# Learning to Track: Online Multi-Object Tracking by Decision Making

Yu Xiang<sup>1,2</sup>, Alexandre Alahi<sup>1</sup>, and Silvio Savarese<sup>1</sup>

<sup>1</sup>Stanford University, <sup>2</sup>University of Michigan

ICCV 2015



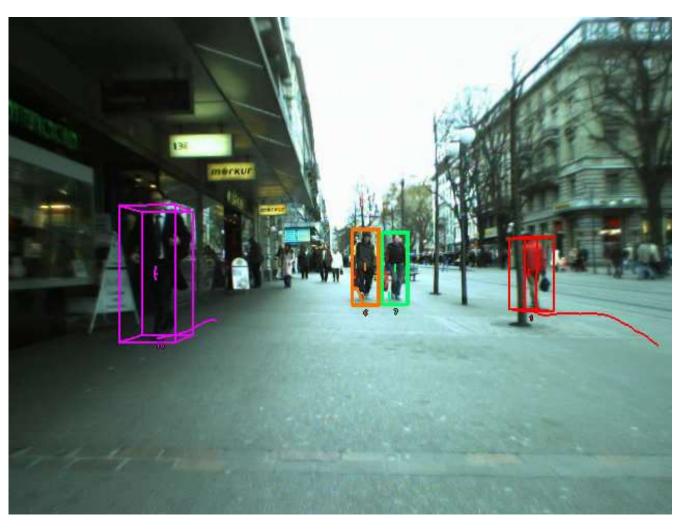
# Multi-Object Tracking



Visual surveillance

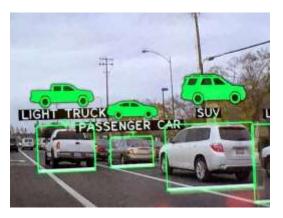


**Sport Analysis** 





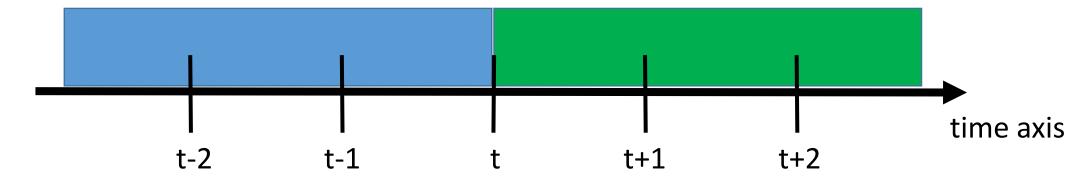
Robot navigation

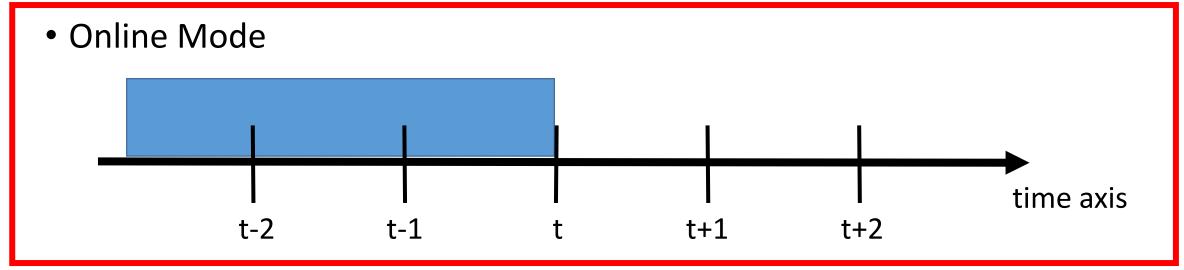


Autonomous driving

#### Batch Mode vs. Online Mode

• Batch Mode

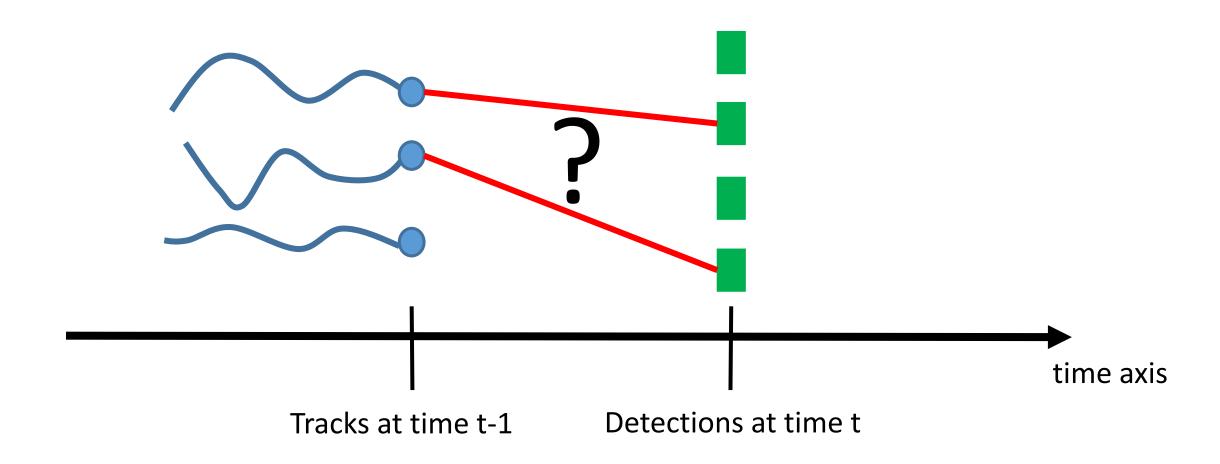




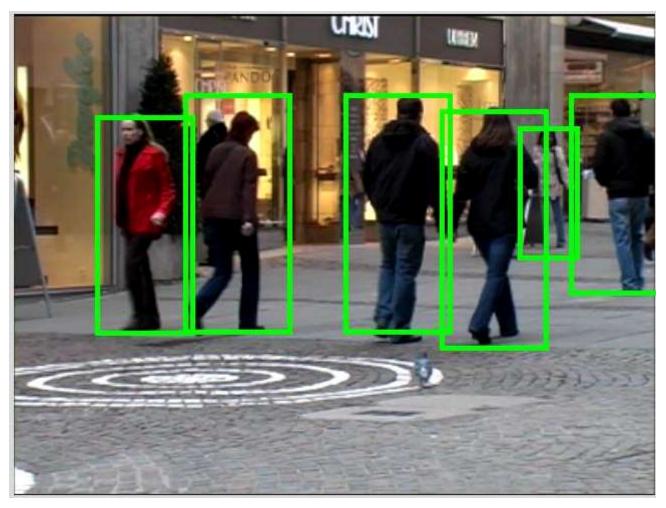
# Tracking by Detection



#### Data Association



### Challenges



Noisy detection: false alarms and missing detections

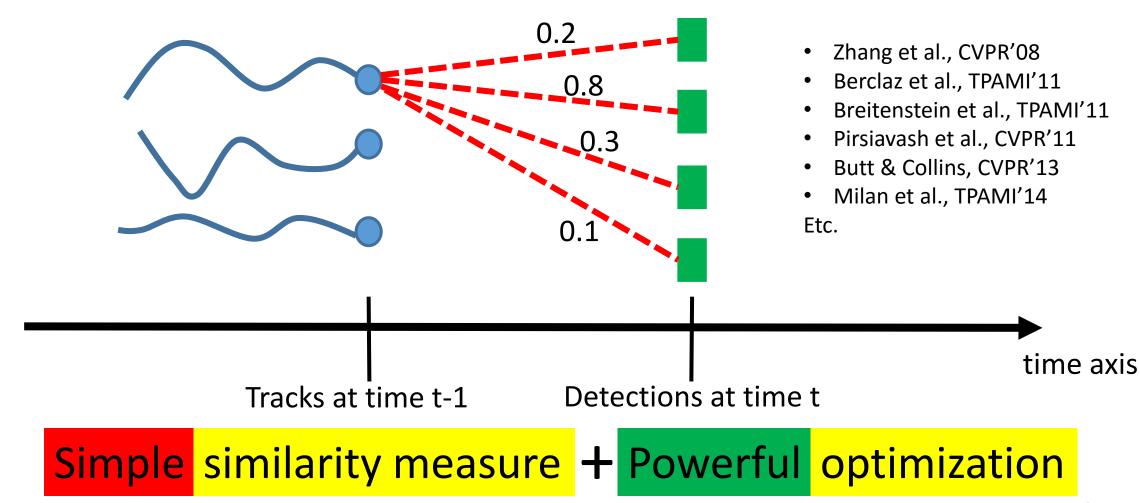
# Challenges



Occlusion

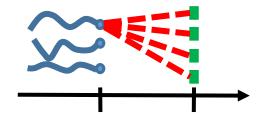
#### Similarity Function for Data Association

**Ours** 



8

# Learning to Track



Different features/cues between targets and detections

- Appearance
- Location
- Motion Etc.

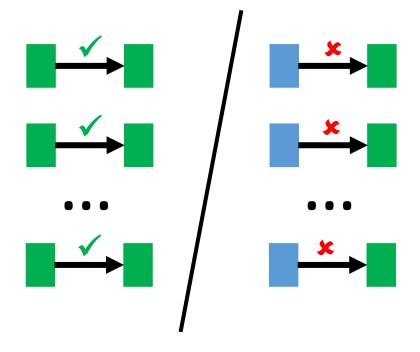
Similarity = 
$$w_1 \phi_1(\mathbf{w}, \mathbf{l}) + \cdots + w_n \phi_n(\mathbf{w}, \mathbf{l})$$

Weights to combine different cues (to be learned)

# Offline-learning vs. Online-learning

#### Offline-learning vs. Online-learning

	Offline- learning
Training time	Before Tracking
With supervision	
Use history of the target	*



- Li et al., CVPR'09
- Kim et al., ACCV'12 Etc.

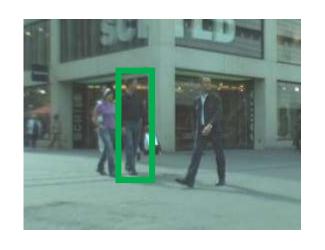
#### Offline-learning vs. Online-learning

	Offline- learning	Online- learning
Training time	Before Tracking	During Tracking
With supervision		*
Use history of the target	*	



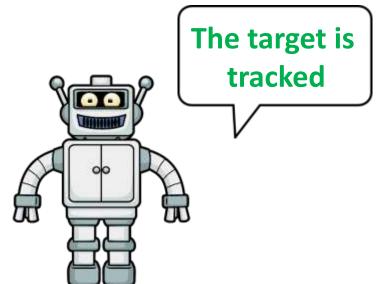
- Song et al., ECCV'08
- Kuo et al., CVPR'10
- Bae et al., CVPR'14 Etc.

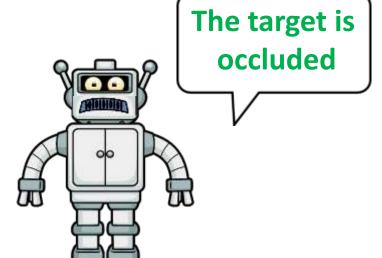
#### Our Solution: Tracking by Decision Making





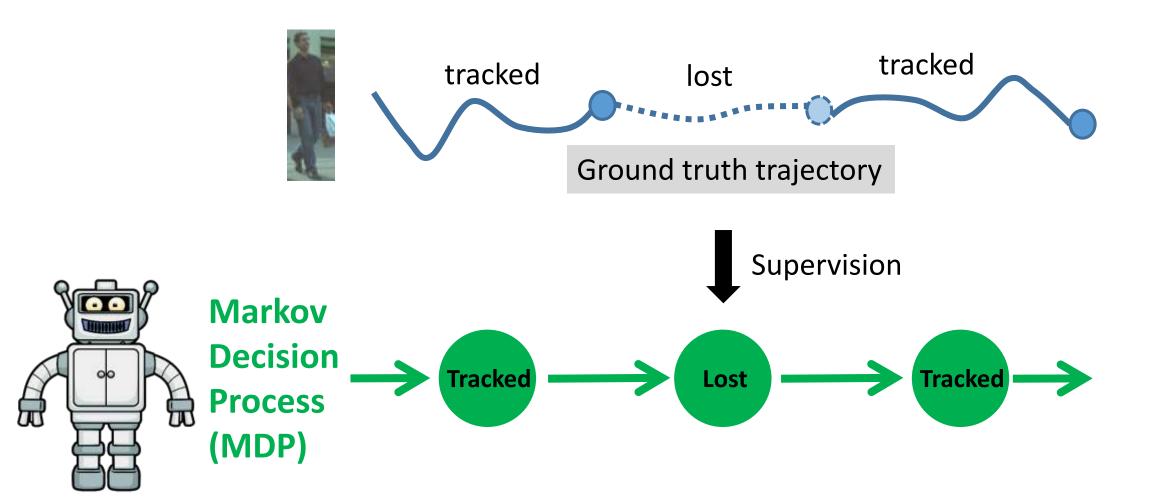








#### Inverse Reinforcement Learning

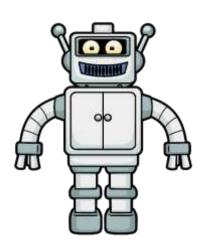


### Comparison between Different Learning Strategies

	Offline- learning	Online- learning
Training time	Before Tracking	During Tracking
With supervision		*
Use history of the target	*	

### Comparison between Different Learning Strategies

	Offline- learning	Online- learning	Ours
Training time	Before Tracking	During Tracking	Before Tracking
With supervision		*	
Use history of the target	*		



#### Outline

Markov Decision Process (MDP) for a Single Target

Online Multi-Object Tracking with MDPs

Experiments

Conclusion

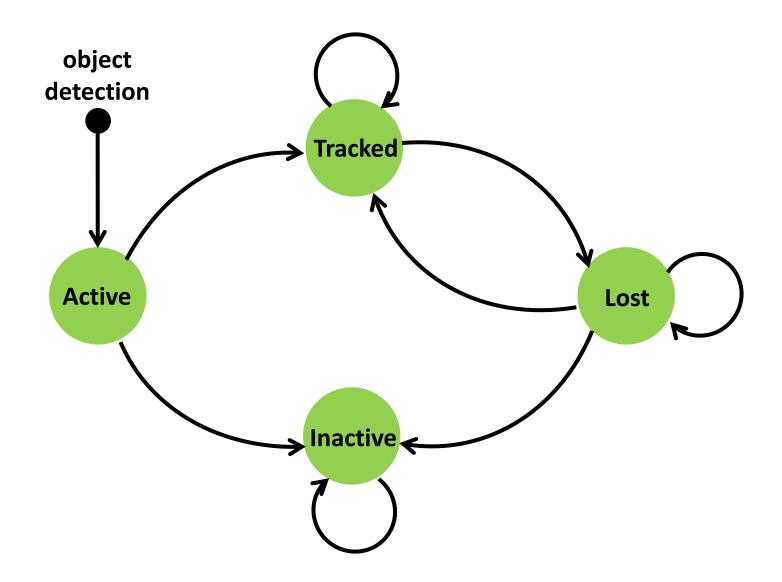
#### Outline

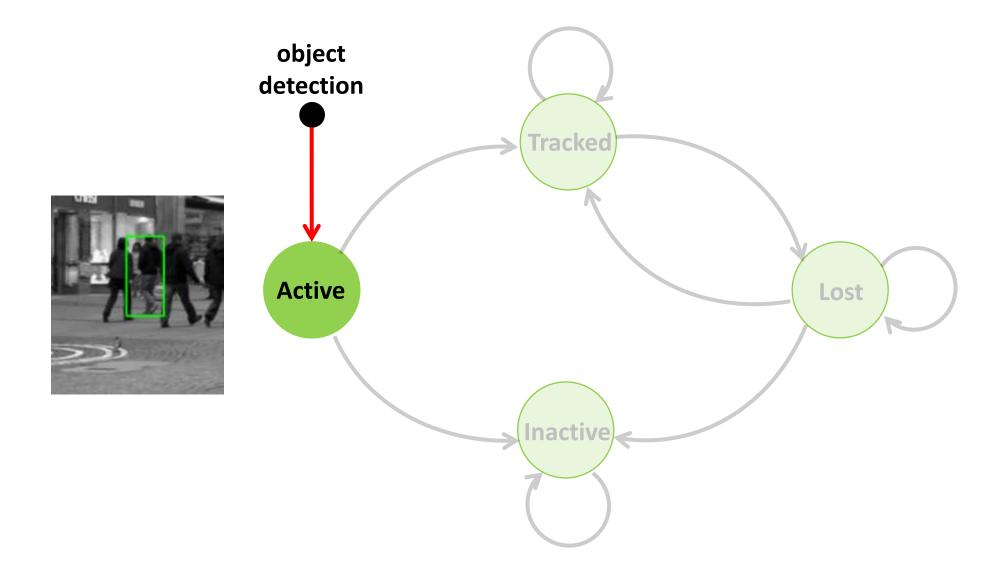
Markov Decision Process (MDP) for a Single Target

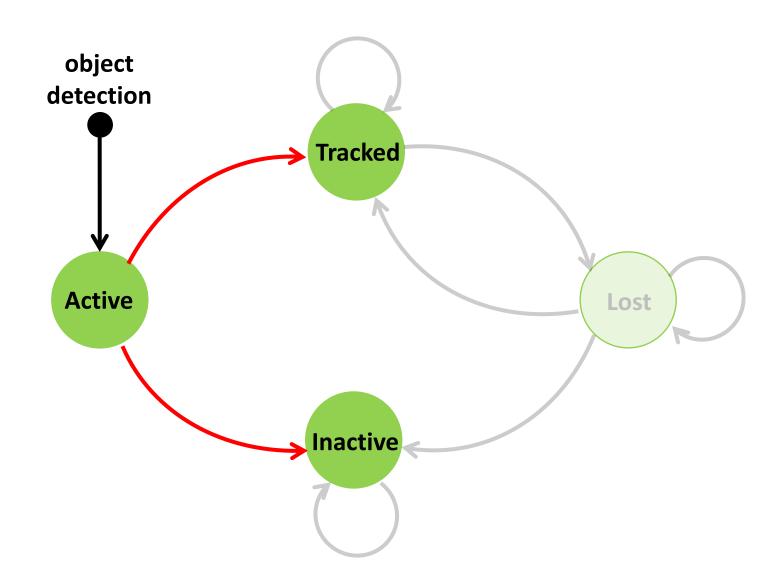
Online Multi-Object Tracking with MDPs

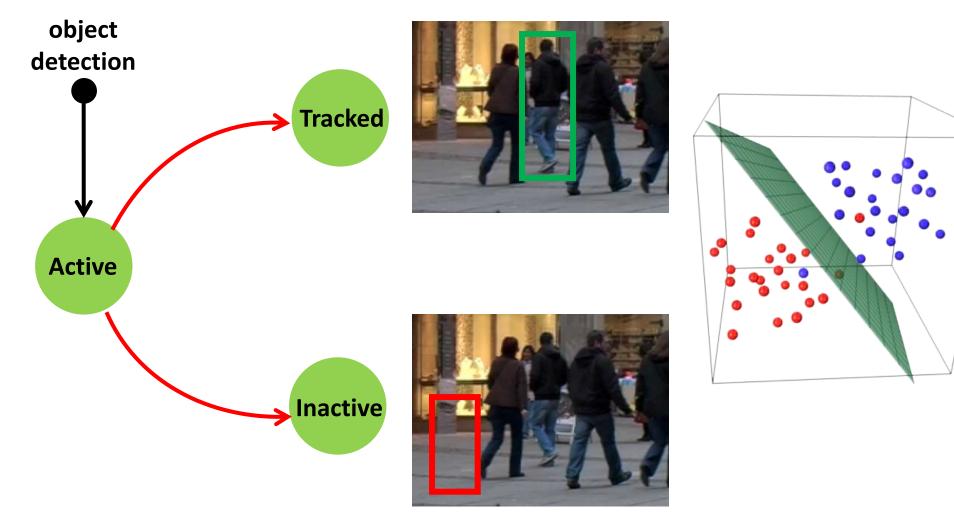
Experiments

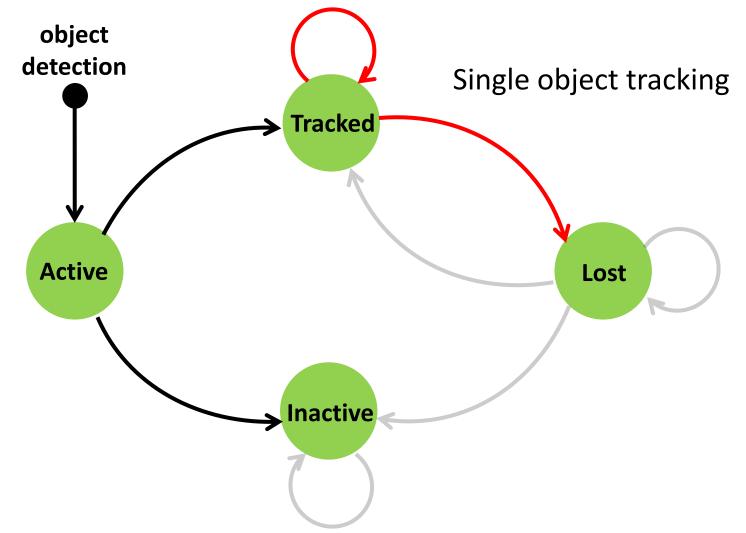
Conclusion



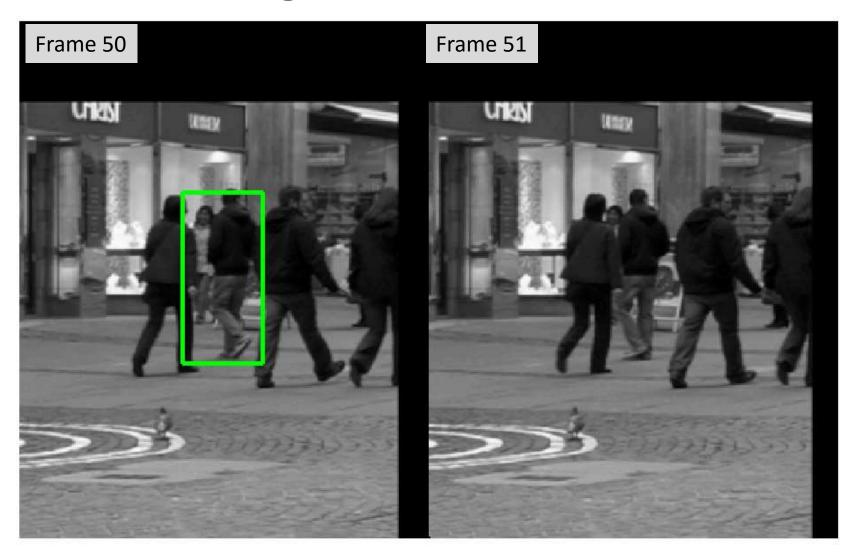




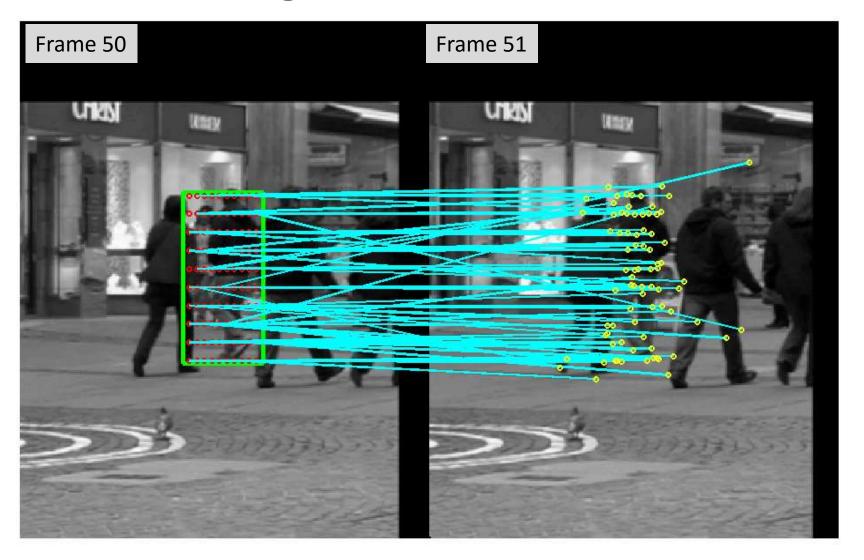


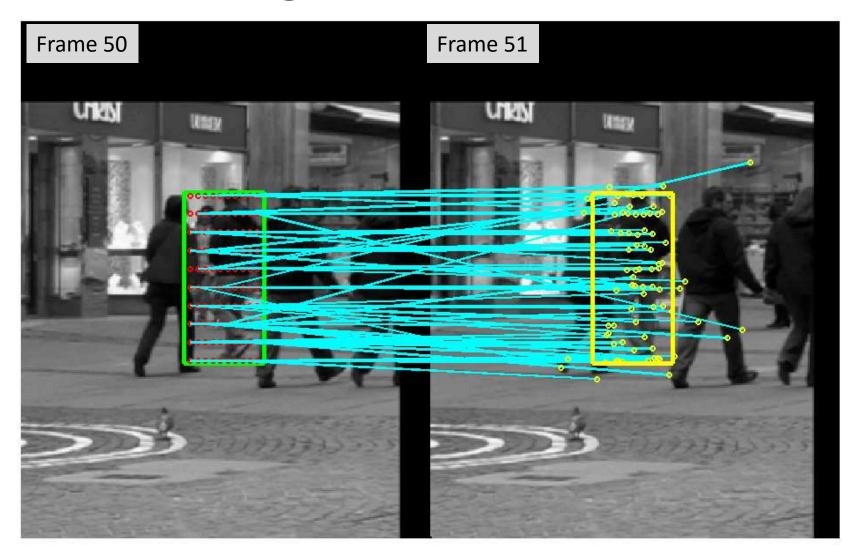


TLD Tracker. Z. Kalal, K. Mikolajczyk, and J. Matas. Tracking-learning-detection. TPAMI, 34(7):1409–1422, 2012.<sup>23</sup>





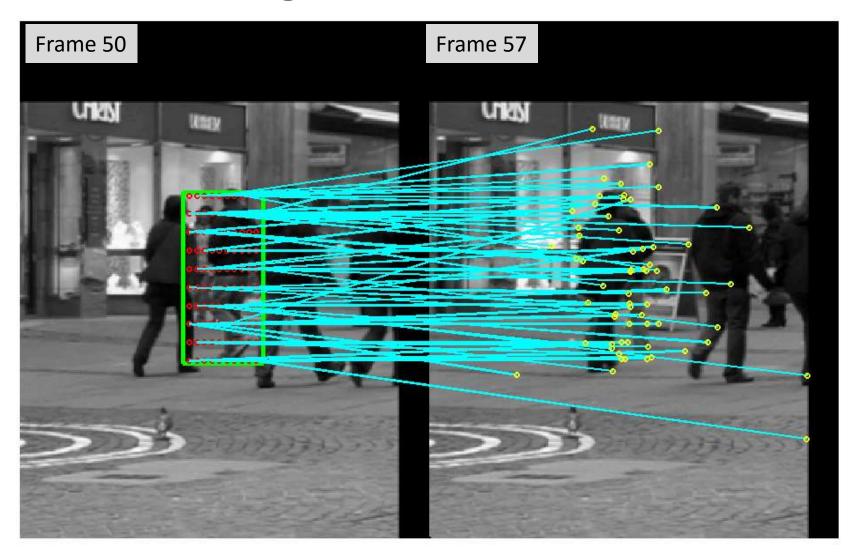


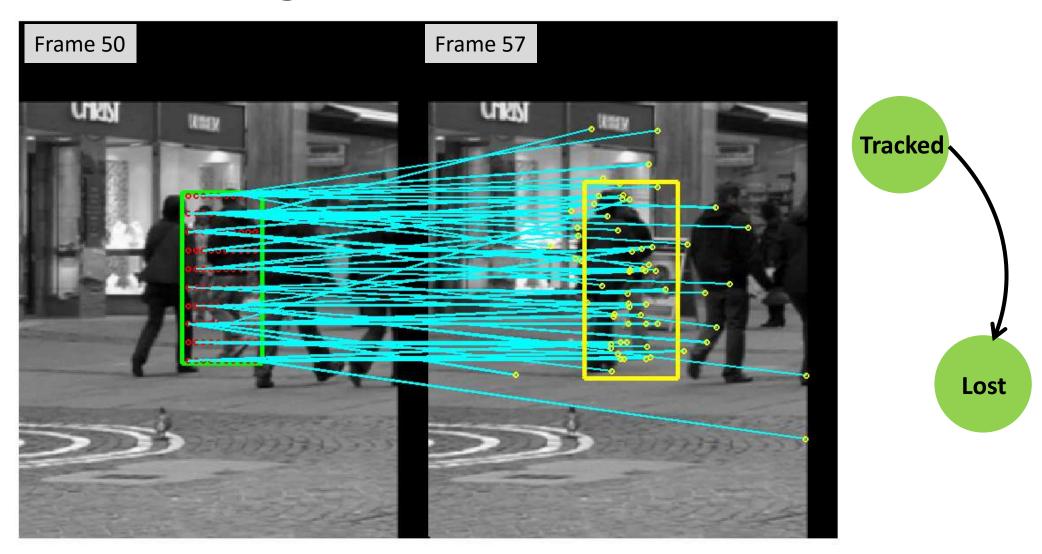


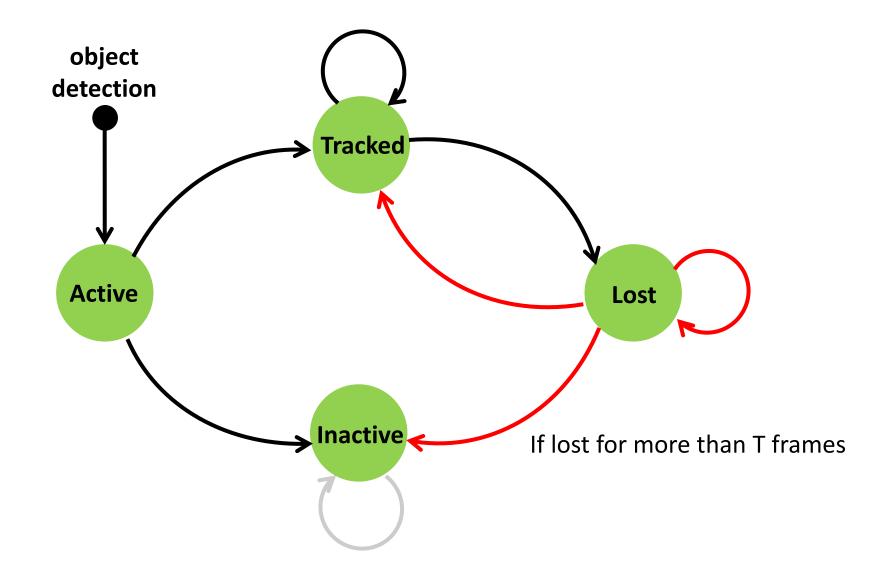




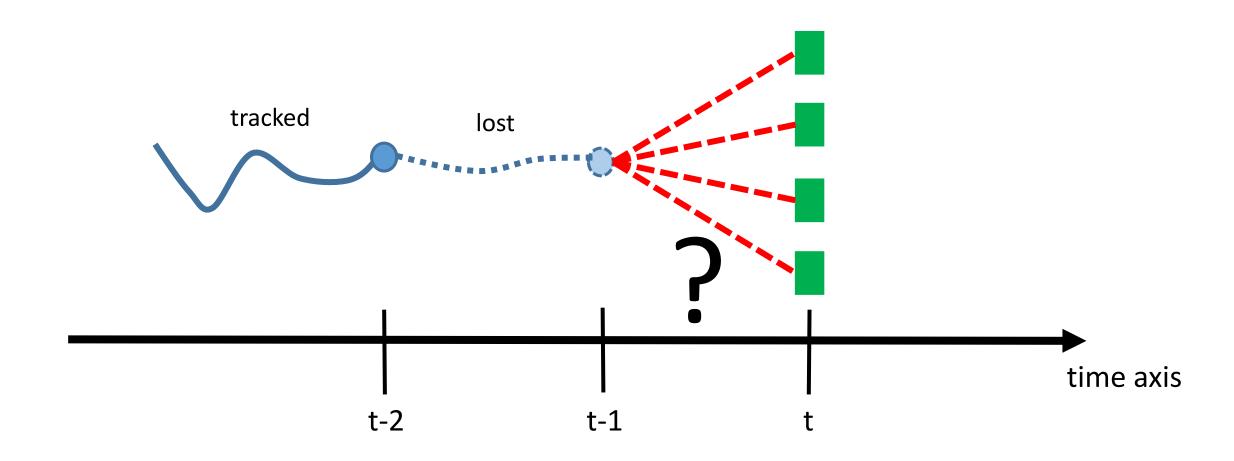








#### Data Association in Lost States

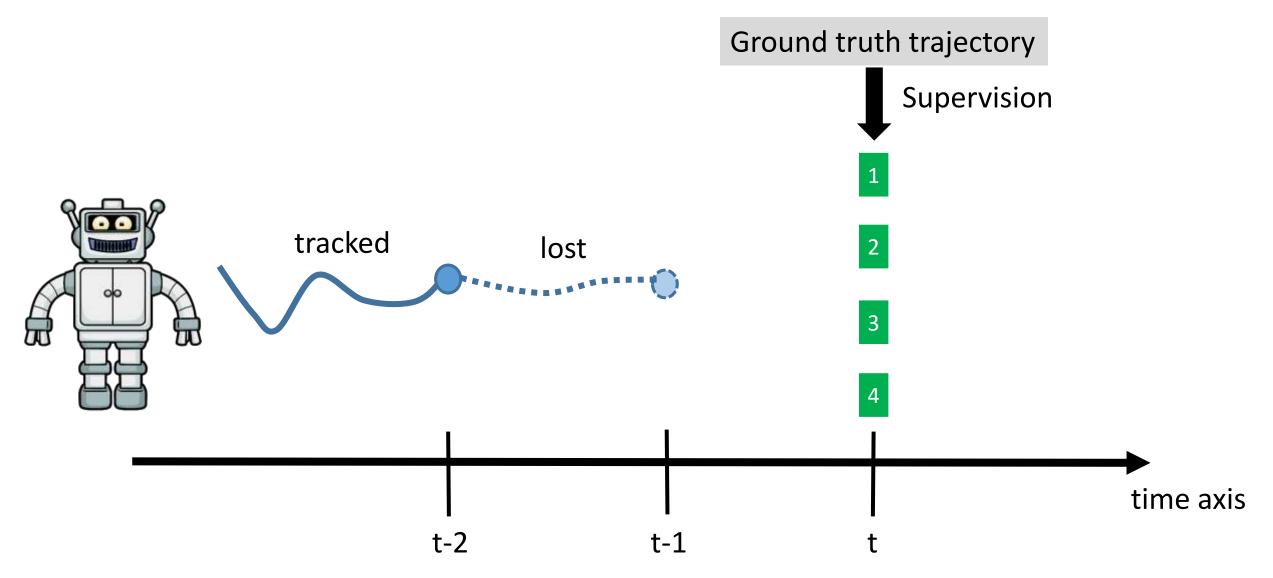


#### Learning the Similarity Function

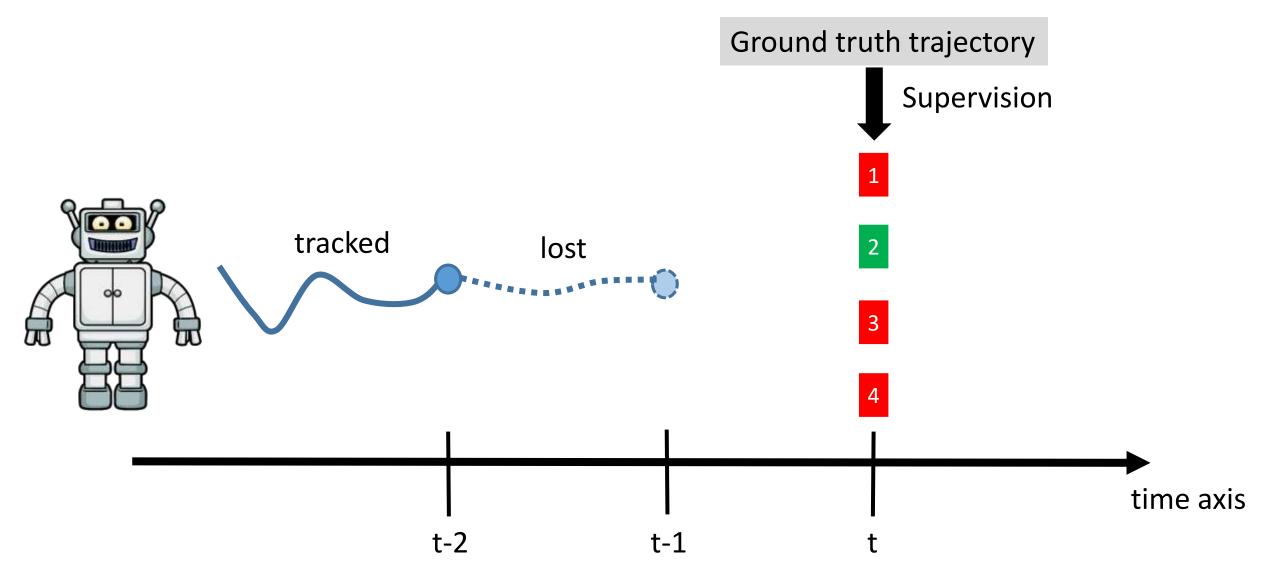
Similarity = 
$$w_1\phi_1(\mathbf{w}, \mathbf{l}) + \cdots + w_n\phi_n(\mathbf{w}, \mathbf{l}) + b$$

Inverse reinforcement learning: tracking objects in training videos!

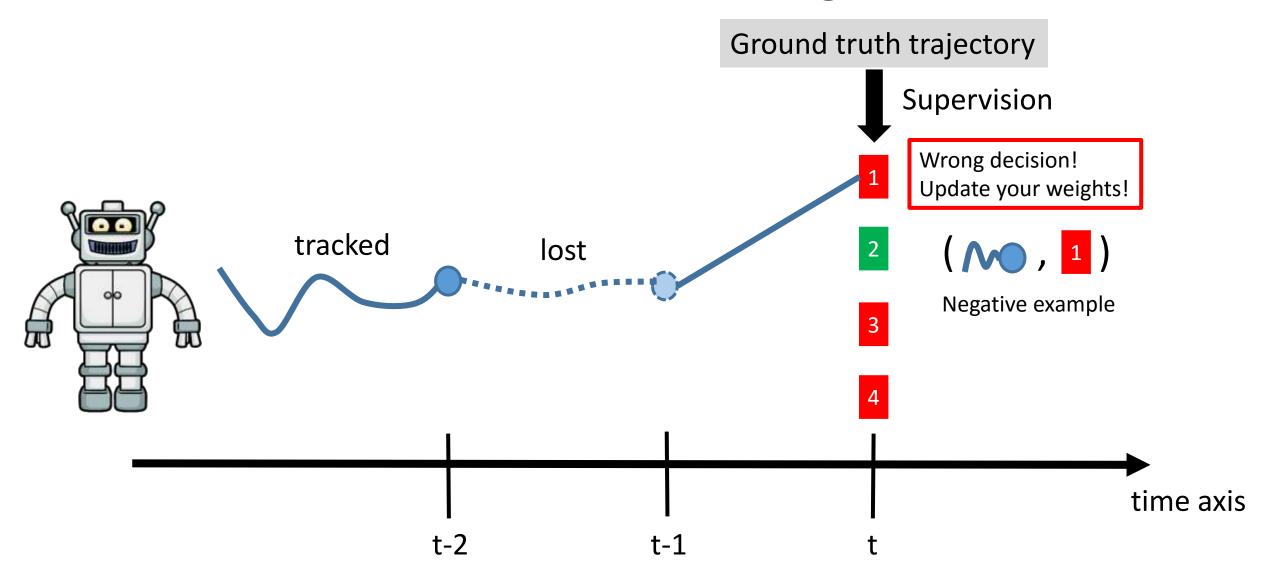
#### Inverse Reinforcement Learning



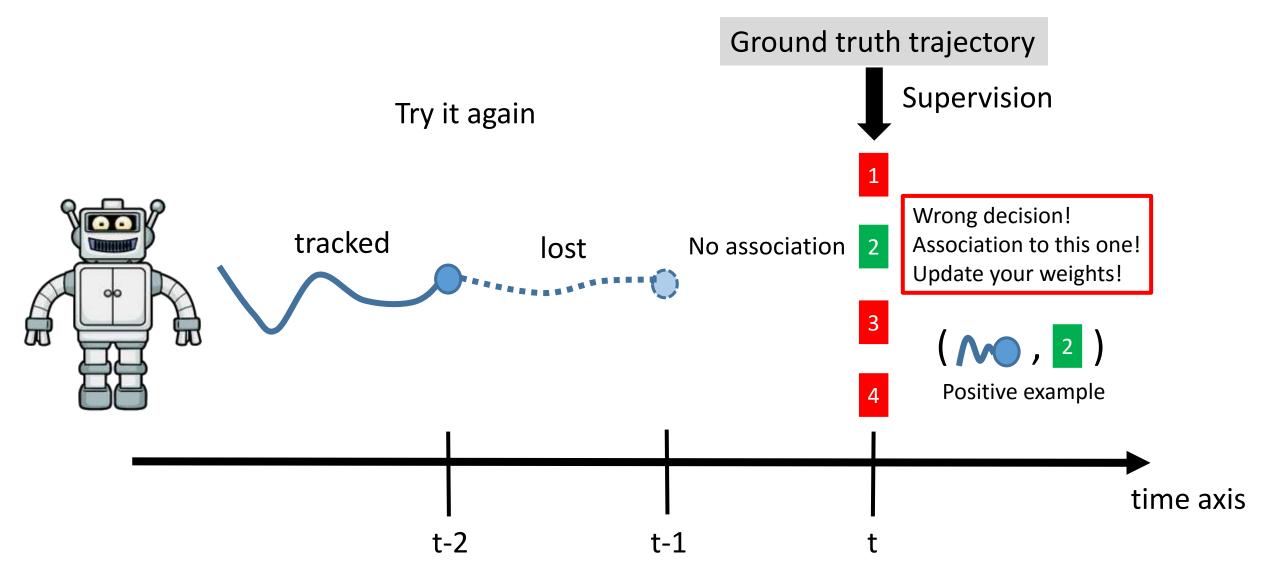
#### Inverse Reinforcement Learning



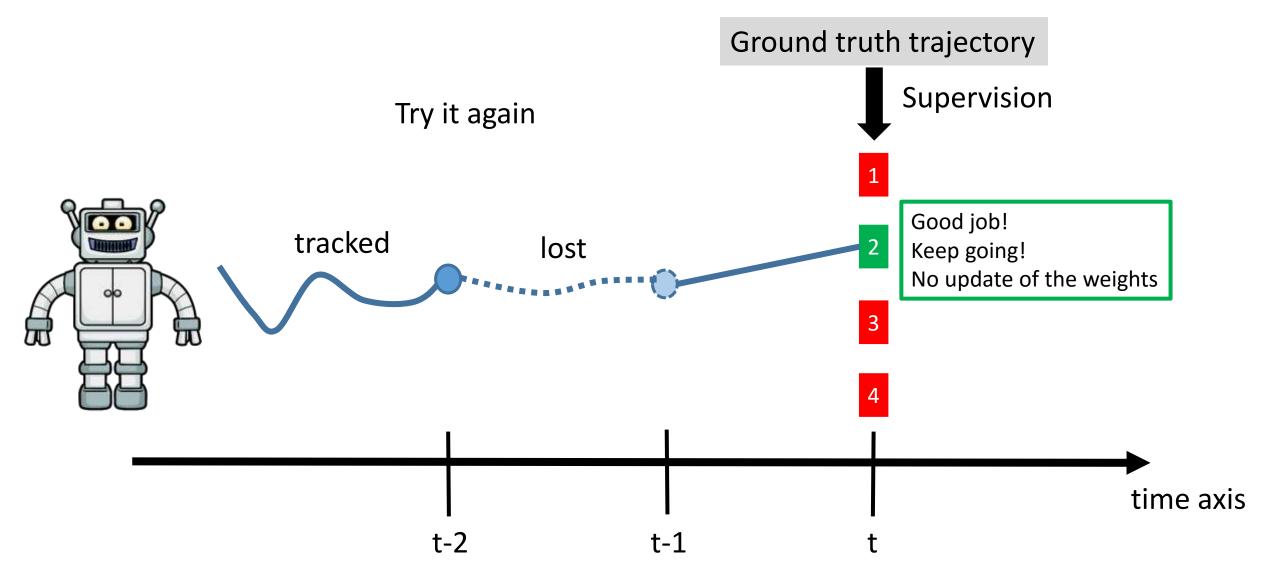
#### Inverse Reinforcement Learning



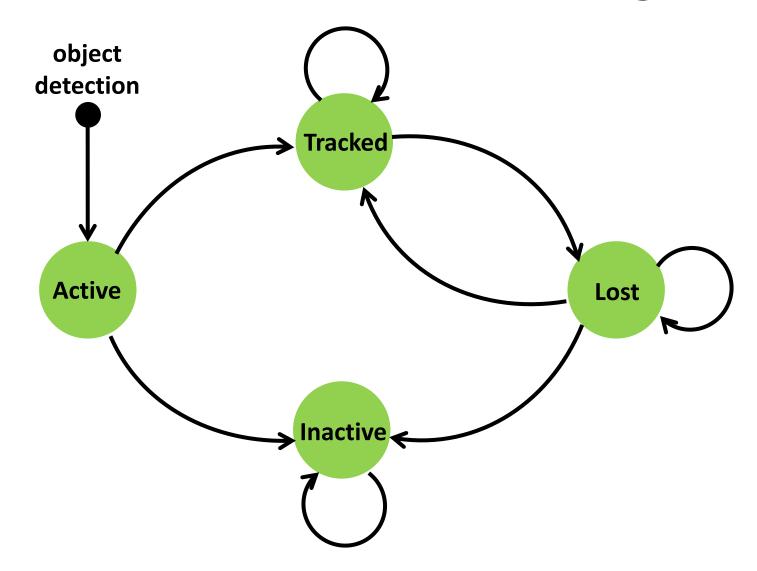
#### Inverse Reinforcement Learning



#### Inverse Reinforcement Learning



#### Markov Decision Process for a Single Target



#### Outline

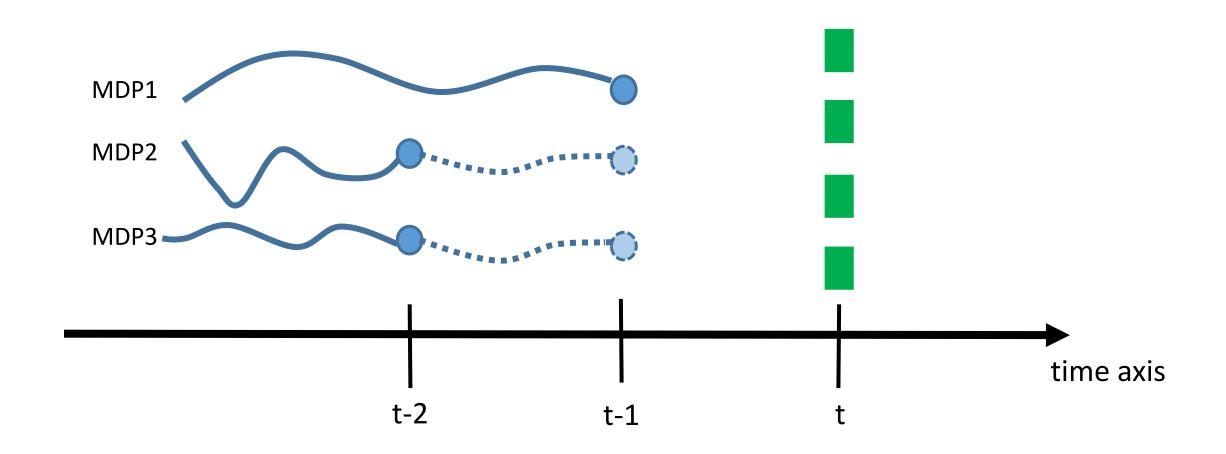
Markov Decision Process (MDP) for a Single Target

Online Multi-Object Tracking with MDPs

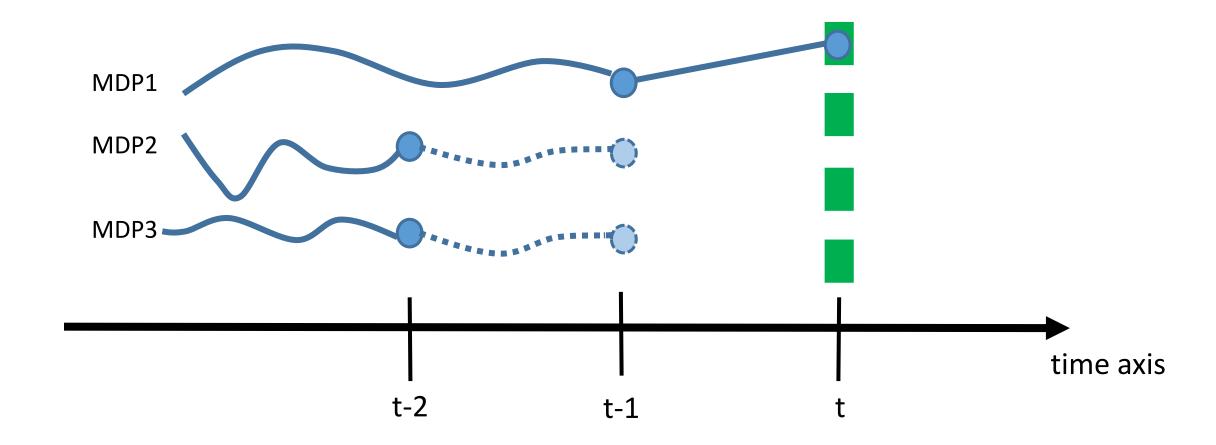
Experiments

Conclusion

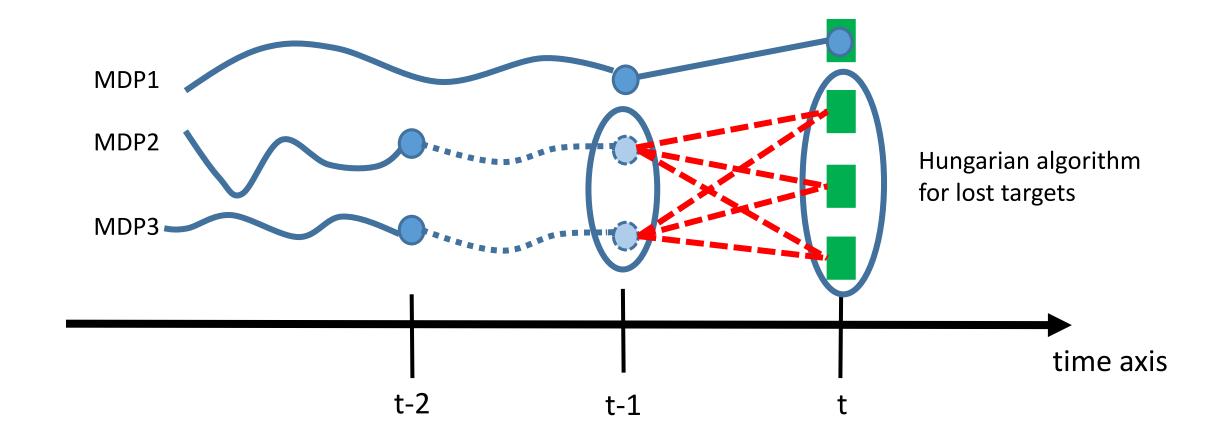
### Ensemble MDPs for Online Multi-Object Tracking



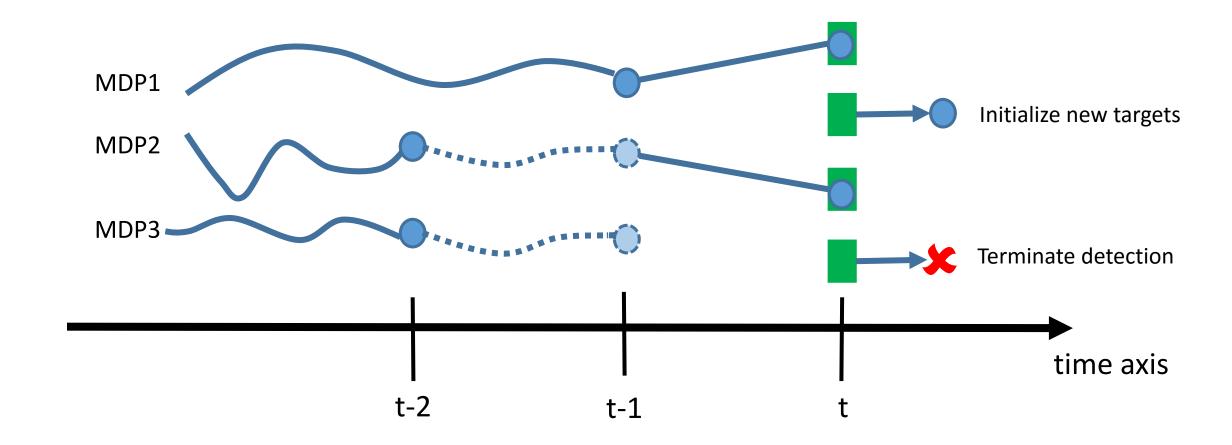
## Step 1: Process tracked targets



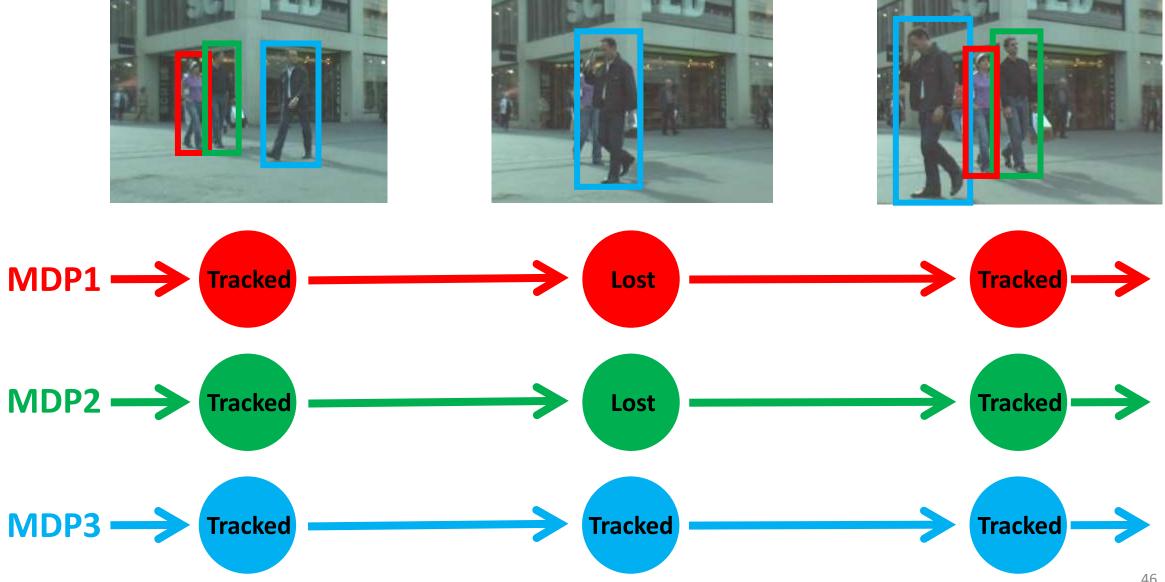
#### Step 2: Process lost targets



#### Step 3: Initialize new targets



#### Online Multi-Object Tracking with MDPs



#### Outline

Markov Decision Process (MDP) for a Single Target

Online Multi-Object Tracking with MDPs

Experiments

Conclusion

#### Experiments: Dataset

- Multiple Object Tracking Benchmark [1]
  - 11 training sequences
  - 11 test sequences
  - Object detections from the ACF detector [2]



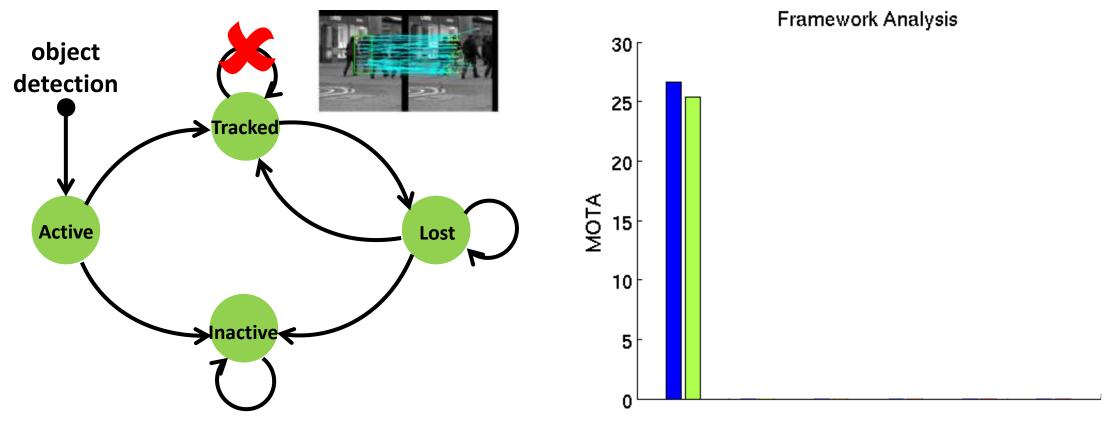






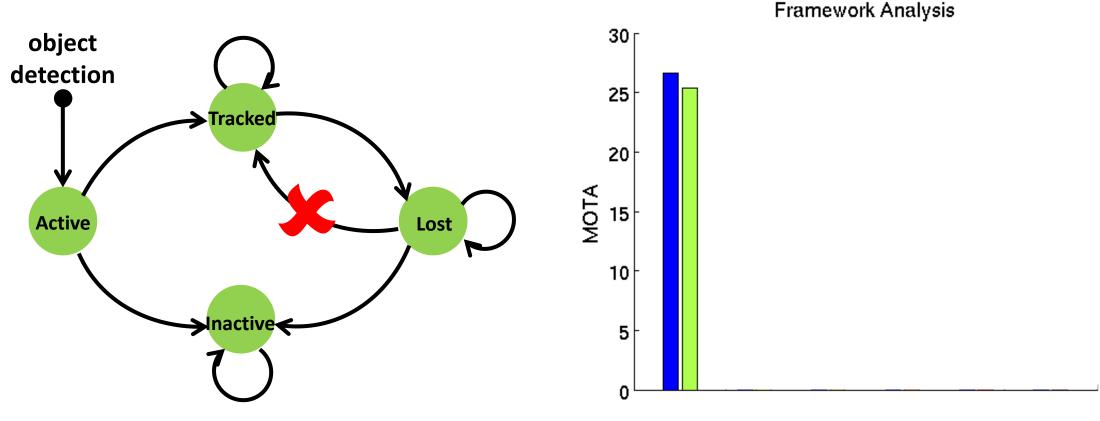
<sup>[1]</sup> L. Leal-Taixé, A. Milan, I. Reid, S. Roth, and K. Schindler. MOTChallenge 2015: Towards a Benchmark for Multi-Target Tracking. arXiv:1504.01942 [cs], 2015.

Contribution of different components

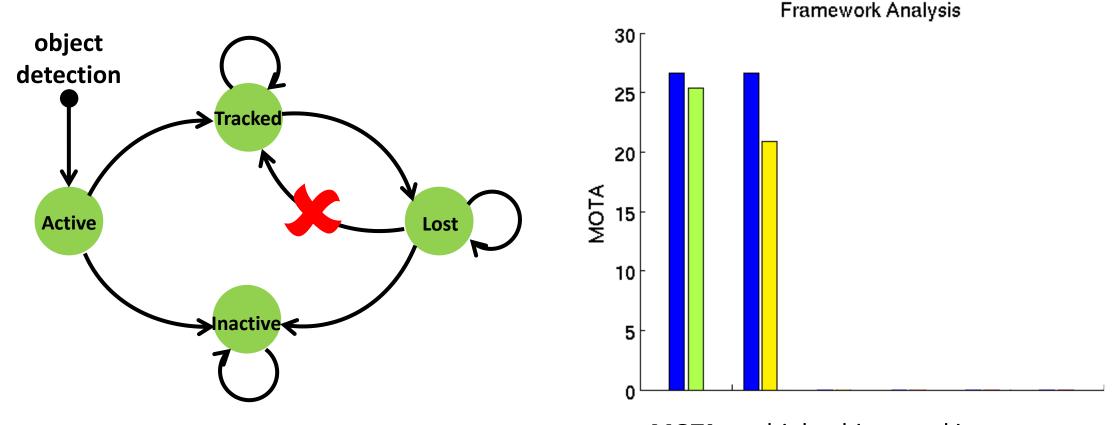


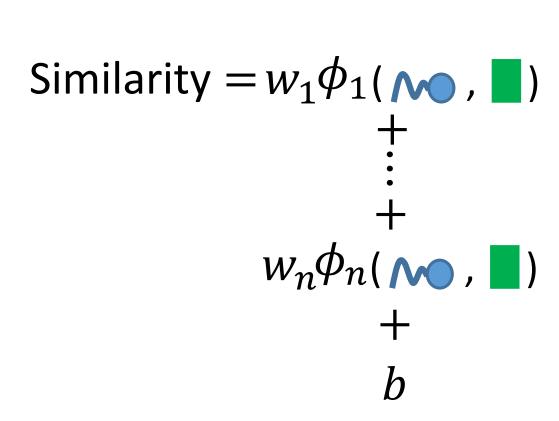
**MOTA**: multiple object tracking accuracy

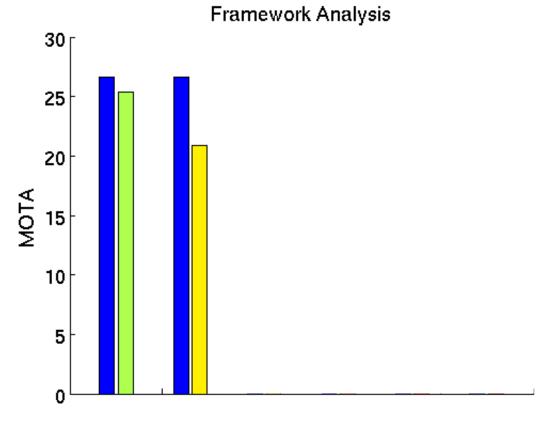
Contribution of different components

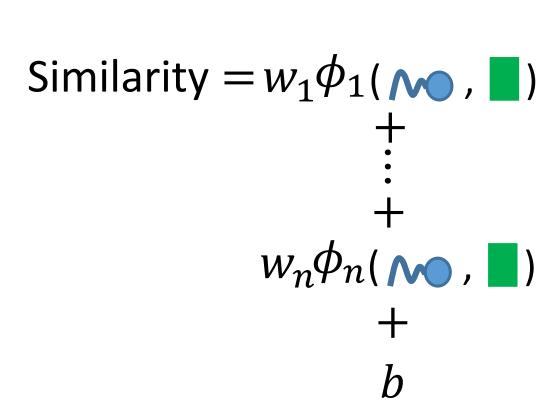


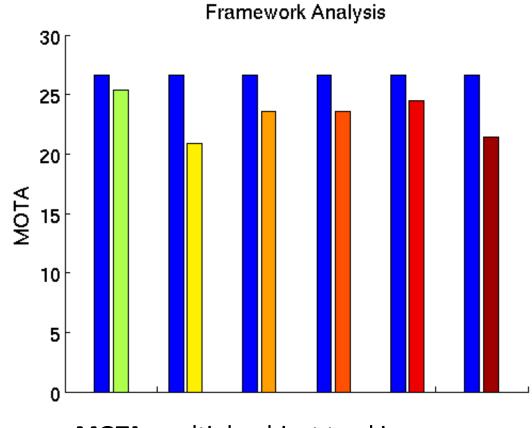
**MOTA**: multiple object tracking accuracy



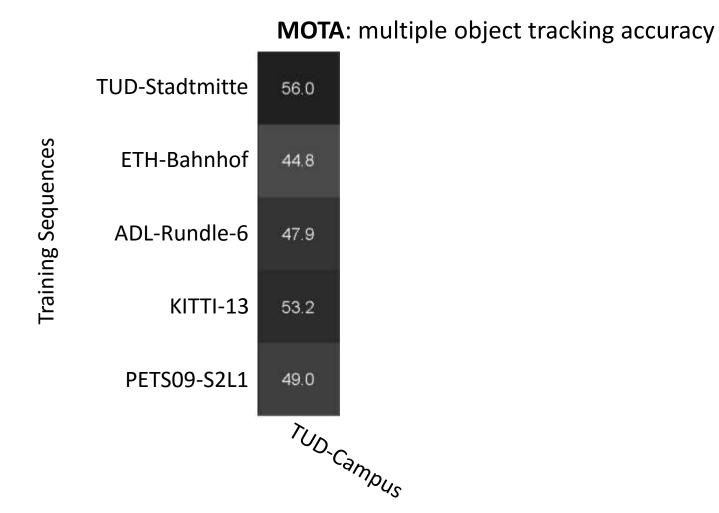




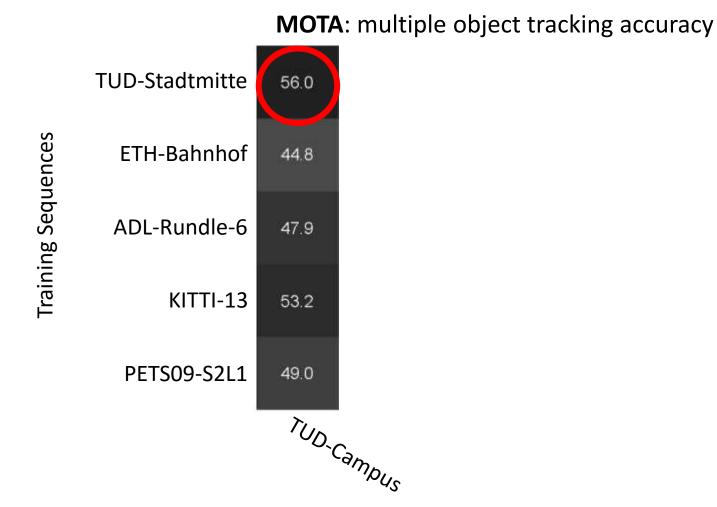




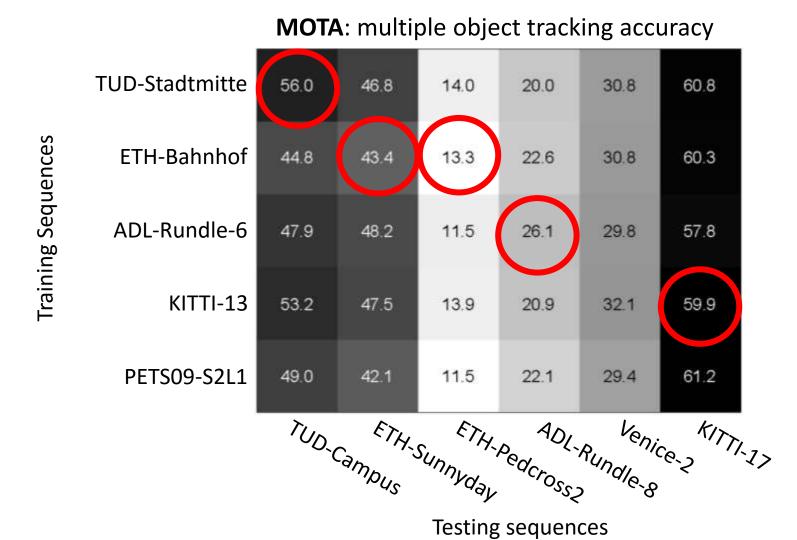
Cross-domain tracking



Cross-domain tracking



Cross-domain tracking



#### Experiments: Evaluation on Test Set

Tracker	Tracking	Learning	MOTA
DP_NMS [1]	Batch	N/A	14.5
TC_ODAL [2]	Online	Online	15.1
TBD [3]	Batch	Offline	15.9
SMOT [4]	Batch	N/A	18.2
RMOT [5]	Online	N/A	18.6
CEM [6]	Online	N/A	19.3
SegTrack [7]	Batch	Offline	22.5
MotiCon [8]	Batch	Offline	23.1
MDP (Ours)	Online	Online	30.3

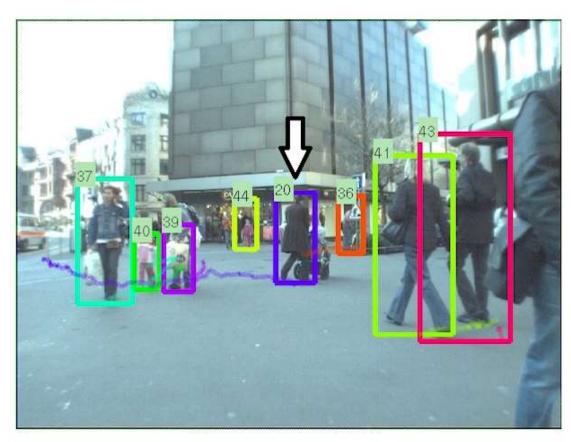
**MOTA**: multiple object tracking accuracy

- [1] Pirsiavash et al., CVPR' 11
- [2] Bae et al., CVPR'14
- [3] Geiger et al., TPAMI'14

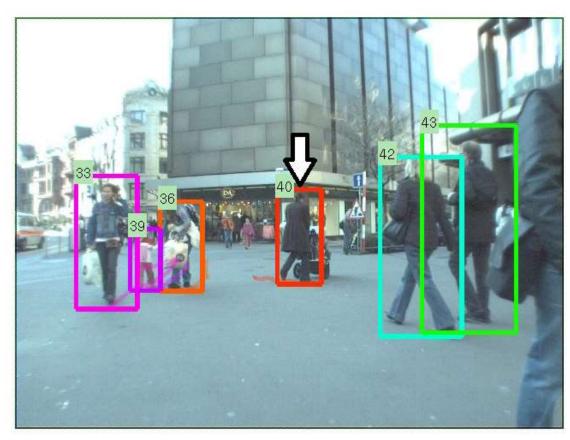
- [4] Dicle et al., ICCV'13
- [5] Yoon et al., WACV'15
- [6] Milan et al., TPAMI'14

- [7] Milan et al., CVPR'15
- [8] Leal-Taixé et al., CVPR'14

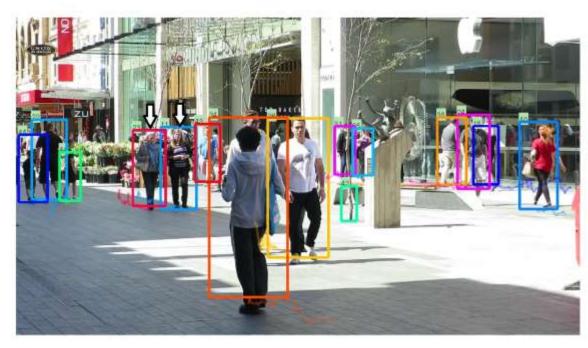
# Tracking Results



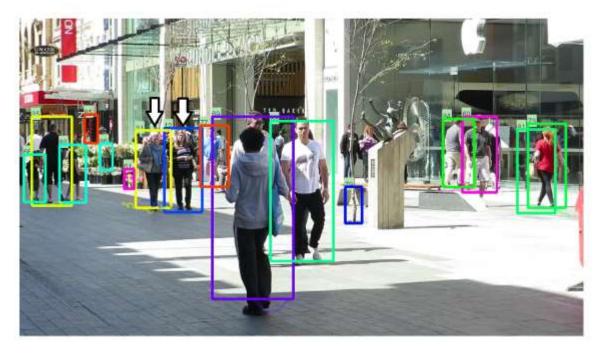
MDP [Ours]



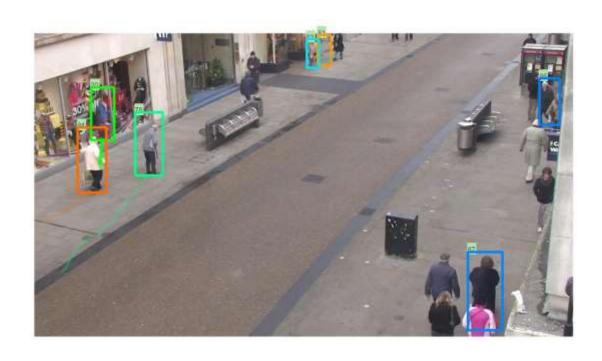
MotiCon [Leal-Taixé et al., CVPR'14]



MDP [Ours]



MotiCon [Leal-Taixé et al., CVPR'14]



MDP [Ours]



MotiCon [Leal-Taixé et al., CVPR'14]

#### Outline

Markov Decision Process (MDP) for a Single Target

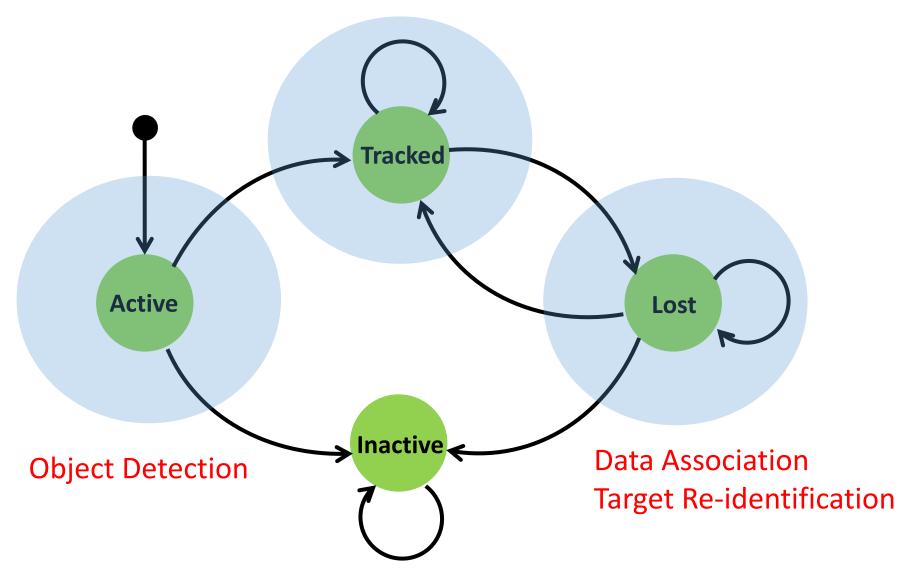
Online Multi-Object Tracking with MDPs

Experiments

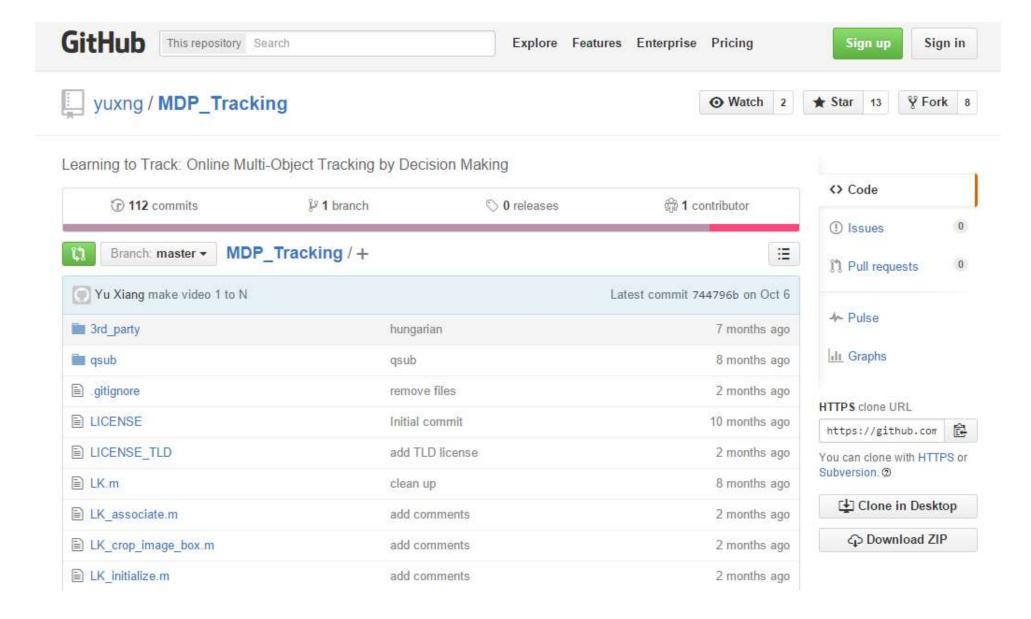
Conclusion

#### Conclusion

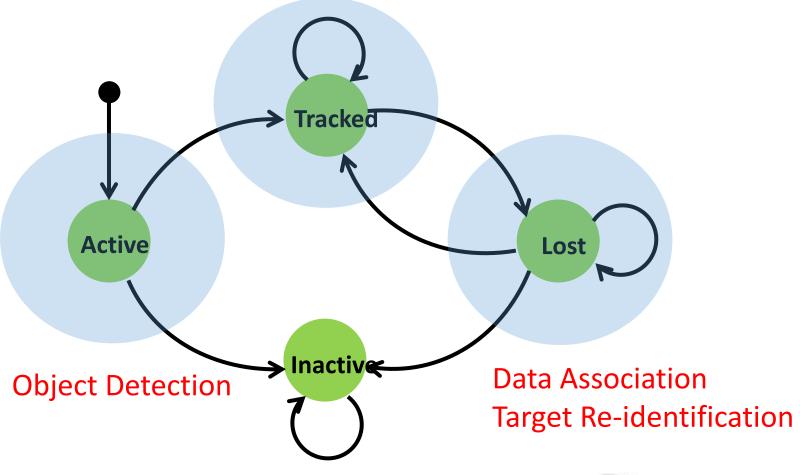
#### **Single Object Tracking**



#### Code



#### **Single Object Tracking**



# Thank you!

