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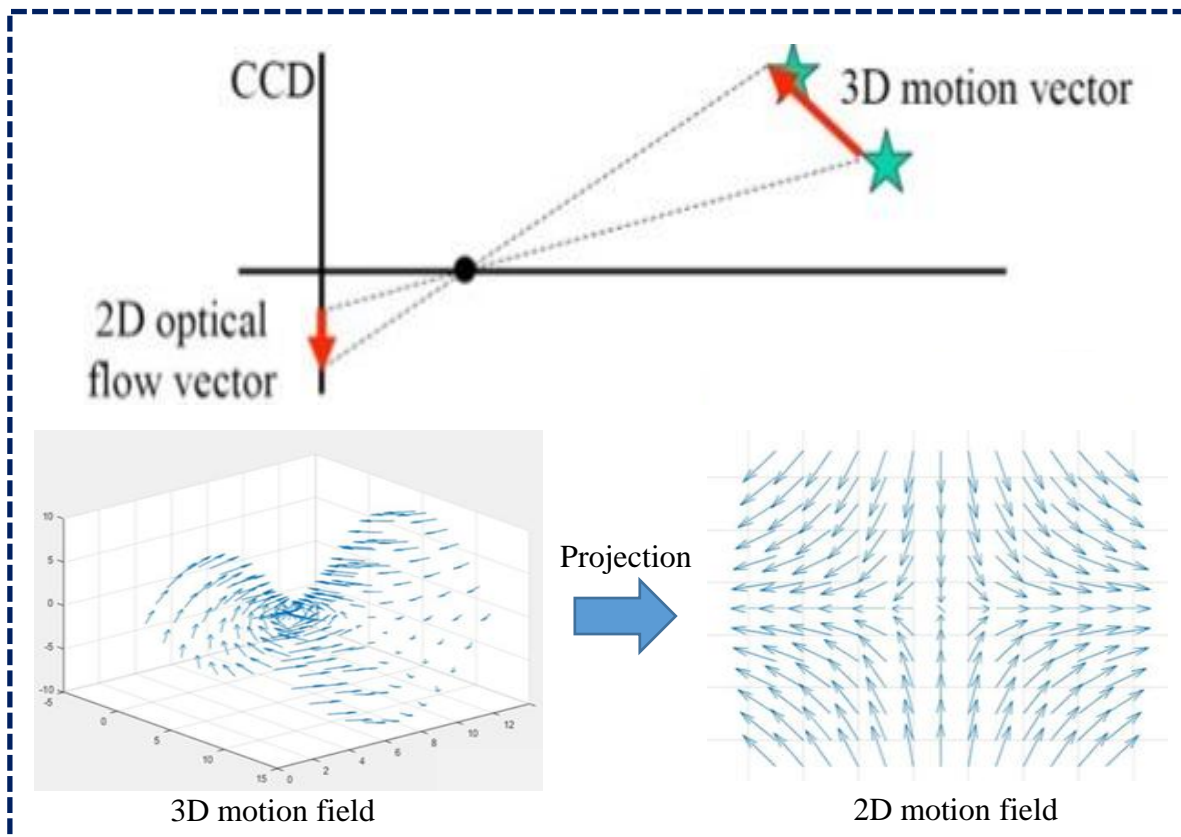
# Exploring the Common Appearance-Boundary Adaptation for Nighttime Optical Flow

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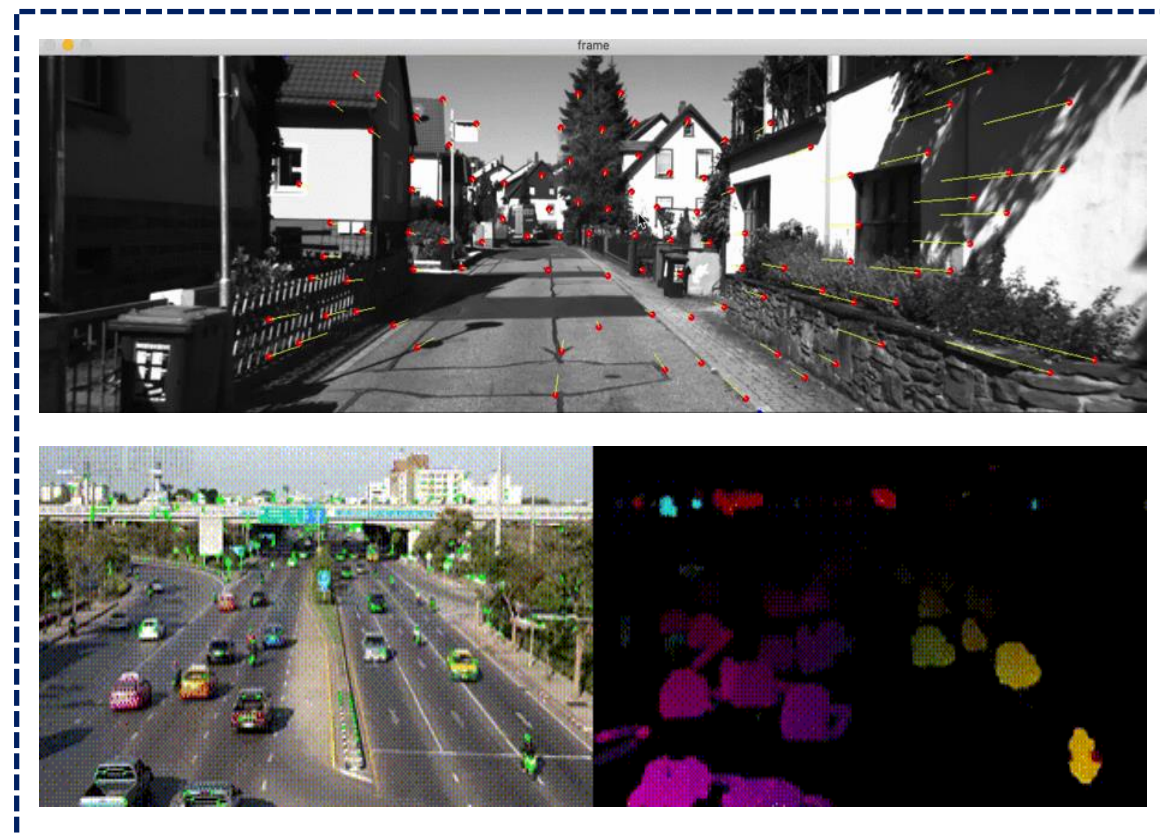
<sup>1</sup> Huazhong University of Science and Technology

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ICLR 2024



**Optical Flow Schematic**



**Optical Flow Visualization**

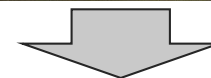
### Clean Scene



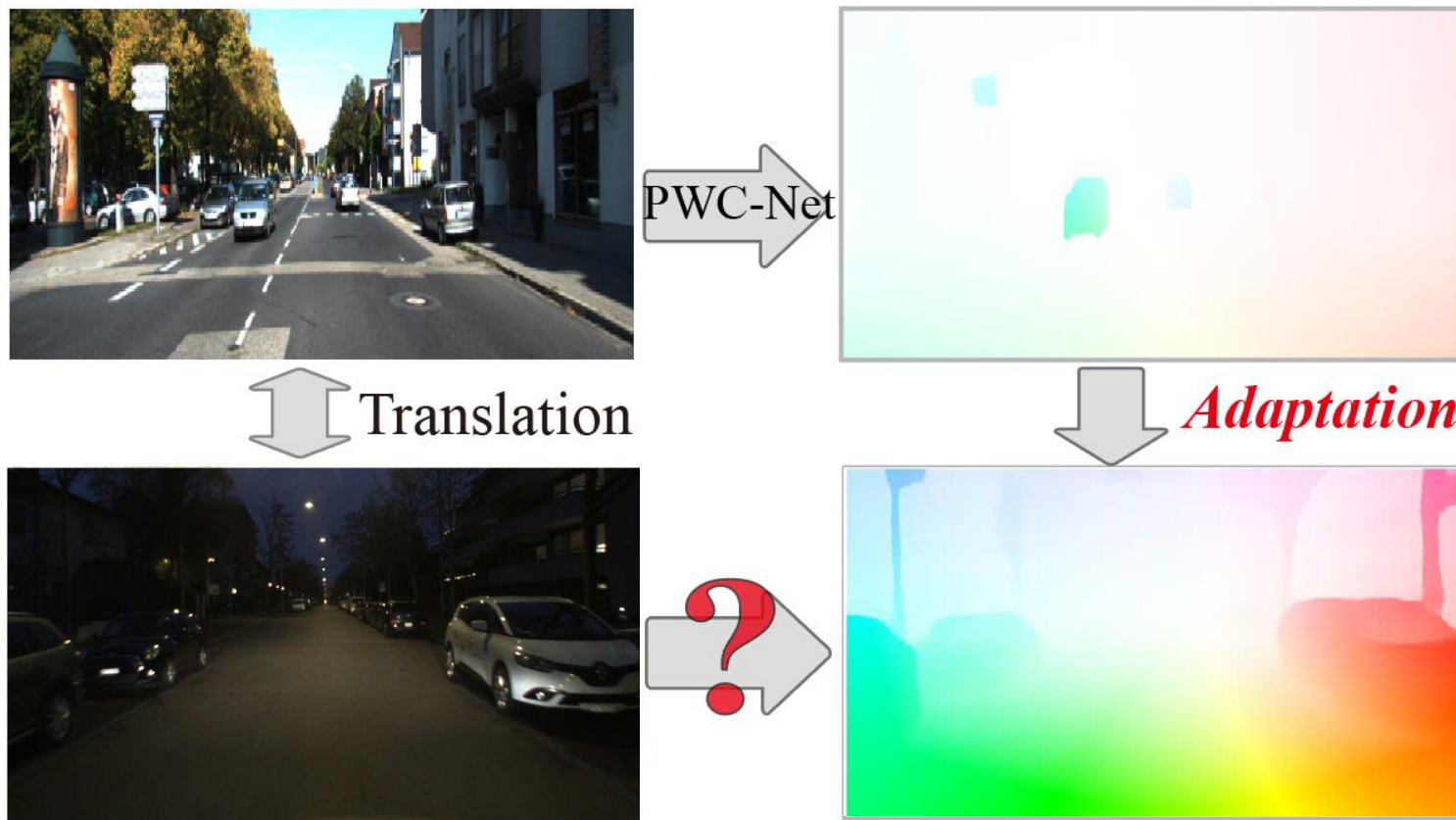
### Nighttime Scene



### Enhanced Nighttime Scene

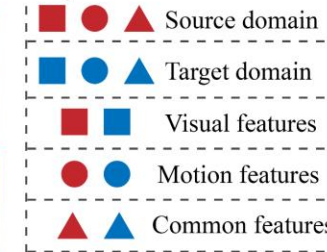
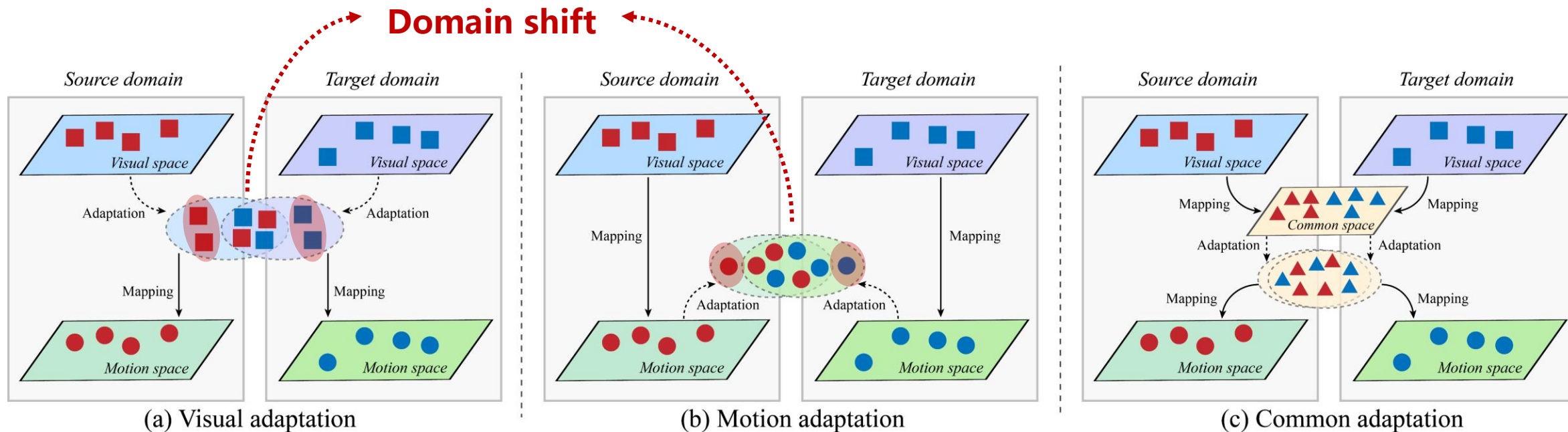


**Optical flow suffers degradation under nighttime scenes**



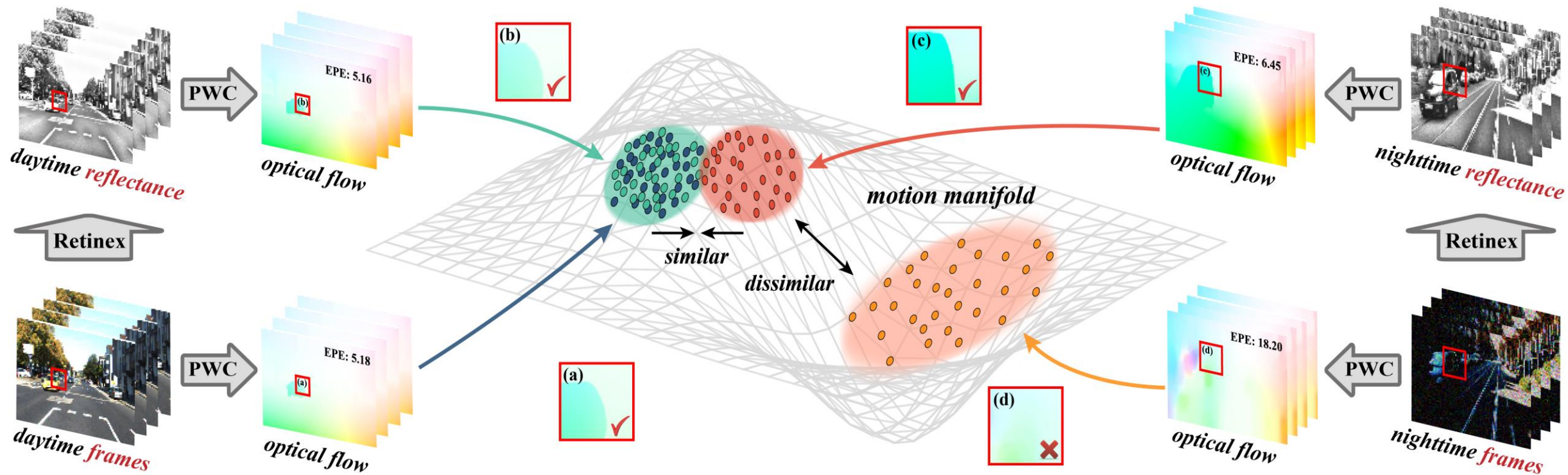
□ **Solution: Transfer the knowledge from source clean domain to target nighttime domain.**





(d) Optical flows estimated by (a) (b) (c)

Exploring a **common space** to reinforce feature distribution alignment



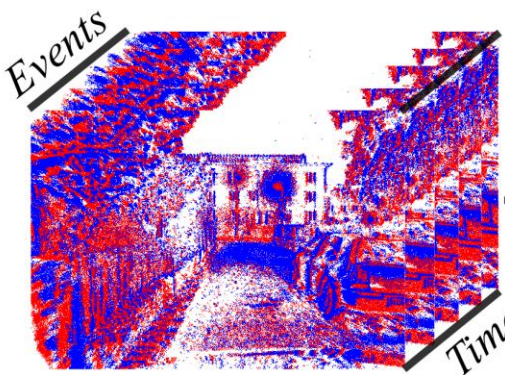
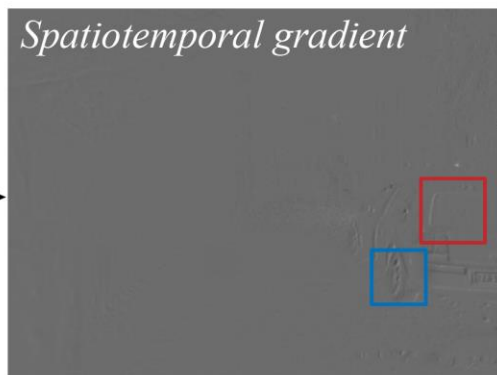
**Motion distributions of daytime/nighttime reflectance are similar**



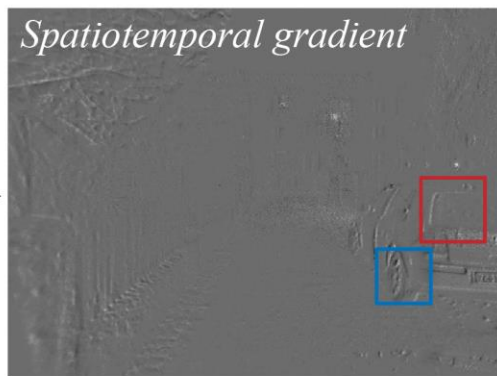
# Common space: Boundary Space



$$-\nabla I \cdot U$$

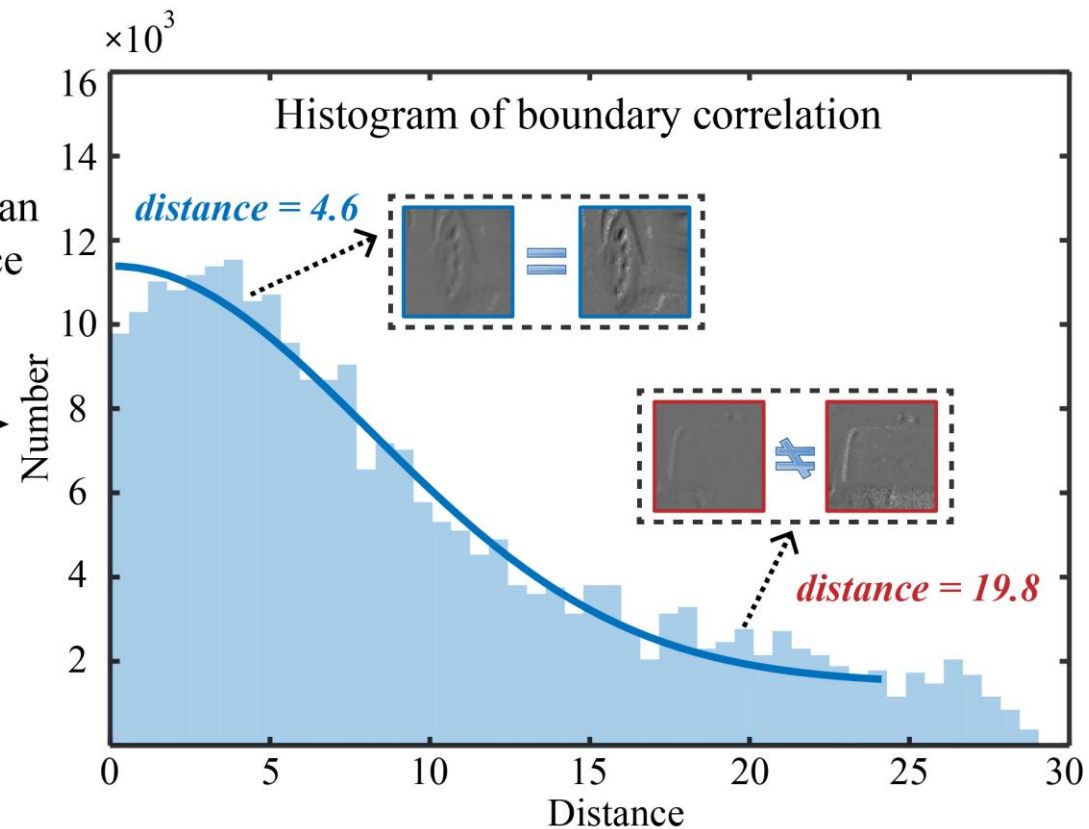


$$U \cdot \sum_i P C$$

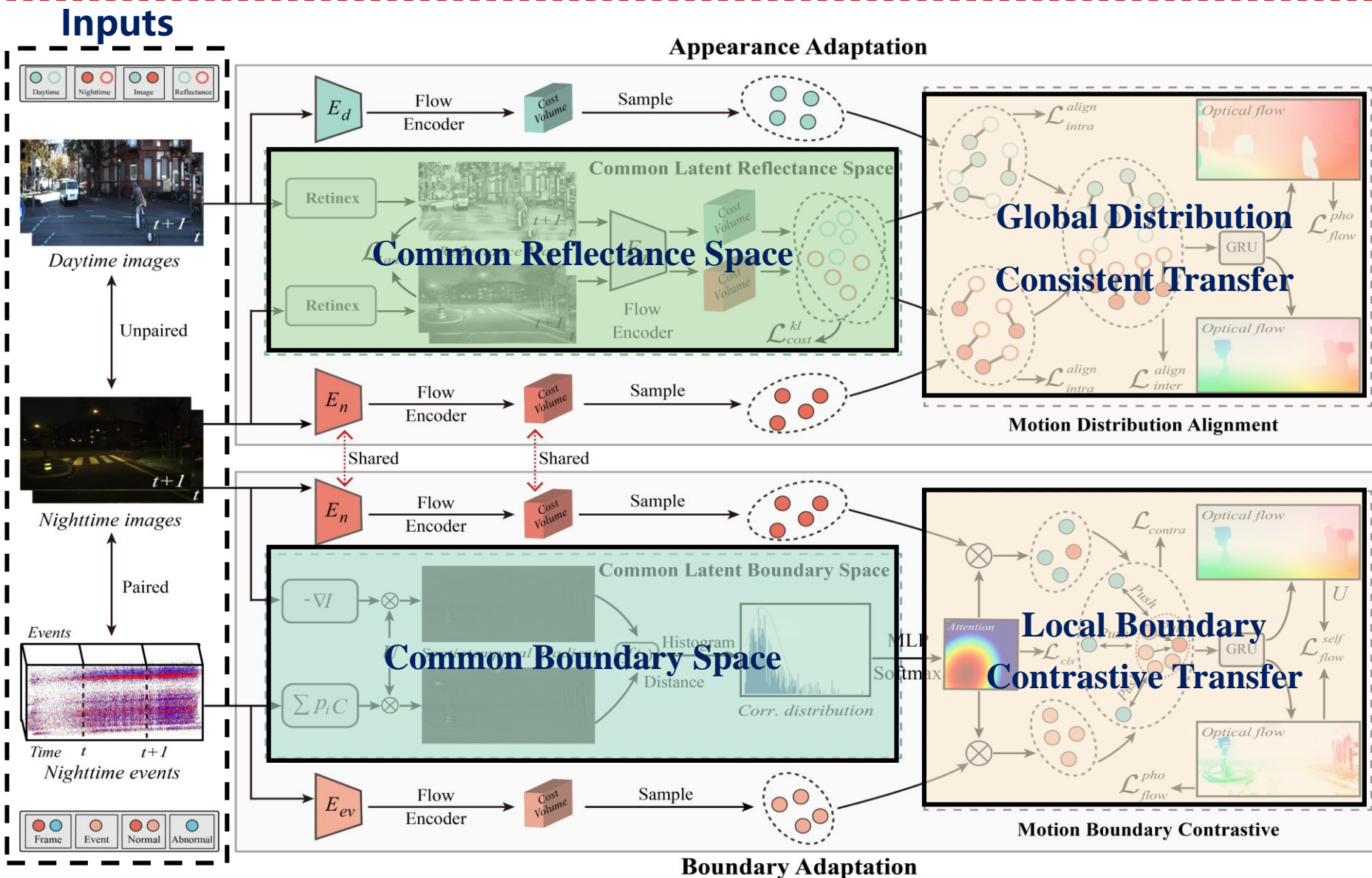


Euclidean distance

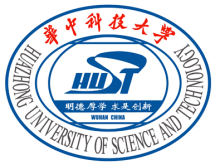
$$f(\cdot)$$



Correlation discrepancy makes up boundary similarity of frame/event domains





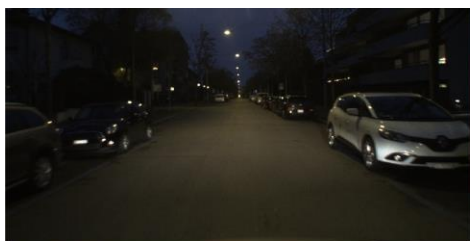


# Comparison on Synthetic/Real datasets



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Method		DarkFlow-PWC	Selfflow			SMURF			ABDA
			-	(KinD++) +	AGLLNet +	-	(KinD++) +	AGLLNet +	
D-KITTI	EPE	7.56	14.22	12.58	11.70	11.36	10.03	8.42	<b>3.47</b>
	Fl-all	35.75%	55.87%	48.69%	46.31%	45.88%	44.65%	39.25%	<b>16.13%</b>
ND-KITTI	EPE	8.56	18.01	16.75	14.54	13.40	11.95	10.26	<b>4.35</b>
	Fl-all	41.28%	65.43%	59.55%	55.26%	54.21%	45.91%	45.60%	<b>23.86%</b>



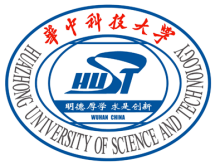
(a) Nighttime Images

(b) SMURF

(c) DarkFlow-PWC

(d) GMA

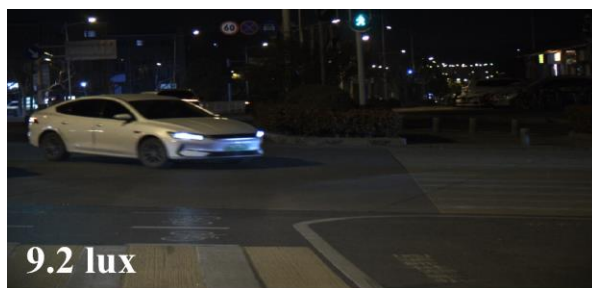
(e) ABDA-Flow



# Comparison on Unseen Nighttime Scenes



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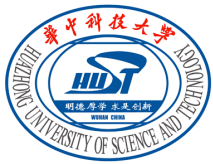
(a) Nighttime images

(b) DarkFlow-PWC

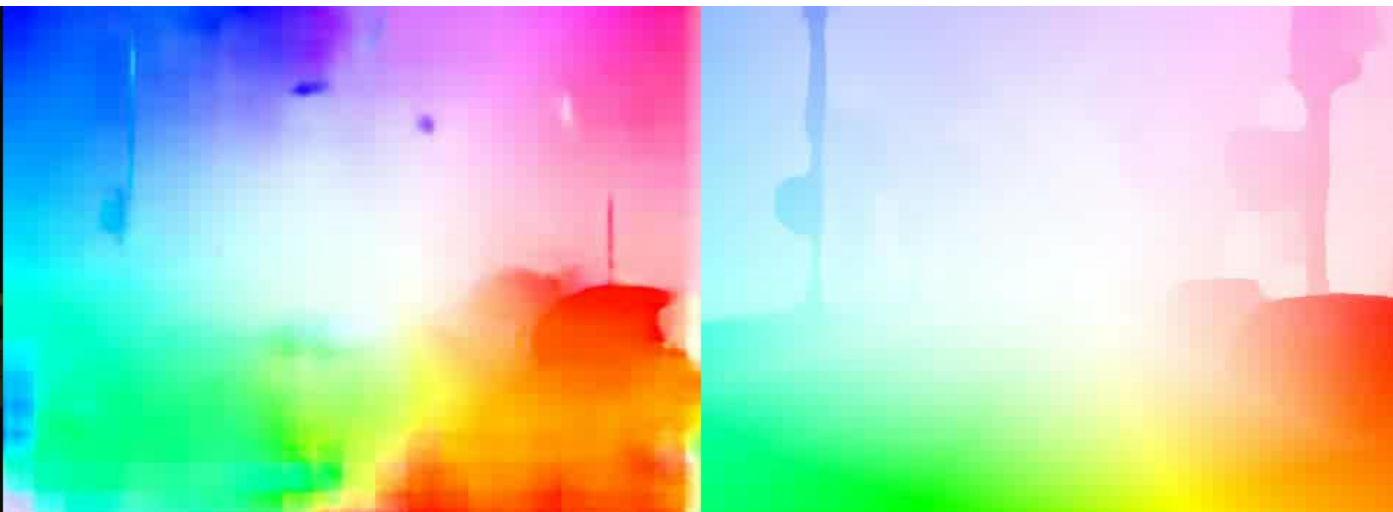
(c) GMA

(d) ABDA-Flow

**ABDA-Flow could generalize well for various illumination**



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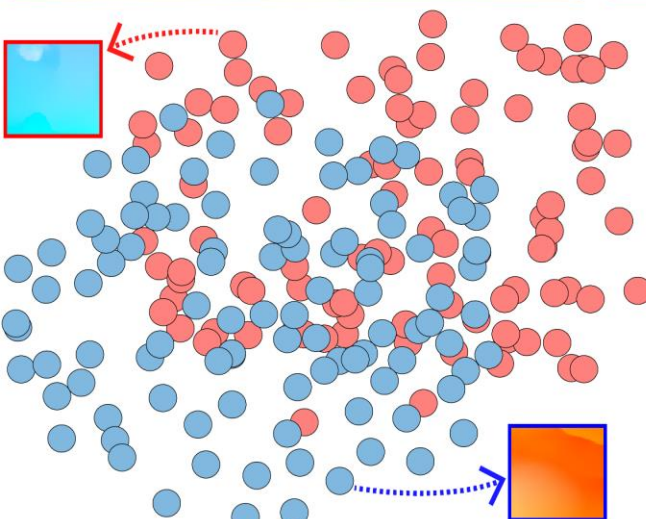
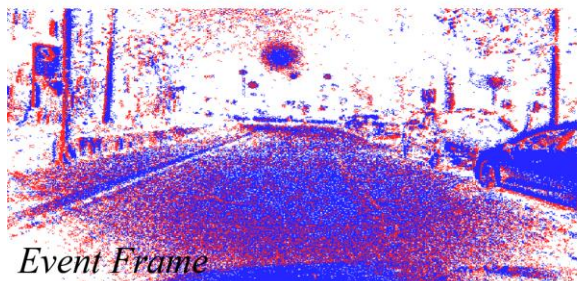
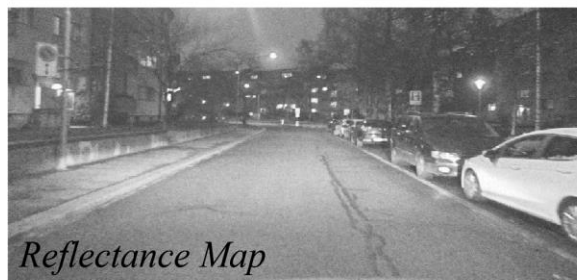
**Nighttime images**

**DarkFlow**

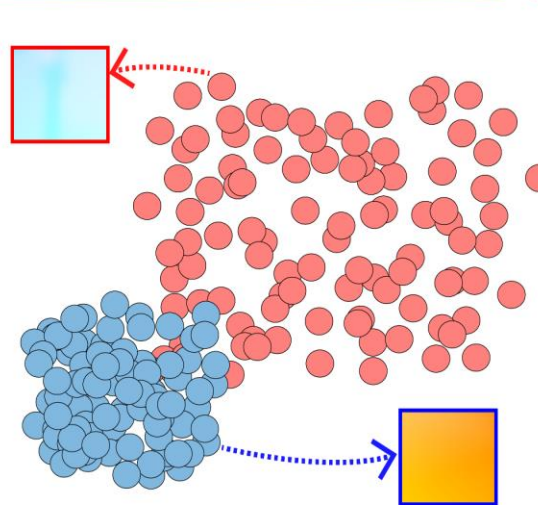
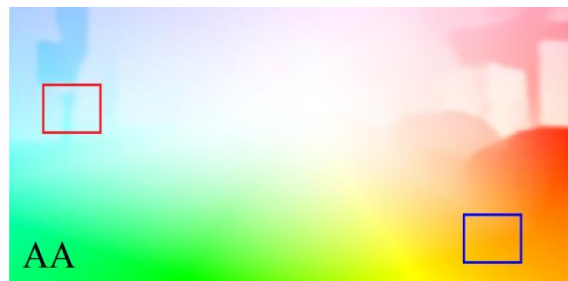
**ABDA-Flow (Ours)**



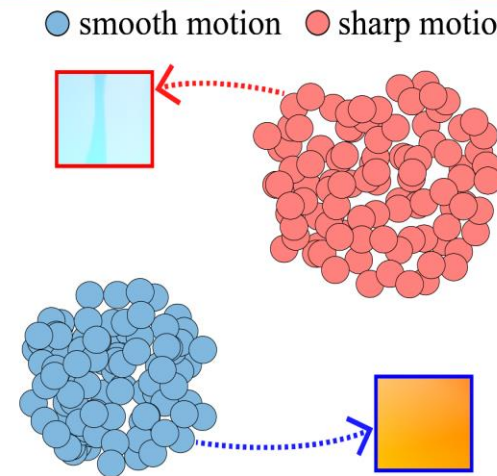
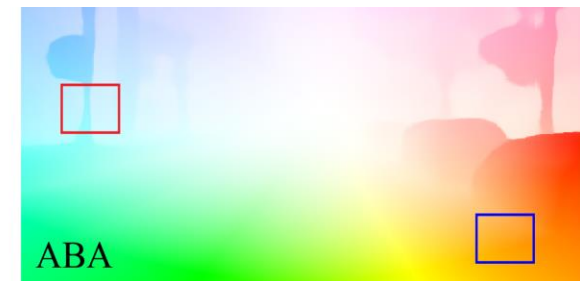
# Direct Adaptation v.s. Common Adaptation



(a) Direct adaptation (DA)

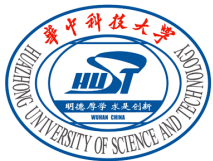


(b) Appearance adaptation (AA)



(c) Appearance-boundary adaptation (ABA)

Common adaptation transfers global and local motion



# Discussion



Table 3: Discussion on effect of common space.

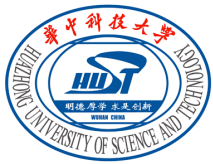
Strategy	EPE
w/o motion/reflect./bound.	1.51
w/ motion, w/o reflect./bound.	1.09
w/ motion/reflect., w/o bound.	0.87
w/ motion/bound., w/o reflect.	0.95
w/ motion/reflect./bound.	<b>0.74</b>

Table 4: Ablation study on adaptation losses.

$\mathcal{L}_{intra}^{align}$	$\mathcal{L}_{inter}^{align}$	$\mathcal{L}_{contra}$	$\mathcal{L}_{flow}^{self}$	EPE
×	×	×	×	1.45
✓	×	×	×	1.24
✓	✓	×	×	1.05
✓	✓	✓	×	0.85
✓	✓	✓	✓	<b>0.74</b>

Table 5: Discussion on training data and optical flow backbone.

Training data	Method	EPE
daytime, nighttime	CycleGAN + our baseline	1.41
	Our appearance adaptation	0.87
daytime, nighttime, event	CycleGAN + our baseline + E-RAFT	1.33
	Ours w/ CNN backbone	0.77
	Ours w/ Transformer backbone	<b>0.74</b>



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*Thanks*