### Genlmage: A Million-Scale Benchmark for Detecting Al-Generated Image

Eight generators are used to generate the images. The number of fakes images are the same as the real images from ImageNet

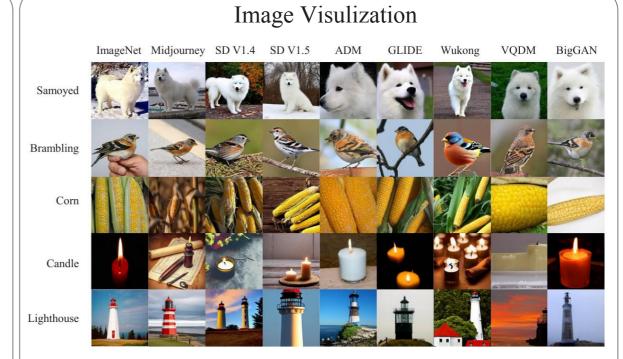
Dataset	Image	Genera	tor Category	Public	Real	Fake
Dataset	Content	GAN	Diffusion	Avalibility	Images	Images
UADFV 30	Face	1	Х	X	241	252
FakeSpotter [25]	Face	1	X	×	6,000	5,000
DFFD 3	Face	1	X	1	58,703	240,336
APFDD 6	Face	1	X	×	5,000	5,000
ForgeryNet 9	Face	1	×	1	1,438,201	1,457,861
DeepArt [27]	Art	X	1	1	64,479	73,411
CNNSpot 26	General	1	X	1	362,000	362,000
IEEE VIP Cup 24	General	/	✓	×	7,000	7,000
DE-FAKE 22	General	X	✓	×	20,000	60,000
CiFAKE [1]	General	X	/	/	60,000	60,000
GenImage	General	1	1	✓	1,331,167	1,350,000

1000 classes images

Over 1 million generated images Containing GAN and Diffusion Generator GitHub

https://github.com/GenImage-Dataset/GenImage Project Address

https://genimage-dataset.github.io/



Human eye cannot classify real and fake images.

The content is rich.

The variance of each class issignificant.

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Verify the significance of increasing the number of images. Increasing the classes or the number of each class can efficiently increase accuracy.

Total	Image Classes	Images Per Class	Identical-Generator Acc. (%)	Cross-Generator Acc. (%)
1600	10	160	73.6	60.9
1600	100	16	81.8	62.7
16000	100	160	97.7	68.7
16000	1000	16	96.3	68.4
162000	1000	162	99.9	72.1

Training the model on GenImage, and testing it on face and art images.

Image	Image S	Image Source				
Content	Real	Fake	(%)			
Face	LFW	SD V1.4	99.9			
Art	Laion-Art	SD V1.4	95.0			

Real

Fake











Verify the significance of collecting a large number of image categories
Using image category subsets for training can effectively generalize
to other categories

Image Classes	Fake Images Per Class	Acc.(%)
10	1280	78.8
50	256	80.3
100	128	81.2
100	1300	98.4

Detectors trained with GenImage can accurately detect AI generated images from the internet

#### 用Midjourney画"球迷冲进球场拥抱梅西"事件

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# Genlmage: A Million-Scale Benchmark for Detecting Al-Generated Image

#### Task1: Cross-generator Evaluation

Examining the generalization of detectors under different generator scenarios

Stable Diffusion V1.4 is used for training, and eight different generators are used for testing

Method	Testing Subset								
Method	Midjourney	SD V1.4	SD V1.5	ADM	GLIDE	Wukong	VQDM	BigGAN	Acc.(%)
ResNet-50 [8]	54.9	99.9	99.7	53.5	61.9	98.2	56.6	52.0	72.1
DeiT-S [23]	55.6	99.9	99.8	49.8	58.1	98.9	56.9	53.5	71.6
Swin-T [13]	62.1	99.9	99.8	49.8	67.6	99.1	62.3	<b>57.6</b>	74.8
CNNSpot [26]	52.8	96.3	95.9	50.1	39.8	78.6	53.4	46.8	64.2
Spec [31]	52.0	99.4	99.2	49.7	49.8	94.8	55.6	49.8	68.8
F3Net [19]	50.1	99.9	99.9	49.9	50.0	99.9	49.9	49.9	68.7
GramNet [14]	54.2	99.2	99.1	50.3	54.6	98.9	50.8	51.7	69.9

#### Different generators are used for training and testing

-	Testing Subset								
Method	Midjourney	SD V1.4		_		Wukong	VQDM	BigGAN	Avg Acc.(%)
ResNet-50 [8]	59.0	72.3	72.4	59.7	73.1	71.4	60.9	66.6	66.9
DeiT-S [23]	60.7	74.2	74.2	59.5	71.1	73.1	61.7	66.3	67.6
Swin-T [13]	61.7	76.0	76.1	61.3	76.9	75.1	65.8	69.5	70.3
CNNSpot [26]	58.2	70.3	70.2	57.0	57.1	67.7	56.7	56.6	61.7
Spec [31]	56.7	72.4	72.3	57.9	65.4	70.3	61.7	64.3	65.1
F3Net [19]	55.1	73.1	73.1	66.5	57.8	72.3	62.1	56.5	64.6
GramNet [14]	58.1	72.8	72.7	58.7	65.3	71.3	57.8	61.2	64.7

Cross validation using ResNet-50 on different generation models Using the same generator for training and testing results in high accuracy. Using different generators for training and testing results in low accuracy

Training Subset	Testing Subset								
Training Subset	Midjourney	SD V1.4	SD V1.5	ADM	GLIDE	Wukong	VQDM	BigGAN	Acc.(%)
Midjourney	98.8	76.4	76.9	64.1	78.9	71.4	52.5	50.1	71.1
SD V1.4	54.9	99.9	99.7	53.5	61.9	98.2	56.6	52.0	72.1
SD V1.5	54.4	99.8	99.9	52.7	60.1	98.5	56.9	51.3	71.7
ADM	58.6	53.1	53.2	99.0	97.1	53.0	61.5	88.3	70.4
GLIDE	50.7	50.0	50.1	56.0	99.9	50.3	51.0	74.0	60.2
Wukong	54.5	99.7	99.6	51.4	58.3	99.9	58.7	50.9	71.6
VQDM	50.1	50.0	50.0	50.7	60.1	50.2	99.9	66.8	59.7
BigGAN	49.9	49.9	49.9	50.6	68.4	49.9	50.6	99.9	58.6

Task 2: Image degradation verification

The test uses low resolution, compressed, and blurred images. Examining the Generalization of Detectors for Degenerated Images

Method	Testing Subset								
Method	LR (112)	LR (64)	JPEG (q=65)	JPEG (q=30)	Blur ( $\sigma$ =3)	Blur ( $\sigma$ =5)	Acc(%)		
ResNet-50 8	96.2	57.4	51.9	51.2	97.9	69.4	70.6		
DeiT-S 23	97.1	54.0	55.6	50.5	94.4	67.2	69.8		
Swin-T 13	97.4	54.6	52.5	50.9	94.5	52.5	67.0		
CNNSpot [26]	50.0	50.0	97.3	97.3	97.4	77.9	78.3		
Spec [31]	50.0	49.9	50.8	50.4	49.9	49.9	50.1		
F3Net [19]	50.0	50.0	89.0	74.4	57.9	51.7	62.1		
GramNet 14	98.8	94.9	68.8	53.4	95.9	81.6	82.2		