

Generative Adversarial Text to Image Synthesis

Reed S, Akata Z, Yan X, et al. ICML 2016

the flower has petals that are bright pinkish purple with white stigma

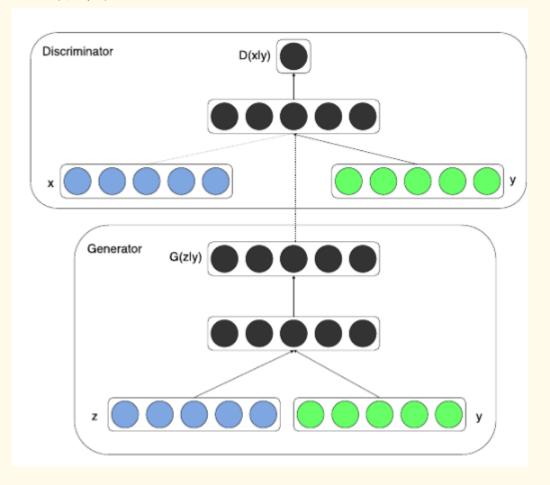


this white and yellow flower have thin white petals and a round yellow stamen



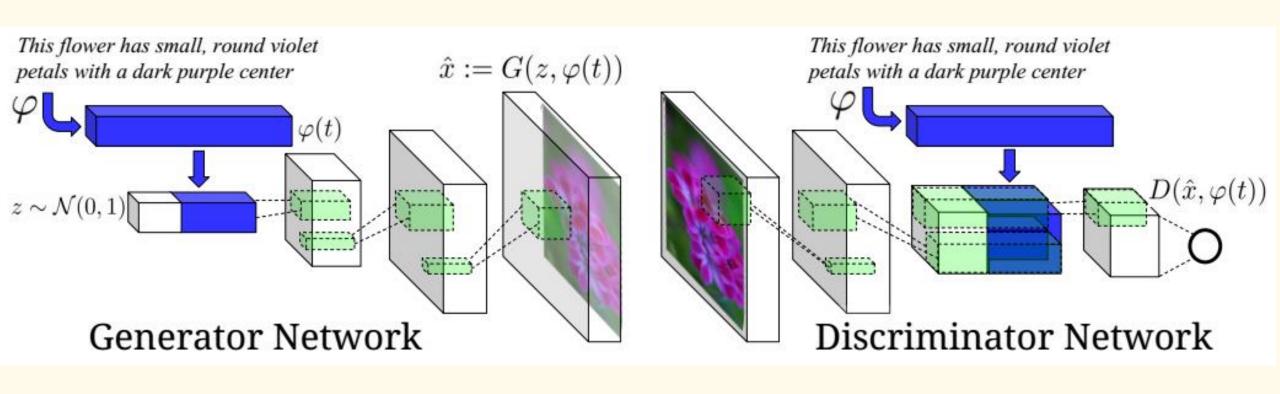
DCGAN + CGAN , 文字描述由一个现成的词嵌入方法生成向量*

$$\begin{split} \min_{G} \ \max_{D} V(D,G) &= \mathbb{E}_{x \sim p_{data}(x)}[\log D(x)] + \\ &\mathbb{E}_{x \sim p_{z}(z)}[\log(1 - D(G(z)))] \end{split}$$



* Reed, S., Akata, Z., Lee, H., and Schiele, B. Learning deep representations for fine-grained visual descriptions. In CVPR, 2016.

 $\mathsf{DCGAN} + \mathsf{CGAN}$,文字描述 t 由一个现成的词嵌入方法生成向量 $\varphi(t)$





an all black bird

GT with a distinct thick, rounded bill.



this small bird has a yellow breast, brown crown, and black superciliary



a tiny bird, with a tiny beak, tarsus and feet, a blue crown, blue coverts, and black cheek patch











GAN-CLS 通过加入反例来训练D(与G无关)

Intuition: the discriminator has no explicit notion of whether real training images match the text embedding context.

Algorithm 1 GAN-CLS training algorithm with step size α , using minibatch SGD for simplicity.

- 1: **Input:** minibatch images x, matching text t, mismatching \hat{t} , number of training batch steps S
- 2: **for** n = 1 **to** S **do**
- 3: $h \leftarrow \varphi(t)$ {Encode matching text description}
- 4: $\hat{h} \leftarrow \varphi(\hat{t})$ {Encode mis-matching text description}
 - 5: $z \sim \mathcal{N}(0,1)^Z$ {Draw sample of random noise}
 - 6: $\hat{x} \leftarrow G(z, h)$ {Forward through generator}
 - 7: $s_r \leftarrow D(x, h)$ {real image, right text}
 - 8: $s_w \leftarrow D(x, \hat{h})$ {real image, wrong text}
 - 9: $s_f \leftarrow D(\hat{x}, h)$ {fake image, right text}
 - 10: $\mathcal{L}_D \leftarrow \log(s_r) + (\log(1-s_w) + \log(1-s_f))/2$
 - 11: $D \leftarrow D \alpha \partial \mathcal{L}_D / \partial D$ {Update discriminator}
 - 12: $\mathcal{L}_G \leftarrow \log(s_f)$
 - 13: $G \leftarrow G \alpha \partial \mathcal{L}_G / \partial G$ {Update generator}
 - 14: **end for**

- 一张图片和他的
- 一个文字描述是
- 一个样本对





GAN-INT 通过插值法增加文本样本数量(针对G)

Deep networks have been shown to learn representations in which interpolations between embedding pairs tend to be near the data manifold

$$\mathbb{E}_{t_1,t_2 \sim p_{data}} [\log(1 - D(G(z,\beta t_1 + (1-\beta)t_2)))]$$

Because the interpolated embeddings are synthetic, the discriminator D does not have "real" corresponding image and text pairs to train on. However, D learns to predict whether image and text pairs match or not. Thus, if D does a good job at this, then by satisfying D on interpolated text embeddings G can learn to fill in gaps on the data manifold in between training points.

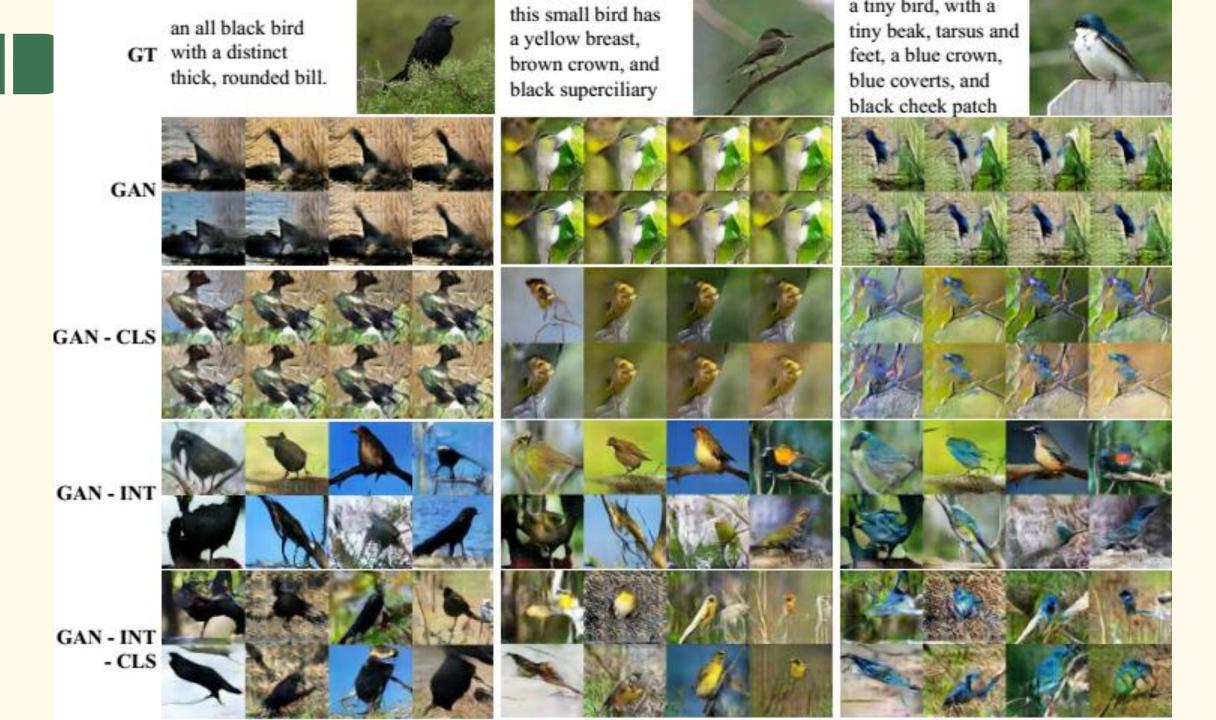
原来的样本数量不足以使得判别器D判断图文的关系,通过插值法增加样本数使得D能学得他们之间的关系。

GAN-INT-CLS 两种改进方法结合

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- 13: $G \leftarrow G \alpha \partial \mathcal{L}_G / \partial G$ {Update generator}
- 14: **end for**

差值得到的文字描述当作真实的对应文本





CUB contains 200 bird species with 11,788 images.

Oxford-102 contains 8,189 images of flowers from 102 different categories.

Solution: train a convolutional network S to invert G to regress from samples $x \hat{}$ back onto z.

$$\mathcal{L}_{style} = \mathbb{E}_{t,z \sim \mathcal{N}(0,1)} ||z - S(G(z,\varphi(t)))||_2^2$$

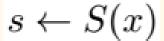
where S is the style encoder network. With a trained generator and style encoder, style transfer from a query image x onto text t proceeds as follows:

$$s \leftarrow S(x), \hat{x} \leftarrow G(s, \varphi(t))$$

where \hat{x} is the result image and s is the predicted style.

Text descriptions Images (content) (style)





The bird has a **yellow breast** with **grey** features and a small beak.

This is a large white bird with black wings and a red head.

A small bird with a black head and wings and features grey wings.

This bird has a **white breast**, brown and white coloring on its head and wings, and a thin pointy beak.

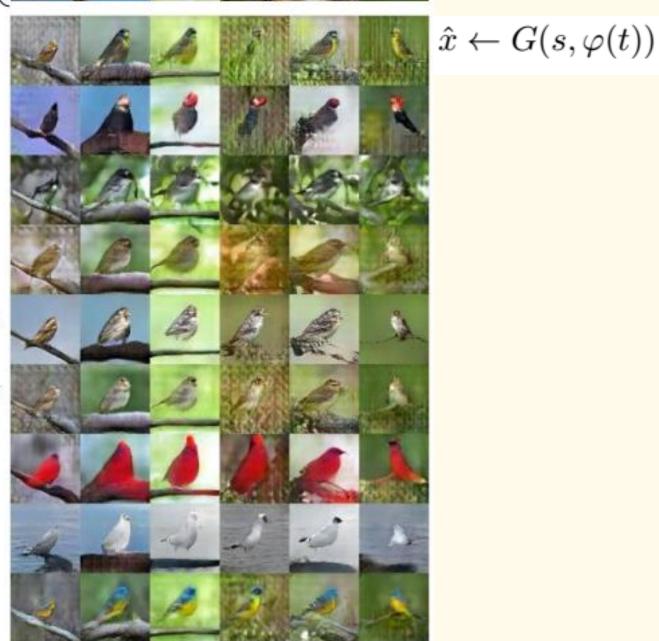
A small bird with **white base** and **black stripes** throughout its belly, head, and feathers.

A small sized bird that has a cream belly and a short pointed bill.

This bird is completely red.

This bird is completely white.

This is a **yellow** bird. The **wings are bright blue**.



插值法实验 - 左边噪声z固定,对文字描述插值,右边相反

'Blue bird with black beak' → 'Red bird with black beak'



'Small blue bird with black wings' → 'Small yellow bird with black wings'



'This bird is bright.' → 'This bird is dark.'



'This bird is completely red with black wings'



'this bird is all blue, the top part of the bill is blue, but the bottom half is white'



'This is a yellow bird. The wings are bright blue'



Problem

the generated scenes are not usually coherent; for example the human-like blobs in the baseball scenes lack clearly articulated parts.

Low resolution (64x64像素)

GΤ

Ours

a pitcher is about to throw the ball to the batter.





StackGAN: Text to Photo-realistic Image Synthesis with Stacked Generative Adversarial Networks

Zhang H, Xu T, Li H, et al. arXiv preprint. 2016.

the space of plausible images given text descriptions is multimodal. There are a large number of images that correctly fit the given text description

给定文字描述作为条件,输入是噪声z拼上文字向量。一阶段只生成大概的轮廓和颜色的低分辨率图像

(a) Stage-I images

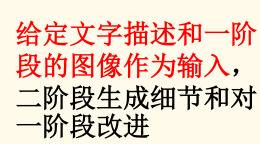
This bird has a yellow belly and tarsus, grey back, wings, and brown throat, nape with a black face

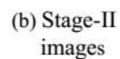
This bird is white with some black on its head and wings, and has a long orange beak This flower has overlapping pink pointed petals surrounding a ring of short yellow filaments







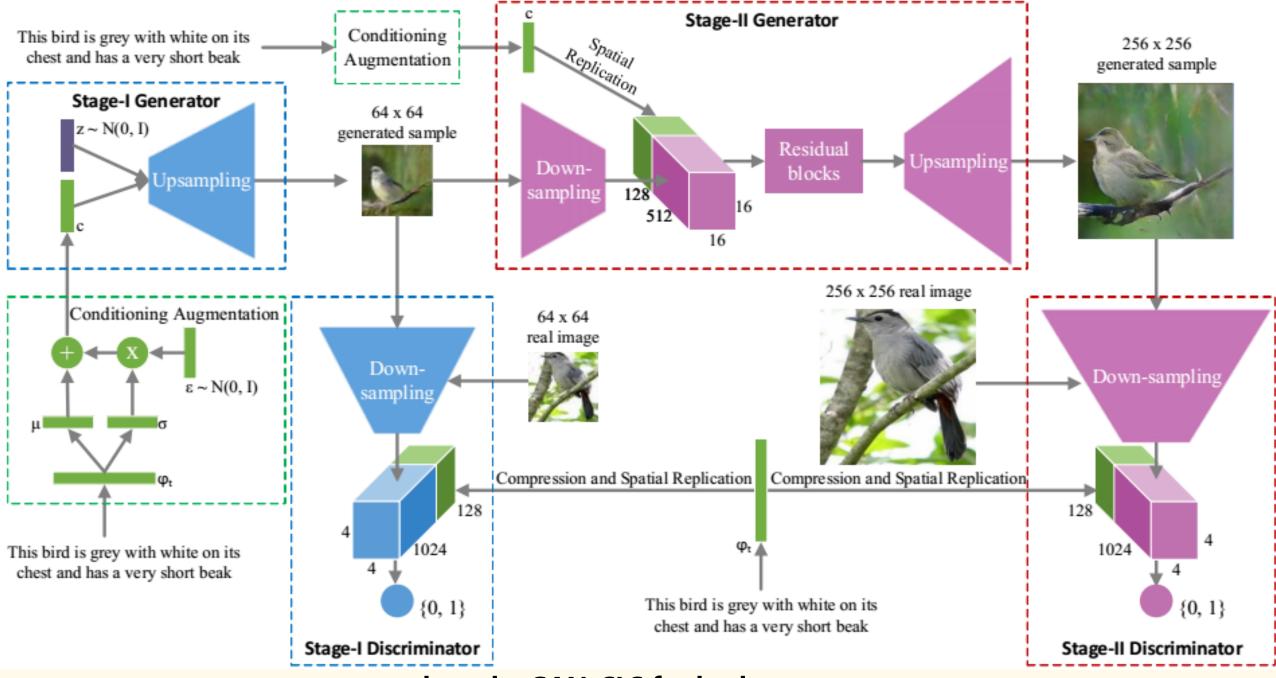






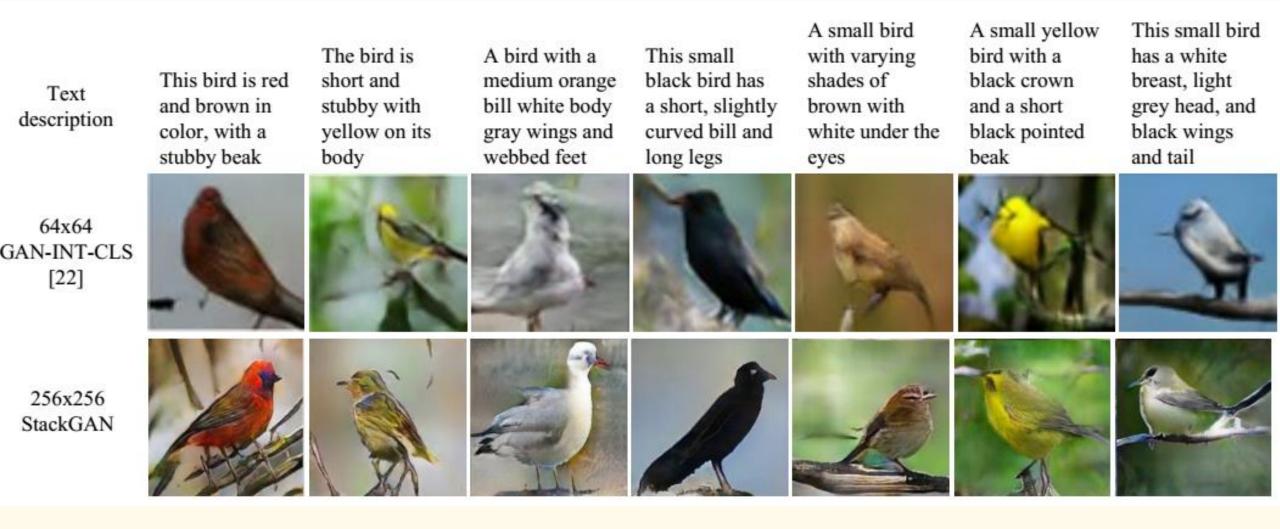






we adopt the GAN-CLS for both stages.





CUB contains 200 bird species with 11,788 images.
Oxford-102 contains 8,189 images of flowers from 102 different categories.

Text description	This flower has petals that are white and has pink shading	This flower has a lot of small purple petals in a dome-like configuration	This flower has long thin yellow petals and a lot of yellow anthers in the center	This flower is pink, white, and yellow in color, and has petals that are striped	This flower is white and yellow in color, with petals that are wavy and smooth	This flower has upturned petals which are thin and orange with rounded edges	This flower has petals that are dark pink with white edges and pink stamen
64x64 GAN-INT-CLS [22]							
256x256 StackGAN					S CONTRACTOR OF THE PARTY OF TH		

一阶段和二阶段生成的图像有显著进步,说明分阶段生成图片的可行性

This bird is The bird has This is a small, This bird is This bird is This bird has A white bird white, black, white black and small beak, black bird with Text with a black and brown in yellow in color, with reddish blue with white wings that are a white breast description brown and has crown and color, with a with a short and has a very brown crown and white on black beak yellow beak brown beak short beak a yellow belly and gray belly the wingbars. Stage-I images Stage-II images

通过将生成图片从训练 集中找出5个最近邻, 发现生成的图片和训练 集中的图片差别较大,

说明了stackGAN 不是 记住了训练样本,而是 找到了语言和图像之间 的联系(更强的泛化能 力) Images generated from text in test sets

Five nearest neighbors from training sets



Figure 6. For generated images (column 1), retrieving their nearest training images (columns 2-6) by utilizing Stage-II discriminator D to extract visual features. The L2 distances between features are calculated for nearest-neighbor retrieval.

由于类似VAE组建的存在,对输入的文字描述加入了一定的扰动,使得即使固定噪声和文字输入,也能输出多样化的图片

This small blue bird has a short pointy beak and brown on its wings

This bird is completely red with black wings and pointy beak

A small sized bird that has a cream belly and a short pointed bill

A small bird with a black head and wings and features grey wings



Figure 7. Birds with different poses and viewpoints generated with the same input text embedding by our StackGAN. The noise vector z and text embedding are fixed for each row.

To demonstrate that our StackGAN learns a smooth latent data manifold, we use it to generate images from linearly interpolated sentence embeddings

The bird is completly red → The bird is completly yellow



This bird is completely red with black wings and pointy beak → this small blue bird has a short pointy beak and brown on its wings



This bird is completely red with black wings and pointy beak →
The bird has a yellow breast with grey features and a small beak



Figure 8. (Left to right) Images generated by interpolating two sentence embeddings. Gradual appearance changes from the first sentence's meaning to that of the second sentence can be observed. The noise vector z is fixed to be zeros for each row.

More example

This bird sits close to the ground with his short yellow tarsus and feet; his bill is long and is also yellow and his color is mostly white with a black crown and primary feathers



images

A large bird has large thighs and large wings that have white wingbars



The small bird has a red head with feathers that fade from red to gray from head to tail



This bird is black with green and has a very short beak



This flower has white petals with a yellow tip and a yellow pistil



A flower with small pink petals and a massive central orange and black stamen cluster



Failure cases

The main reason for failure cases is that Stage-I GAN fails to generate plausible rough shapes or colors of the objects

This medium

sized bird is

This bird has a primarily black The medium Colored bill dark brown and has a large Bird has brown sized bird has a with a white overall body wingspan and a body feathers, long black bill This particular Grey bird with dark grey color, a ring around color, with a bird has a black flat beak brown breast black downward it on the small white with a strip of Text white at the patch around the brown body with grey and feathers, and curved beak, and description upper part brown beak base of the bill and brown bill white big wings long wings near the bill beginning of it Stage-I images Stage-II images

Failure cases

The main reason for failure cases is that Stage-I GAN fails to generate plausible rough shapes or colors of the objects

This flower The flower A flower that This flower is is yellow have large has white petals A unique yellow This is a light petals that are The petals of flower with no pink and yellow colored flower and green in with some this flower are visible pistils in color, with tones of yellow with many color, with pink with Text escription white with a petals that and green protruding from petals that are different petals yellow on some large stigma oddly shaped are ruffled of the petals the center filaments on a green stem Stage-I images Stage-II images

