



Changing Background to Foreground: an Augmentation Method Based on Conditional Generative Network for Stingray Detection



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Introduction

Objective

Improving the performance of automatic stingray detection in the aerial images captured by an UAV to help ecological researchers in counting the number of stingray.

Problems

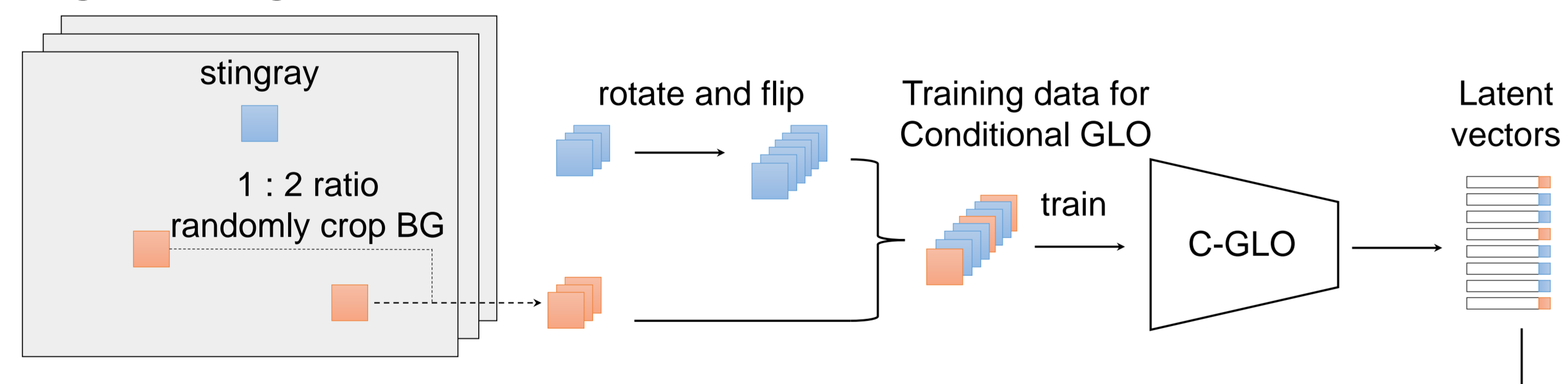
- A wide range aerial imagery (high resolution) limits the image batch number used for training in a deep learning object detection model (Faster R-CNN in this paper) due to memory issue.
- Stingray is always sparse in each image. It limits the batch number of the object for training.
- Using limited amount of training images results in worse detection results.

Solution

We propose a new data augmented method, called Conditional Generative Latent Optimization (C-GLO), to generate sufficient amount of stingray samples for reaching better object detection on marine aerial images.

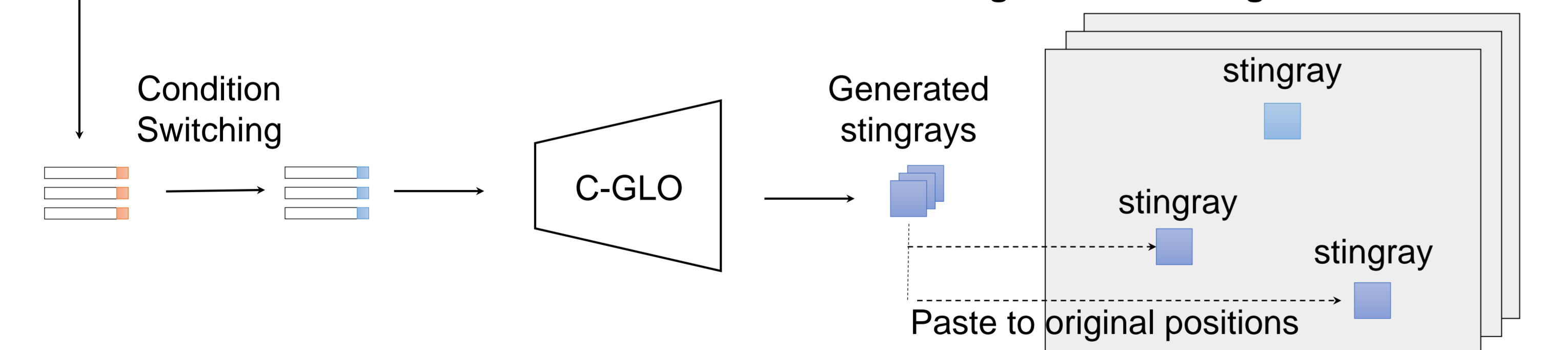
Method Overview

Original training data for detection



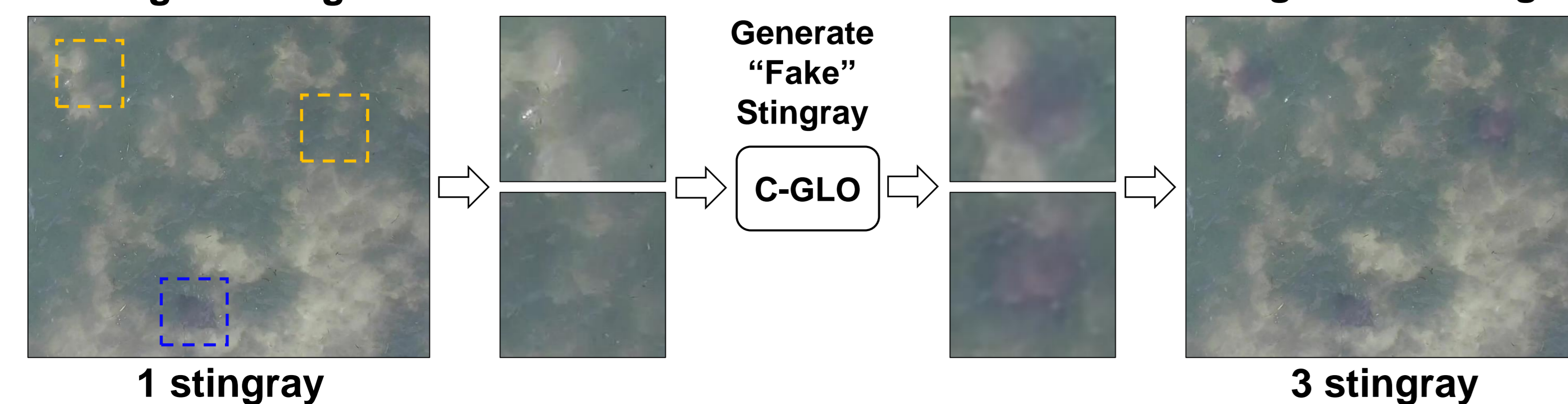
Remove FG vectors

Augmented training data for detection

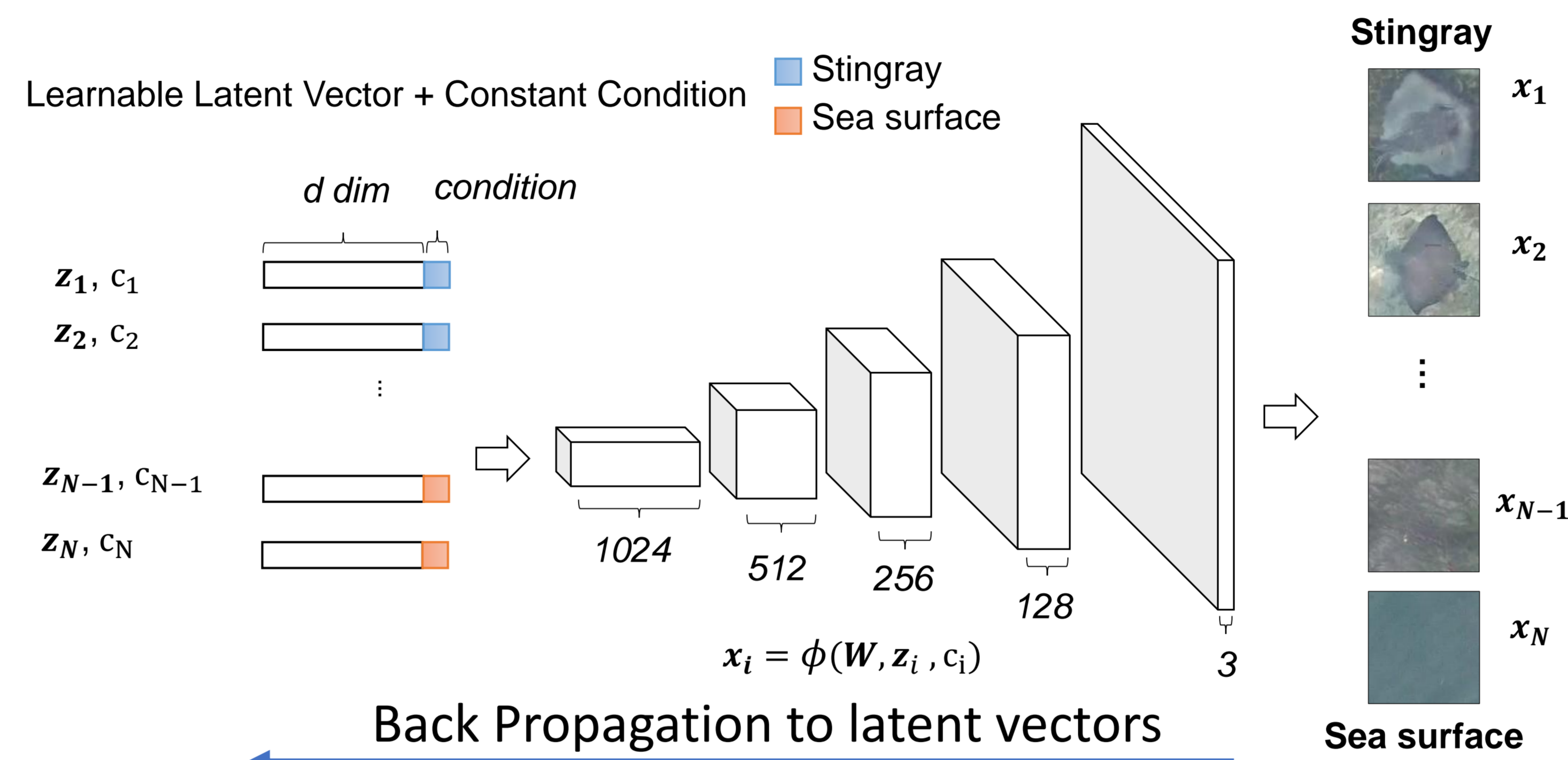


Original Image

Augmented Image



The Architecture of C-GLO



Given unsupervised training images $I = \{I_1, \dots, I_N\}$, C-GLO trains a generator ϕ (with the input z, c and network weights W), such that the following objective is minimized:

$$e(W, z, c) = \sum_{i=1}^N \text{loss}(\phi(W, z_i, c_i) - I_i)$$

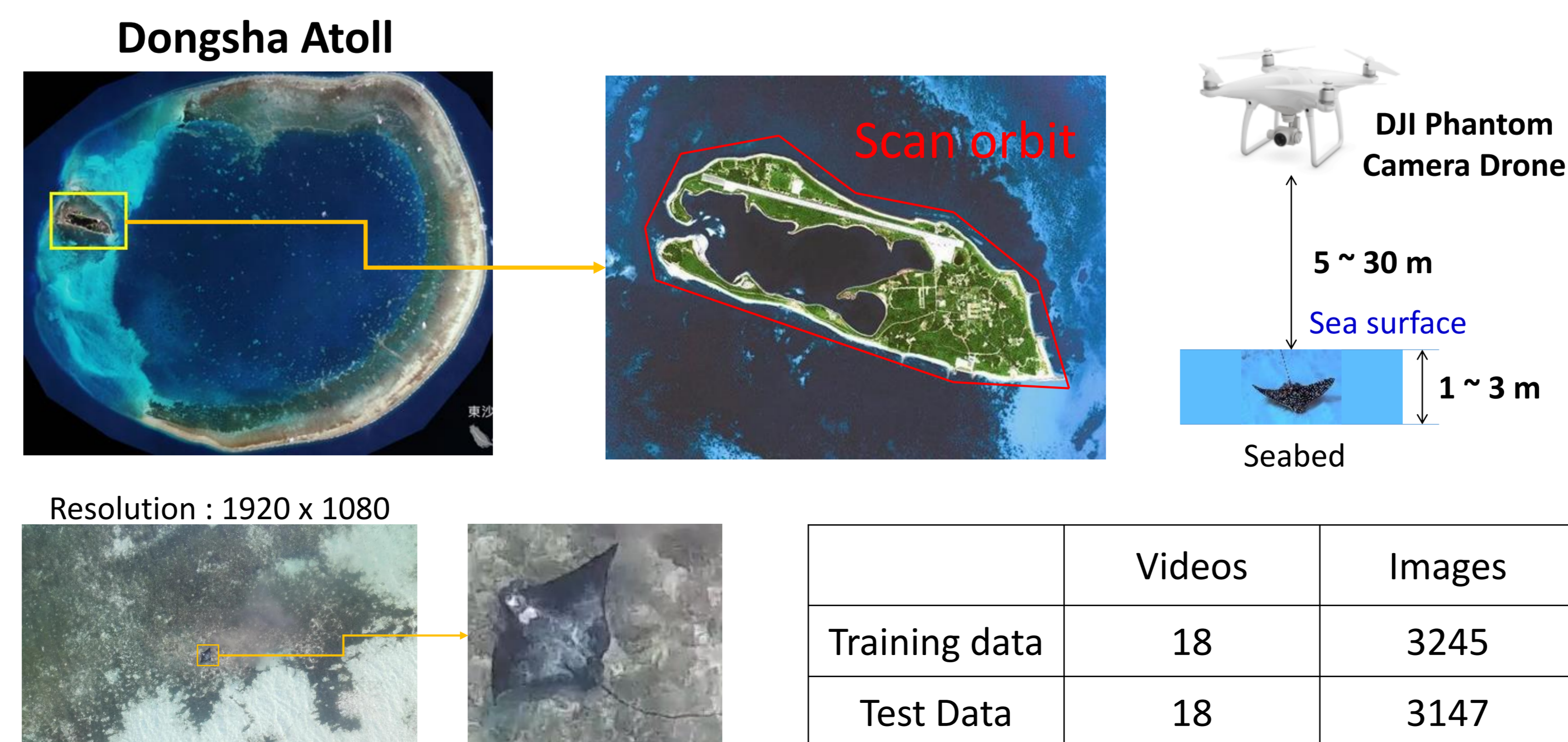
where $z = \{z_1, z_2, \dots, z_N\}$. $c = 0$ indicates sea surface, and $c = 1$ indicates stingray.

The training process of C-GLO is:

1. Given z, c , find W to reduce the total reconstruction loss of I .
2. Given W, c_i , find z_i to reduce the reconstruction loss of $I_i, \forall I$.

The above two steps are executed iteratively.

Data Description



Results & Comparison

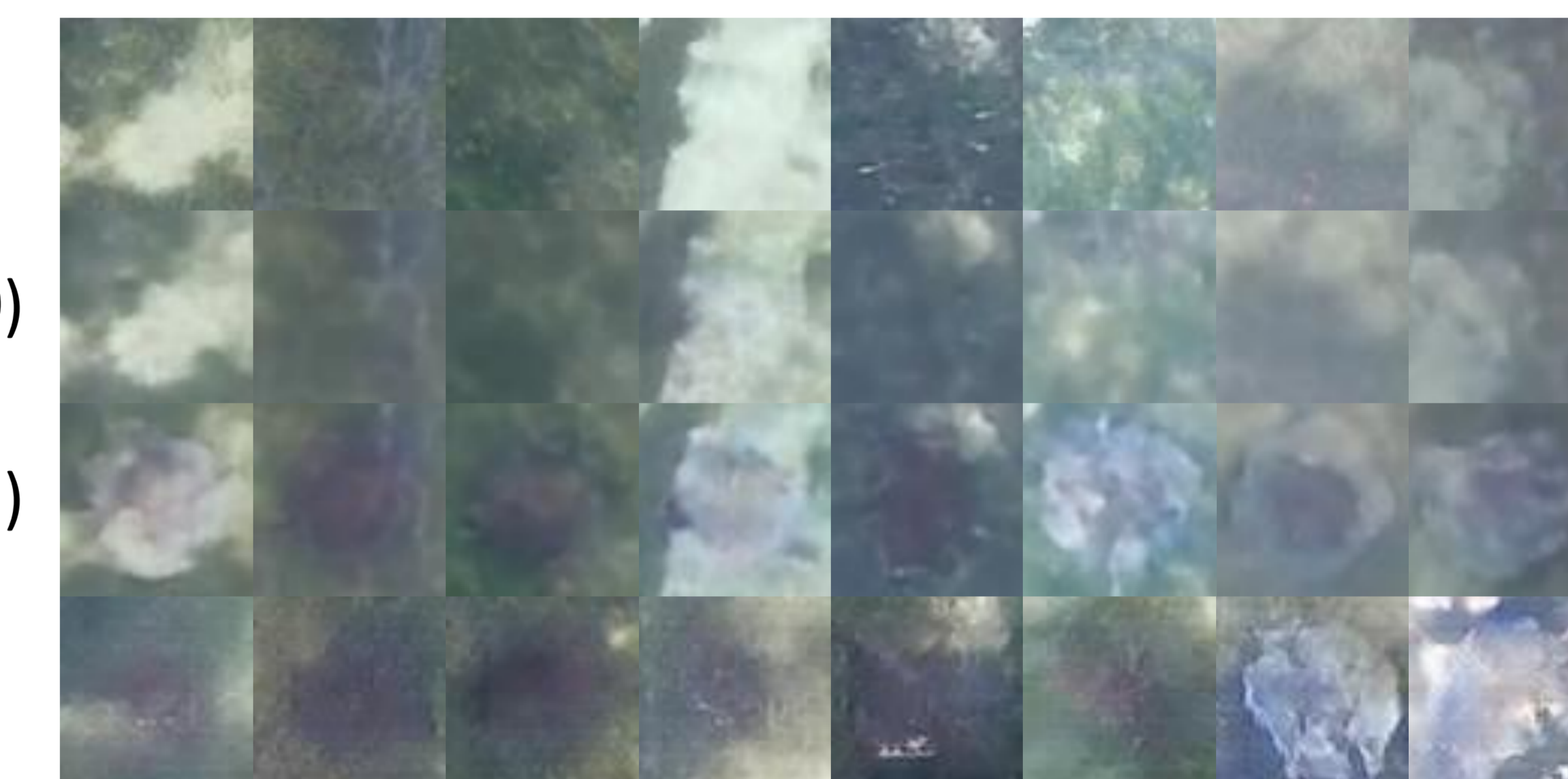
Generated Images

Input

C-GLO ($c = 0$)

C-GLO ($c = 1$)

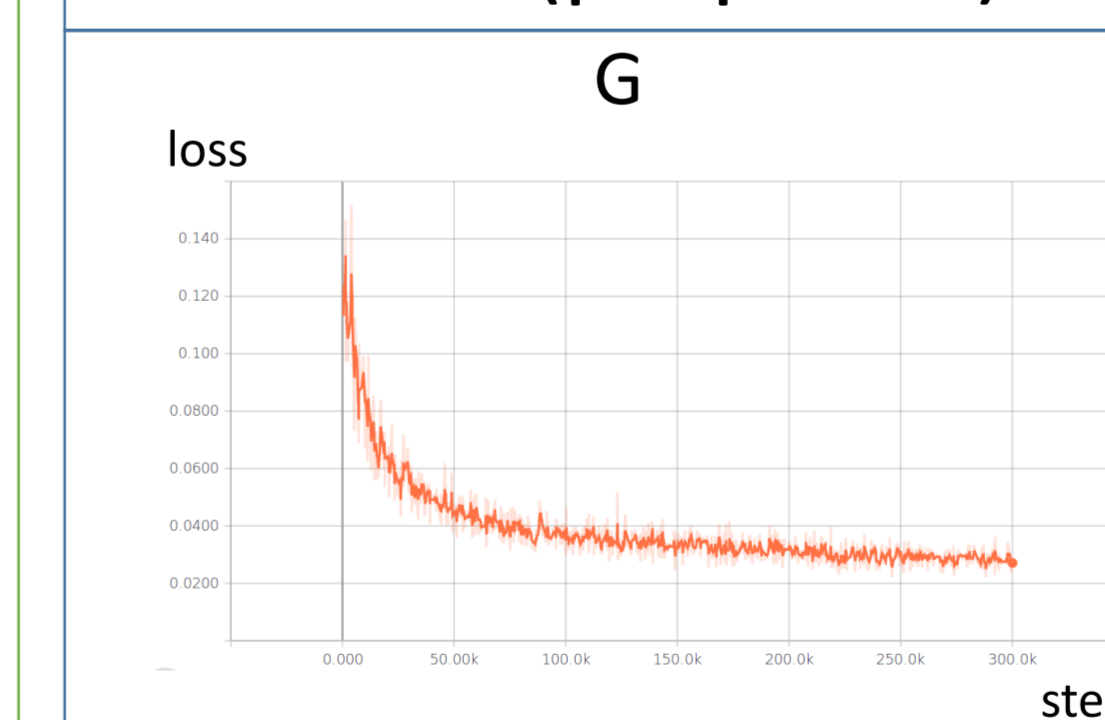
ICGAN



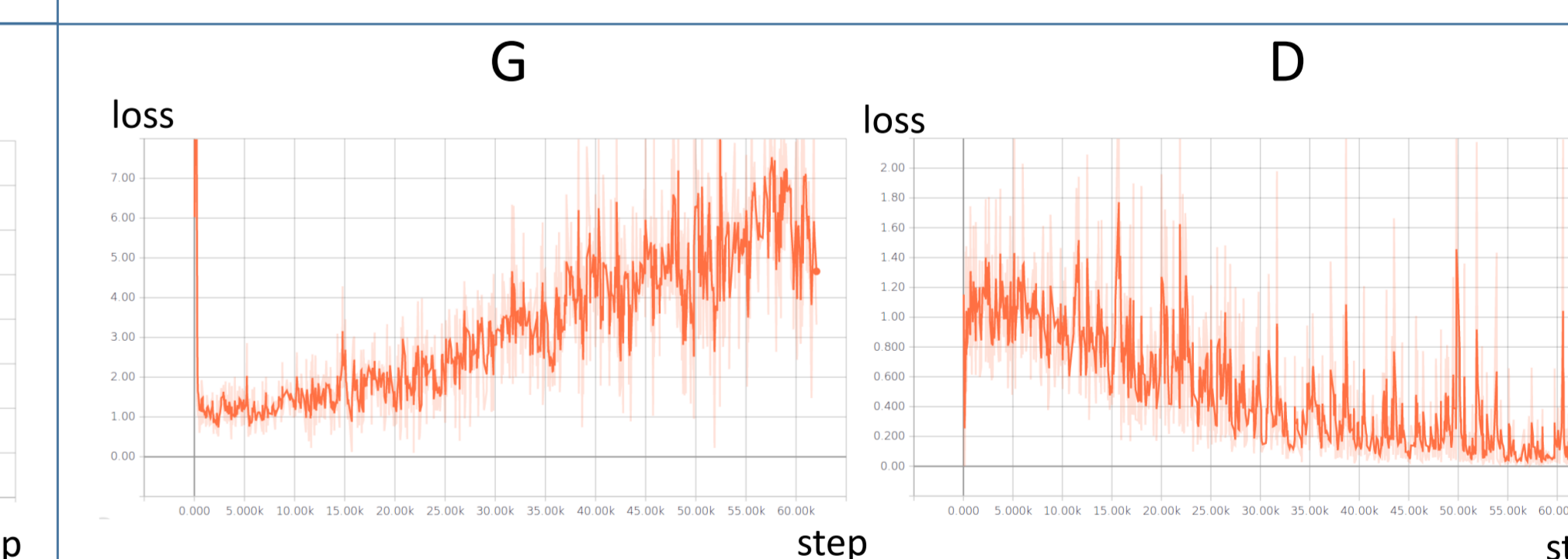
Comparison with Conditional GAN method

	Model Structure			Property	
	Encoder	Generator	Discriminator	Memory latent vectors	Invertible
C-GLO (proposed)	×	✓	×	✓	✓
C-GAN	×	✓	✓	×	×
ICGAN	✓	✓	✓	×	✓

C-GLO (proposed)



C-GAN



Faster R-CNN Detection Performance (AP)

Network	Baseline	Ours-128	Ours-256	Ours-512
ZF	78.89	82.75	82.42	83.04
VGG-16	84.59	86.14	86.61	86.43

Conclusion

- We introduce a data augmentation method, C-GLO to fuse background patches and foreground objects.
- Experimental results reveal that the detection performance can be improved our approach.
- C-GLO performs more stable than GAN-based methods do in convergence of training.