



硕士生学术交流

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内容安排

CVPR2020线上会议参会感受分享

研究问题概述 Problem Definition

Paper Reading 1: CVPR20 MixNMatch

Paper Reading 2: CVPR20 SEAN

Future Works

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● 基本情况介绍

- 西雅图时间6月14日至19日在线上成功召开。
- 有7600名与会者，25个赞助商，5025个活动项目（论文/演讲等），以及1,497,800分钟的ZOOM讨论。
- 6月14日、15日、19日为Workshops和Tutorials
- 16日~18日为Main Conference（Oral和Poster）

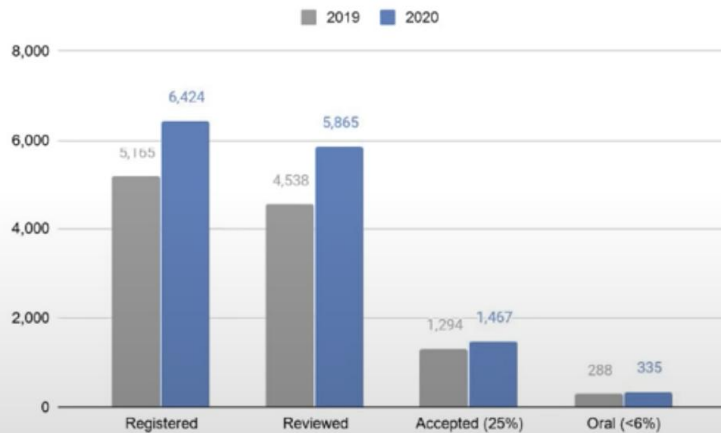


● 基本情况介绍

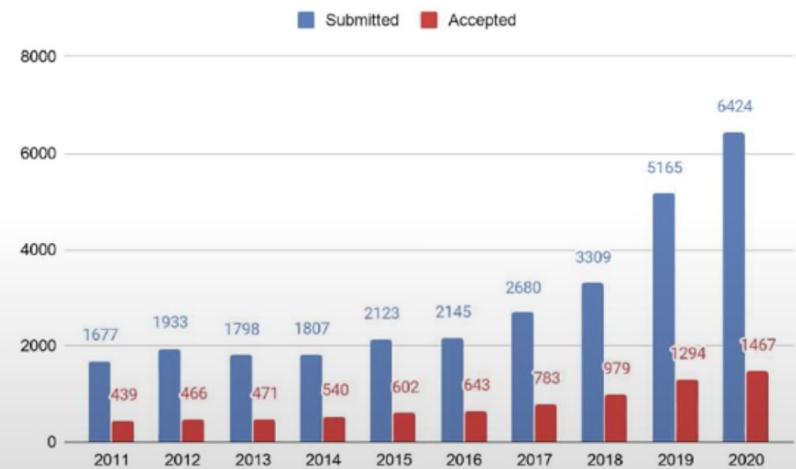
Posters/Orals

- **6,424 registered** (vs. 5,165 in 2019)
- **5,865 valid submissions** (vs. 4,538 in 2019)
- **1,467 accepted (25.0%)**
- **335 orals (5.7%)**

As before, papers were accepted as orals and posters purely based on the quality. There were no caps set in the paper decision process.

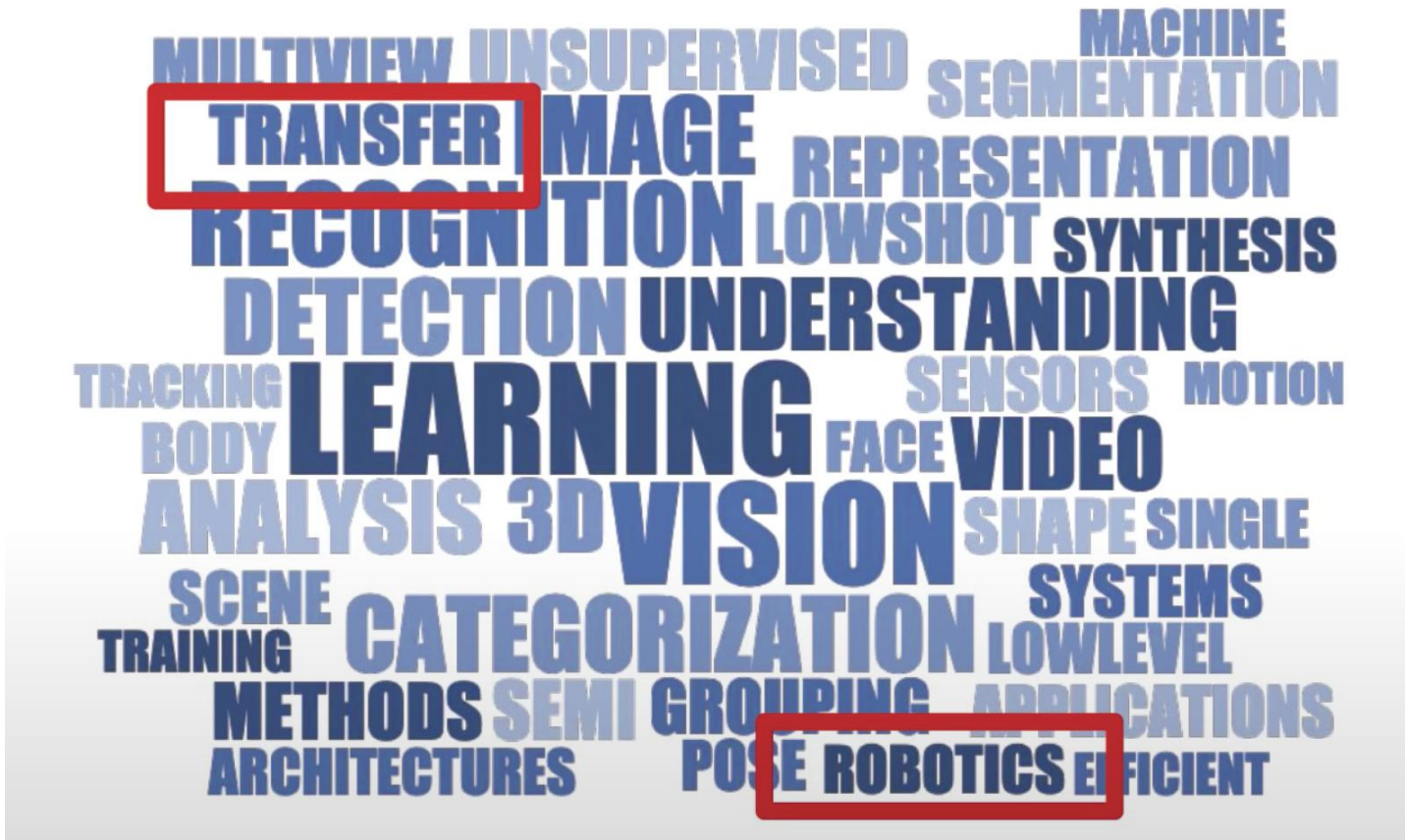


Registered vs accepted last 10 years



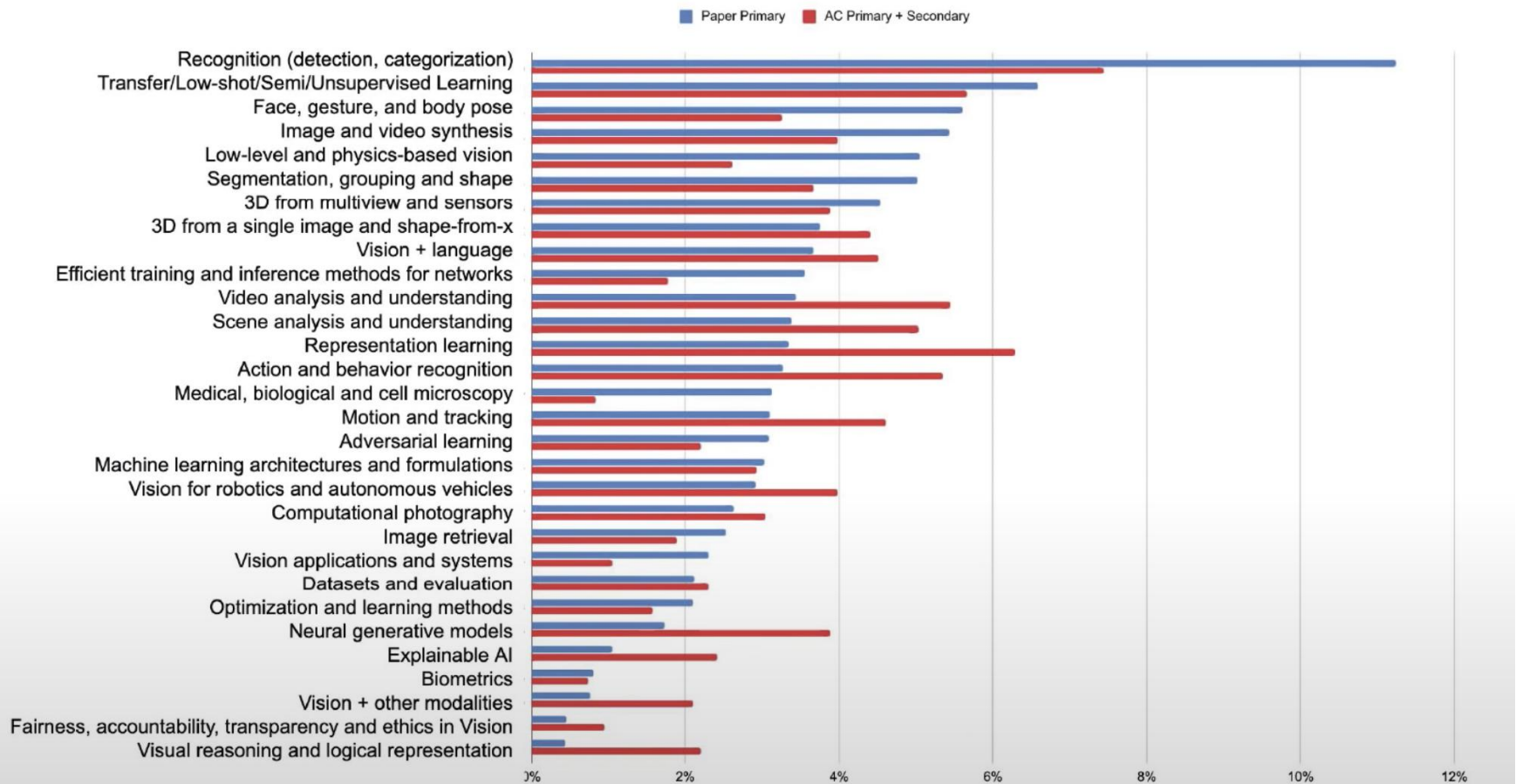
- 基本情况介绍

Popular Areas



● 基本情况介绍

Subject areas vs AC expertise



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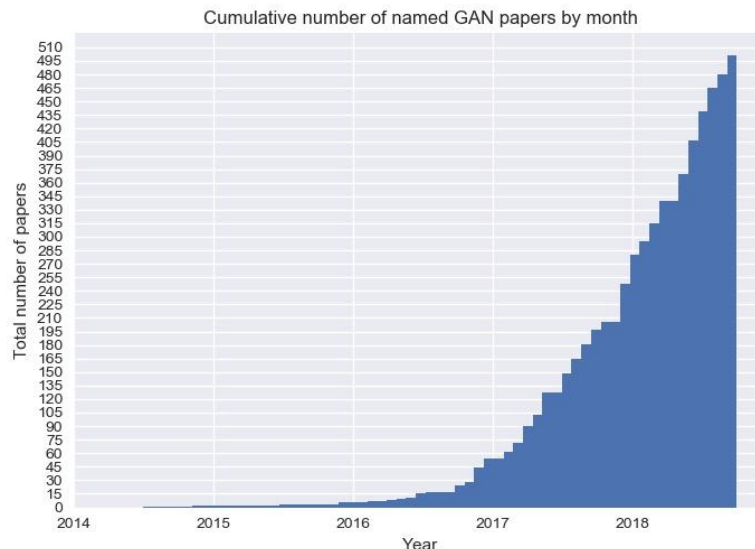
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Future Works

● 生成对抗网络GAN

由于能够从训练数据中学习数据分布从而生成全新样本，GAN为机器学习提供了一个全自动自学习的损失函数。



- 各种图像生成任务取得了成功
 - 包括人脸和自然图像生成，图像跨域转换，图像超分辨率和文本到图像合成等。
- 相关理论和技术快速发展
 - ✓ 训练不稳定及模式坍塌
 - ✓ 评价指标
 - ✗ 局部与全局收敛性及时间效率
 - ✗ 与其他生成模型的互补组合

Generator	The Evolution of GANs	Discriminator
DCGAN Generator	ACGAN IS:28.5	AC Discriminator
Class Batch Norm	SNGAN IS:36.8 FID: 27.2	Projection Discriminator Hinge Loss Spectral Normalization
Spectral Normalization Self-Attention Module	SAGAN IS:52.5 FID: 18.7	Self-attention Module
Large Batch Training Shared Embedding in ClassBN Truncated Trick	BigGAN IS:98.8 FID: 8.7	Large Batch Training Truncated Trick
Deeper Architecture	BigGAN-deep IS:124.5 FID: 5.7	Deeper Architecture

● 样式（风格）编码

- 样式（风格）控制是各种图像合成和处理应用的重要组成部分，是可控的条件生成的重要手段。
- 样式通常不是由用户手动设计的，而是从参考图像中提取的。

Content Image



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Style Image



Starry Night by Van Gogh

+

=

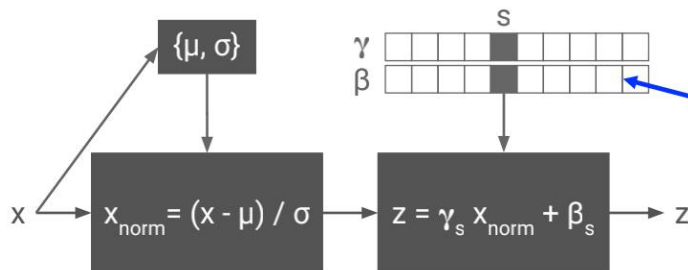
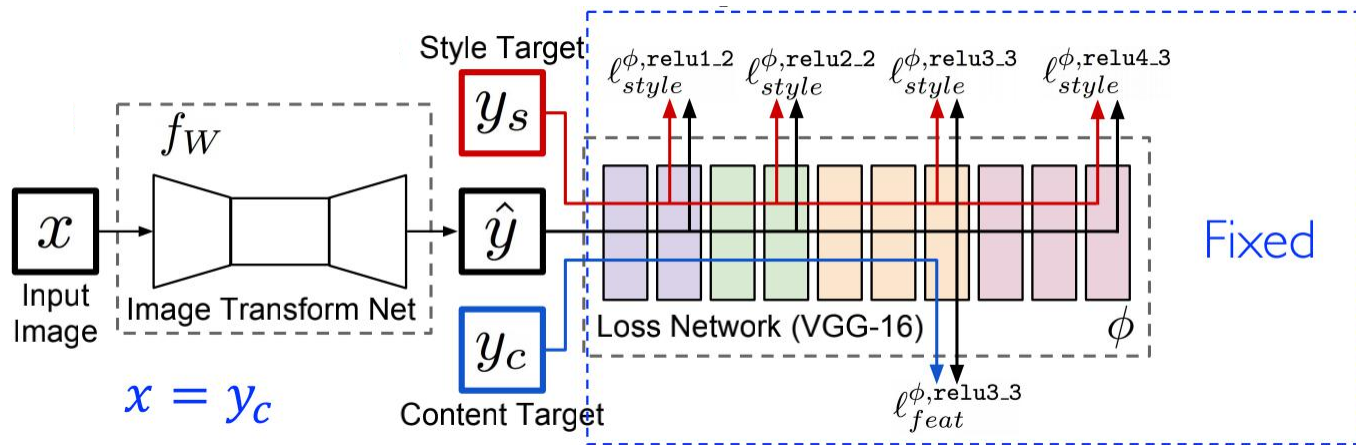
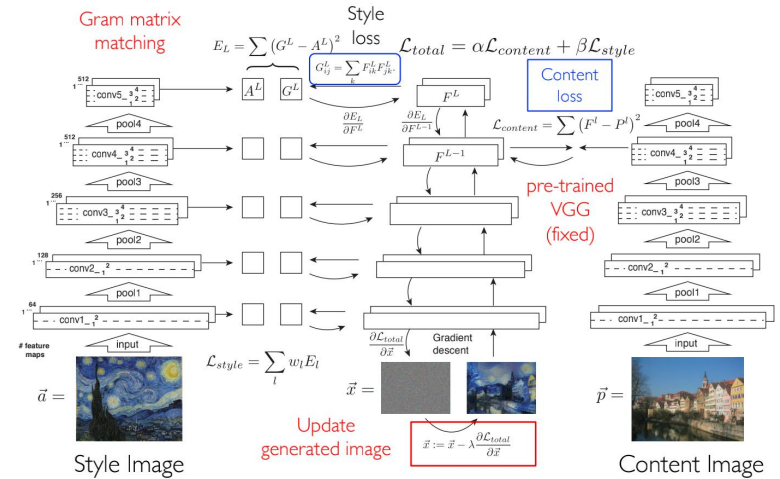
Style Transfer



● 样式（风格）编码

■ 样式编码三大方式：

- ▶ 编码为图像特征的统计量；
- ▶ 编码为神经网络的权重；
- ▶ 编码为网络规一化层的参数。

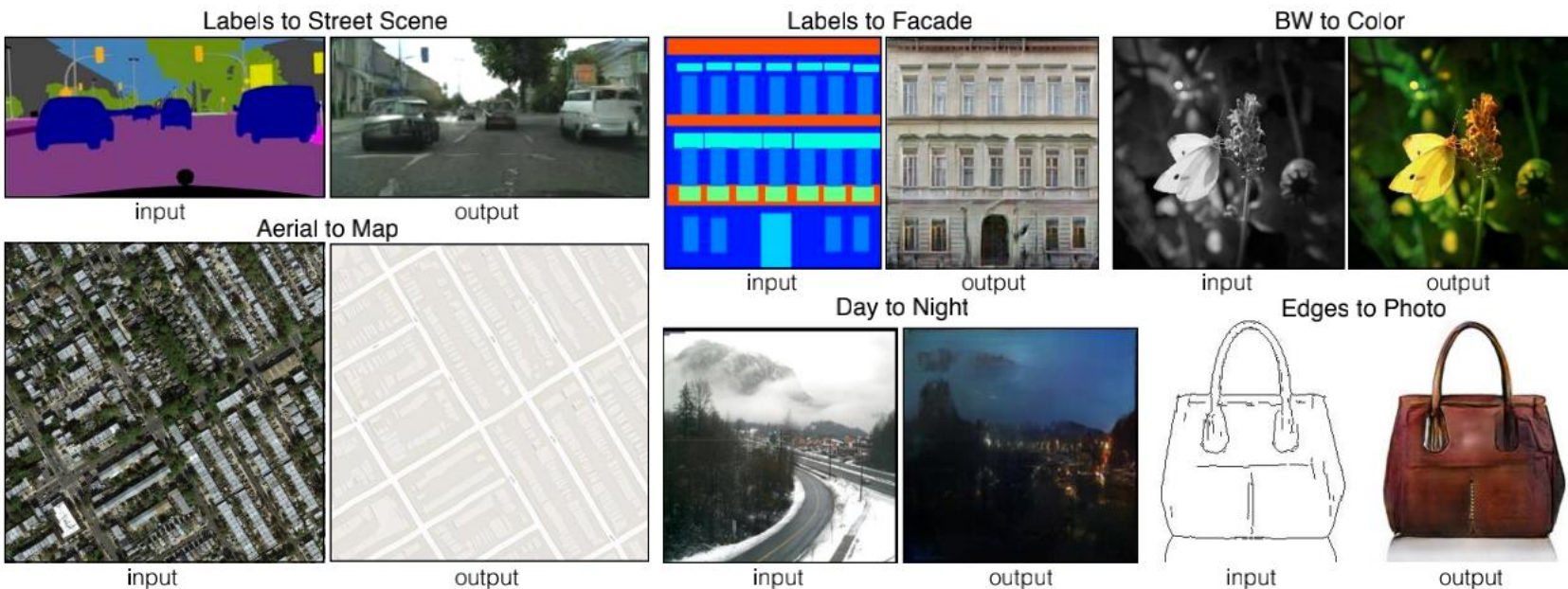
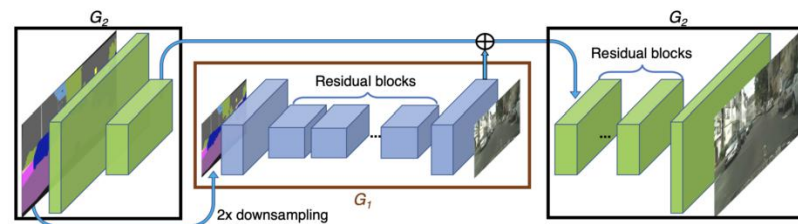
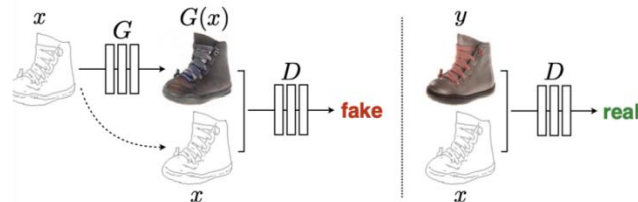


Each style image is corresponding to a group of γ_s and β_s .

● 跨域图像转换I2I

计算机图形学和计算机视觉中的许多问题都可被视为将输入图像“翻译”为相应的输出图像的图像处理问题，例如着色，超分辨率，以及逼真的语义图像合成（标签→图像），它是一种特别有用的类型，可以通过修改输入的语义布局图像轻松地进行用户控制。

成对的训练数据



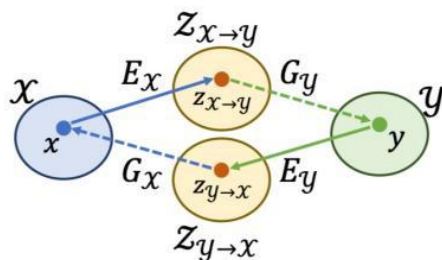
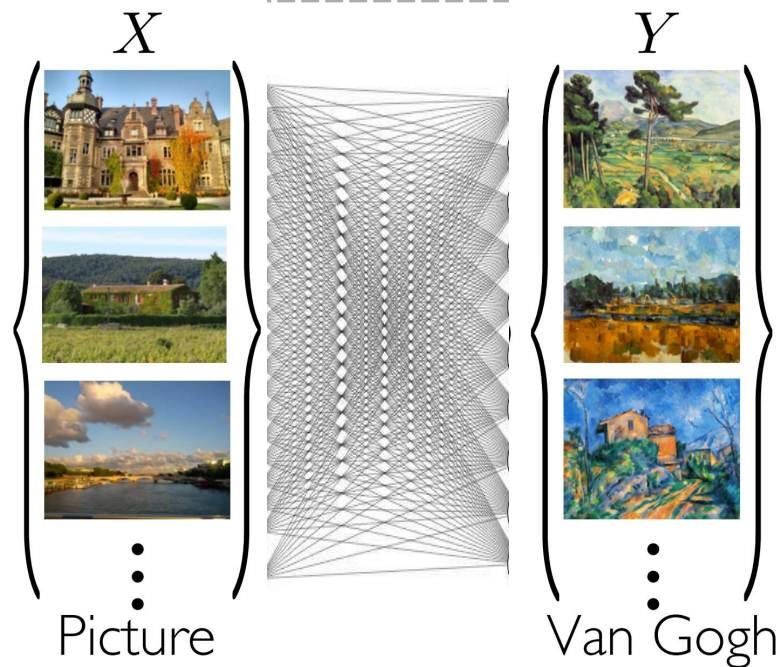
● 跨域图像转换I2I

✗ 对齐的训练图像对难以获得，
限制其应用

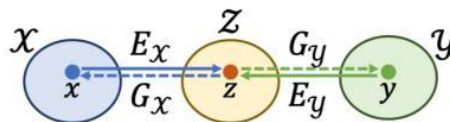
■ 无监督图像转换，无需成对数据

- ✓ 循环一致性损失
- ✓ 多对多映射，多模式生成
- ✓ 域共享隐层空间
- ✓ 解耦特征空间
- ✓ 注意力模块特征选择

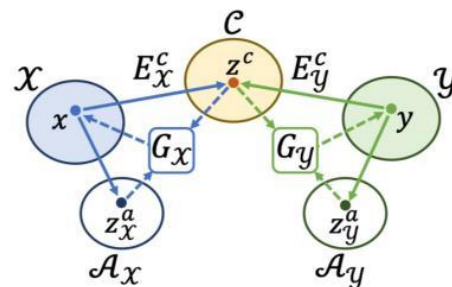
不成对的训练数据



(a) CycleGAN



(b) UNIT



(c) MUNIT , DRIT

● 语义分割域自适应

- 语义分割是计算机视觉中最具挑战性的任务之一，它试图预测给定图像或视频帧的像素级语义标签，以方便下游应用，例如自动驾驶，视频监控和图像编辑。
- 基于CNN方法的关键是标注大量涵盖可能的场景变化的图像。

无监督的语义分割

Training

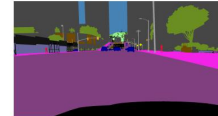
source image



Labeller NN



label mask



Testing

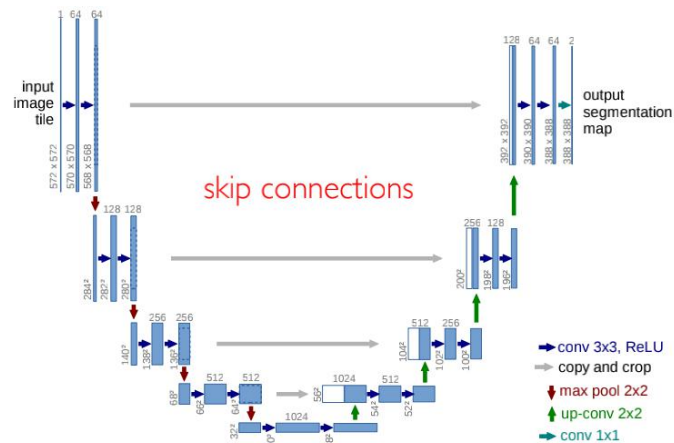
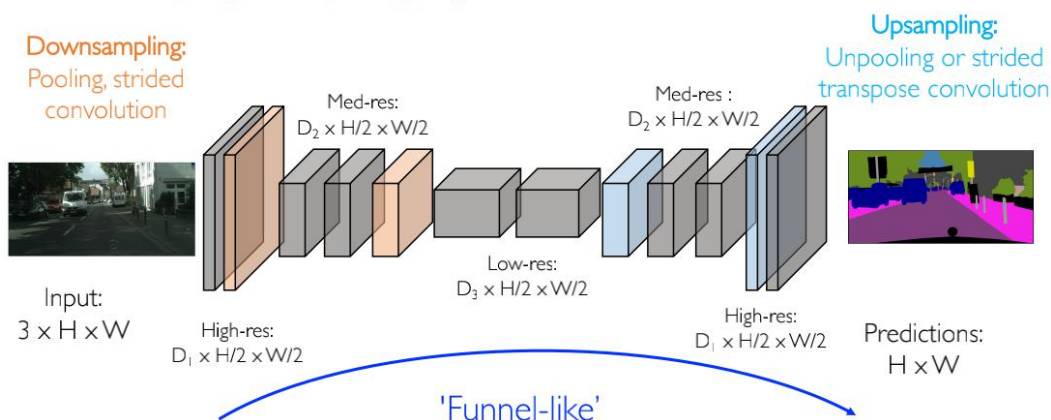
target image



Labeller NN



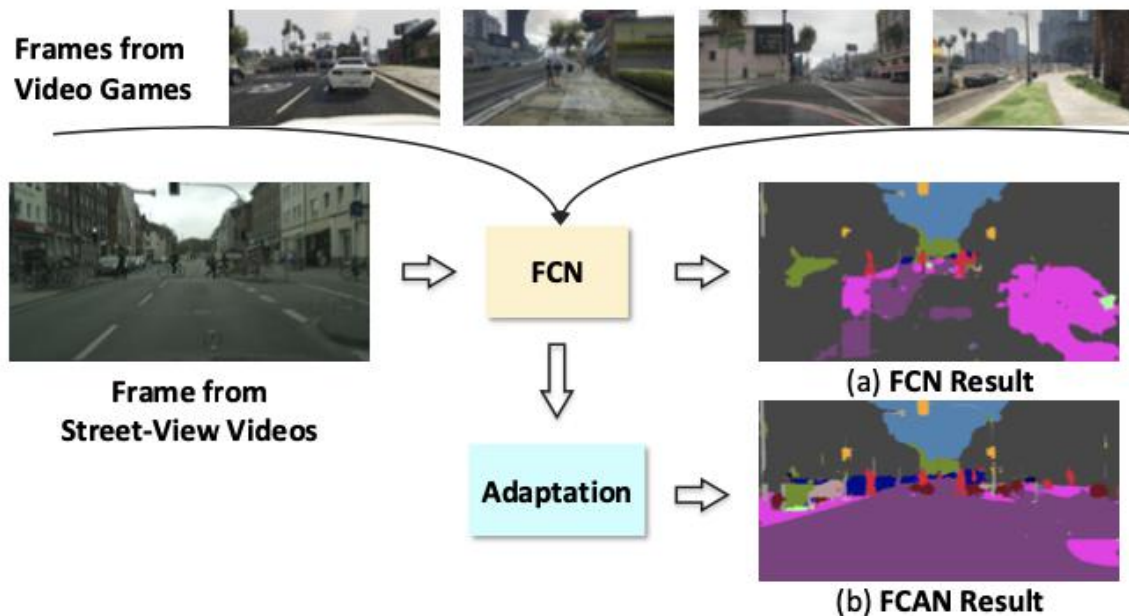
?



● 语义分割域自适应

- 利用计算机图形技术来模拟虚拟环境并自动生成图像和标签是一种经济的选择。

- 例如GTA5和SYNTHIA 是两个热门的城市街道合成数据集，与真实数据集（例如Cityscapes）相似且具有重叠的类别。尽管相似，但在纹理，布局，颜色和照明条件上仍然存在细微差异，这导致数据分布不同或域差异。因此，在合成数据集上训练的某个模型的性能在应用于现实场景时会急剧下降。



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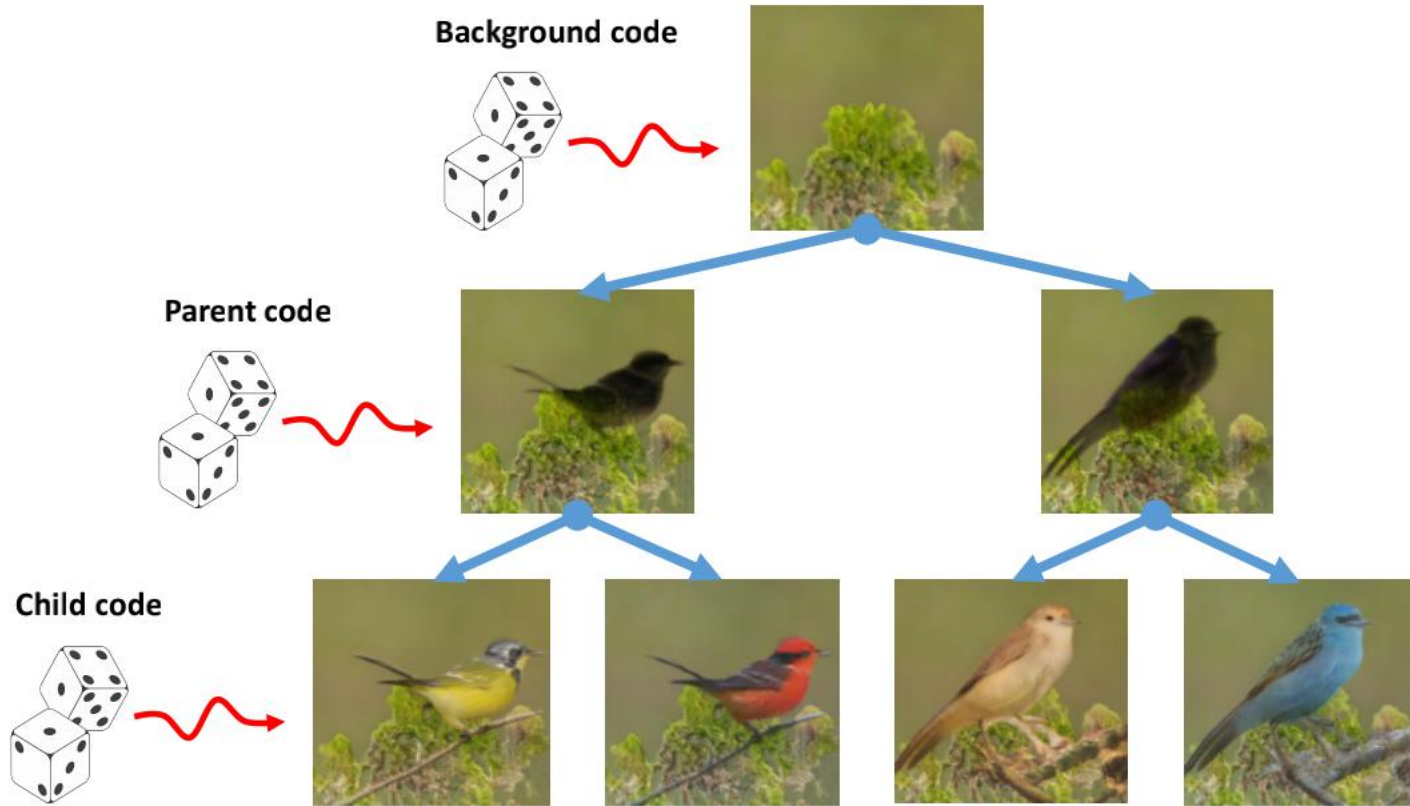
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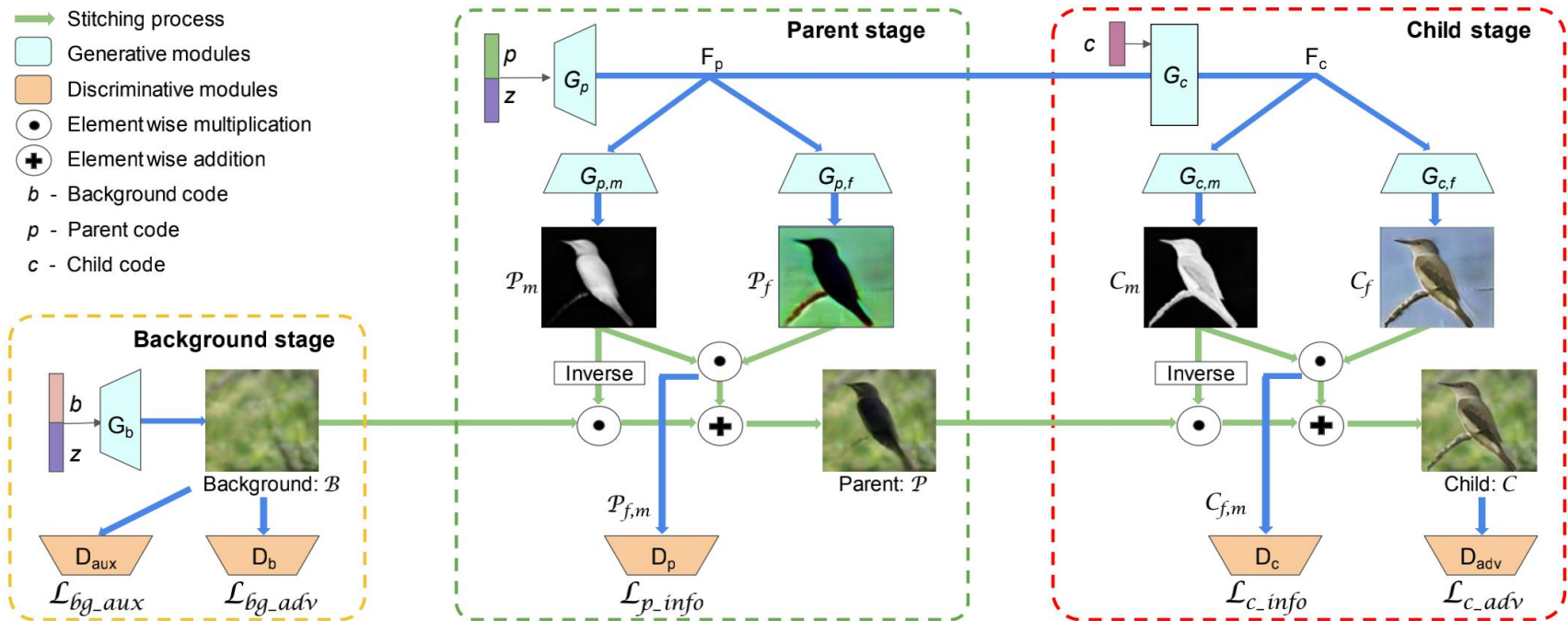
研究问题概述 Problem Definition

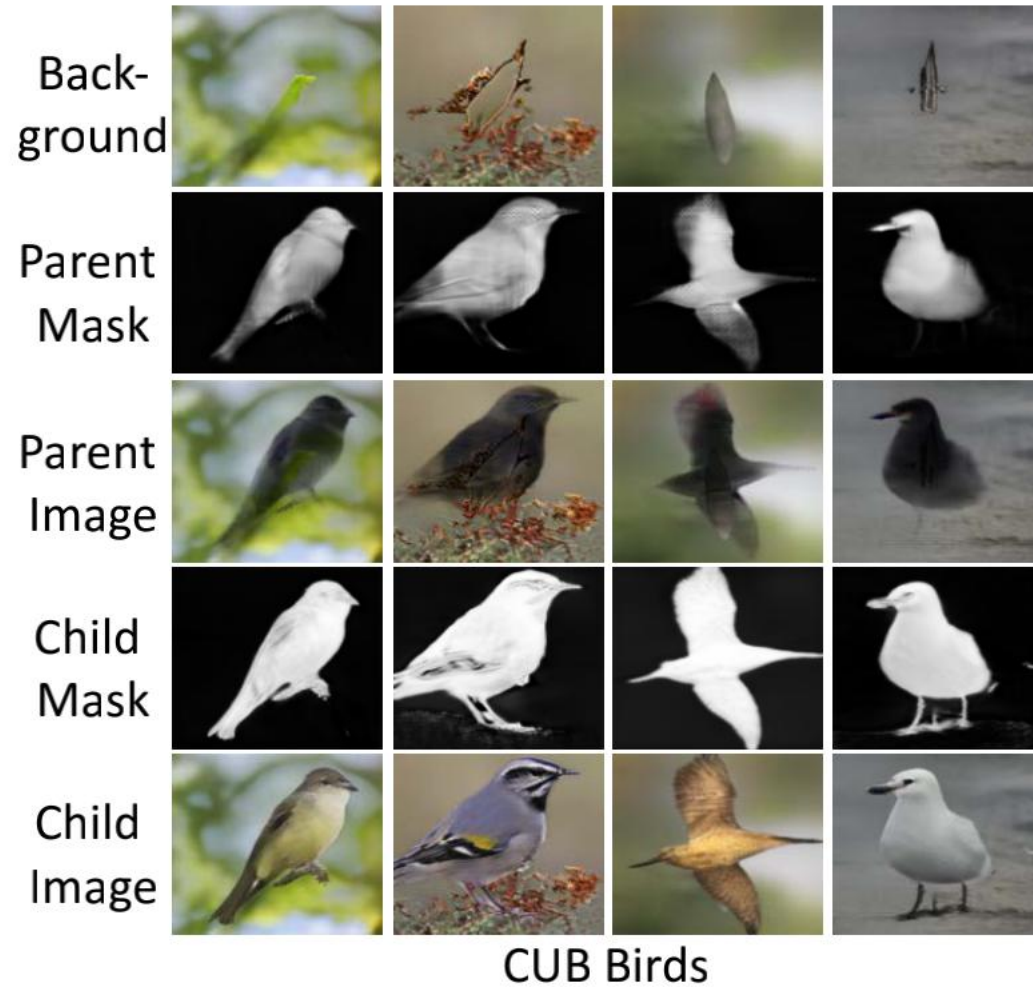
Paper Reading 1: CVPR20 MixNMatch

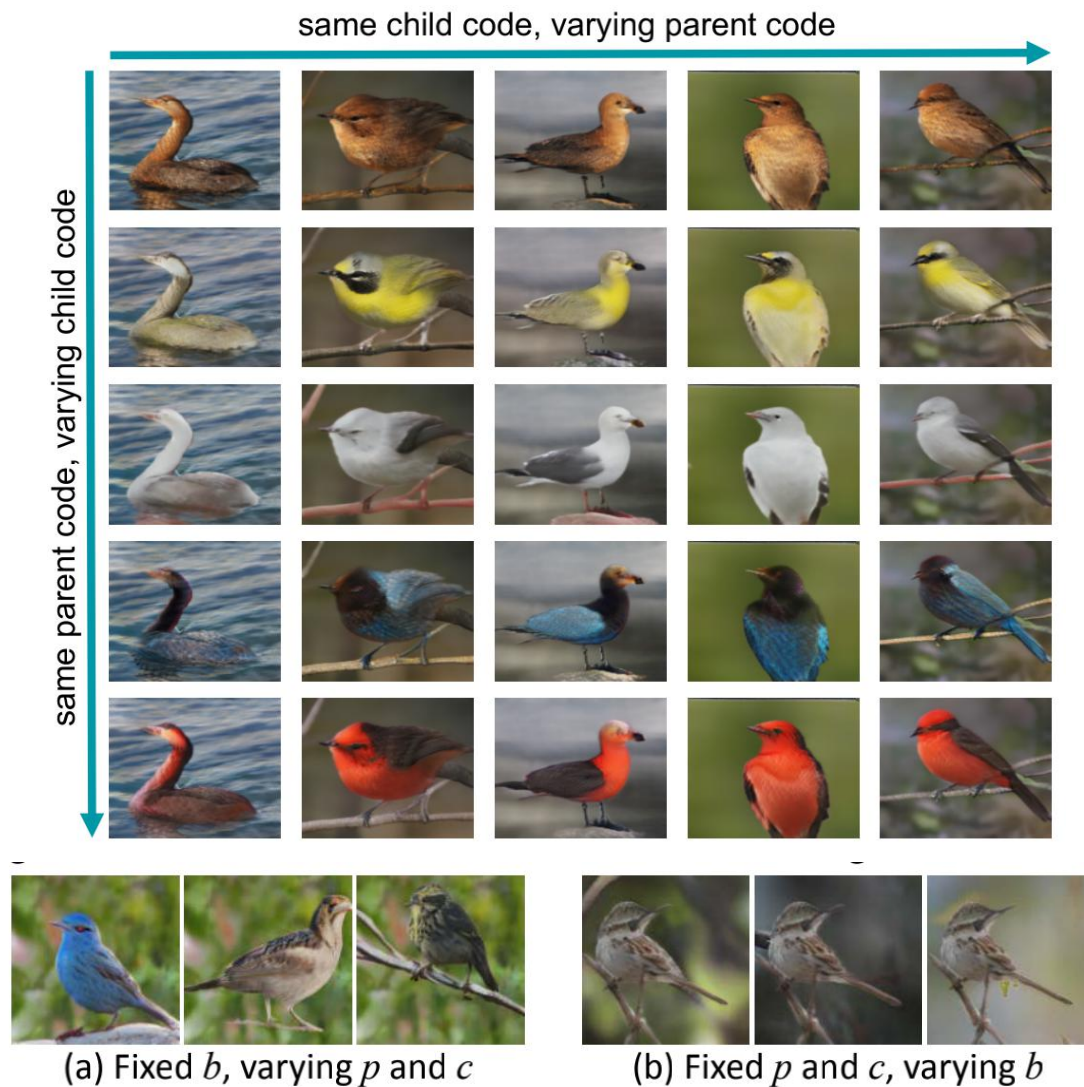
Paper Reading 2: CVPR20 SEAN

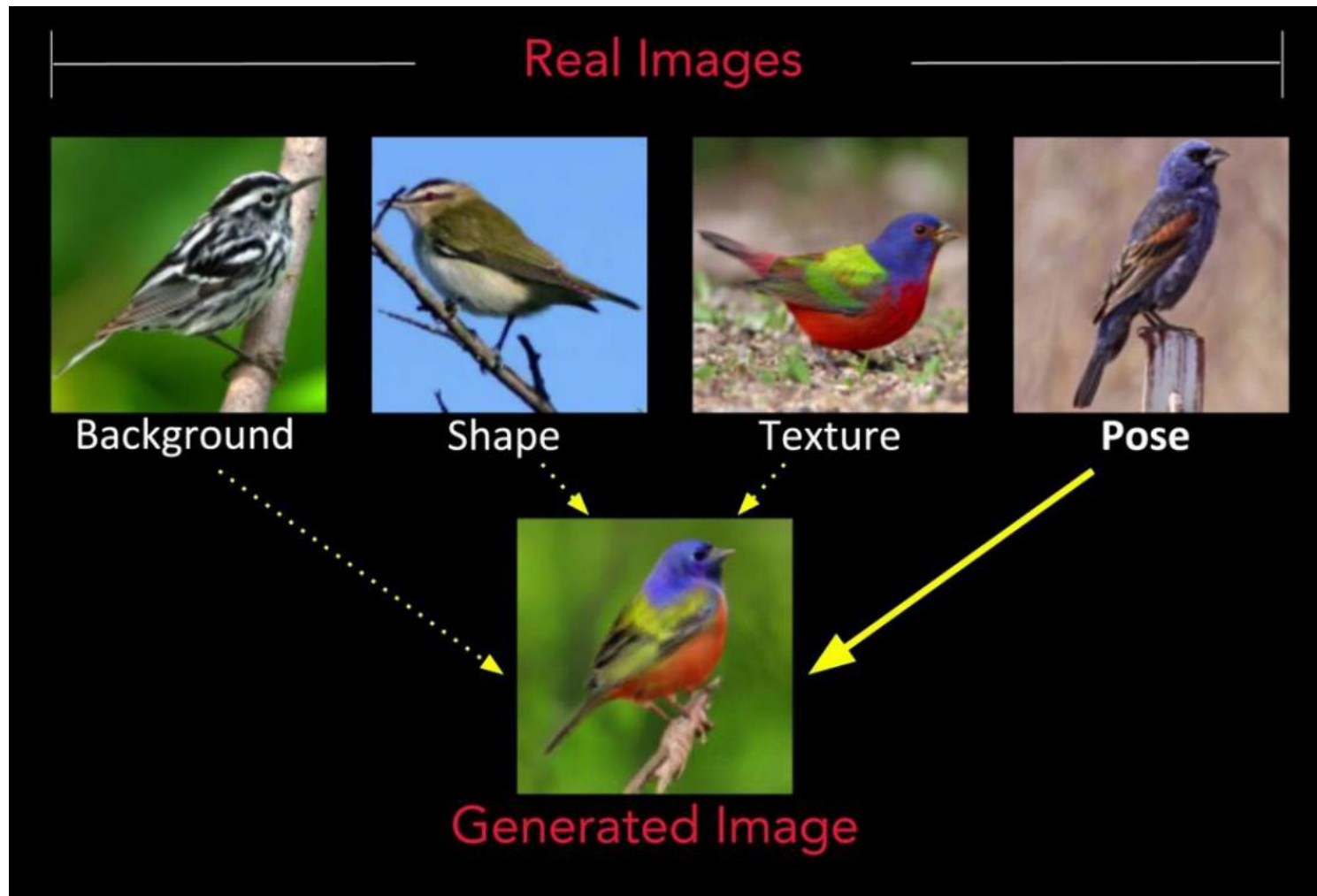
Future Works











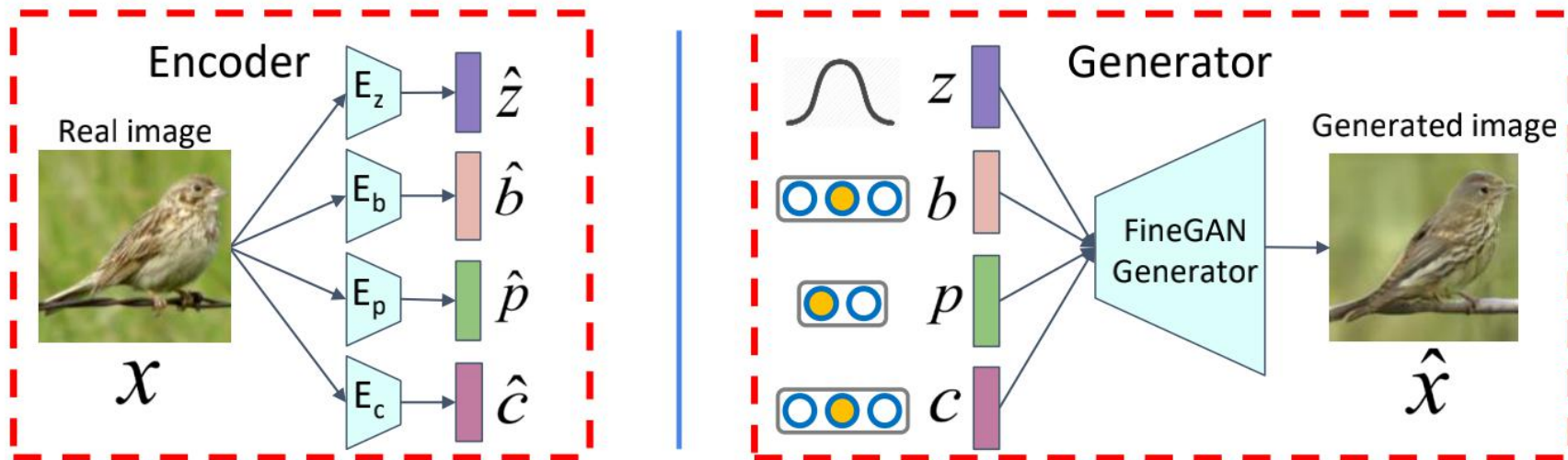
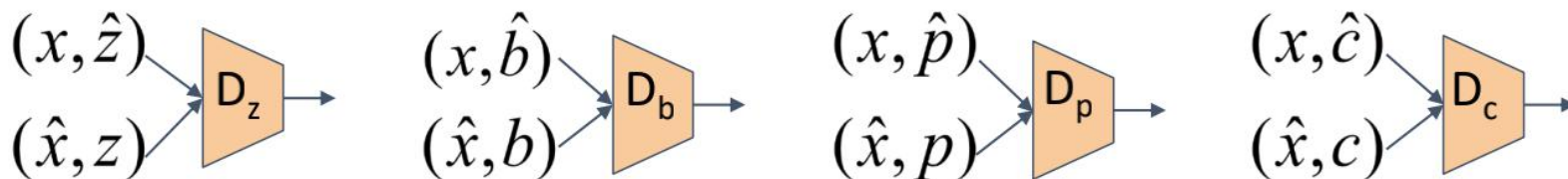


Image-code *paired* distributions need to be matched



Adversarial loss used to match *paired* distributions

Cartoon/Sketch \rightarrow Image

shape+pose



background



texture



generation



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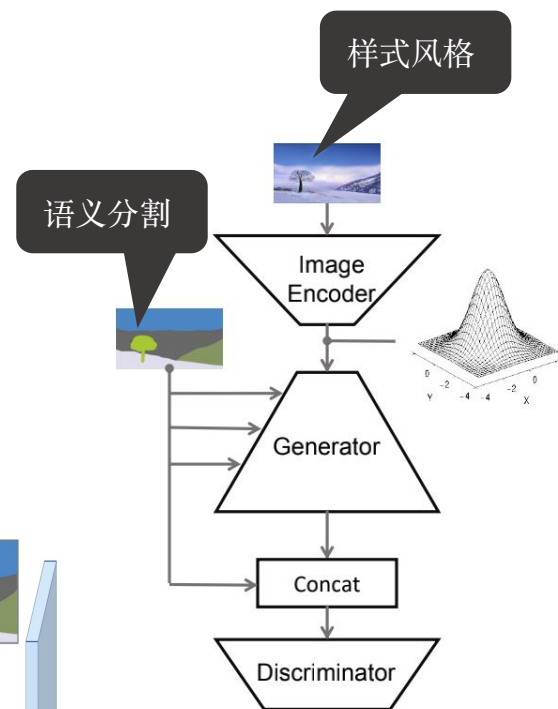
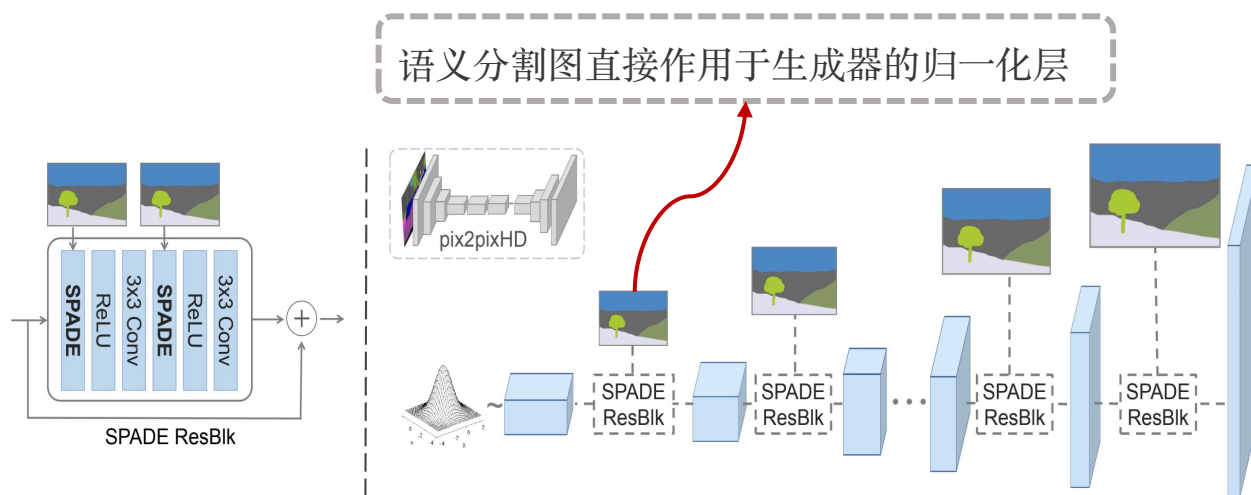
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Future Works

■ 成对监督图像转换工作 SPADE(CVPR'19Oral)

- 支持多样化输出
- 输出满足分割语义和场景风格的合成图

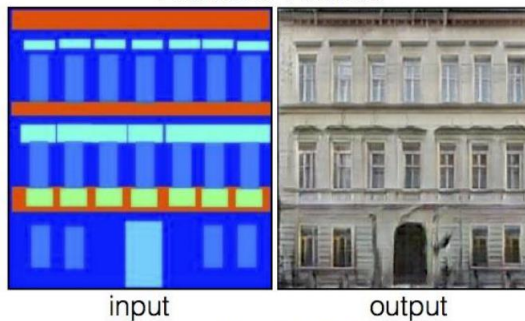




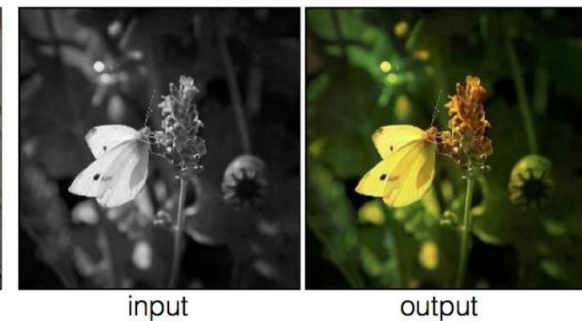
Labels to Street Scene



Labels to Facade



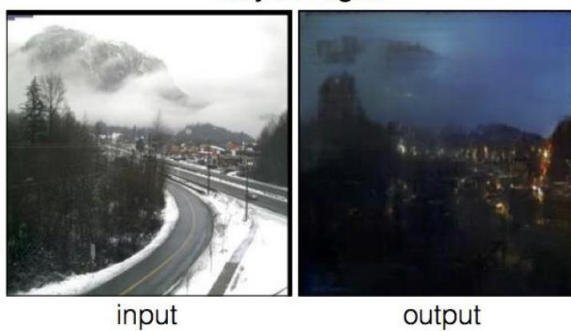
BW to Color



Aerial to Map



Day to Night



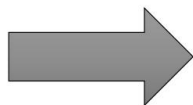
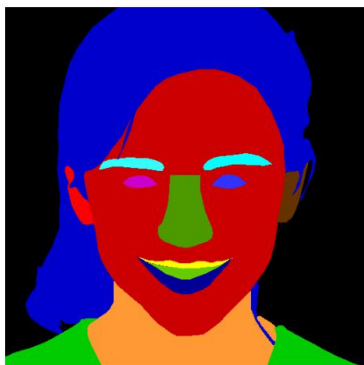
Edges to Photo



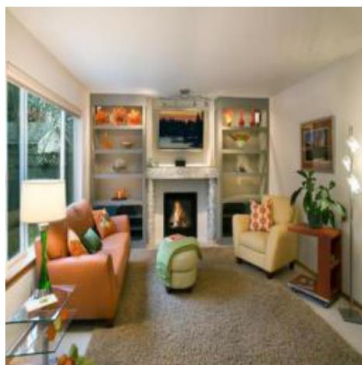
pix2pix (Isola et al.)



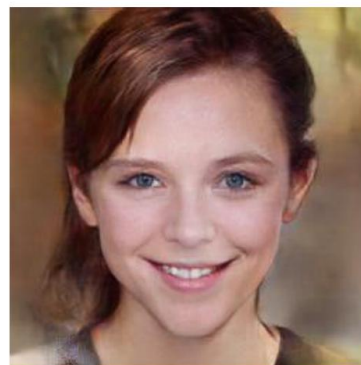
Semantic Map



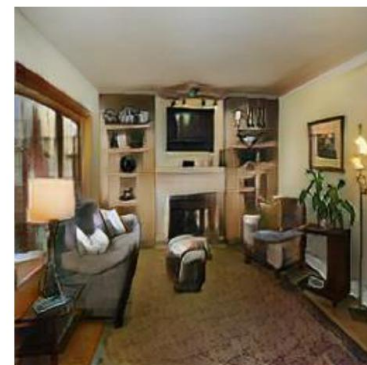
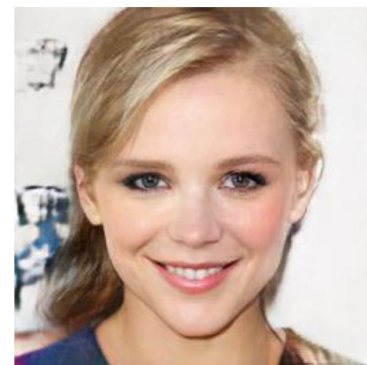
Photo



pix2pixHD



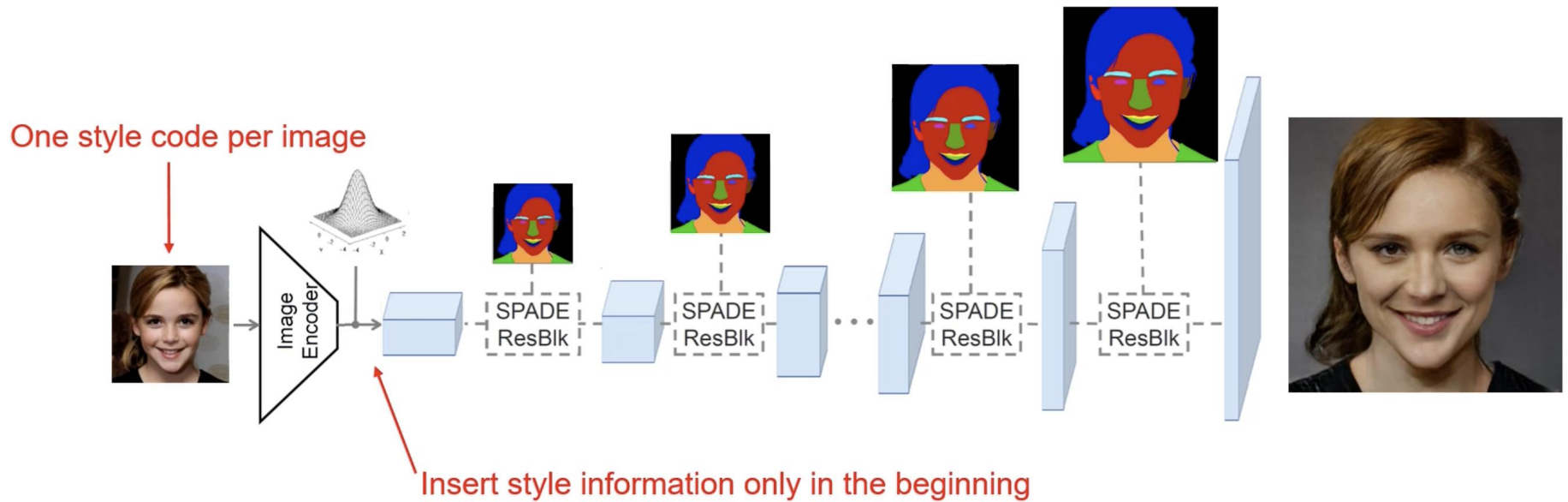
SPADE



pix2pixHD (Wang et al.) & SPADE (Park et al.)

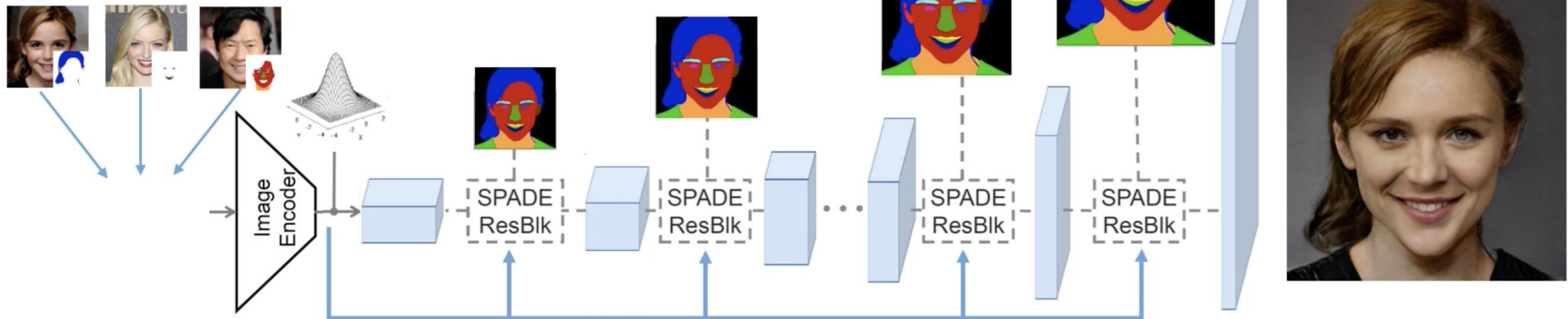
SPADE (Park et al.)

Two shortcomings

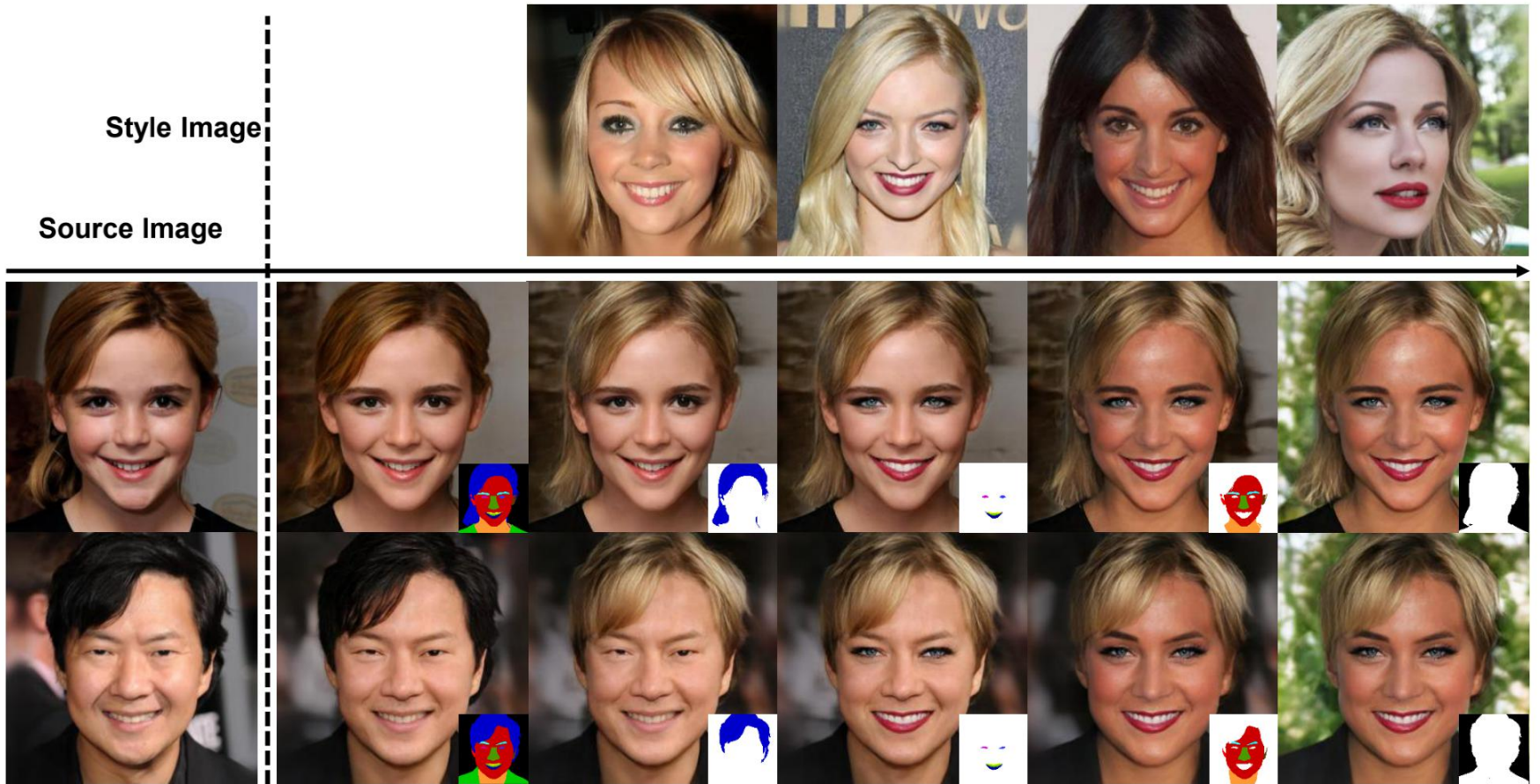


Two modifications

One style code per region



Inject style information at multiple locations



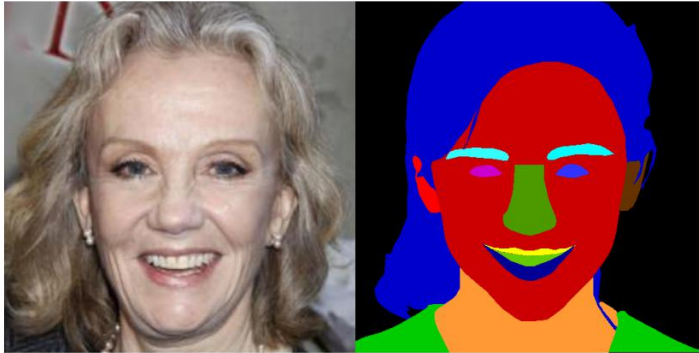


Style

Semantic Map

SPADE

Ours





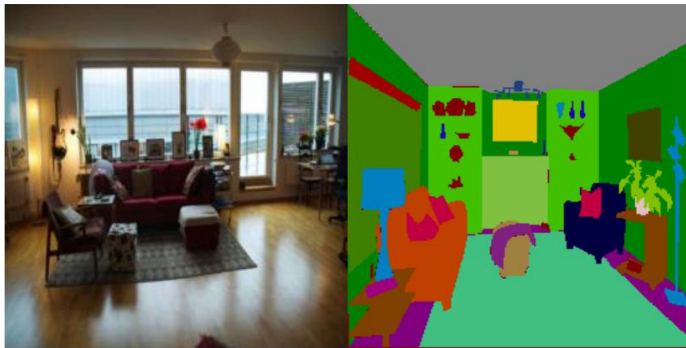
Style

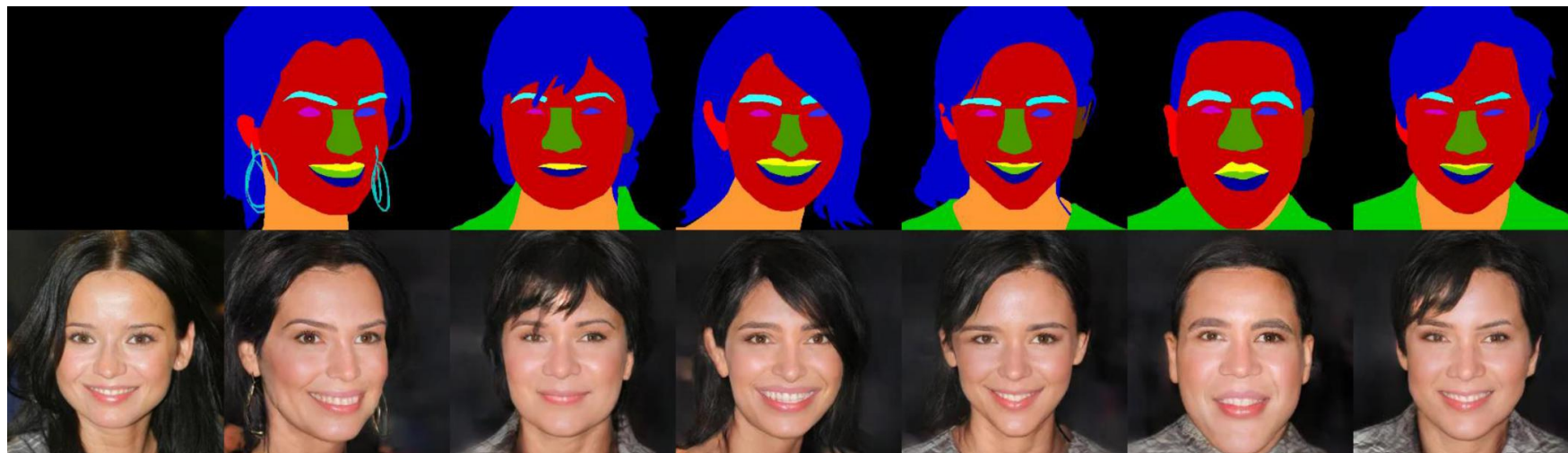
Semantic Map



SPADE

Ours





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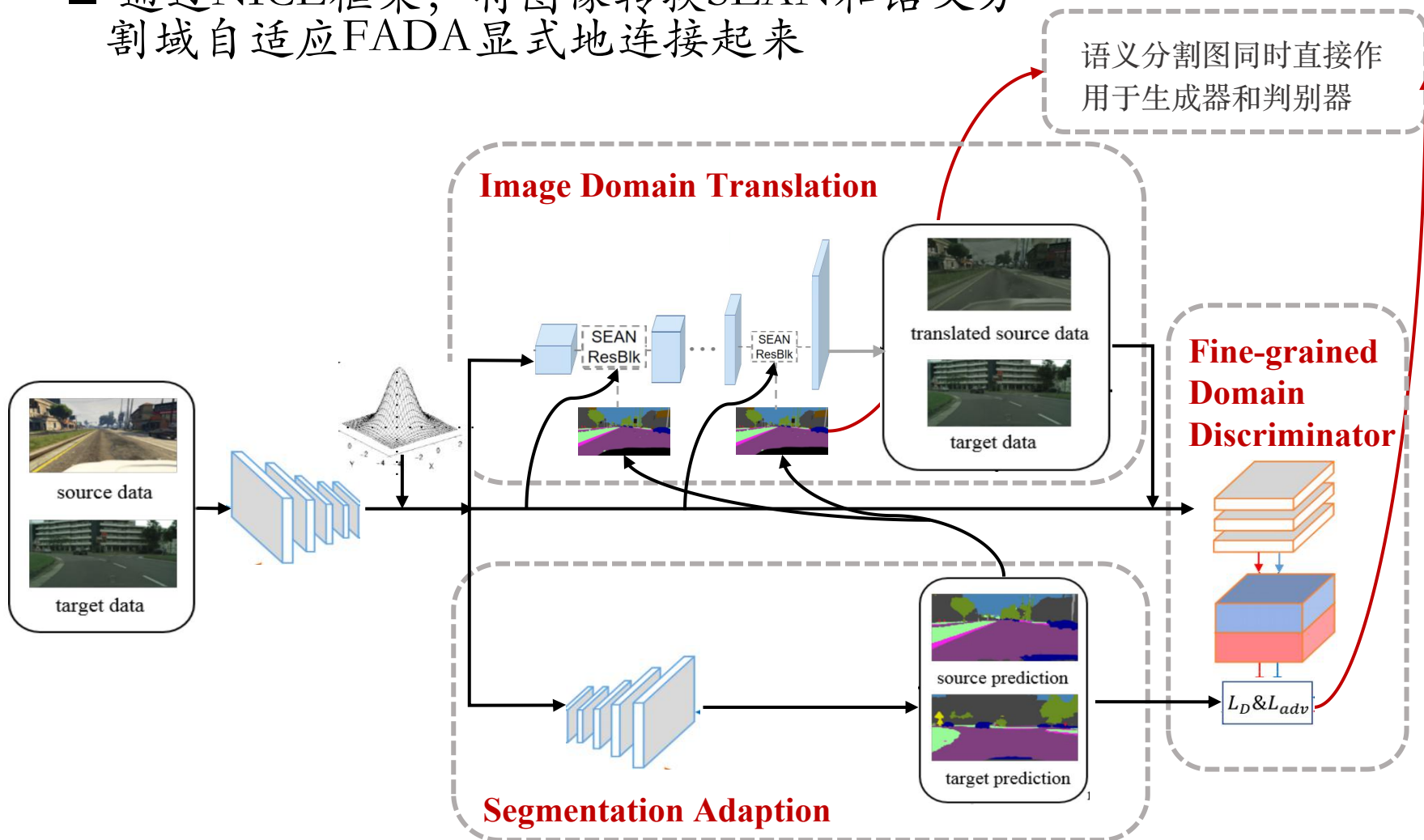
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Future Works

- 通过NICE框架，将图像转换SEAN和语义分割域自适应FADA显式地连接起来





谢谢！
请老师同学们批评指正！