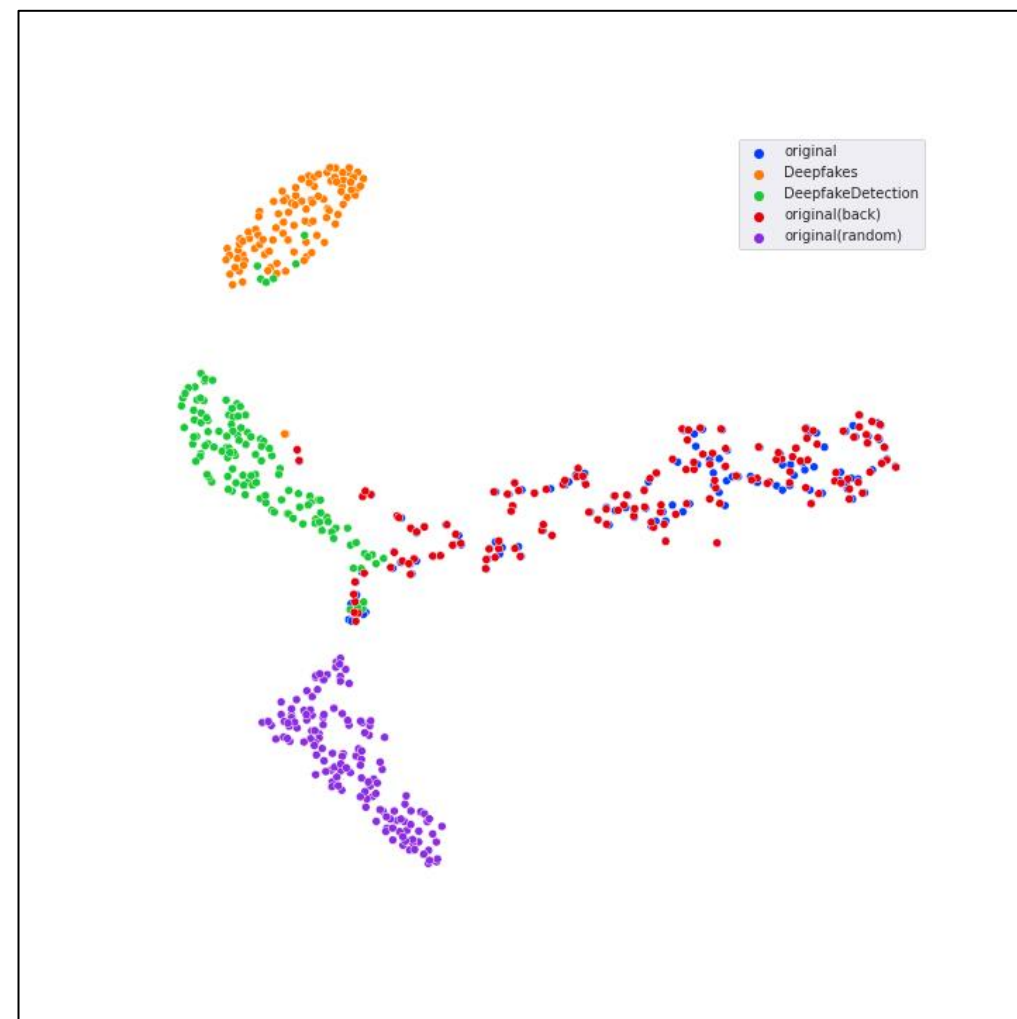


Experiment

Dataset

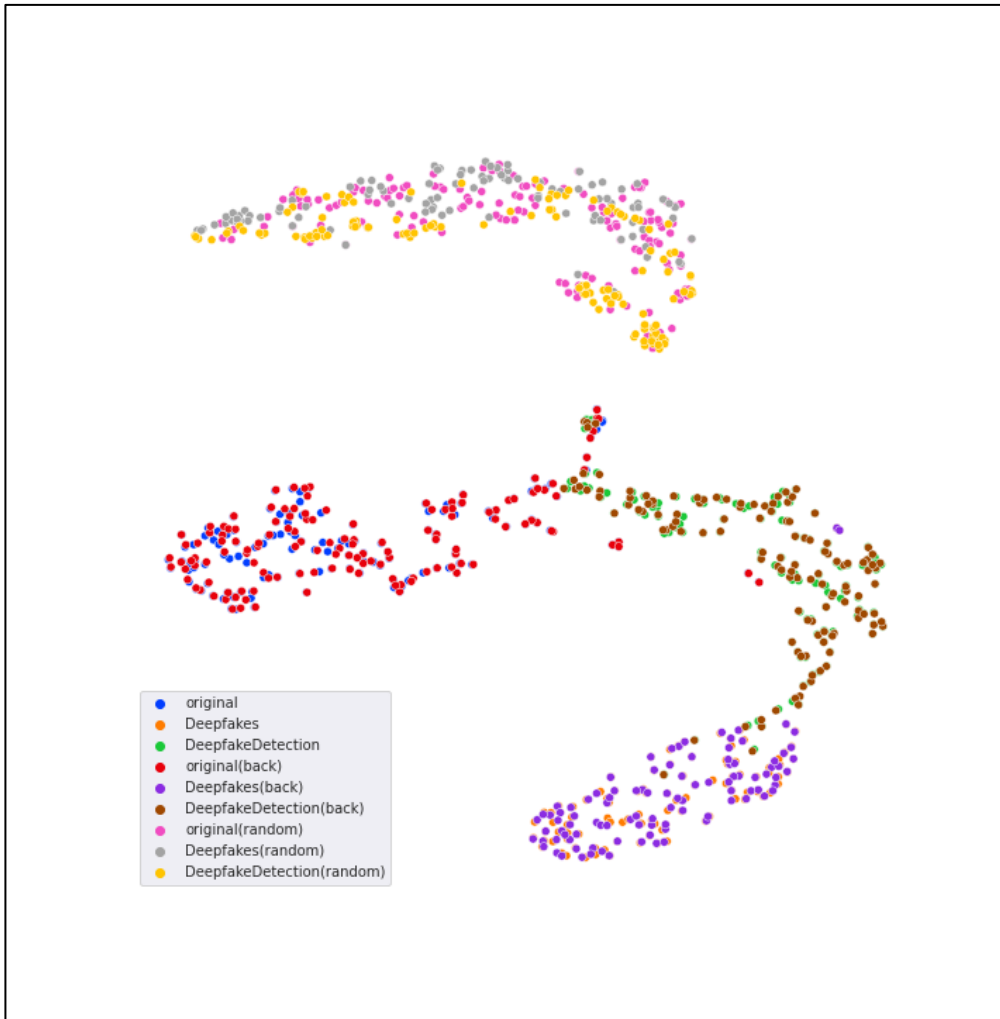


Temporal Consistency



Experiment

Dataset + Temporal Consistency



4. Conclusion

Conclusion

- FTCN lacks generalization performance for 2nd generation Dataset
- Frame back
- Frame random

Reference

- [1] Exploring Temporal Coherence for More General Video Face Forgery Detection (ICCV 2021)
- [2] Detecting Deepfakes with Self-Blending Images (CVPR 2022 oral)

감사합니다