

# Exploiting Distilled Learning for Deep Siamese Tracking

Chengxin Liu<sup>1</sup>, Zhiguo Cao<sup>1</sup>, Wei Li<sup>2</sup>, Yang Xiao<sup>1</sup>, Shuaiyuan Du<sup>1</sup> and Angfan Zhu<sup>1</sup>

<sup>1</sup>Huazhong University of Science and Technology <sup>2</sup>Queen Mary University of London

## Motivation

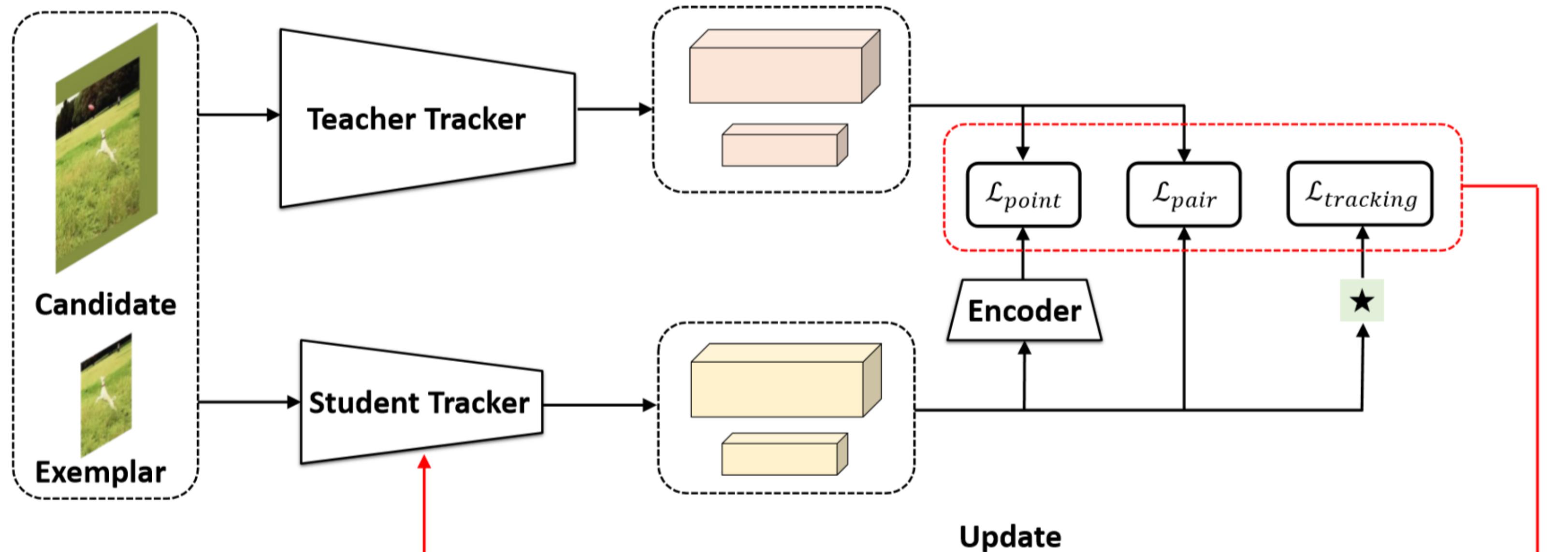
### ➤ Existing deep Siamese trackers :

- huge power consumption
- high memory usage

### ➤ Our aim: Learning tracking model with

- efficiency
- high accuracy

## Method



An overview of our proposed Distilled Learning Framework (DLF).

During training, student tracker is optimized under the guidance of teacher tracker.

## Formulation

### ➤ Point-wise Distillation

$$\mathcal{L}_{point} = \|\mathbf{F}_T - E(\mathbf{F}_S)\|_2$$

where  $\mathbf{F}_T$  and  $\mathbf{F}_S$  are teacher and student features

### ➤ Pair-wise Distillation

$$S_{\mathbf{F}}^{ij} = \frac{1}{h \times w} \cdot \frac{\mathbf{F}_i}{\|\mathbf{F}_i\|_2} \cdot \frac{\mathbf{F}_j}{\|\mathbf{F}_j\|_2} \quad \mathcal{L}_{pair} = \sum_i \sum_j \|S_{\mathbf{F}_S}^{ij} - S_{\mathbf{F}_T}^{ij}\|^2$$

where  $S \in \mathbb{R}^{hw \times hw}$  is the pair-wise similarity matrix

### ➤ Optimization

$$\mathcal{L} = \mathcal{L}_{tracking} + \lambda_0 \mathcal{L}_{point} + \lambda_1 \mathcal{L}_{pair}$$

## Ablation Study

Teacher: SiamDW

Student: SiamDW-L & SiamDW-S

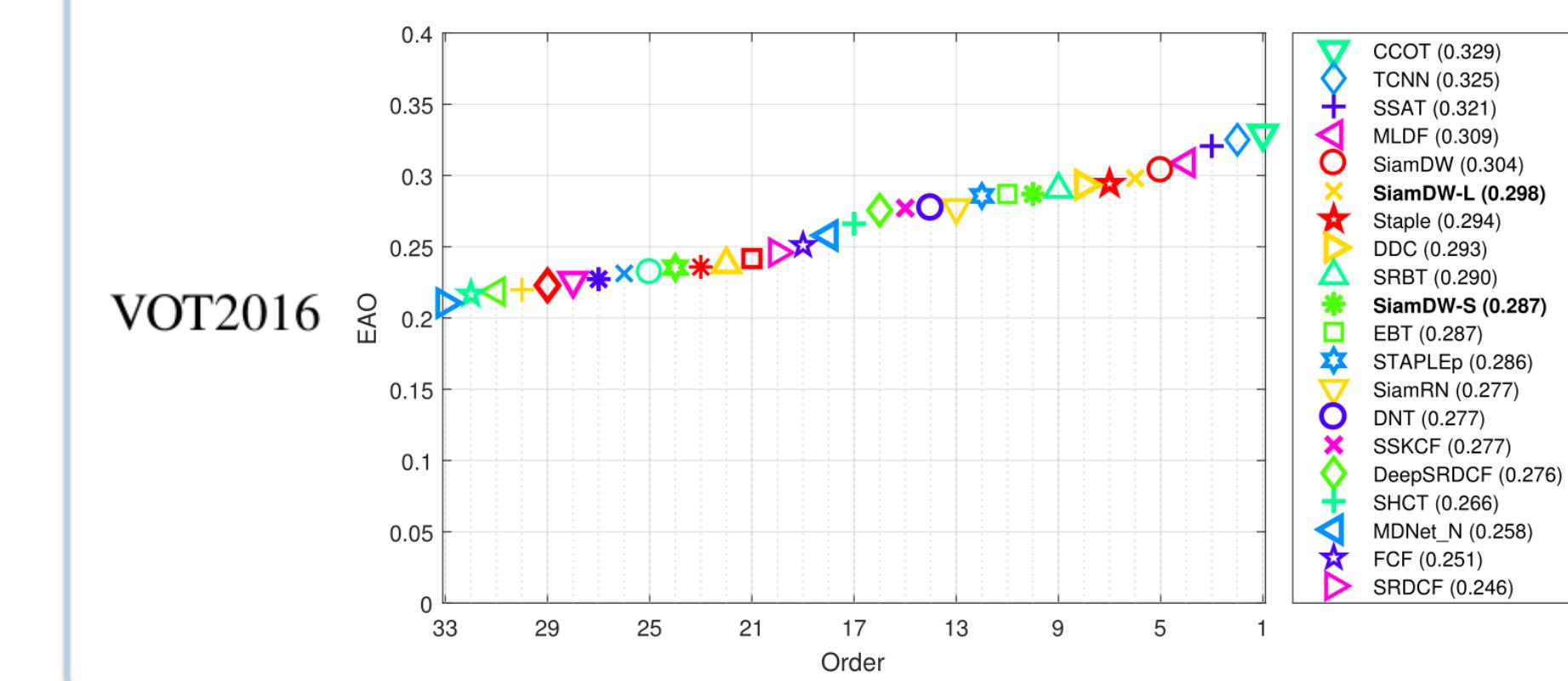
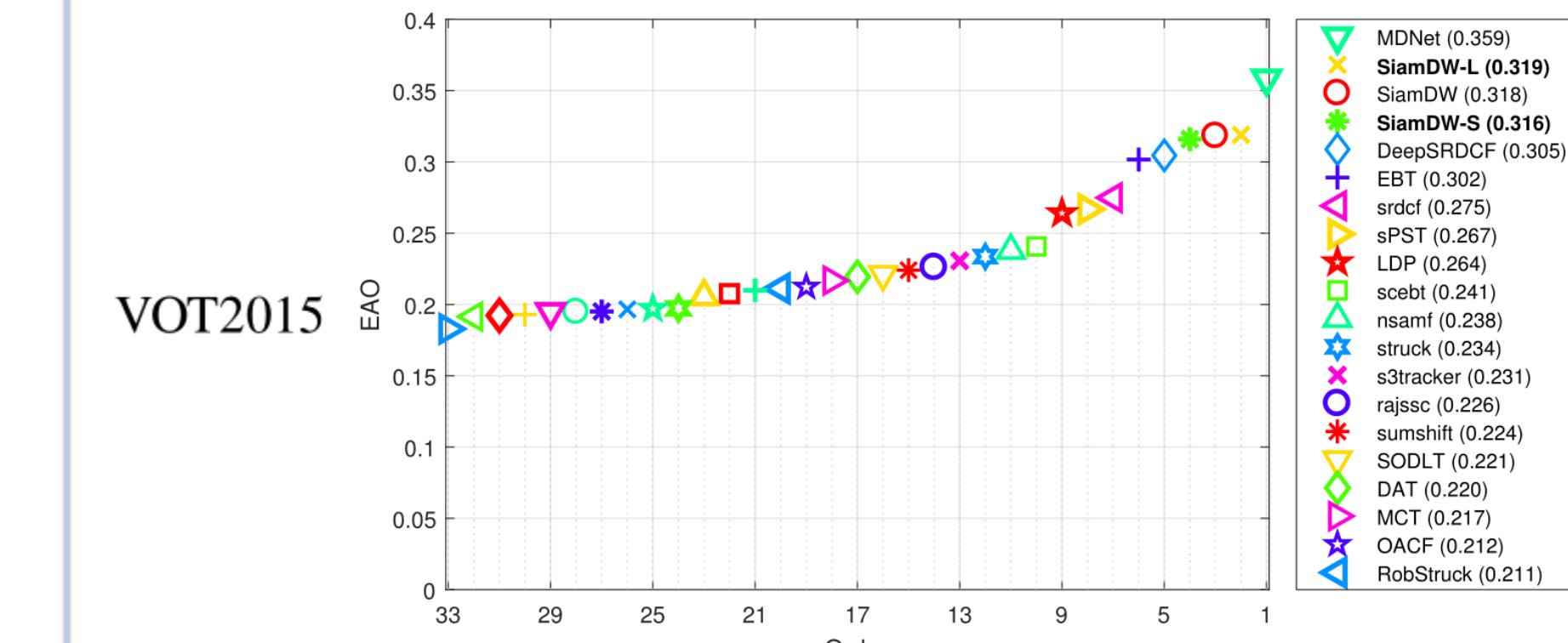
### Ablation Study of distillation

| Model         | SiamDW | SiamDW-L | SiamDW-S |
|---------------|--------|----------|----------|
| Parameters    | 1.445M | 0.354M   | 0.136M   |
| Compress Rate | -      | 4.08     | 10.62    |

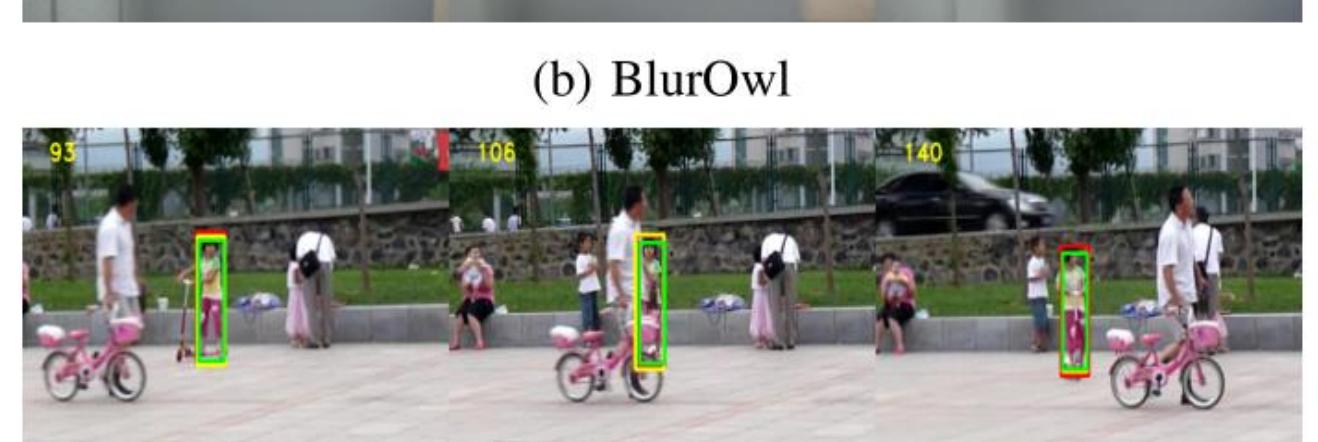
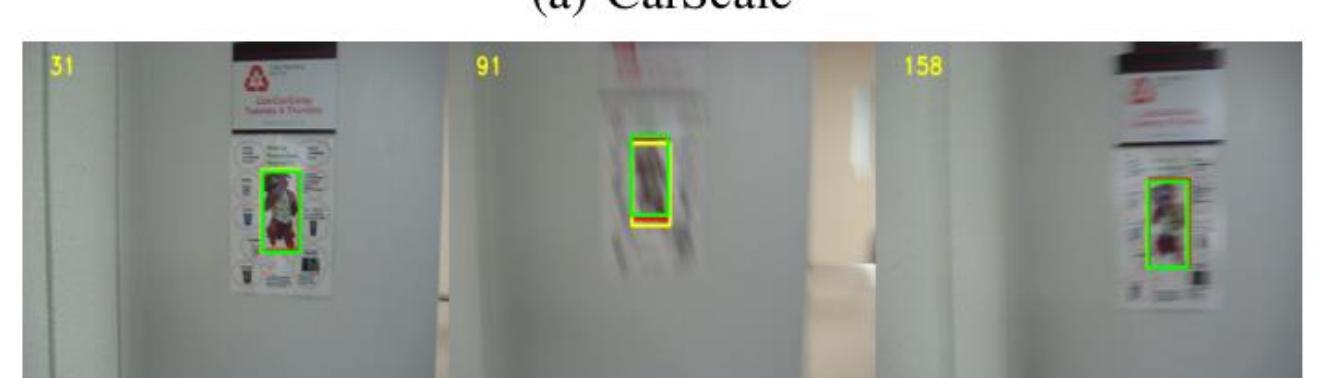
### Speed comparison

| Model | SiamDW | SiamDW-L | SiamDW-S |
|-------|--------|----------|----------|
| FPS   | 53     | 73       | 80       |

## Experiments & Results



### Visualization on OTB2015 benchmark



— SiamDW — SiamDW-L — SiamDW-S