



Memory-Driven Text-to-Image Generation

Bowen Li, Philip H.S. Torr, and Thomas Lukasiewicz



Introduction

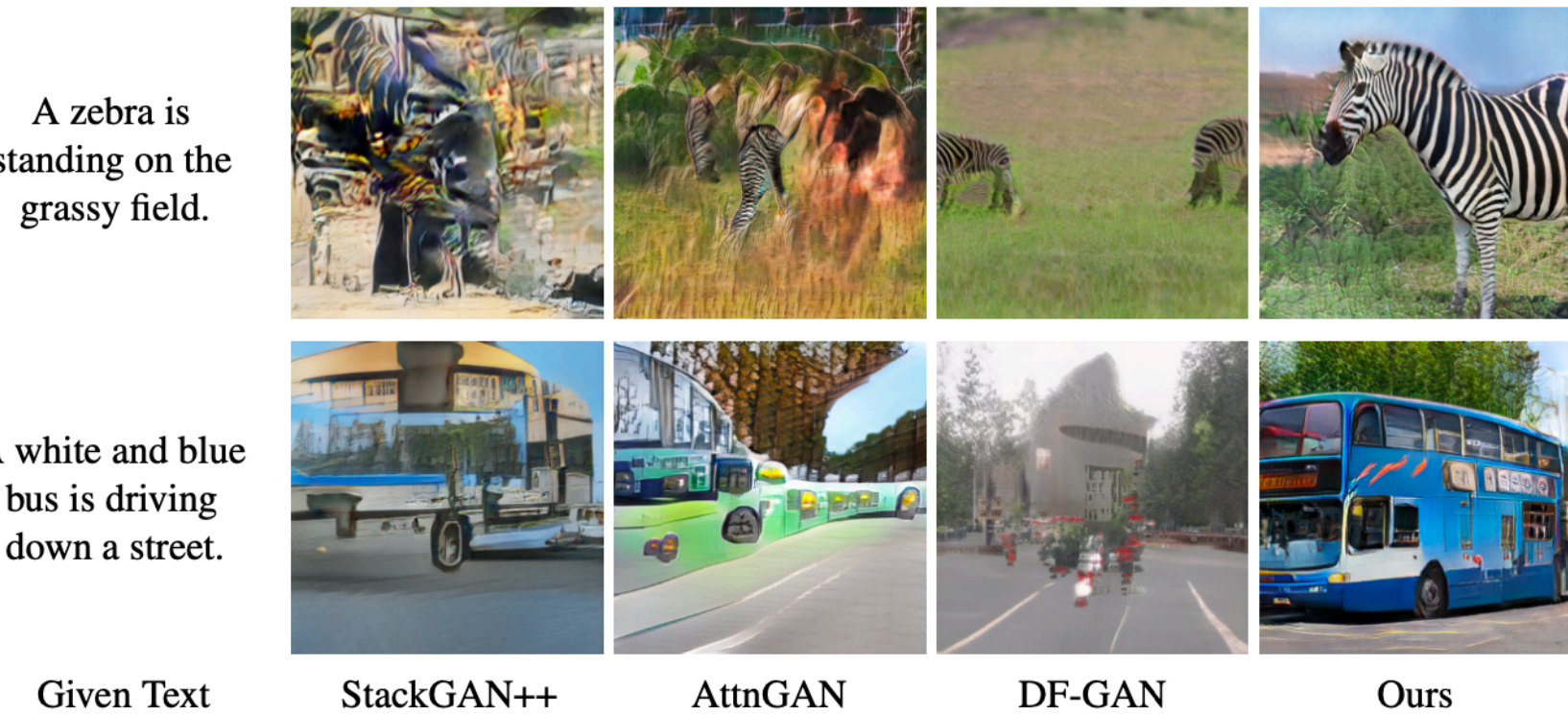


Fig. 1. Examples of text-to-image generation on COCO. Current approaches only generate low-quality images with unrealistic objects. In contrast, our method can produce realistic images, in terms of both visual appearances and geometric structure.

Method

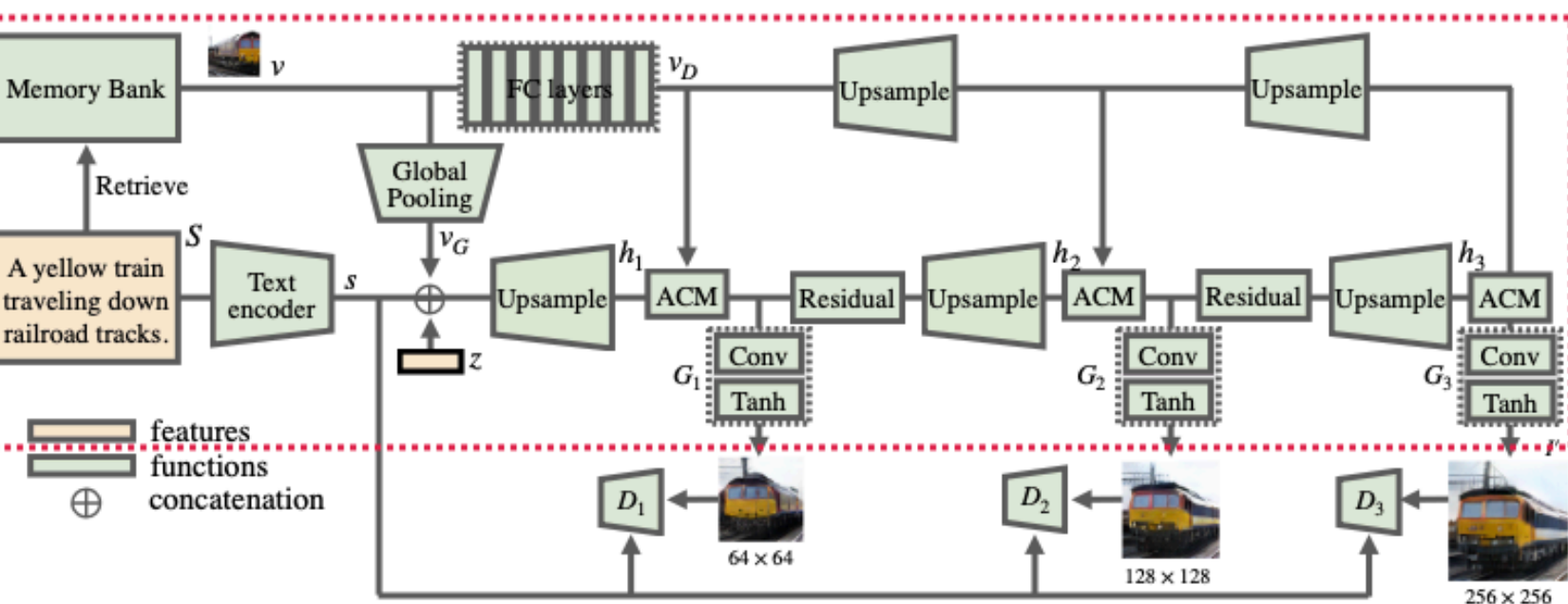


Fig. 2. Top: architecture of our proposed method. The red box indicates the inference pipeline that retrieves image features from the memory bank according to the given text description S ; during training, we directly feed image features from the text-paired training image. z is a random vector drawn from the Gaussian distribution.

Experiments

Table 1. Quantitative comparison: IS, FID, R-precision, SOA-C, and SOA-I of current methods and our approach on the CUB and COCO datasets.

Method	CUB			COCO				
	IS	FID score	R-precision	IS	FID score	R-precision	SOA-C	SOA-I
Real Images	25.34	-	89.17	34.88	-	92.71	74.97	80.84
AttnGAN [49]	4.36	23.98	67.82	25.89	32.32	85.47	25.88	39.01
ControlGAN [22]	4.58	13.92	69.33	24.06	33.58	72.43	-	-
MirrorGAN [40]	4.56	-	57.67	26.47	-	74.52	-	-
DM-GAN [56]	4.75	16.09	72.31	32.32	32.64	88.56	33.44	48.03
DF-GAN [46]	5.10	14.81	-	-	21.42	-	-	-
XMC-GAN [52]	-	-	-	30.45	9.33	-	50.94	71.33
LAFITE	5.97	10.48	-	32.34	8.12	-	61.09	74.78
DALL-E [42]	-	-	-	-	~20	-	-	-
GLIDE [35]	-	-	-	-	12.89	-	-	-
CP-GAN [29]	-	-	-	52.73	55.82	93.59	77.02	84.55
Obj-GAN [27]	-	-	-	30.29	36.52	87.84	27.14	41.24
OP-GAN [12]	-	-	-	27.88	24.70	89.01	35.85	50.47
Ours	5.91	10.49	73.87	29.36	19.47	90.32	47.46	65.83



Fig. 3. Qualitative results on CUB and COCO.

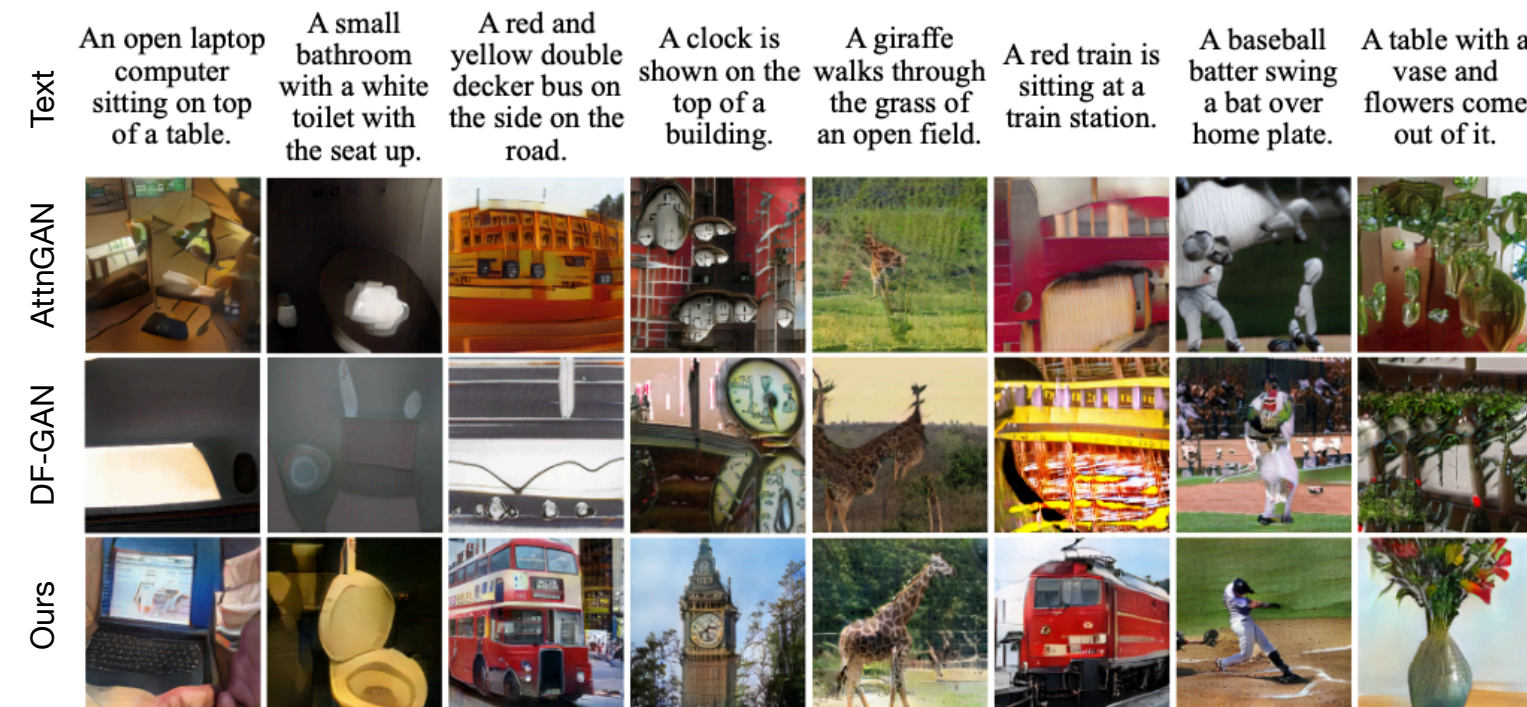


Fig. 4. Qualitative comparison between AttnGAN, DF-GAN, and Ours on COCO.

Experiments



Fig. 5. Diversity of our approach, where we only change the value of noise vector z .

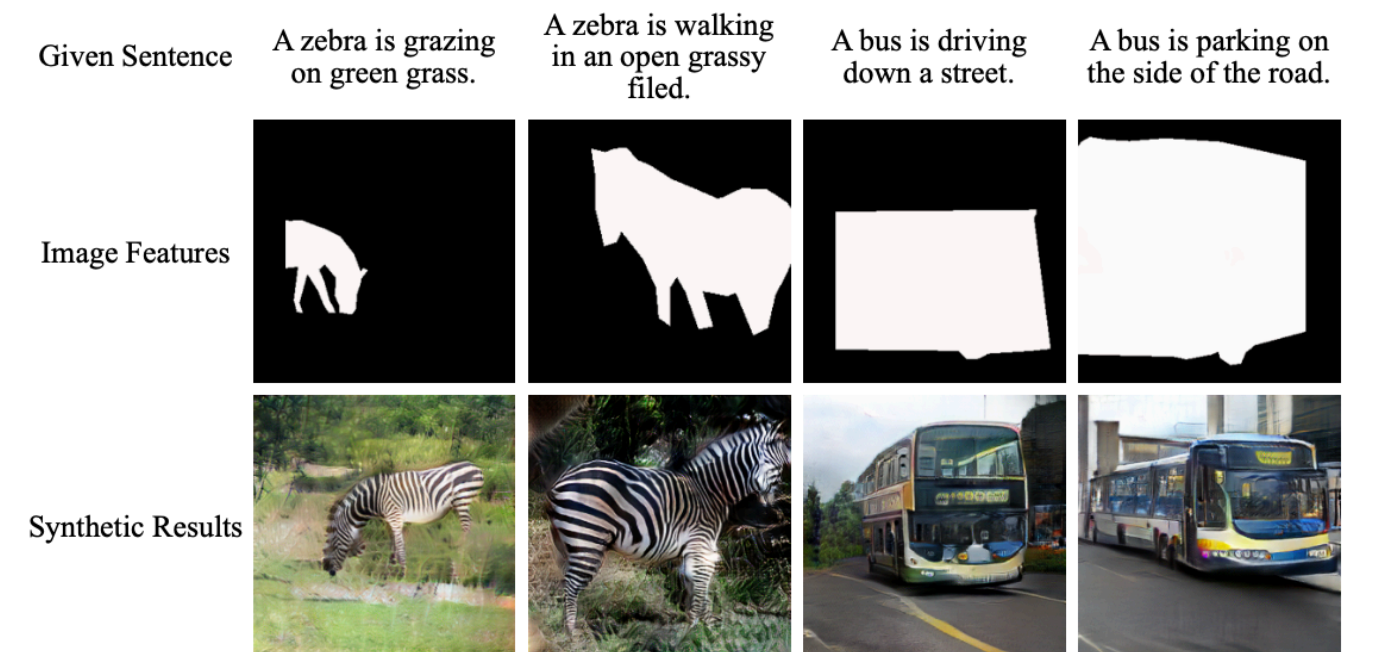


Fig. 6. Semantic information exploration by feeding semantic masks into our network to generate realistic images.

Table 2. Ablation studies of different components used in our approach.

Method	FID	R-psr
Ours w/o Feature	22.20	84.63
Ours w/o Disen.	18.82	92.17
Ours w/o Disen.*	18.80	67.05
Ours w/o Content	20.96	88.95
Ours w/o Reg.	27.12	82.97
Ours w/ Max	26.12	83.11
Ours w/ Aver (Full Model)	19.47	90.32