



Domain Generalized Person Re-Identification via Cross-Domain Episodic Learning

ICPR 2020

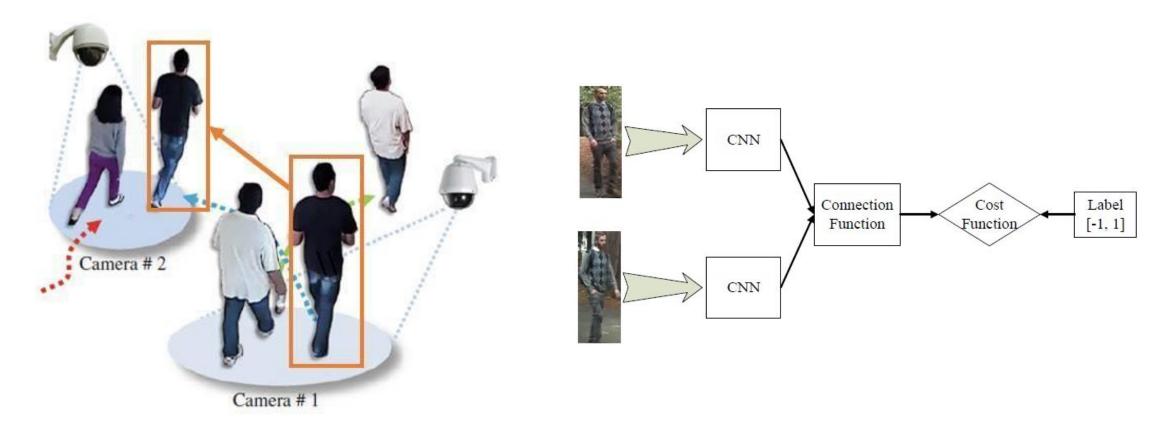
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Person Re-ID

• Definition: Match person images across different cameras.



Person Re-ID

Challenges



Camera/View variation



Occlusion



Clothing similarity



Clutter



Pose variation

Person Re-ID

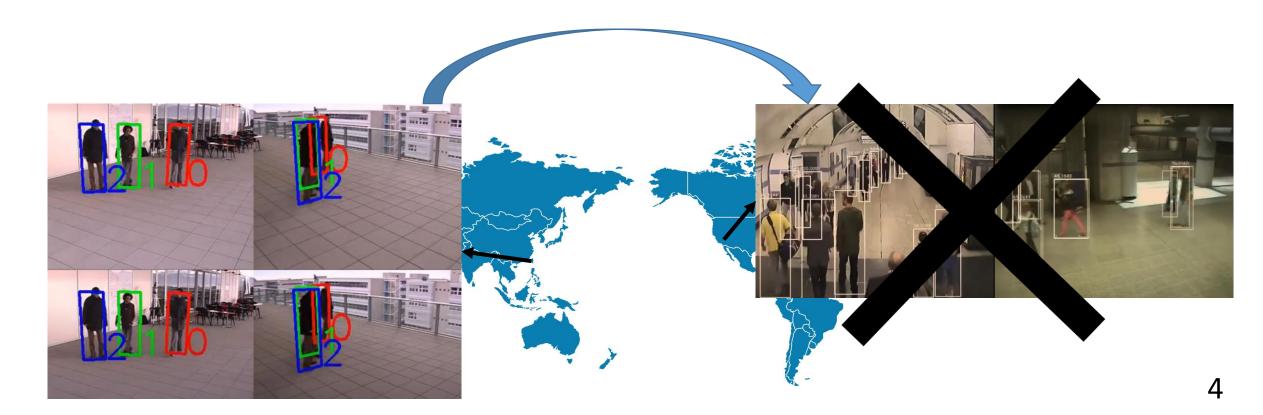
Progress of supervised person re-identification

Person Re-Identification on Market-1501



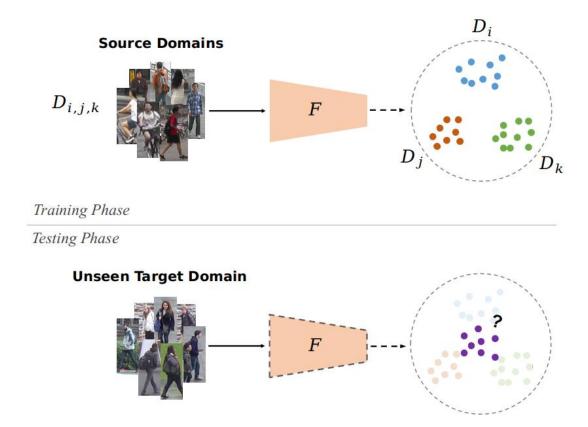
Domain Generalized Person Re-ID

- A more challenging setting, domain generalization, is considered.
- No data (either labeled or unlabeled) from target domain is observed.



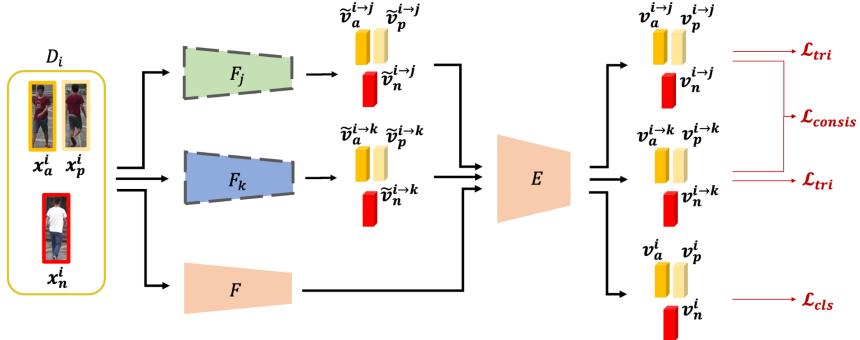
Domain Generalized Person Re-ID

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Domain Generalized Person Re-ID

- Three domains Di, Dj and Dk are random sampled in each episode
- Di mimics the unseen domain, and Dj and Dk mimic source domains



Lin, C. S., Cheng, Y. C., & Wang, Y. C. F.: Domain Generalized Person Re-Identification via Cross-Domain Episodic Learning. In ICPR (2020)

Experiments and Results

Compared to SOTAs

Target	GRID	i-LIDS	PRID	VIPeR	Avg.
DIMN [19]	23.4	44.8	13.1	29.9	27.8
DualNorm [16]	29.2	58.3	54.3	38.6	45.1
Ours	33.0	62.3	57.6	38.5	47.8

Target	GRID	i-LIDS	PRID	VIPeR	Avg.
Baseline	18.8	52.5	14.8	32.0	29.5
DANN [34]	29.0	57.2	56.8	37.8	45.2
Ours	33.0	62.3	57.6	38.5	47.8

Experiments and Results

Ablation Study

Target	GRID	i-LIDS	PRID	VIPeR	Avg.
Ours w/o \mathcal{L}_{tri}	31.3	59.0	55.8	37.3	45.8
Ours w/o \mathcal{L}_{consis}	30.6	60.3	55.7	40.1	46.7
Ours	33.0	62.3	57.6	38.5	47.8

Thank you for listening!