# **Group Fourier Transform**

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# Method

- Data Structure
- Features

#### **Data Structure**

- Load images
- 3 Lists
  - Class 1 (genuine)
  - o Class 0 (synthetic)
  - Sort (here we use spoofed and genuine images)
- Calculate FFT of images
- Apply bandpass filter to images(n bands)
  - O List of ImageFFT objects
  - One n-list per object
  - $\circ$  n = 30 in most of the tests







#### Features and KNN

- Energy (Square sum of absolute values of the fourier coefficients)
- Calculate energy of all bands
- For every image in "sort"
  - Sort genuine and synthetic images by energy difference
  - Take k nearest neighbours
  - O Compare which class is more common
- Leave one subject out / Leave one out Cross Validation

# Results

- Expectations
- PLUS
- VERA (IDIAP)
- SCUT

### **Expectations**

- Energy of spoofed is closer to synthetic than to genuine
- Algorithm distinguishes between spoofed/synthetic and genuine

#### **Findings**

- Energy of spoofed is closer to genuine
- Algorithm distinguishes between real and Al generated images

Distinguishing between spoofed and genuine images (no synthetic images)

baseline		Actual	
		1	0
Predicted	1	46.53%	0.68%
Fredicted	0	3.48%	49.31%

k = 10

Distinguishing between spoofed and genuine (using synthetic images)

003		Actual	
003		1	0
Duadiatad	1	50%	50%
Predicted	0	0%	0%
004		Actual	
00	14	Act	tual
00	)4	Act	tual 0
00 Predicted	)4 1		

Distinguishing only between genuine and synthetic

cycle 003		Actual	
		1	0
Predicted	1	50.00%	0.00%
Fredicted	0	0.00%	50.00%

k = 10

#### **VERA / IDIAP**

Distinguishing between spoofed and genuine images (no synthetic images)

Influence of the k value

baseline (k=10)		Actual	
		1	0
Predicted	1	31.18%	18.68%
Fredicted	0	18.82%	31.32%
baseline (k=13)		Act	tual
		1	0
Predicted	1	30.79%	12.11%
Fredicted	0	19.21%	37.89%
baseline (k=37)		Act	tual
		1	0
Desdistad	1	35.66%	25.39%
Predicted	0	14.34%	24.61%

#### **VERA / IDIAP**

Distinguishing between spoofed and genuine (using synthetic images)

009		Actual	
		1	0
Predicted	1	50%	50%
Fredicted	0	0%	0%

Same results for all GANs

Distinguishing between spoofed and genuine images (no synthetic images)

baseline		Actual	
		1	0
Predicted	1	34.03%	19.72%
Predicted	0	15.97%	30.28%

k = 10

Baseline: Leave One Out vs Leave One Subject Out Cross Validation

k	LOO	LOSO
3	65.30%	65.30%
7	64.25%	64.25%
10	64.31%	64.31%
15	69.40%	69.40%
19	68.18%	68.18%
25	69.09%	69.09%

Distinguishing between spoofed and genuine images (using synthetic images)

007		Actual	
		1	0
Dradiated	1	50%	50%
Predicted	0	0%	0%
00	no	Act	tual
00	08	Act	tual 0
00 Predicted	)8 1		

### **Explanation**

• Synthetic images contain artifacts in the Fourier (frequency) domain

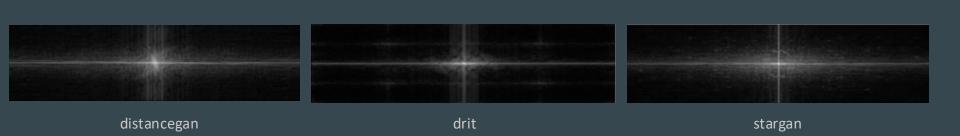
These artifacts are "learned"



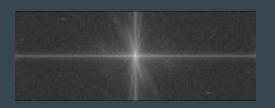
VERA/IDIAP 001\_L\_1

# Visualizations for all Datasets





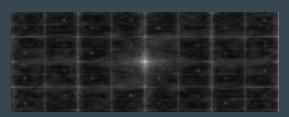
## **VERA / IDIAP**



genuine



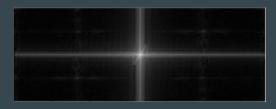
spoofed



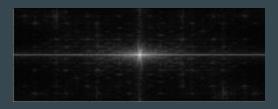
cyclegan



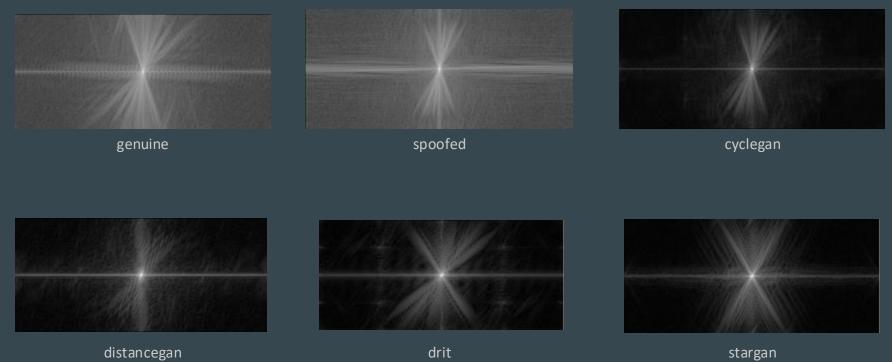
distancegan



drit



stargan



#### **Band Subsets**

- FFT split into 30 bands
- Energy difference differs by band
- Biggest difference between real and synthetic images at
  - O Band 0 15
  - o Band 25 30
- Between genuine and spoofed images
  - Difference is descending
  - O Largest difference: band 1 5
- Testing with different band subsets

#### **Band Subsets**

• Using bands with as few artifacts as possible (e.g. bands between 15 and 25 from the last slide)

cycle 003 (k=25)		Actual	
		1	0
Predicted	1	50.00%	50.00%
Predicted	0	0.00%	0.00%

Input: genuine and synthetic / Sort: genuine and spoofed images

Energy difference genuine-spoofed is always smaller than synthetic-spoofed

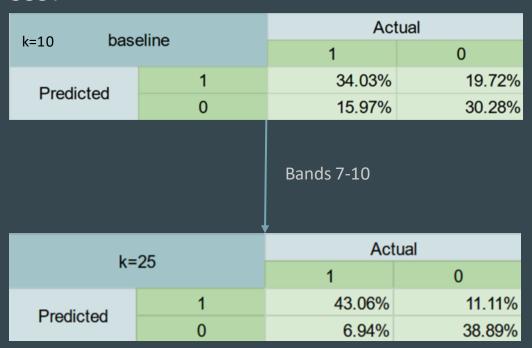
### Band Subsets for Baseline

#### VERA/IDIAP

baseline (k=15)		Actual	
		1	0
Predicted	1	32.37%	15.26%
Fredicted	0	17.63%	34.74%
		Bands 5-10	
baseline (k=15)		Act	ual
		1	0
Predicted	1	46.32%	2.11%
Fredicted	0	3.68%	47.89%

#### **Band Subsets for Baseline**

#### **SCUT**



#### Conclusion

- FT is useful for detecting AI images
- Non synthetic images can also be sorted
- Spoofed and synthetic together could not be distinguished from genuine

#### Sources

- Github Repository
- Result Tables
- Zeit Protokoll (Leon Andrassik)

 https://www.shutterstock.com/de/image-photo/apple-tree-isolated-on-whitebackground-267376982 (Tree on slide 4)

# Thank you for your attention!