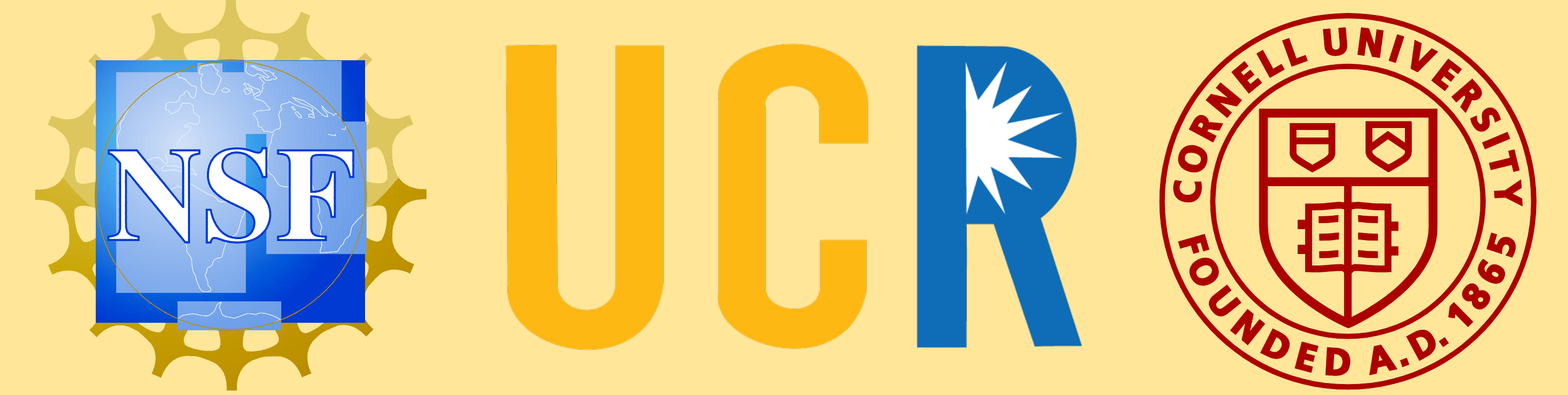


CPS: SYNERGY: DISTRIBUTED SENSING, LEARNING AND CONTROL IN DYNAMIC ENVIRONMENTS

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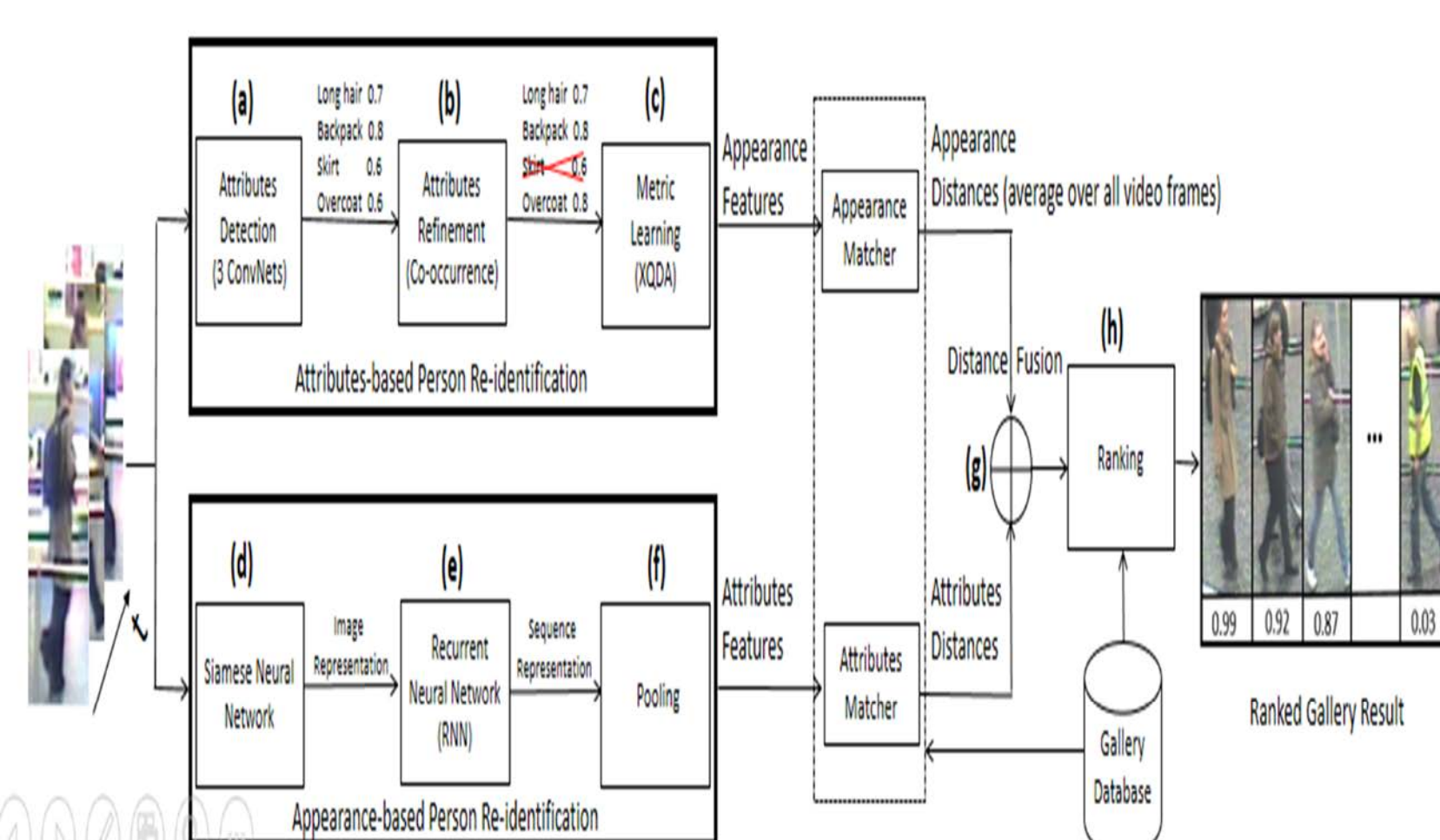


Objectives

- To develop a synergistic framework for fixed and mobile sensors to collaborate on scene understanding
- To perform a tight integration of perception and action and to advance cyber-physical systems by exploring a class of synergies across: control, video understanding, and data management under uncertainty
- To experimentally validate the framework for surveillance domain using a testbed with autonomous agents

Attributes Co-occurrence Pattern Mining for Video-based Person Re-identification

- Visual attributes (e.g., hair and shirt color) offer a human understandable way to recognize people
- Design three ConvNets for attributes detection
- Exploit co-occurrence information to improve attributes' descriptive capabilities
- Propose an attributes-based method and then combine it with an appearance-based model for final prediction
- Achieve better and more consistent results on two public datasets (IEEE AVSS, 2017)



| Methods | r=1 | r=5 | r=10 | r=20 |
|------------------------------------|---------------------|-------------|-------------|-------------|
| Attributes | 3.7 | 9.7 | 14 | 24.2 |
| Ours without refinement | 59.7 (0.034) | 85.3 | 93.6 | 98 |
| Ours | 60.3 (0.035) | 85.3 | 93.6 | 98 |
| RCNN [17] CVPR 2016 | 58.3 (0.035) | 84.6 | 92 | 96.7 |
| TDL [26] CVPR 2016 | 56.3 | 87.6 | 95.6 | 98.3 |
| TAPR [9] ICIP 2016 | 55 | 87.5 | 93.8 | 97.2 |
| SI ² DL [27] IJCAI 2016 | 48.7 | 81.1 | 89.2 | 97.3 |

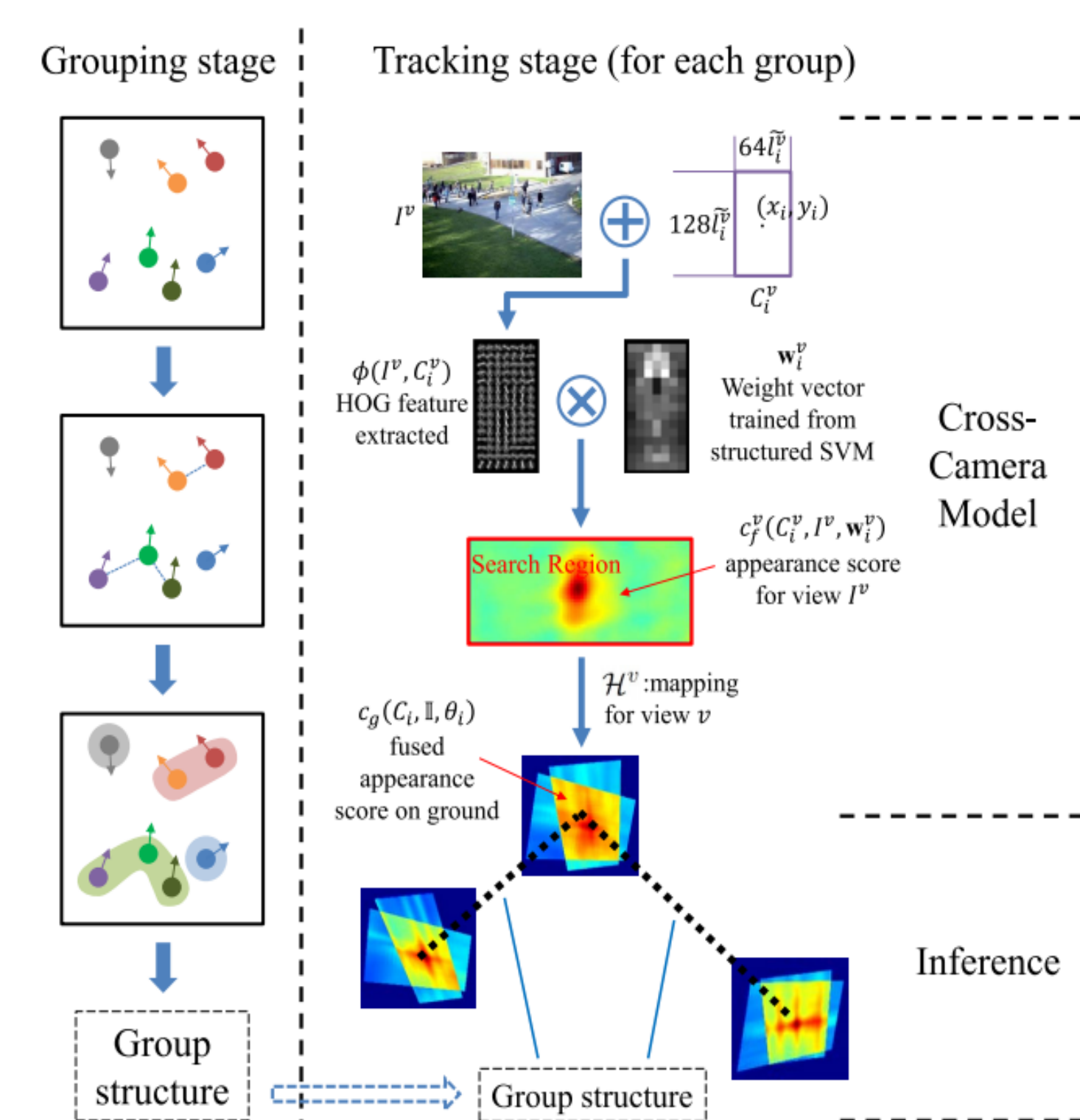
Results on iLIDS-VID dataset

| Methods | r=1 | r=5 | r=10 | r=20 |
|------------------------------------|--------------|-------------|-------------|-------------|
| Attributes | 4.3 | 15.3 | 31 | 43 |
| Ours without refinement | 72.7 (0.041) | 93 | 96.3 | 98.3 |
| Ours | 73.2 (0.038) | 93 | 96.3 | 98.3 |
| RCNN [17] CVPR 2016 | 70.6 (0.051) | 92.3 | 95.3 | 97.3 |
| TDL[26] CVPR 2016 | 56.3 | 87.6 | 95.6 | 98.3 |
| TAPR[9] ICIP 2016 | 55 | 87.5 | 93.8 | 97.2 |
| SI ² DL [27] IJCAI 2016 | 76.7 | 95.6 | 96.7 | 98.9 |

Results on PRID 2011 dataset

Group Structure Preserving Pedestrian Tracking in a Multicamera Video Network

- A new cross-camera model is proposed, which enables the fusion of the confidence information from all camera views
- Group structures on the ground plane provide extra constraints between pedestrians
- The structured support vector machine is adopted to update model
- Excellent results are obtained on challenging data (IEEE Trans. CSVT, 2017)

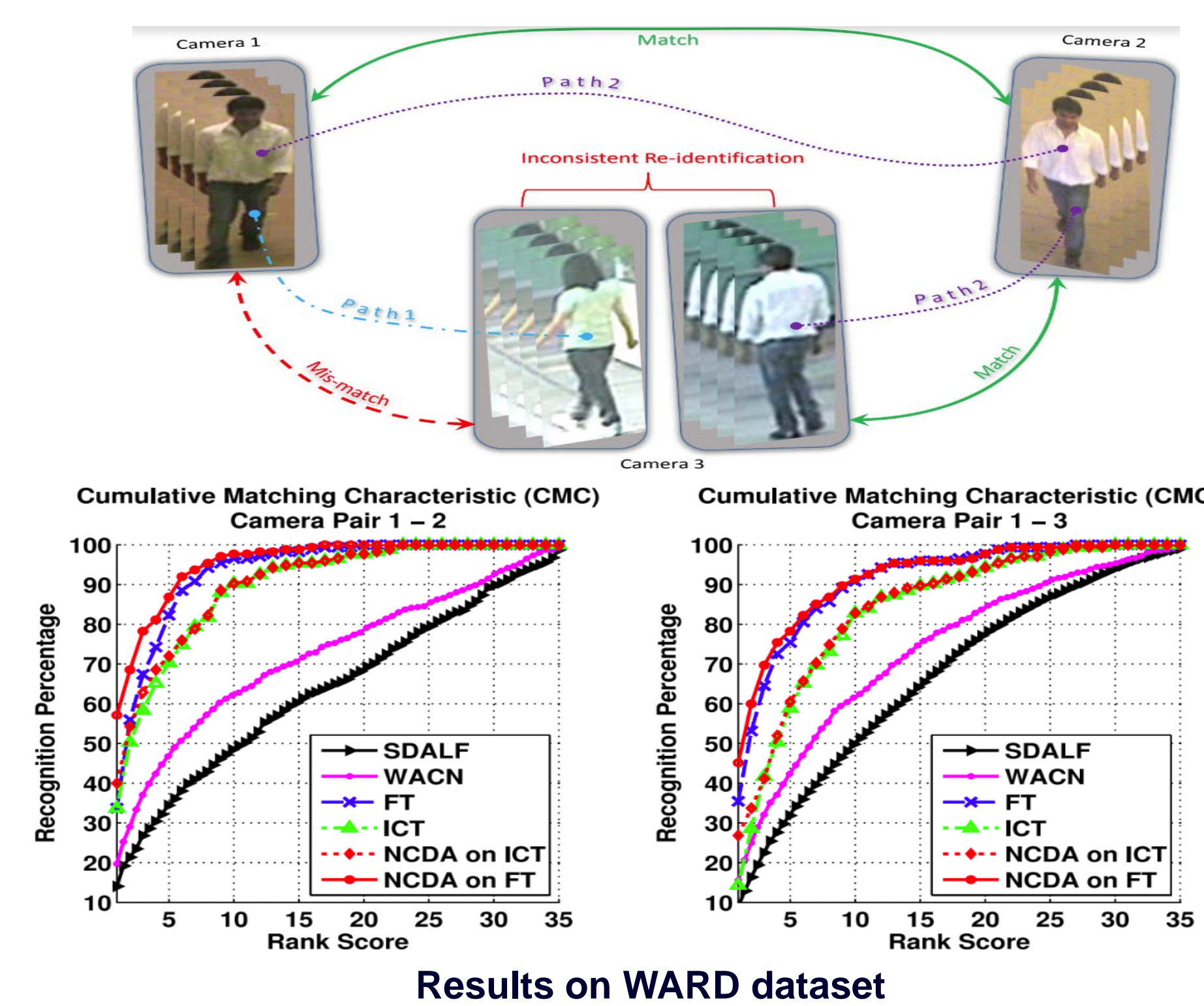


| Evaluation Category | Tracking Individually | | | Tracking in Groups | | |
|---------------------|-----------------------|--------|----------|--------------------|---------------|---------------|
| | MOTP | MOTA | Distance | MOTP | MOTA | Distance |
| View 1 | 82.13% | 76.54% | 167.91 | 82.89% | 79.42% | 136.41 |
| View 2 | 81.89% | 70.75% | 90.36 | 82.42% | 83.42% | 68.17 |
| Ground plane | 84.33% | 59.58% | 36.11 | 85.14% | 72.71% | 26.44 |

Results on PETS 2009 dataset

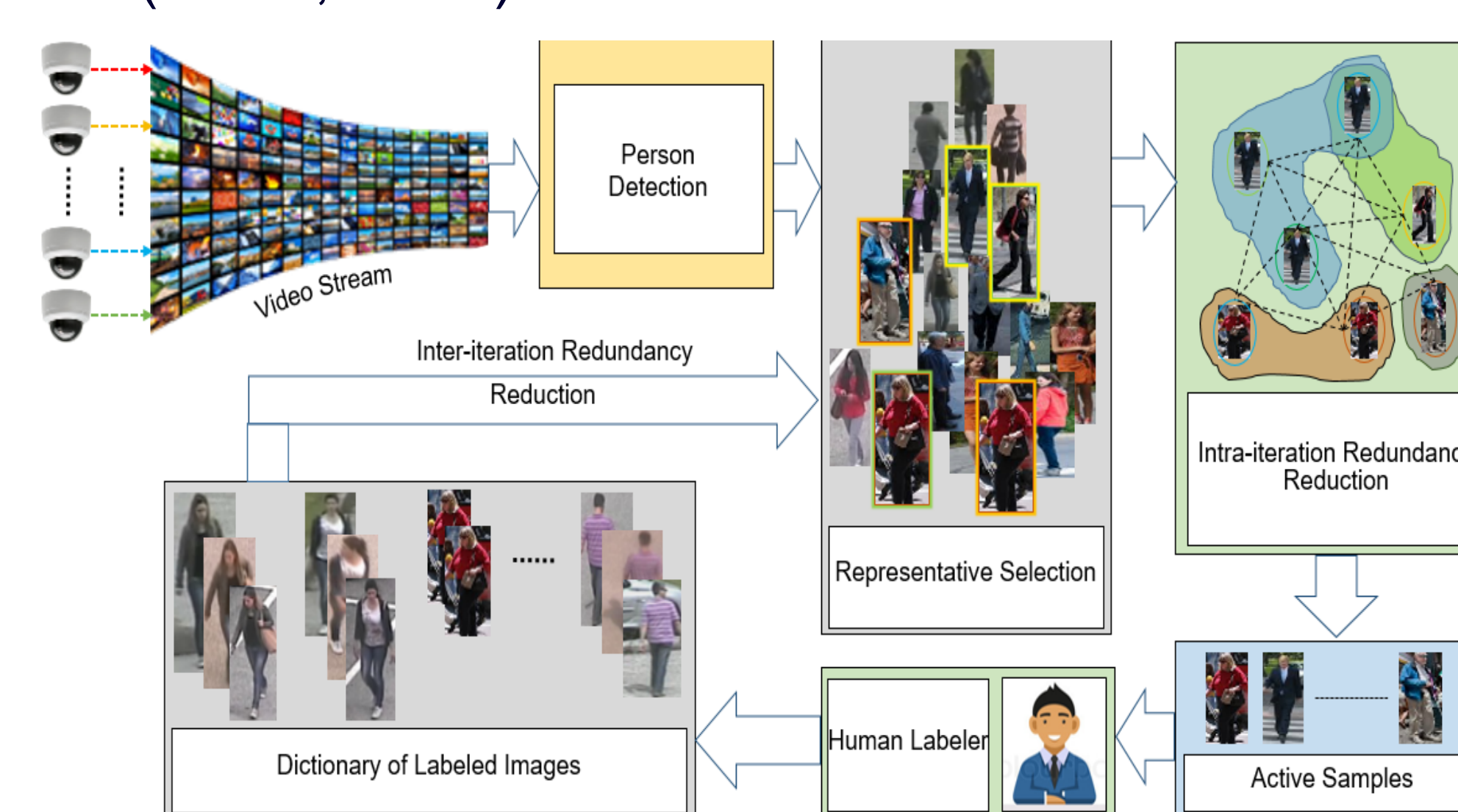
Network Consistent Re-identification & Continuous Adaptation of Person Identification with Minimal Labeling

- Network Consistent Re-identification
 - Address consistent matching as a data-association task over network perspective
 - Formulate an optimization problem to maximize overall pairwise similarity while abiding by network level constraints (IEEE TPAMI, 2016)

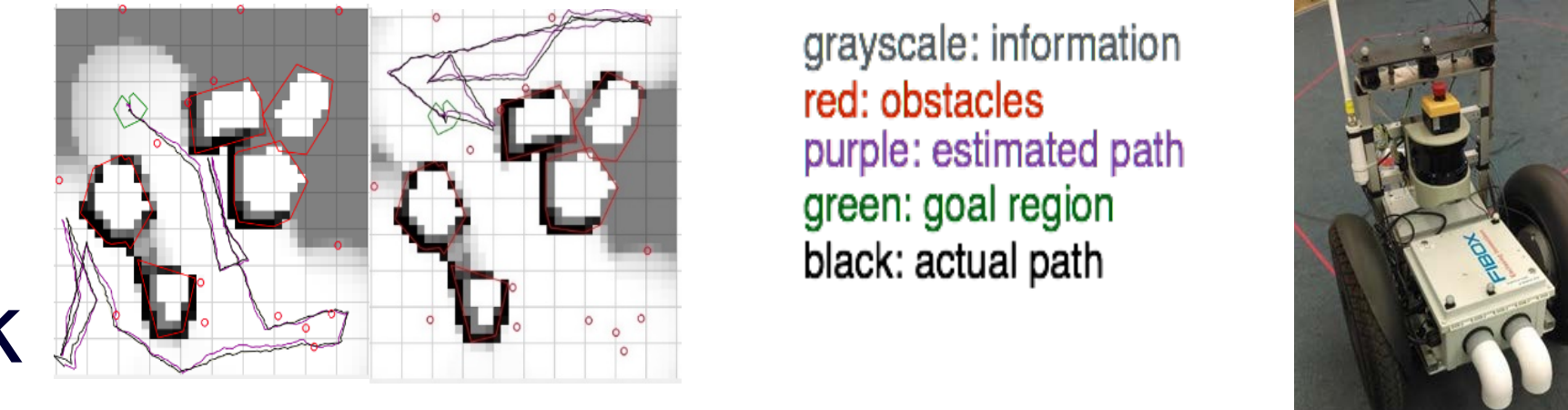


Continuous Adaptation of Person Identification with Minimal Labeling

- Propose a sparse representative selection based approach
- Formulate as a convex optimization problem, which endorses selection of samples with the most variabilities (CVIU, 2017)



Exploration with Localization Guarantees



- Key results:
 - General info gathering framework
 - Maximizing information goals
 - Probabilistic guarantees
 - Asymptotic guarantees as the # of samples increases
 - Bound on the reward for partially known environments
 - Speeds computation (>1000x) to real time
 - Validation by simulation/experiment (IEEE Trans. Control Sys., 2018)

Path Optimization

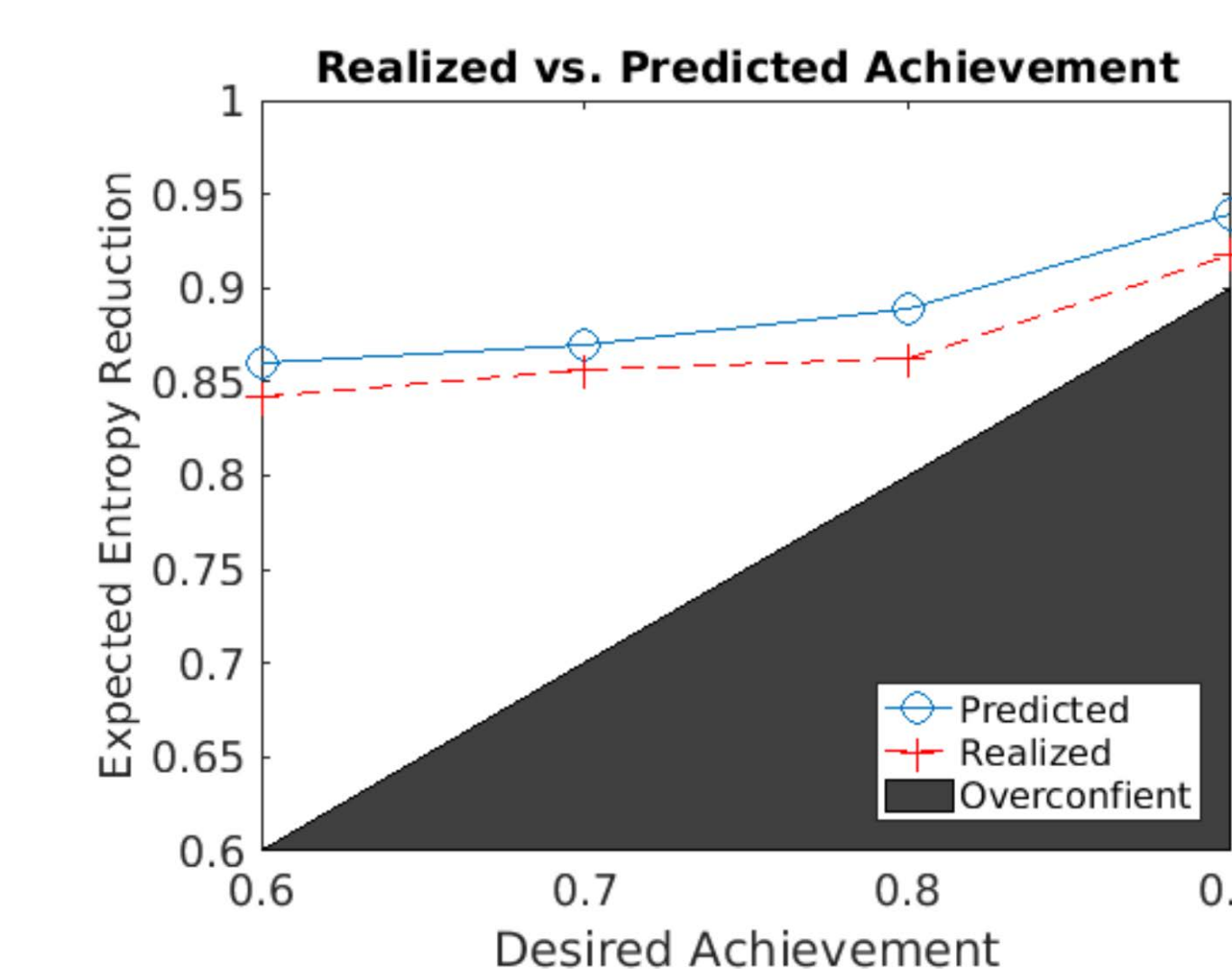
$$X_{t:T}^* = \operatorname{argmax}_{X_{t:T}} R(X_{t:T})$$

Information Reward

$$R(X_{t:T}) = H(Y_{l(t)}) - \mathbb{E}_{Z_{t:T}} [H(Y_{n(T)})]$$

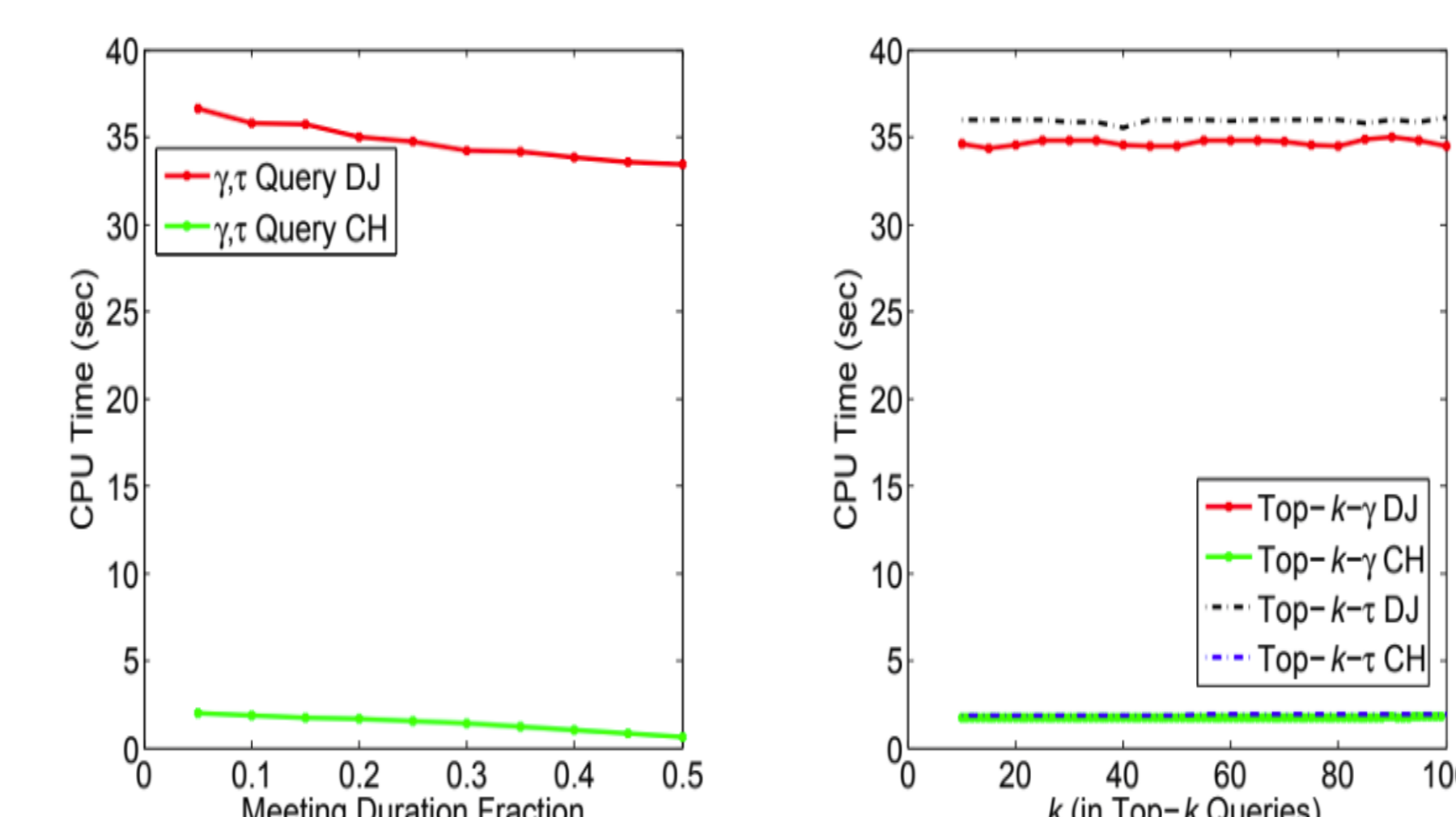
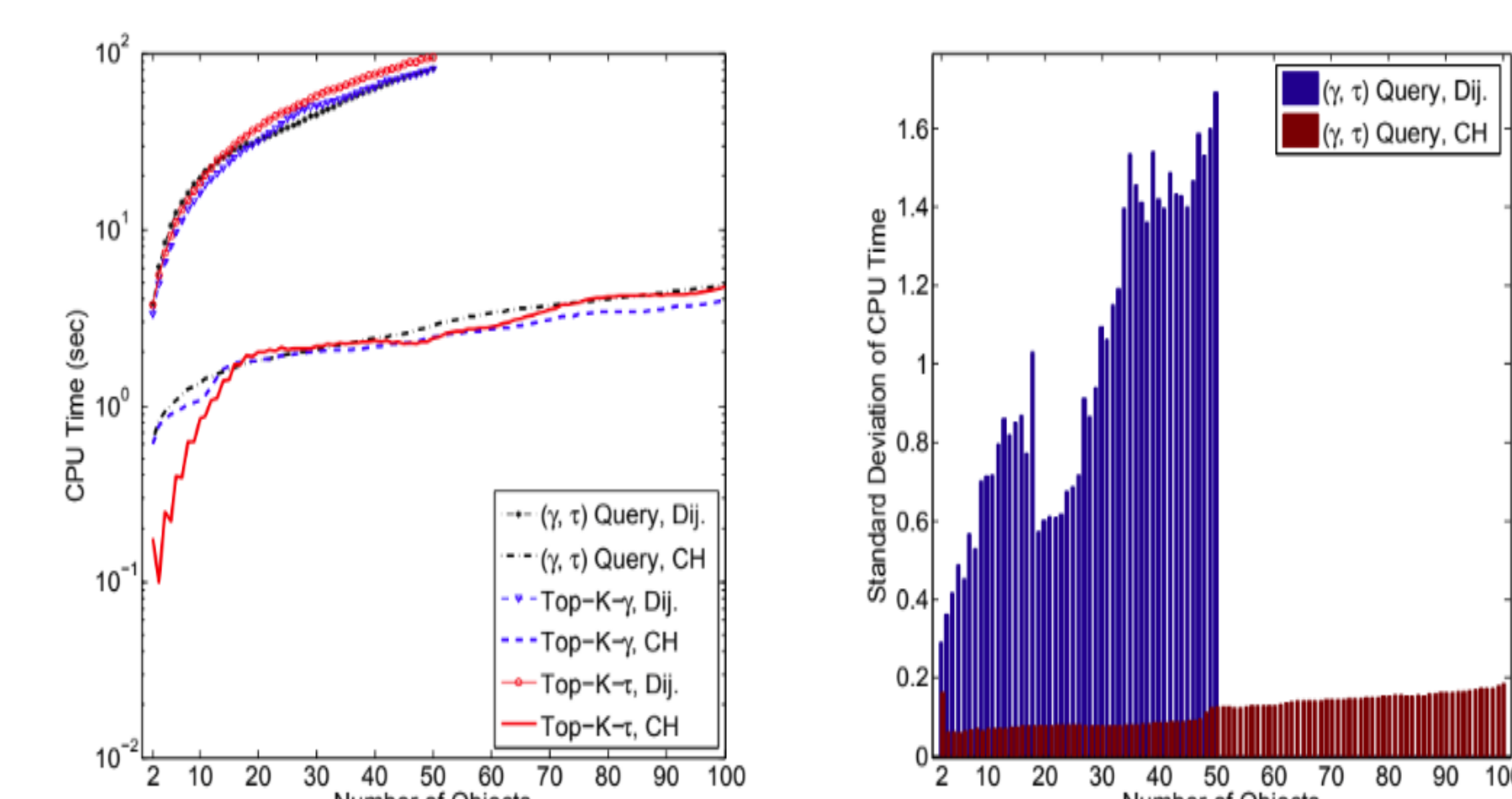
Information constraint

$$f(X_{t:T}) \leq C$$



Assembly Queries: Planning & Discovering Assemblies of Moving Objects Using Partial Information

- Introduce the novel and important class of assembly queries:
 - "Assembly discovery" (determine whether two or more moving objects could have had a meeting within a region of interest)
 - "Assembly planning" (arrange for meetings for a group of friends visiting a city without violating their remaining schedules)
 - Provide efficient solutions given incomplete trajectory information, using the topology of the underlying transportation network
 - Present a formal model for the general problem and prove the correctness of our algorithm
 - Utilize a preprocessing method based on Contraction Hierarchies to gain orders of magnitude speed up over the naïve Dijkstra-based methods (ACM SIGSPATIAL, 2017)



Results on road network data for California and Nevada

Acknowledgment: This work is supported in part by NSF CPS grant 1330110