

# Generative Adversarial Transformers

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facebook AI

Stanford

\*Special Thanks to Christopher D. Manning 🌻





Most GANs use  
Convolution

fake





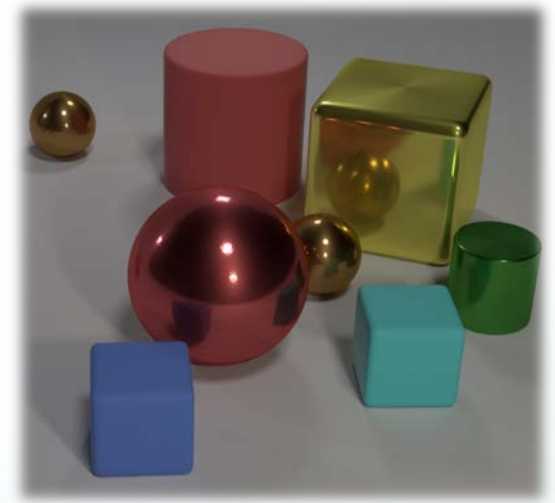
# Convolutional GANs struggle with long-range dependencies



Consistency



Global Structure



Compositionality

# Transformers

$O(n^2)$



use Self-Attention  
to model long-range  
interactions

A diagram consisting of four curved arrows of different colors (green, blue, purple, yellow) connecting the words 'use Self-Attention', 'to model long-range', and 'interactions' in a circular fashion, illustrating the concept of long-range interactions.

**but they don't scale**

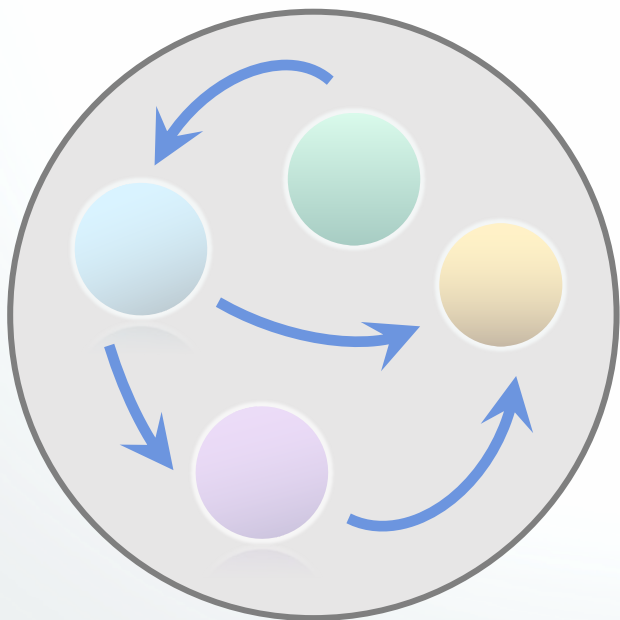


Can we **efficiently** apply  
**transformers**  
for **image generation**?



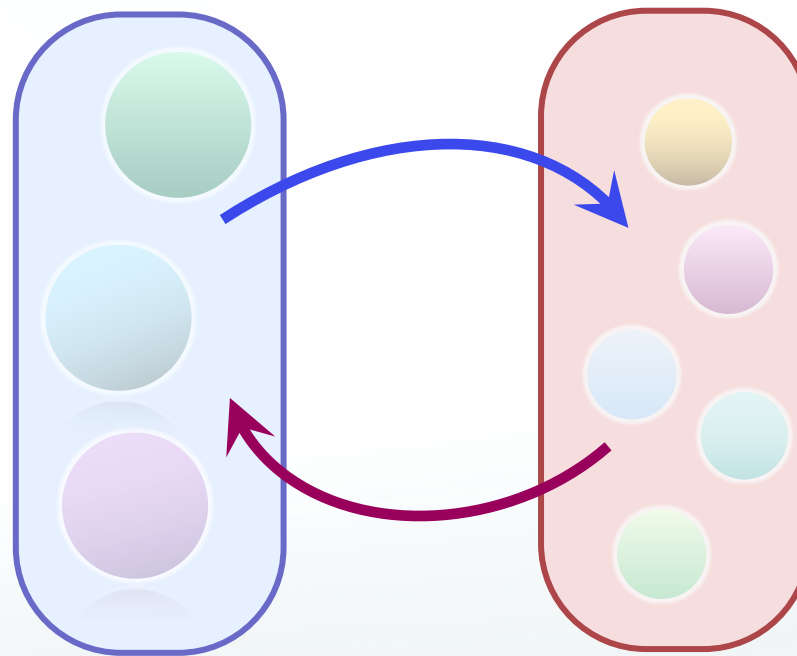


# Generative Adversarial Transformers



Self-Attention

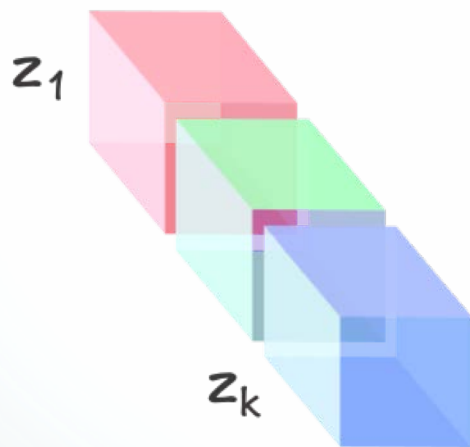
Latents



Image

Bipartite Attention

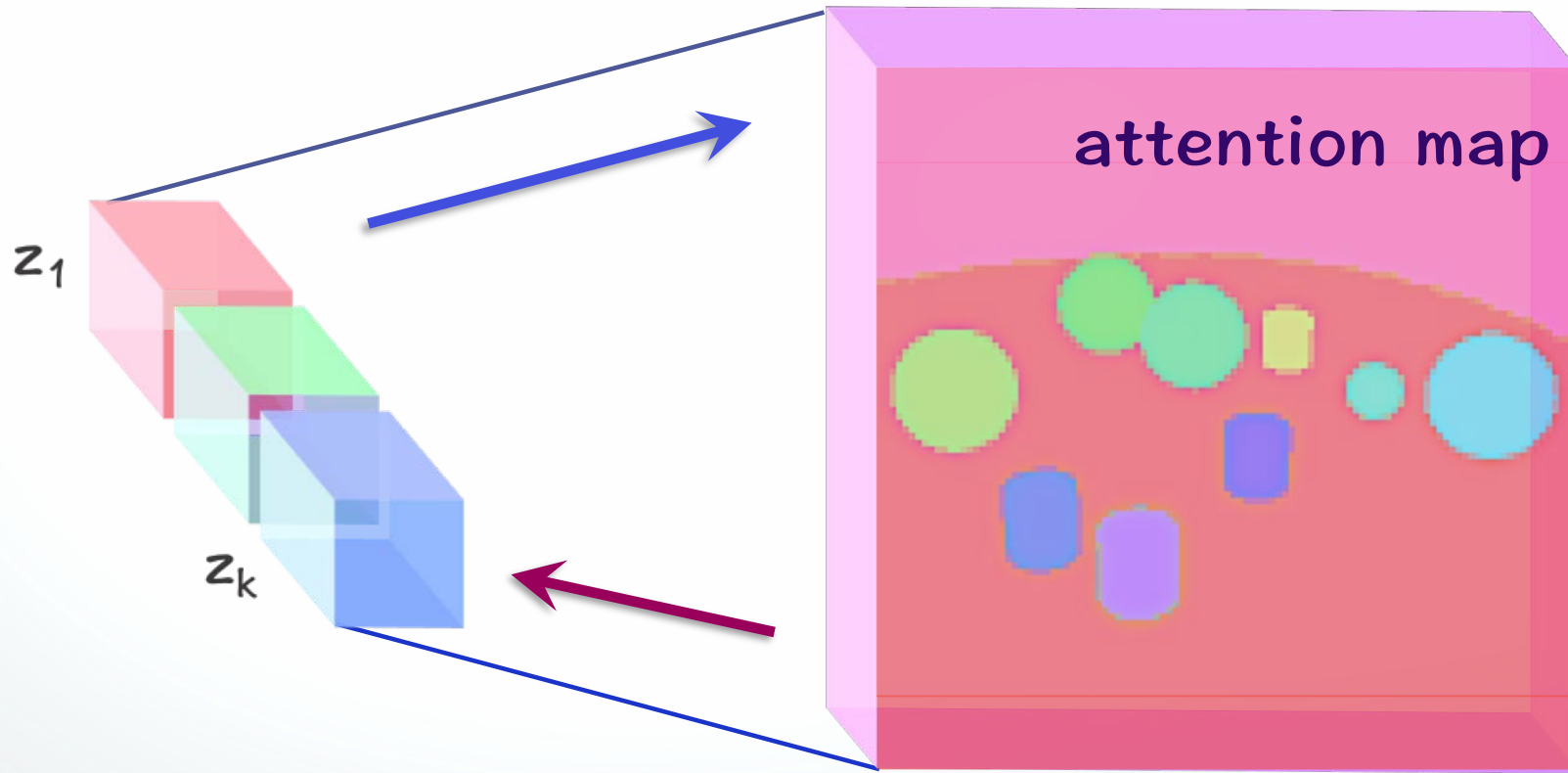
# From Transformer to GANformer



**Compositional latent space with multiple variables**



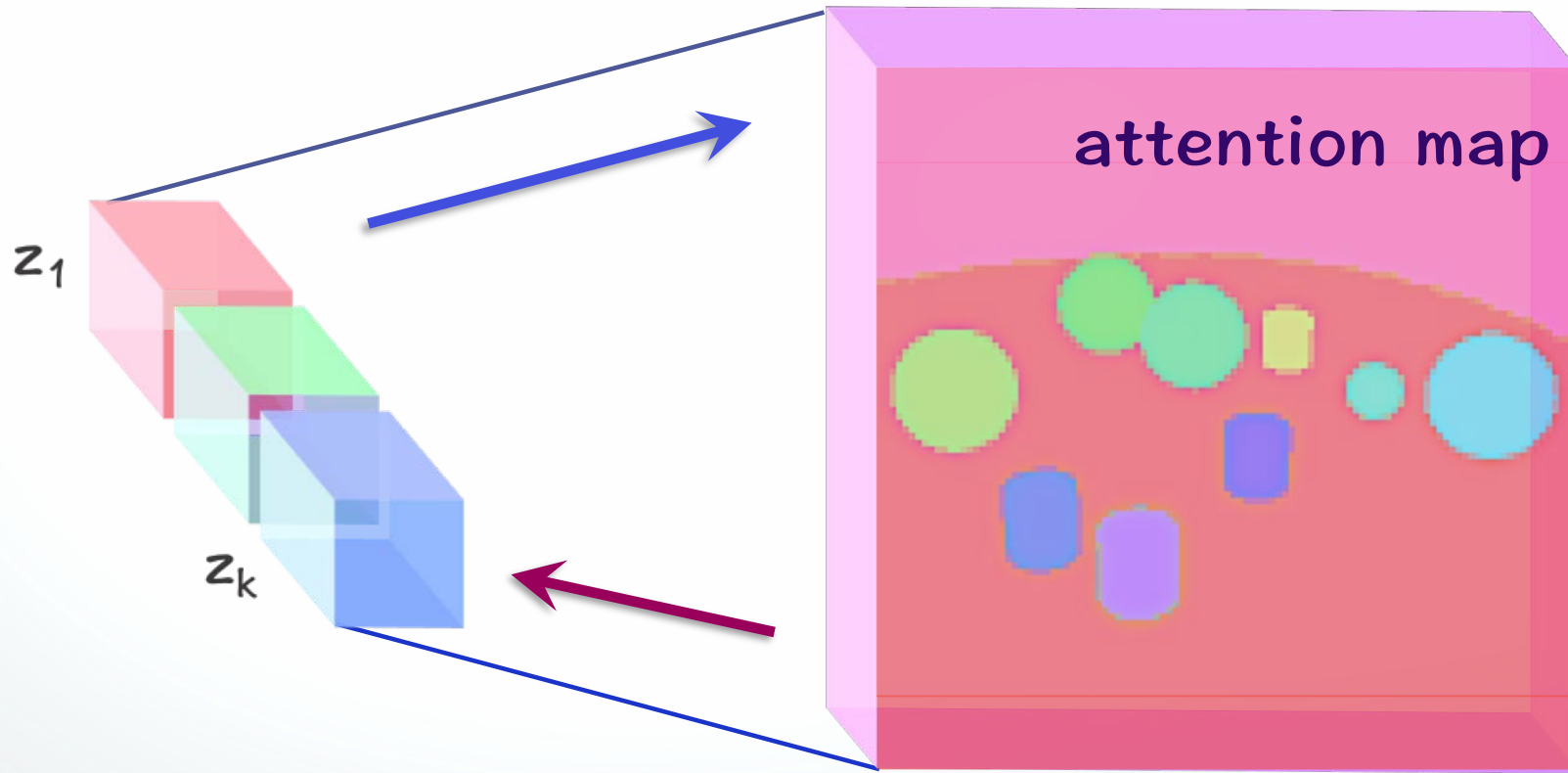
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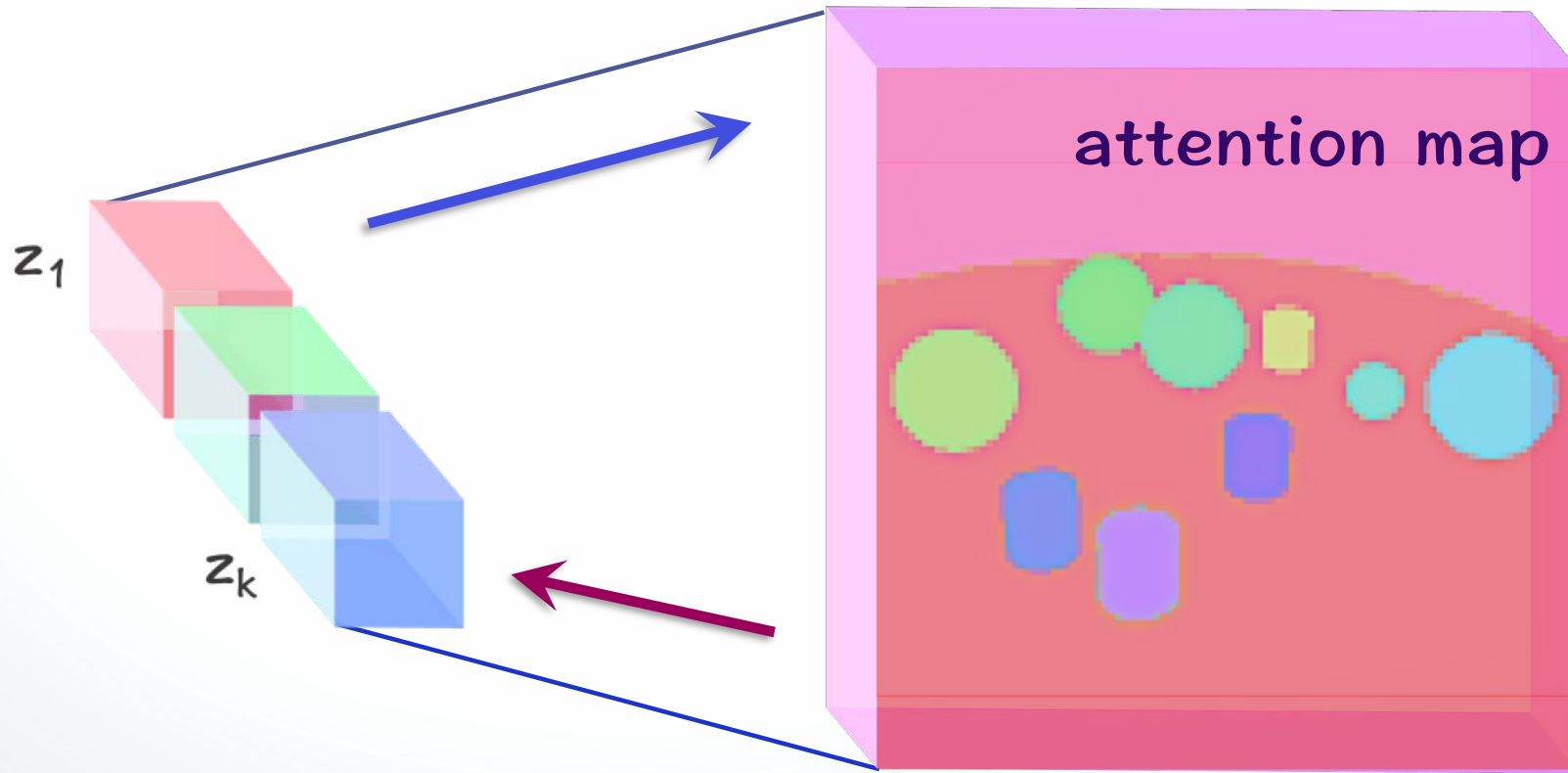
# From Transformer to GANformer



**Bidirectional interaction** between the **latents** and the **image** enables **bottom-up & top-down processing**.

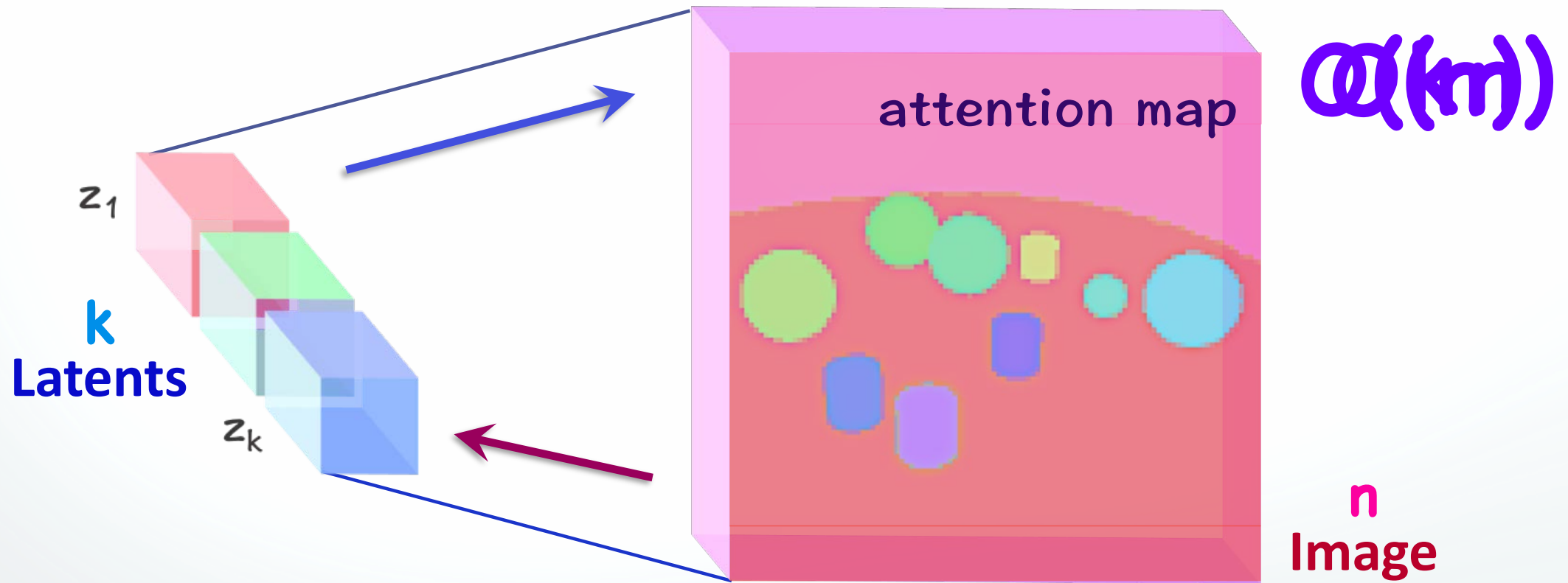


# From Transformer to GANformer



The **latents** guide the image synthesis **cooperatively**,  
**attending** and **modulating** different **objects** or **entities**.

# From Transformer to GANformer



Has **linear efficiency** that **scales to high-resolutions** while capturing **long-range dependencies**.



# GANformer's Attention Maps

The latents attend to **semantic entities** and **cooperatively** generate **compositional** scenes.



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The latents attend to **semantic entities** and **cooperatively generate compositional scenes**.

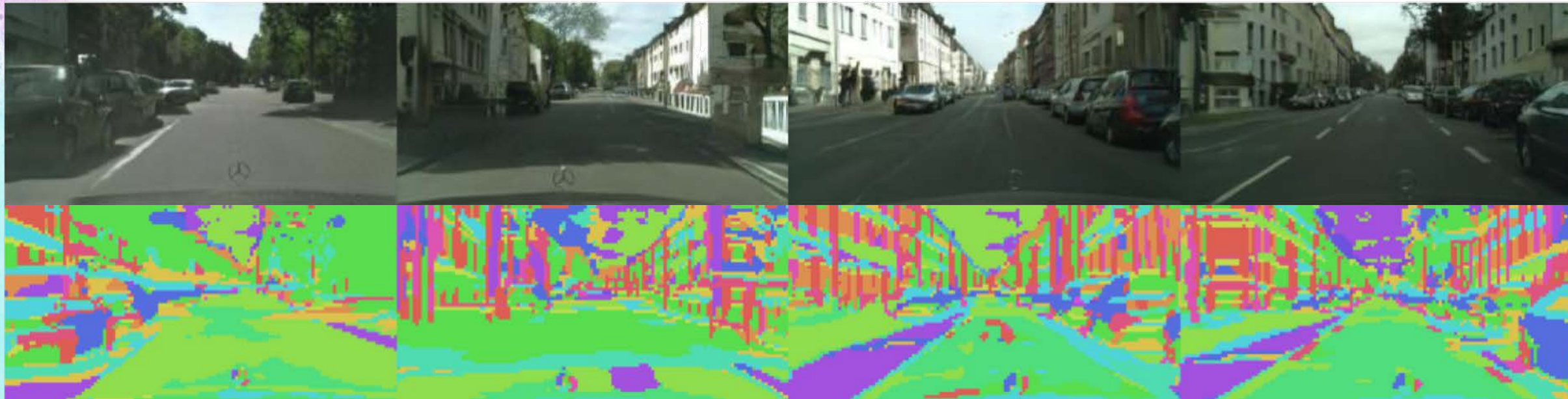




# GANformer's Attention Maps



The latents attend to **semantic entities** and **cooperatively generate compositional scenes.**

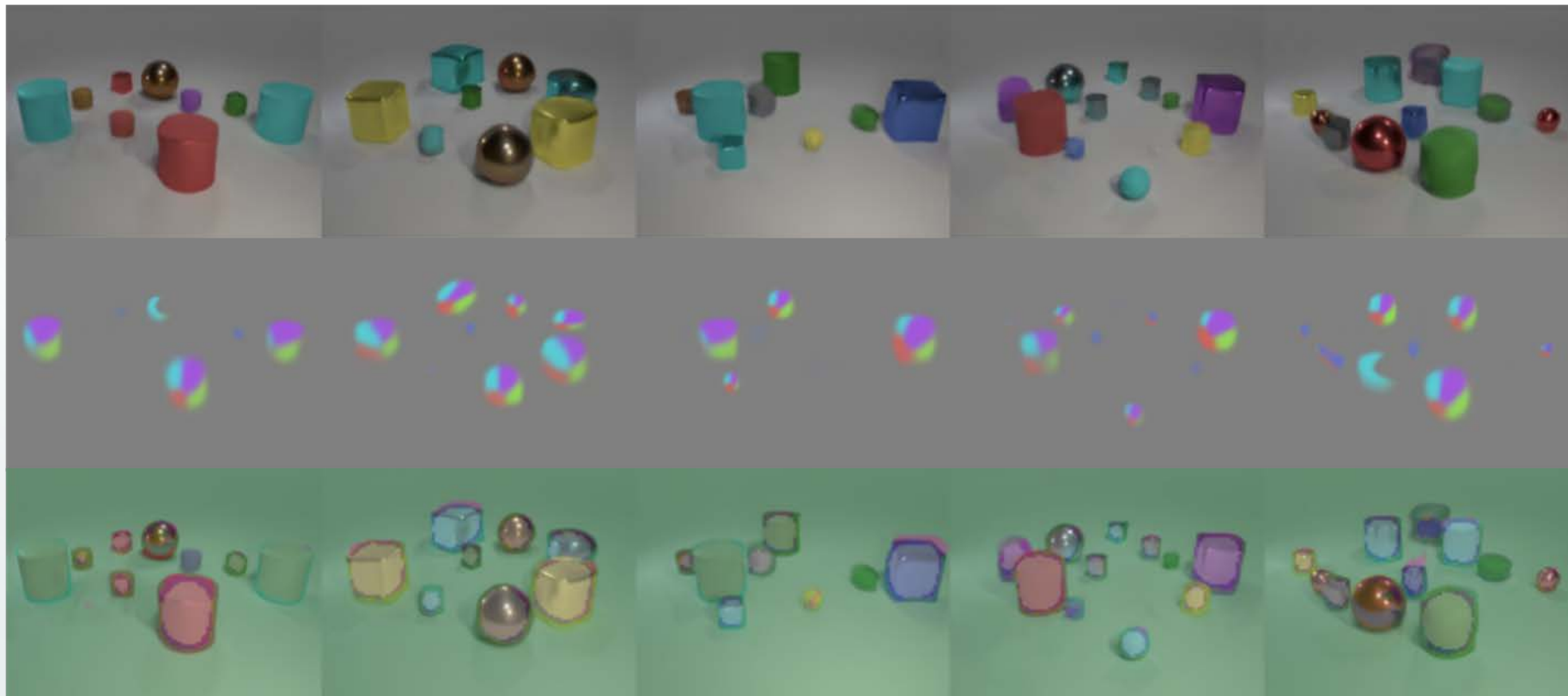




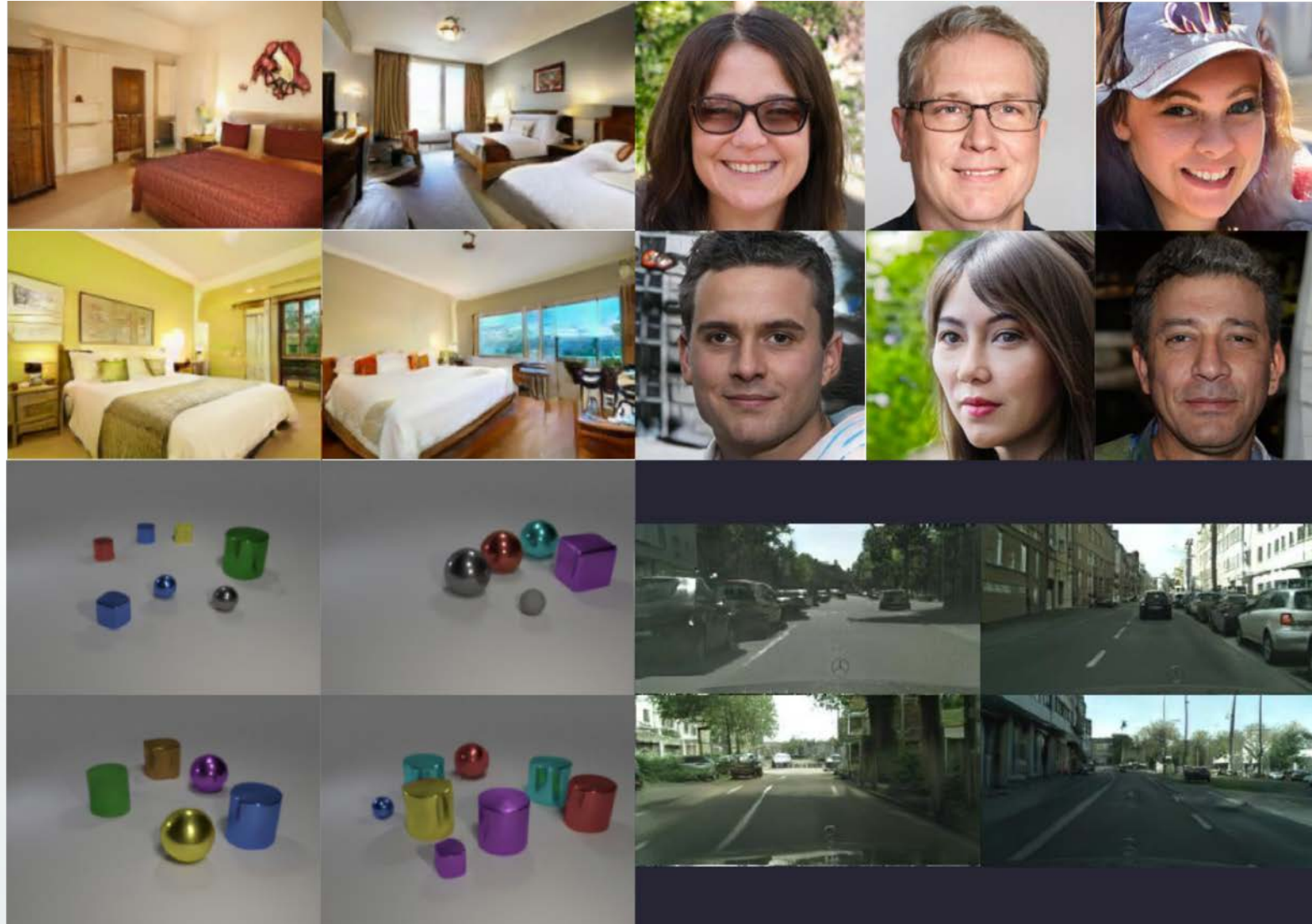
# GANformer's Attention Maps

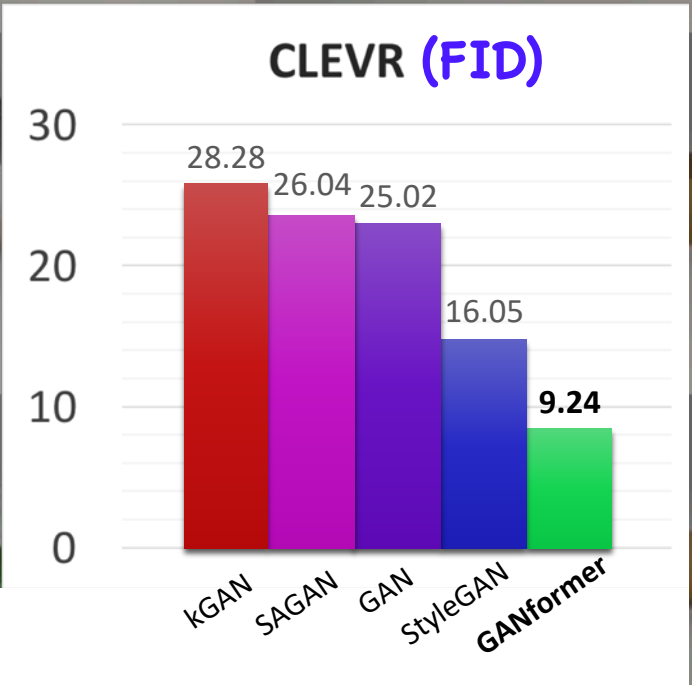
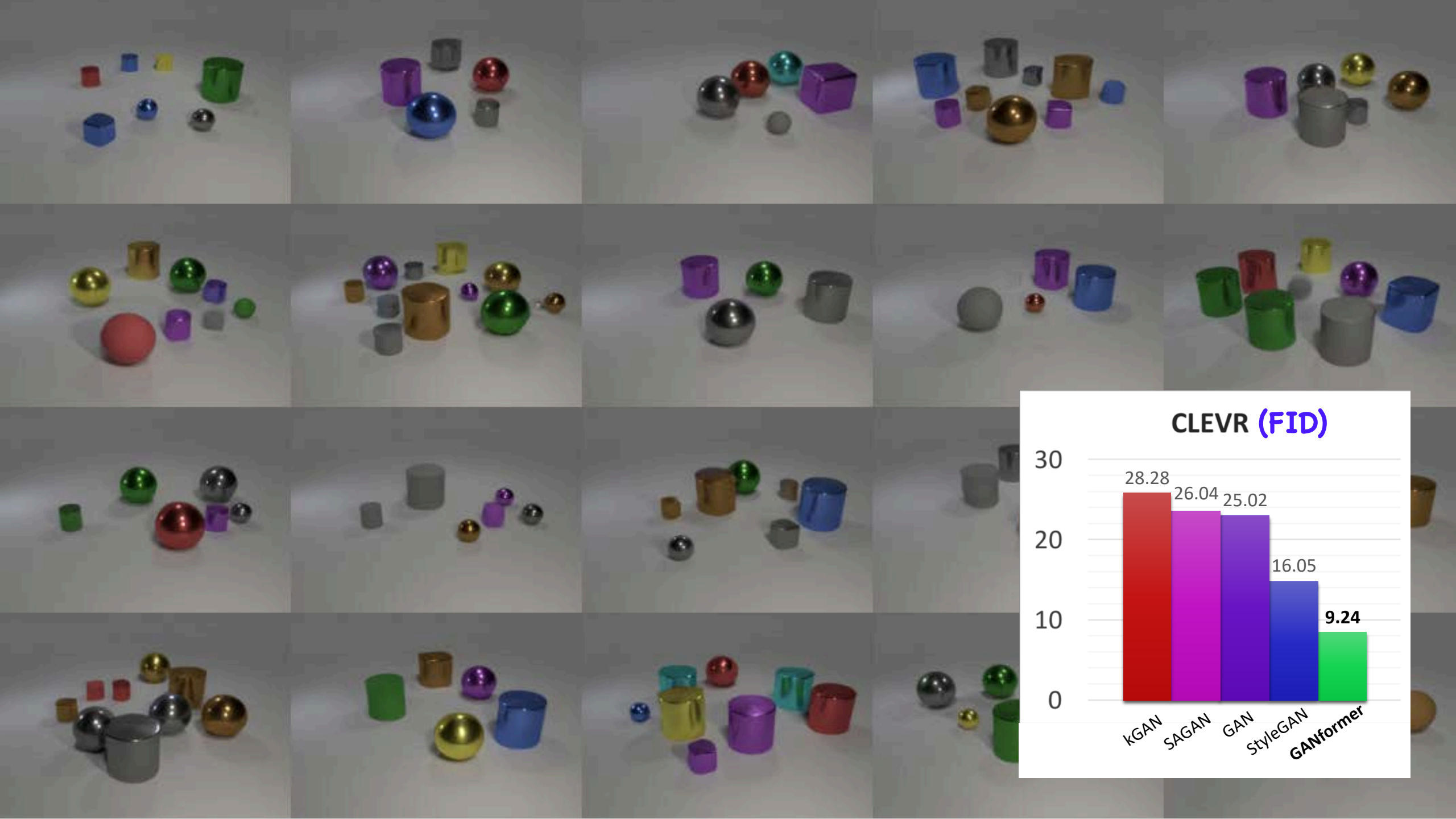


The latents attend to **semantic entities** and **cooperatively** generate **compositional** scenes.

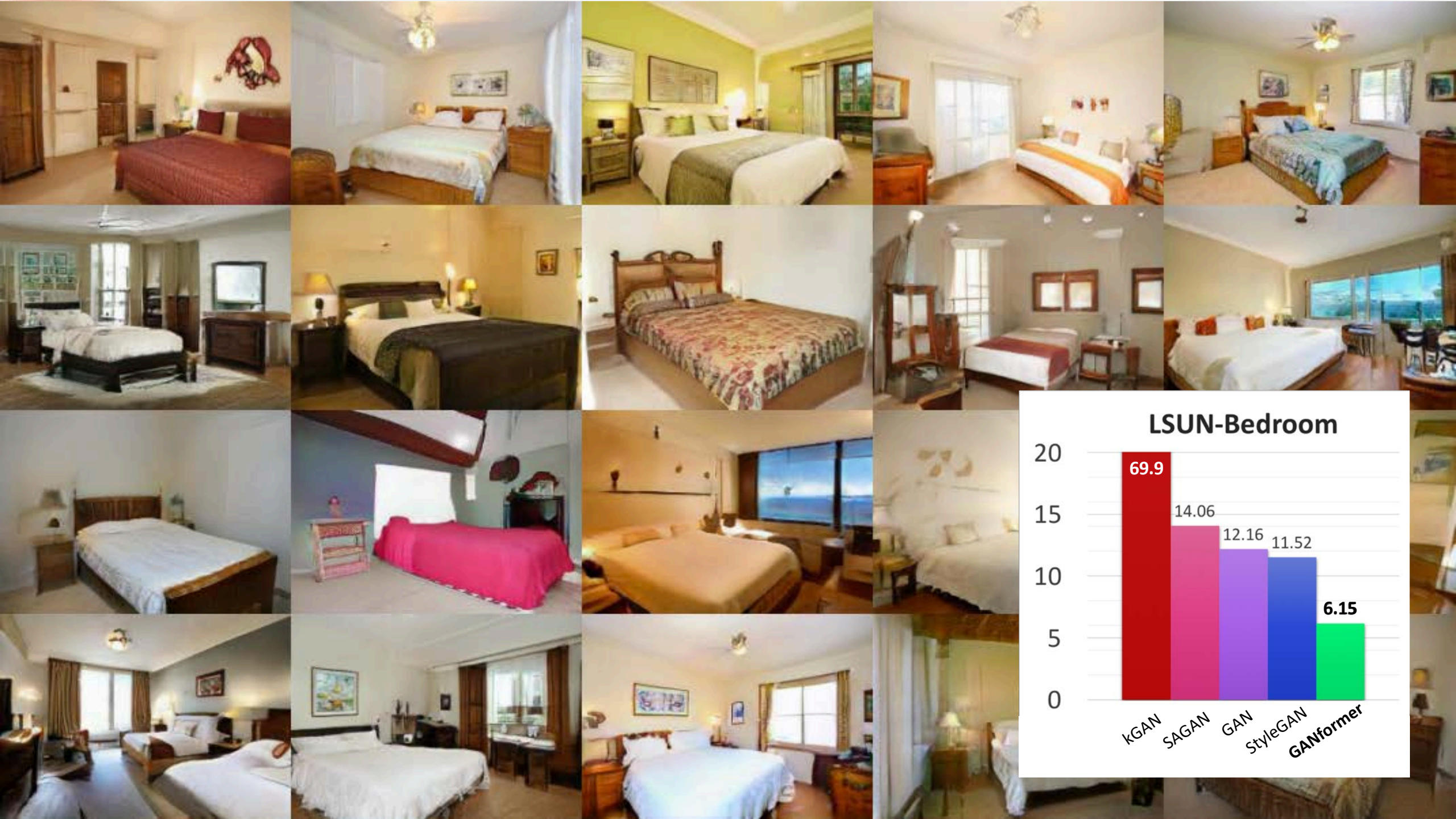


# Excels on highly-structured scenes

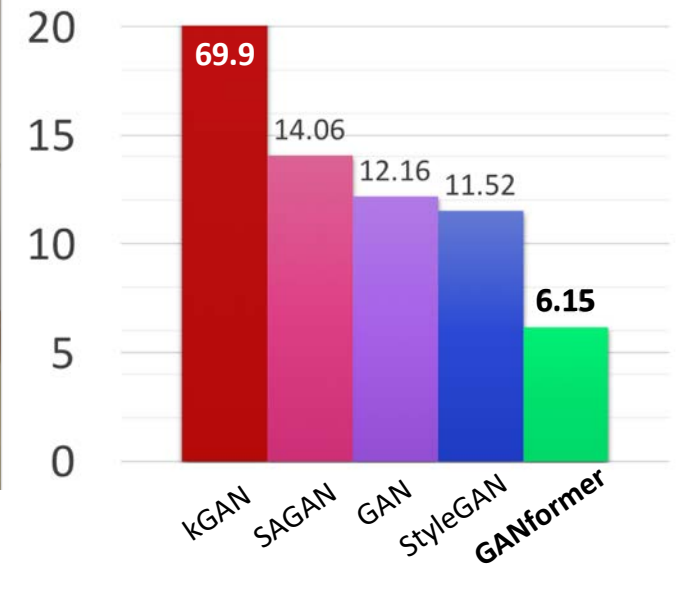








**LSUN-Bedroom**



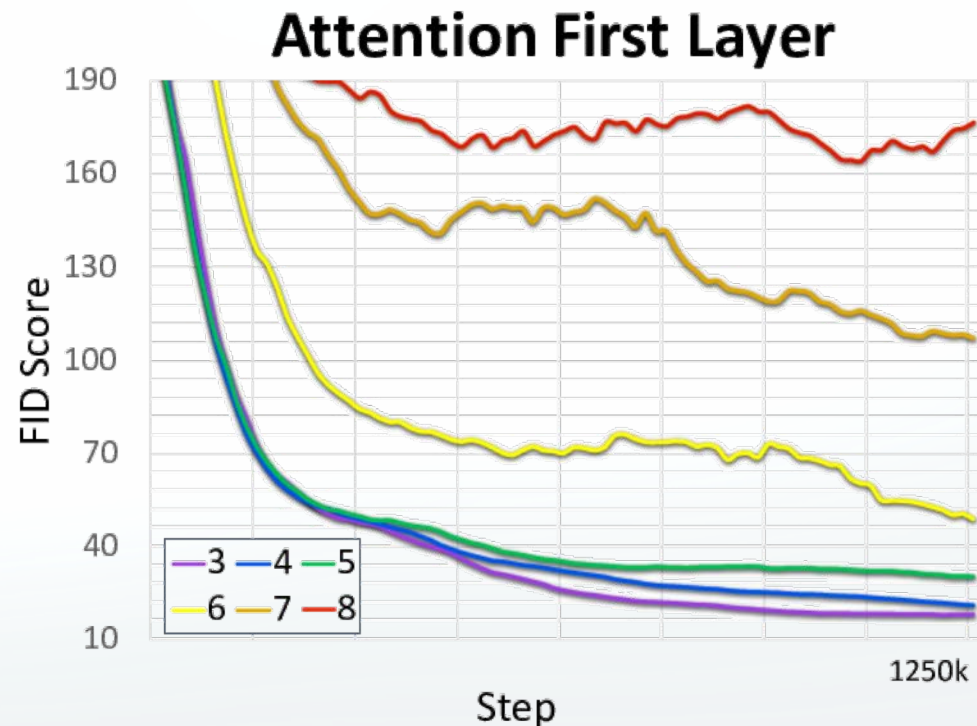
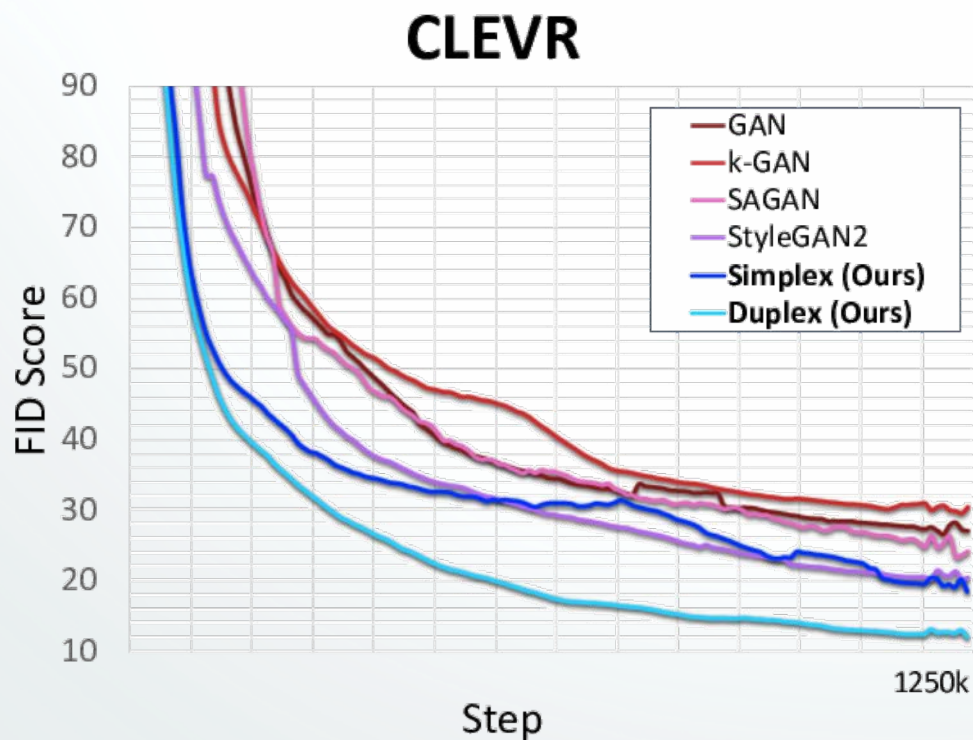




# GANformer Model Analysis

The model **learns faster** and enjoys **higher data-efficiency**.

Latents are more **disentangled**; Images are more **diverse**.





📊 State of the Art Image Generation on CLEVR   📊 State of the Art Image Generation on Cityscapes  
 📊 State of the Art Image Generation on LSUN Bedroom 256 x 256 (FID-10k-training-steps metric)

python 3.7   tensorflow 1.14   cudnn 10.0   license MIT

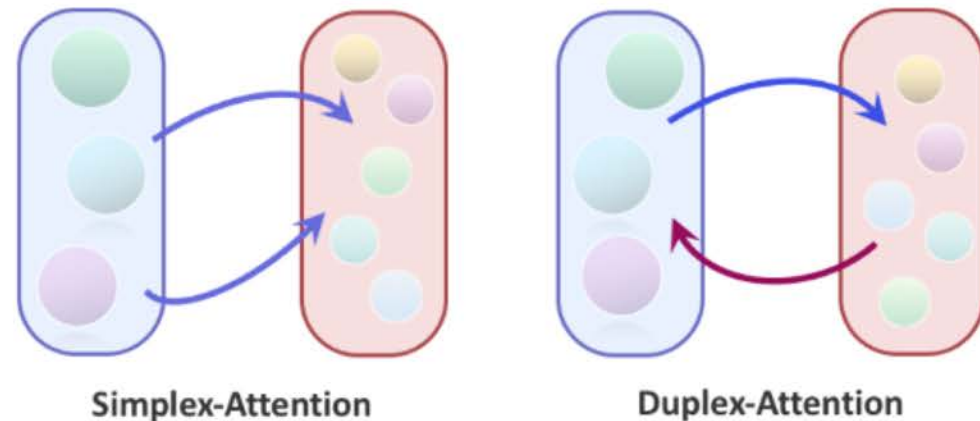
# GANformer: Generative Adversarial Transformers

[github.com / dorarad / gansformer](https://github.com/dorarad/gansformer) 

*\*I wish to thank [Christopher D. Manning](#) for the fruitful discussions and constructive feedback in developing the Bipartite Transformer, especially when explored within the language representation area and also in the visual context, as well as for providing the kind financial support that allowed this work to happen! 🌻*



The Bipartite Transformer





[github.com / dorarad / gansformer](https://github.com/dorarad/gansformer)

**Thank you!** 😊

