



High-Resolution Image Synthesis and Semantic Manipulation with Conditional GANs

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Introduction

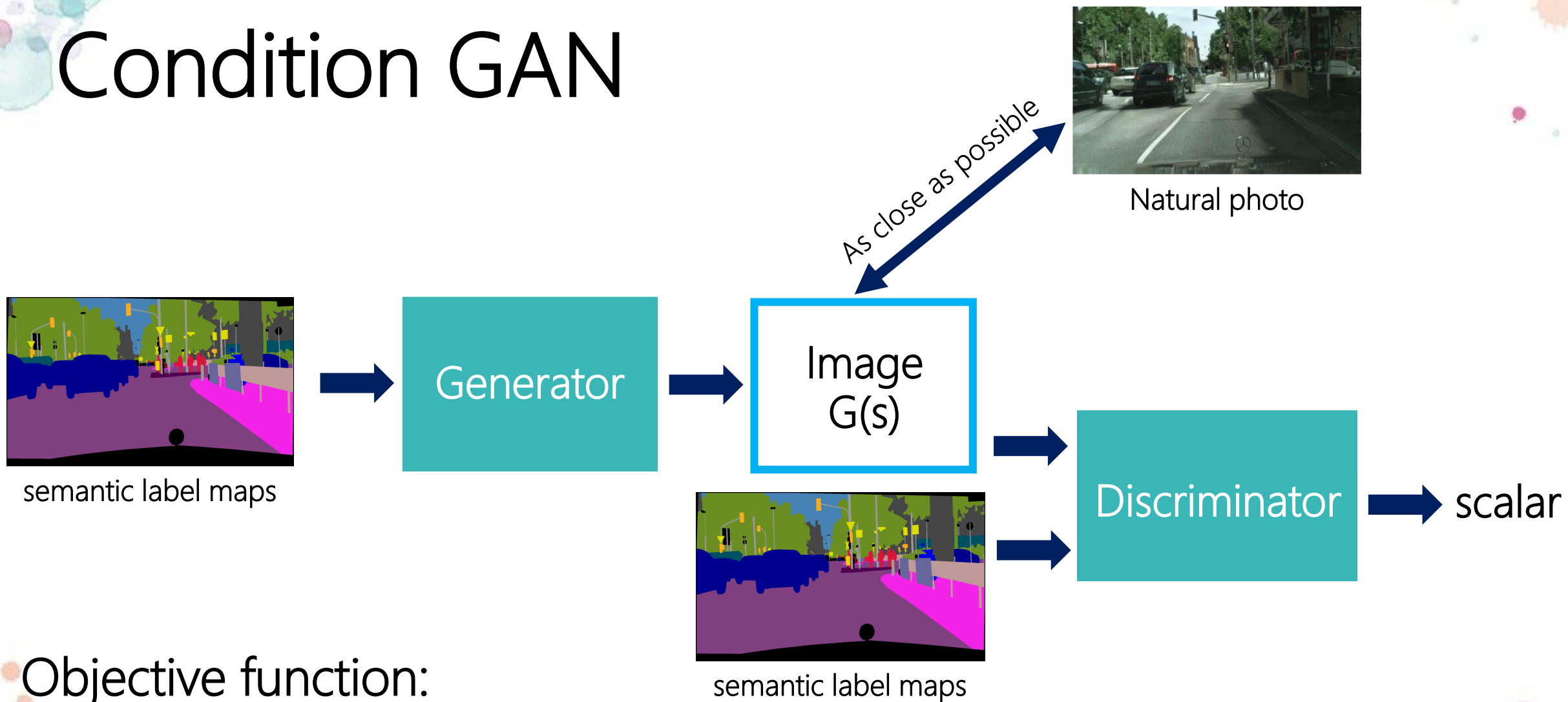


Input
semantic label maps
2048 X 1024



Output
Synthesized image
2048 X 1024

Condition GAN



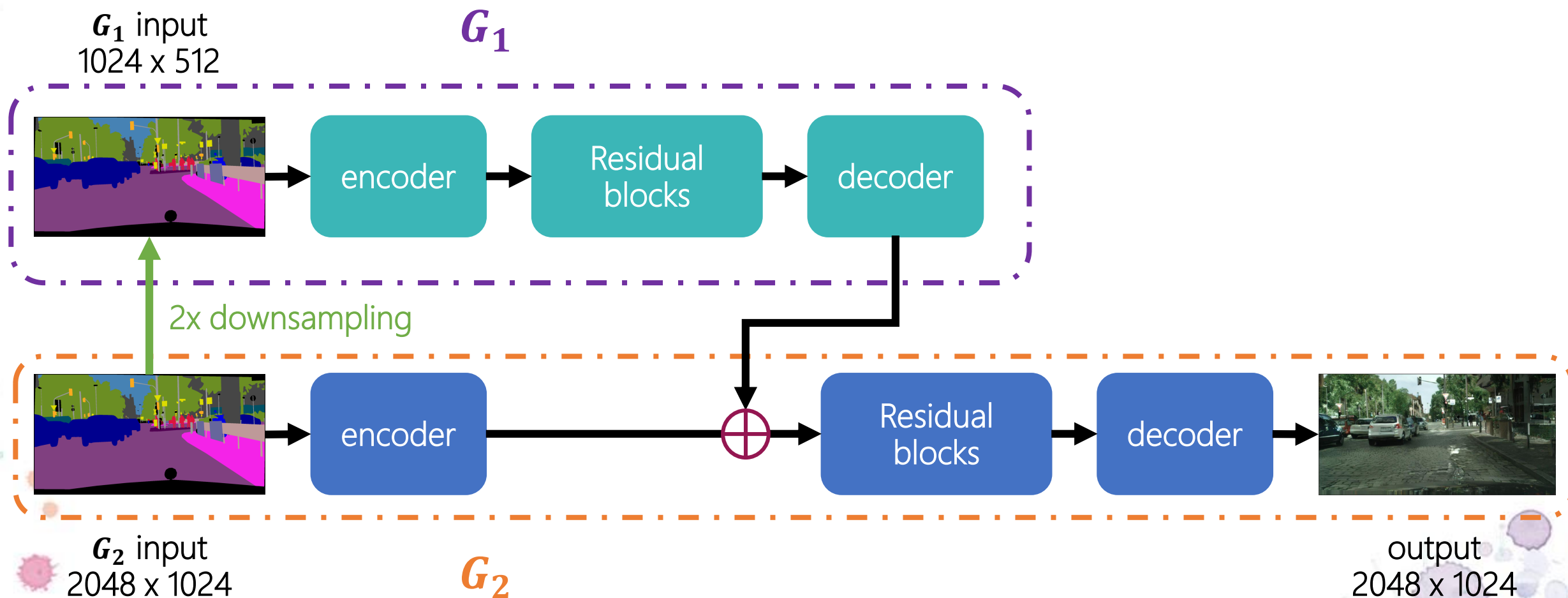
Objective function:

$$\min_G \max_D \mathcal{L}_{GAN}(G, D) = E_{(s,x)} [\log D(s, x)] + E_s [\log(1 - D(s, G(s)))]$$

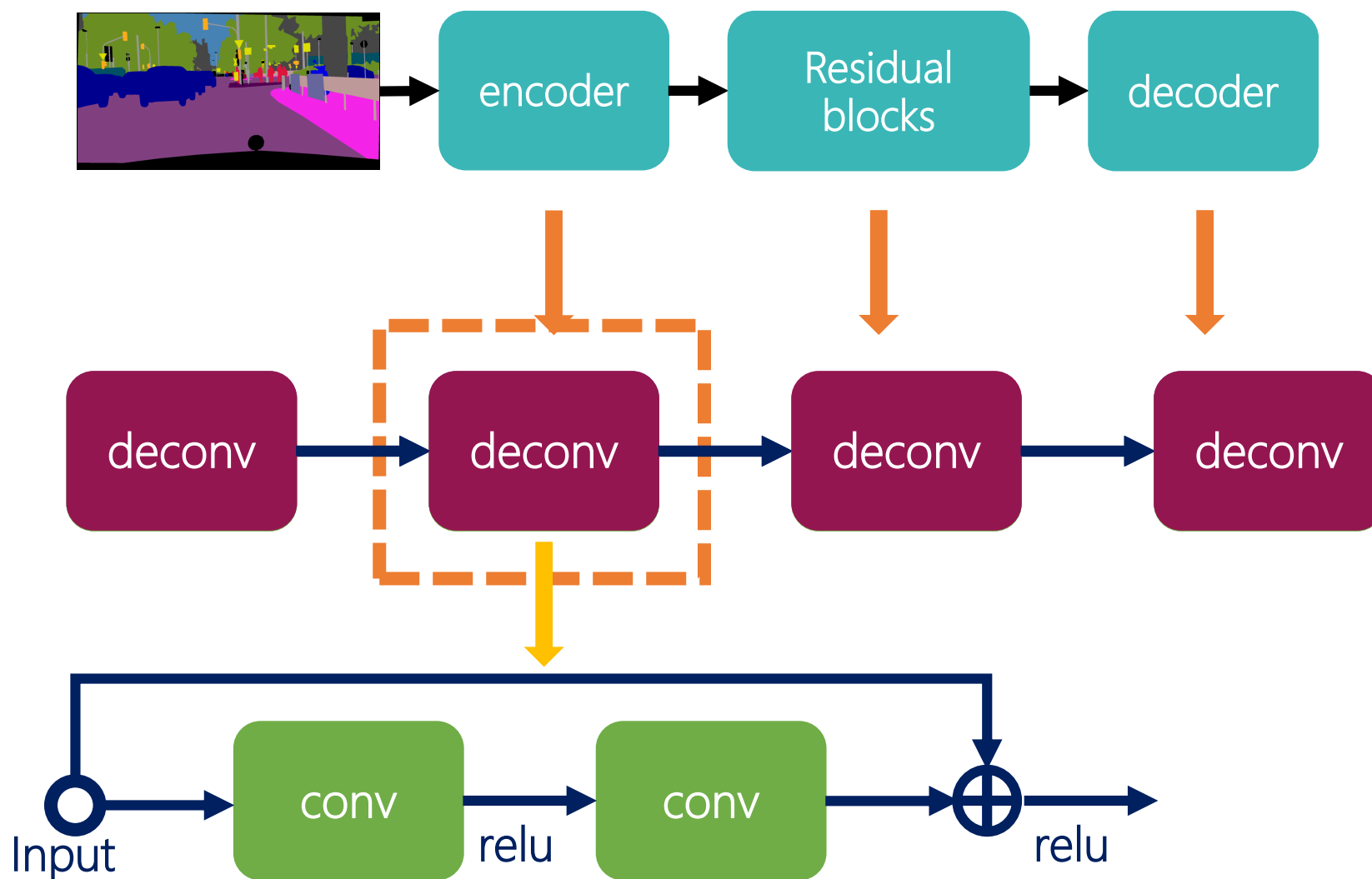
s: semantic label maps

x: natural photo

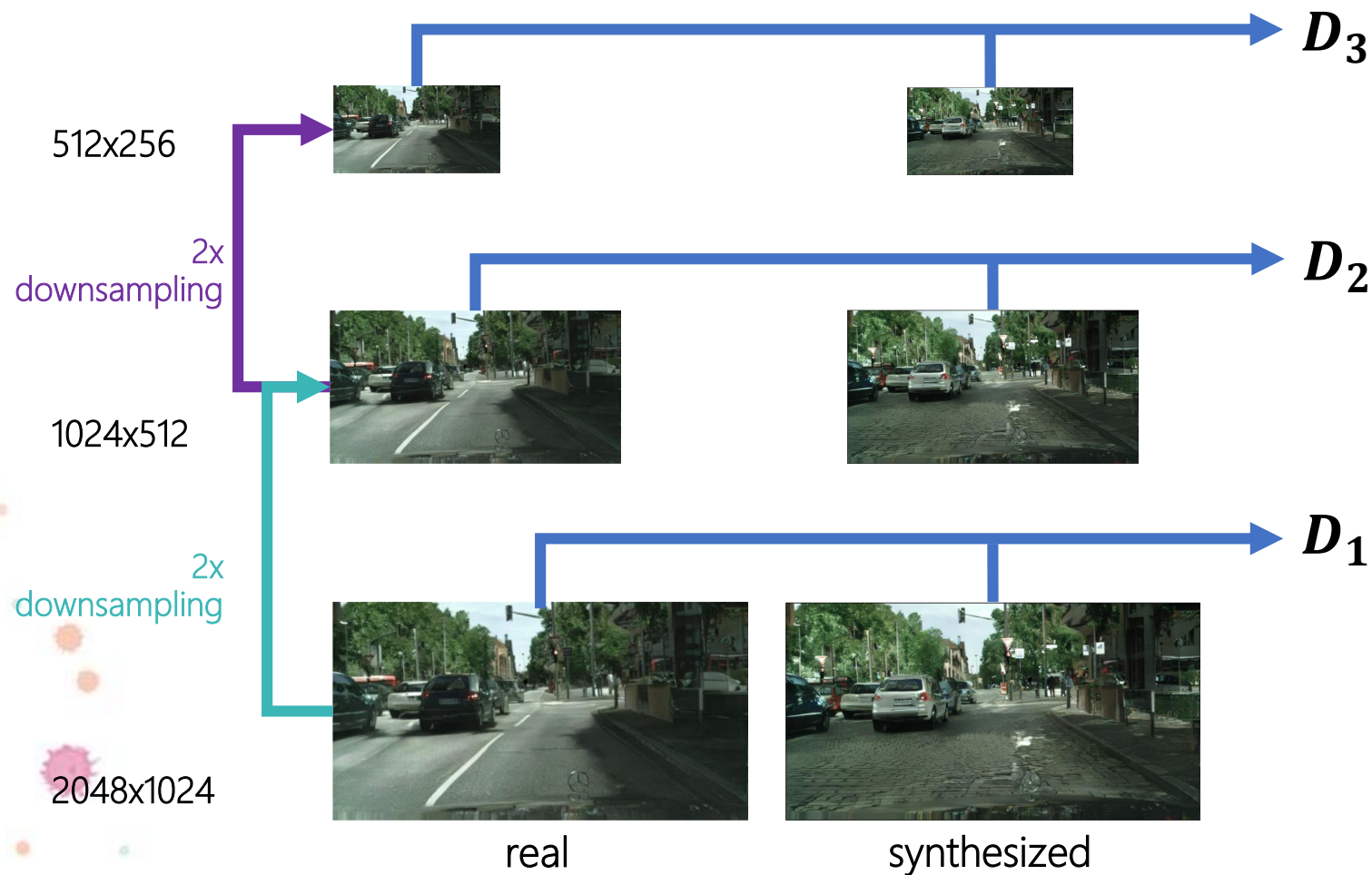
Coarse-to-fine Generator



Generator detail



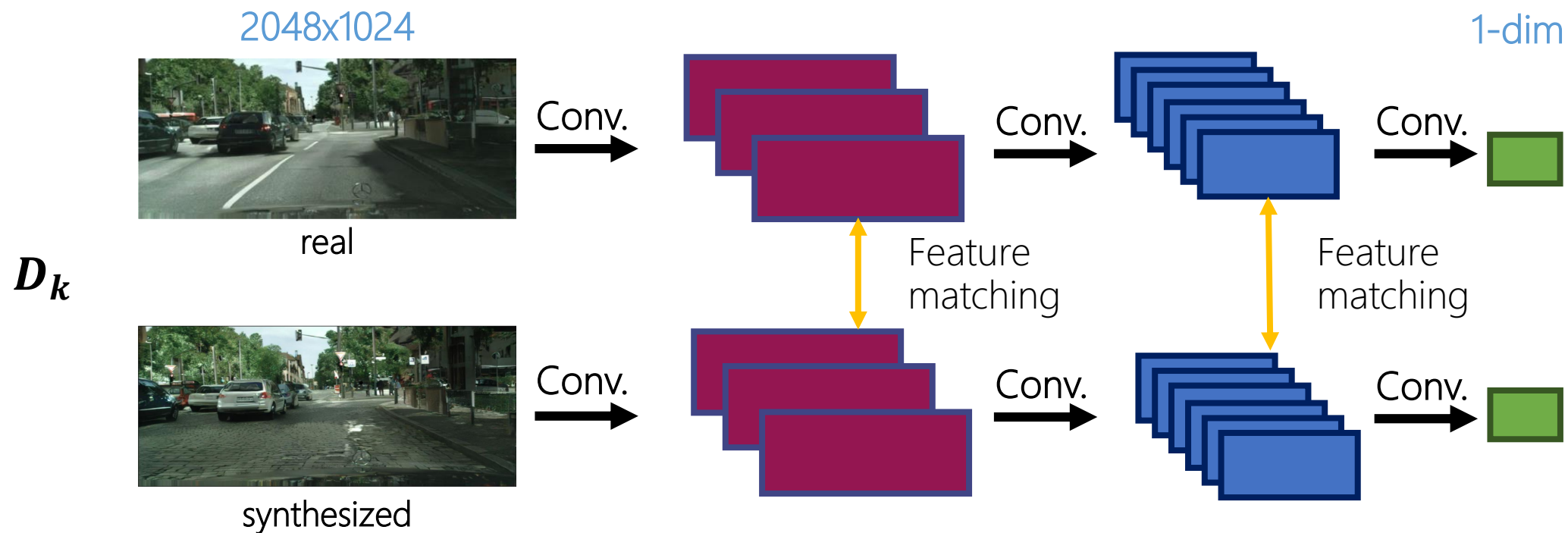
Multi-scale Discriminators



$$\begin{aligned} \mathcal{L}_{GAN}(G, D) &= E_{(s, x)} [\log D(s, x)] \\ &+ E_s [\log(1 - D(s, G(s)))] \end{aligned}$$

$$\min_G \max_{D_1, D_2, D_3} \sum_{k=1,2,3} \mathcal{L}_{GAN}(G, D_k)$$

Improved adversarial loss



$$\mathcal{L}_{FM}(G, D_k) = E_{(s,x)} \sum_{i=1}^T \frac{1}{N_i} \left[\left\| D_k^{(i)}(s, x) - D_k^{(i)}(s, G(s)) \right\|_1 \right]$$

T : total number of layer

N_i : number of elements in i - th layer

$D_k^{(i)}$: i th layer feature extractor of D_k

Full objective

Coarse-to-fine
Generator

Multi-scale
Discriminators

Improved
adversarial loss



$$\mathcal{L}_{GAN}(G, D) = E_{(s,x)}[\log D(s, x)] + E_s[\log(1 - D(s, G(s)))]$$

$$\mathcal{L}_{FM}(G, D_k) = E_{(s,x)} \sum_{i=1}^T \frac{1}{N_i} \left[\left\| D_k^{(i)}(s, x) - D_k^{(i)}(s, G(s)) \right\|_1 \right]$$

$$\min_G \left(\max_{D_1, D_2, D_3} \sum_{k=1,2,3} \mathcal{L}_{GAN}(G, D_k) \right) + \lambda \sum_{k=1,2,3} \mathcal{L}_{FM}(G, D_k)$$