

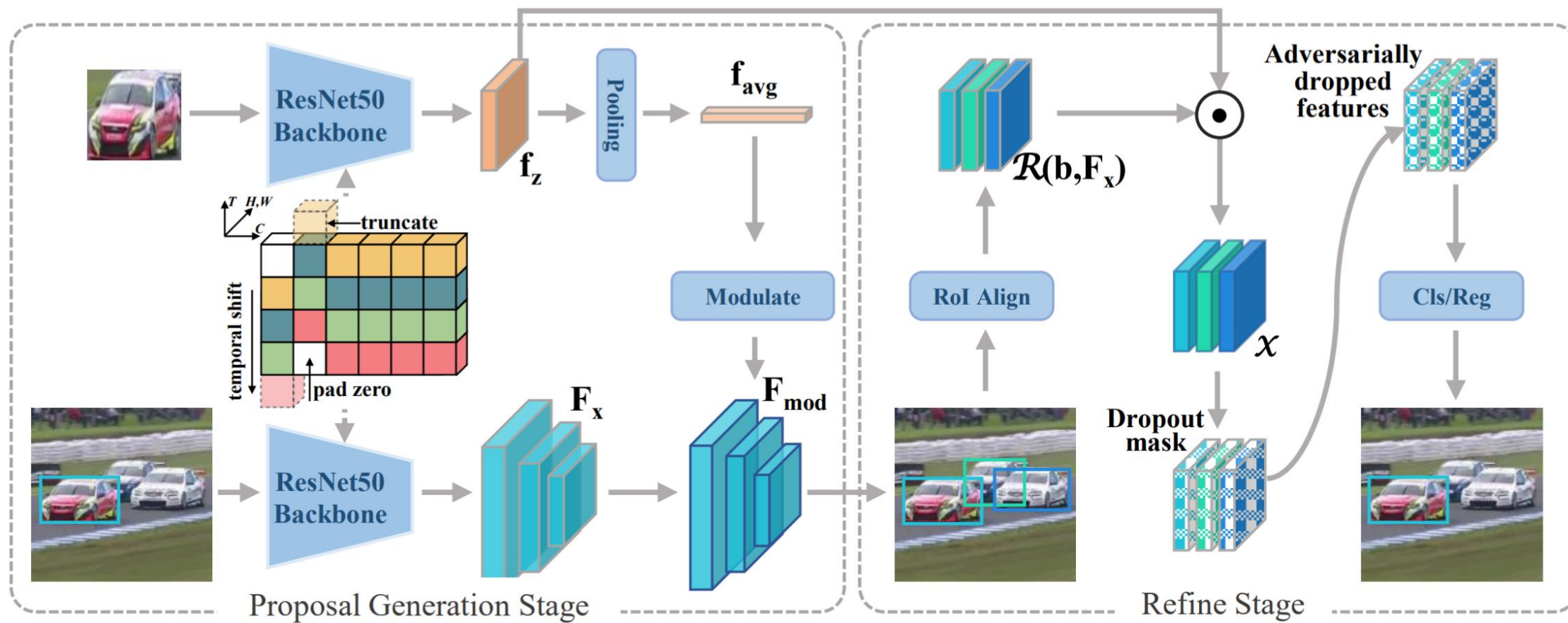
# END-TO-END TEMPORAL FEATURE AGGREGATION FOR SIAMESE TRACKERS

Zhenbang Li, Qiang Wang, Jin Gao, Bing Li, Weiming Hu

National Laboratory of Pattern Recognition, Institute of Automation, Chinese Academy of Sciences  
School of Artificial Intelligence, University of Chinese Academy of Sciences

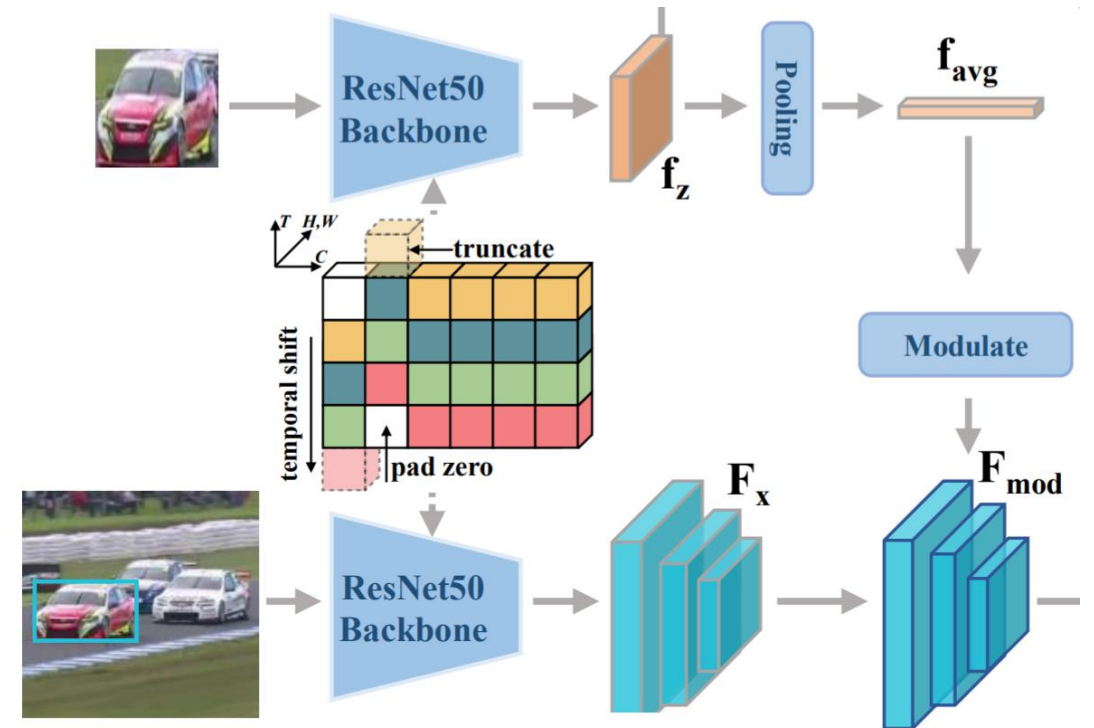


# Introduction



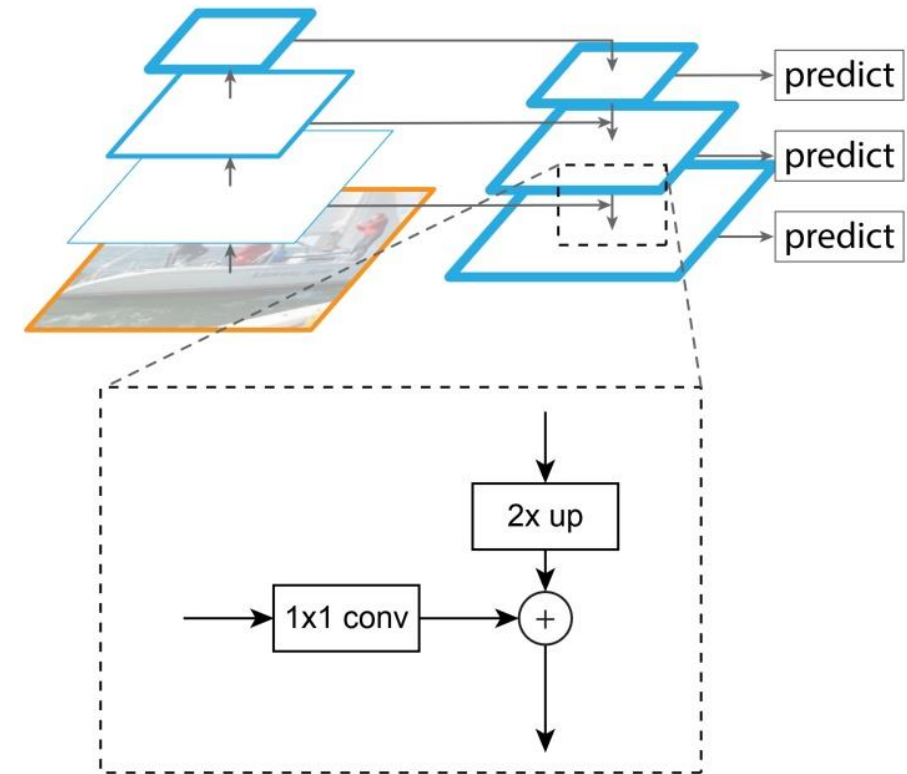
# Proposal Generation Stage

- The proposal generation stage consists of 3 components:
  - (1) feature extractor
  - (2) temporal aggregation module
  - (3) feature modulation module



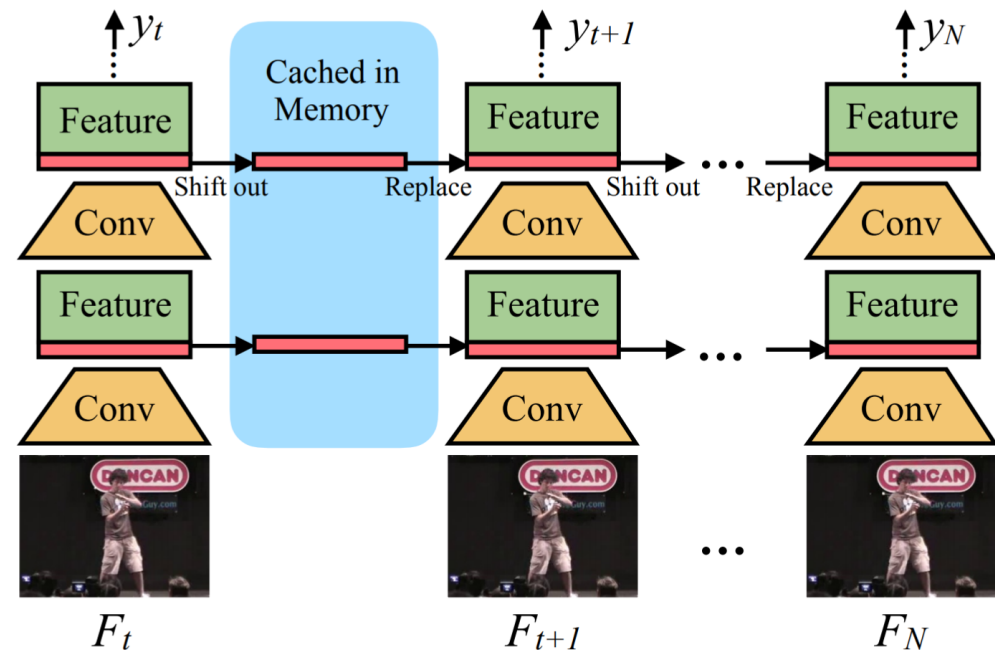
# Feature extractor

- Res50-FPN
- Template feature:  $f_z$
- Search feature pyramid:  $F_x = \{f_x^i\}_{i=1:5}$



# Temporal aggregation module

$$\begin{array}{l}
 \left. \begin{array}{l}
 f^t \in \mathbb{R}^{C \times H \times W} \\
 f_{1:K}^t \in \mathbb{R}^{K \times H \times W} \\
 f_{(K+1):2K}^t \in \mathbb{R}^{\tilde{K} \times H \times W} \\
 f_{(2K+1):C}^t \in \mathbb{R}^{(C-2K) \times H \times W}
 \end{array} \right\} \text{split} \\
 \\
 \left. \begin{array}{l}
 f_{agg}^t = \mathcal{C}(f_{1:K}^{t-1}, f_{(K+1):2K}^{t+1}, f_{(2K+1):C}^t)
 \end{array} \right\} \text{aggregate}
 \end{array}$$



# Refine Stage

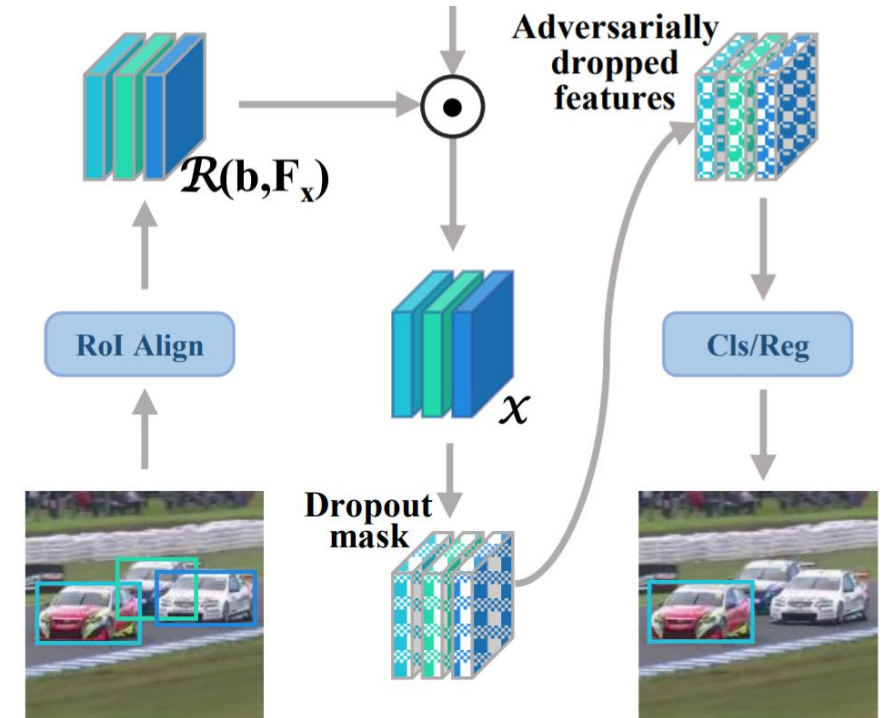
Generate the adversarial dropout mask:

$$\mathbf{m}^{adv} = \arg \max_{\mathbf{m}} D[h^{cls}(\mathcal{X} \odot \mathbf{m}^s), h^{cls}(\mathcal{X} \odot \mathbf{m})]$$

where  $\|\mathbf{m}^s - \mathbf{m}\| \leq \delta_e L$ ,

Minimize the divergence between two predicted distribution:

$$\mathcal{L}_{adv} = \mathbb{E}[D_{KL}[h^{cls}(\mathcal{X} \odot \mathbf{m}^s) || h^{cls}(\mathcal{X} \odot \mathbf{m}^{adv})]]]$$



# Ablation Studies

**Table 1.** Performance of our algorithm with different components on GOT-10k test set.

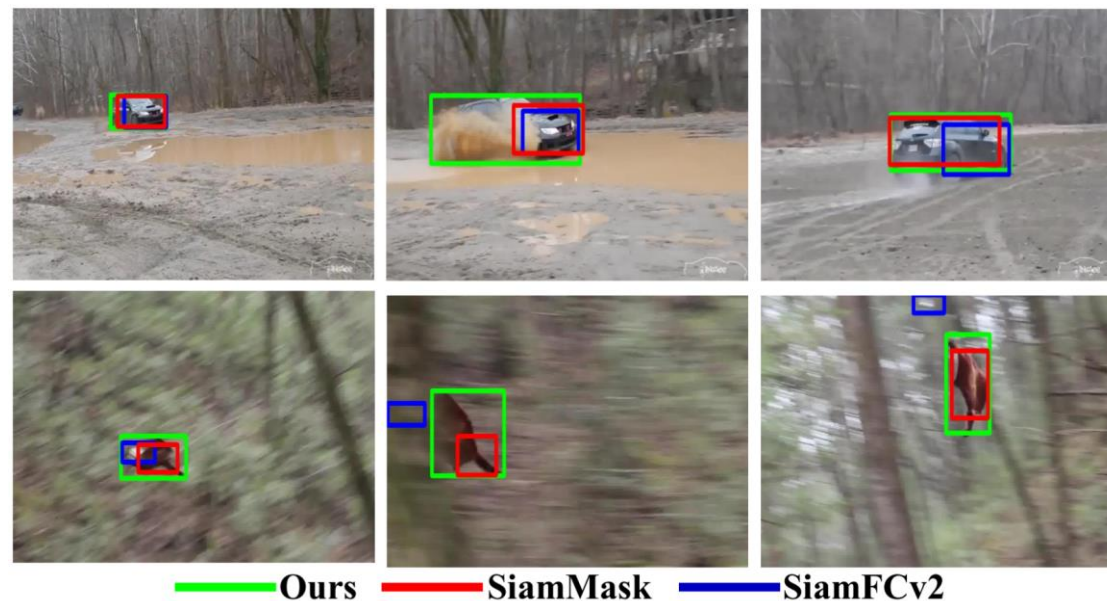
Temporal aggregation	Adversarial dropout	$AO$	$SR_{0.50}$	$SR_{0.75}$
		0.542	0.607	0.456
✓		0.561	0.645	0.480
✓	✓	0.577	0.662	0.509



# Evaluation on GOT-10k Dataset

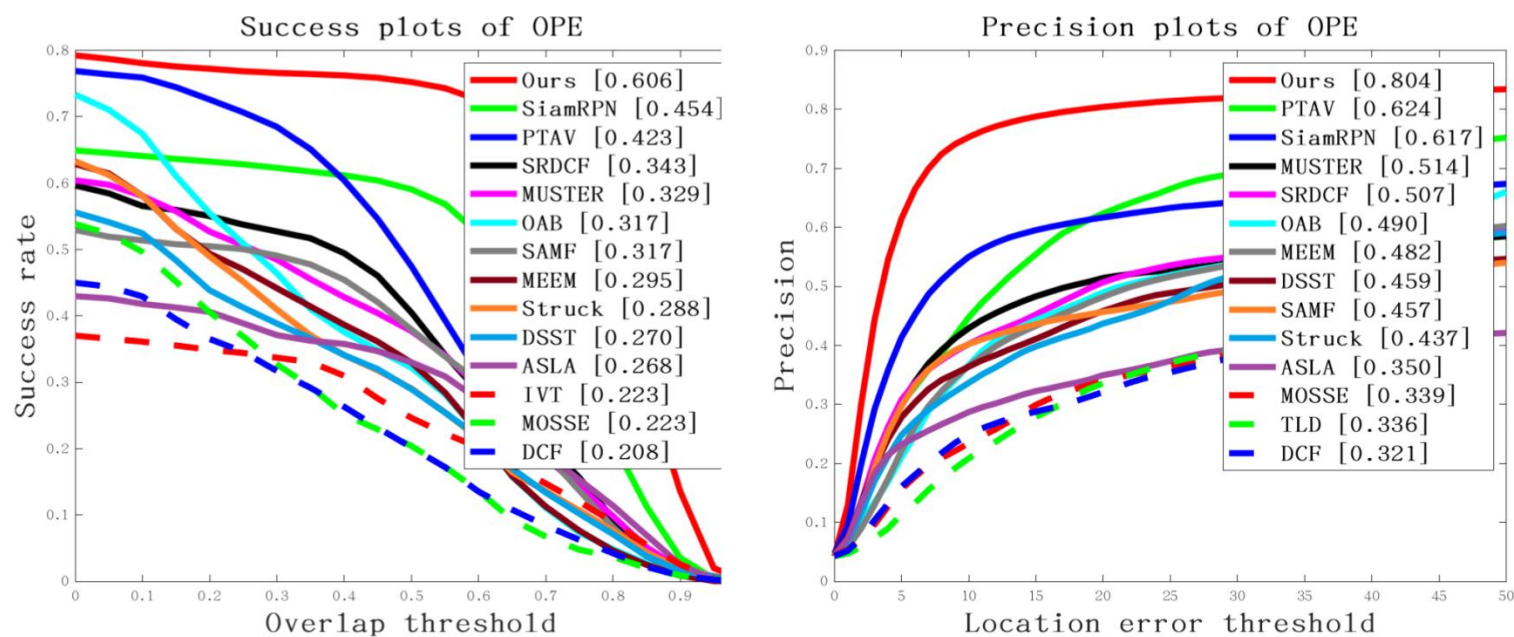
**Table 2.** Comparing the results of our approach against other approaches over the GOT-10k test set.

Method	$AO$	$SR_{0.50}$	$SR_{0.75}$
Ours	<b>0.577<sup>1</sup></b>	<b>0.662<sup>1</sup></b>	<b>0.509<sup>1</sup></b>
SiamMask	0.459	0.560	0.205
SiamFCv2	0.374	0.404	0.144
SiamFC	0.348	0.353	0.098
GOTURN	0.347	0.375	0.124
CCOT	0.325	0.328	0.107
ECO	0.316	0.309	0.111
CF2	0.315	0.297	0.088
MDNet	0.299	0.303	0.099





# Evaluation on UAV20L Dataset



**Fig. 3.** Success and precision plots on UAV20L dataset.