

# Privacy Preserving Multi-target Tracking



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# Visual People Tracking

## Applications and Benefits

- ✓ CCTV: Increased safety
- ✓ Automated video analysis
- ✓ Crowd motion estimation
- ✓ Robotic navigation



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- ✓ CCTV: Increased safety
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**Drawback:**  
**Heavy intrusion of privacy**

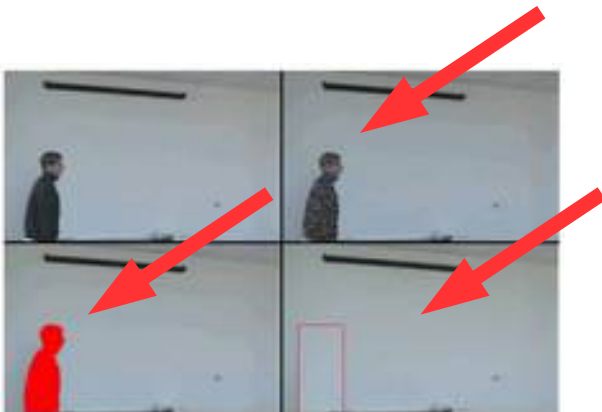
# Existing Alternatives



[Schiff et al., 2009]



[Spindler et al., 2006]



[Wickramasuriya et al., 2005]

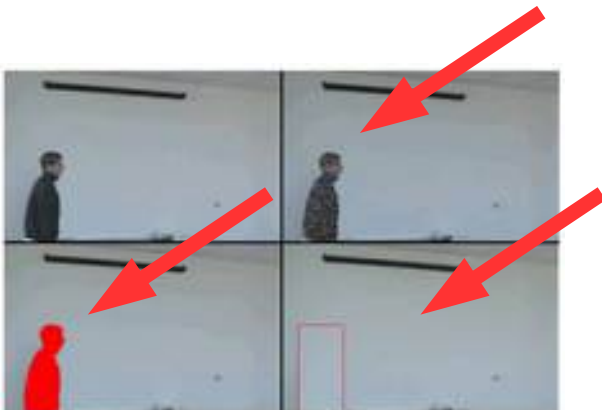
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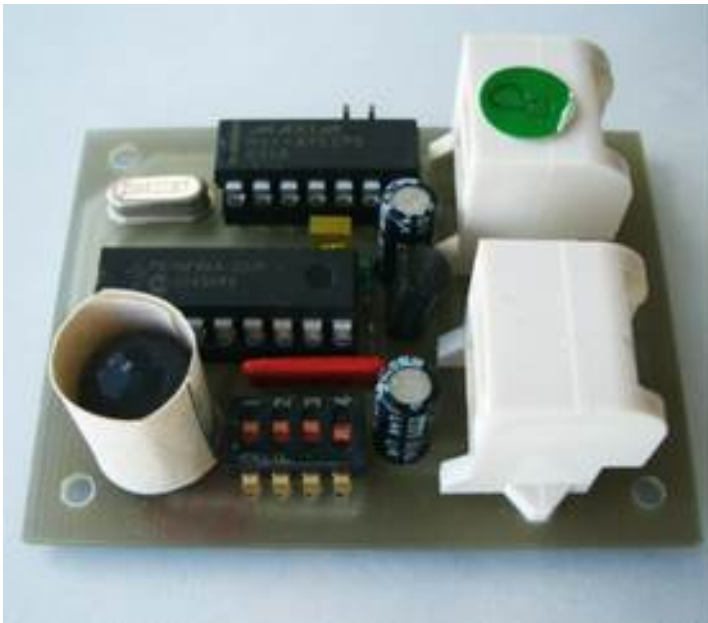


[Wickramasuriya et al., 2005]

Such systems may fail (or be switched off)

# Our Approach

- A different sensor modality
- Existing multi-target tracking techniques



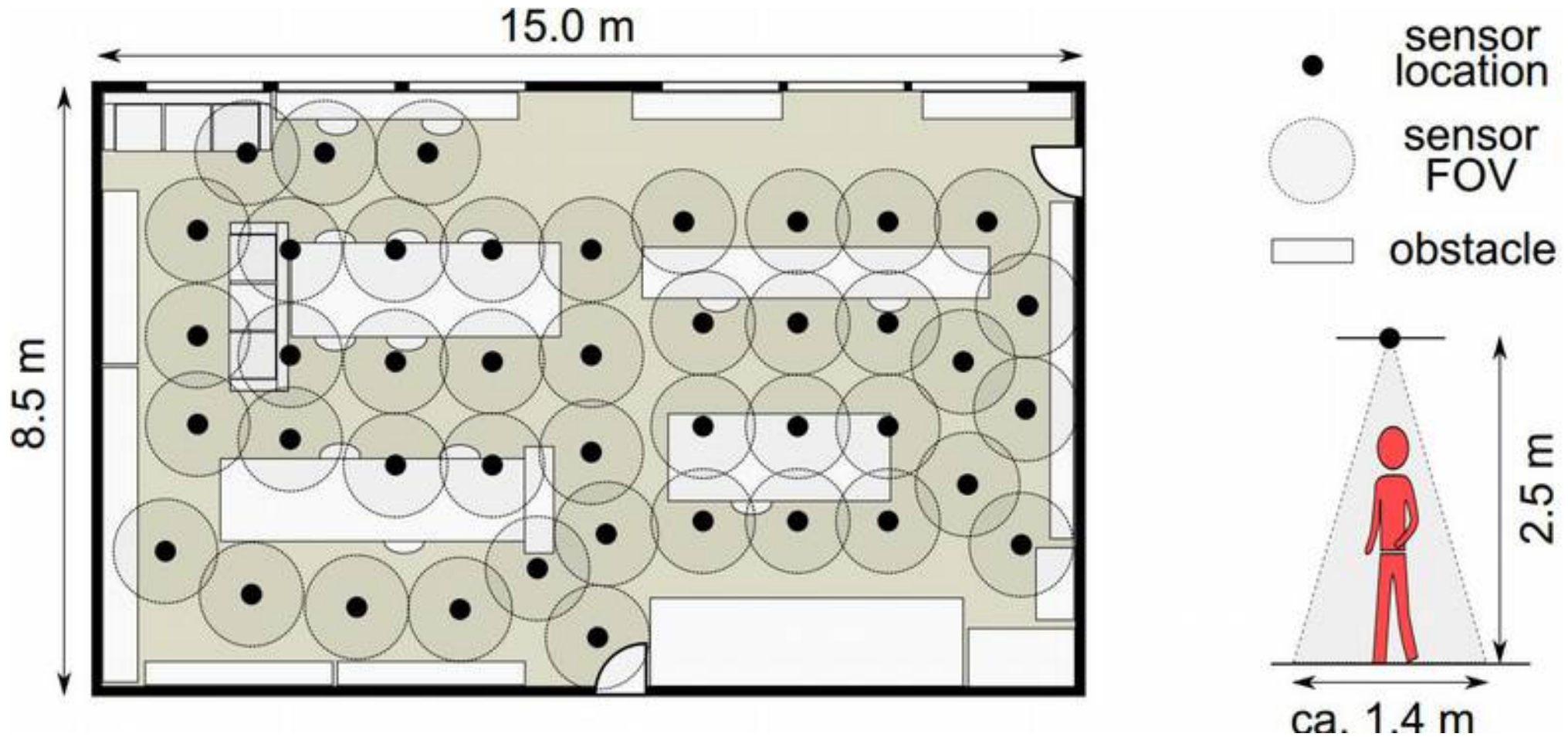
Pyroelectric infrared sensors\* ...



...mounted on a ceiling

\*Also known as: Infrared motion sensors

# The Setup



43 nodes, ca. 3m stride. Total cost:  $\approx$  \$100 USD.

# Tracking with Infrared Sensors

A mostly unexplored research area!

[Luo et al., 2009]

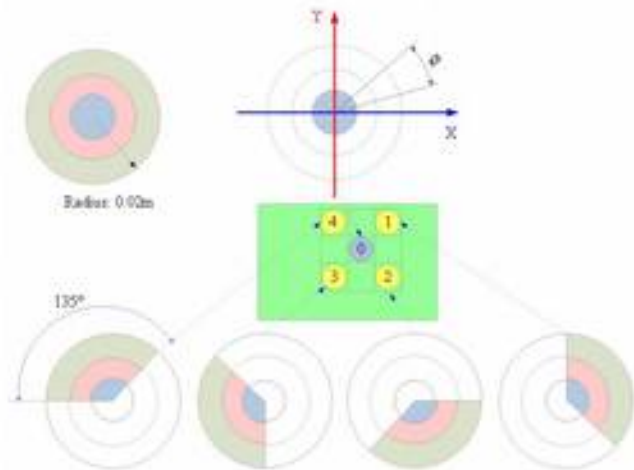
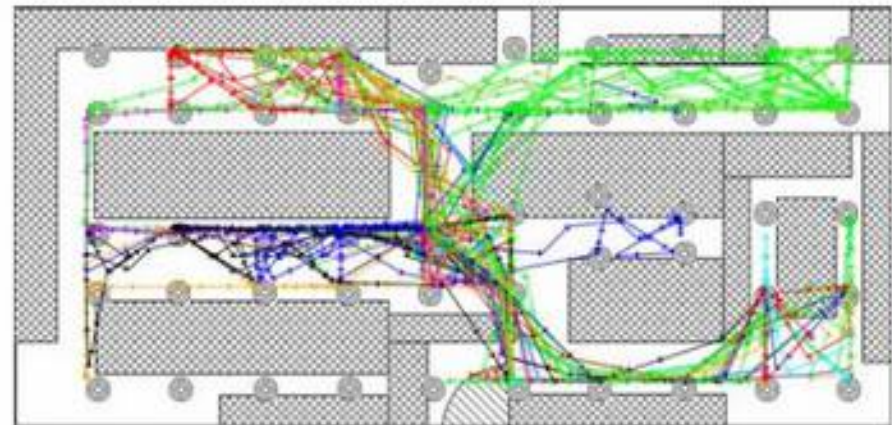


Fig. 2. Sensor Module

[Hosokawa et al., 2009]



- Expensive sensor array with Fresnel lenses

- Limited state space
- Ad hoc algorithm for data association



# Benefits

- Individual identification impossible
  - Respects privacy
- Insensitive to lighting conditions
- Low cost

# Limitations

- Indoor application only
- Less flexible

# Main Challenges

- Extremely low resolution (43 sensors)
- A binary response at 2 Hz per sensor
- No visual (appearance) information
- Poor localization + sensor noise / delay
- Activation by several people
- Multiple measurements by one person

# Main Challenges

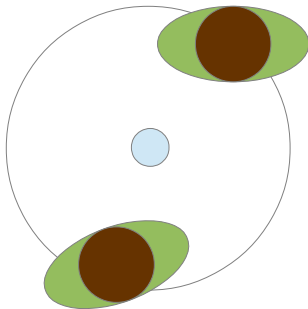
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→ Activation by multiple people

- Multiple measurements by one person

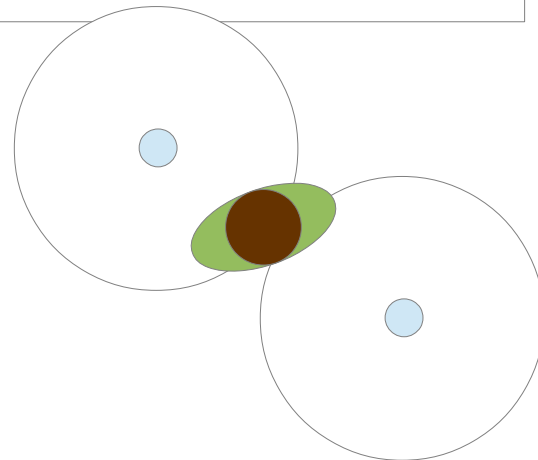
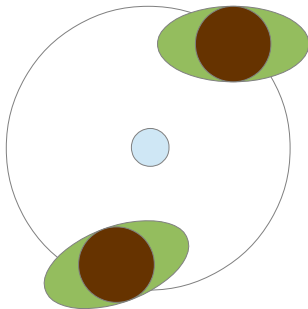


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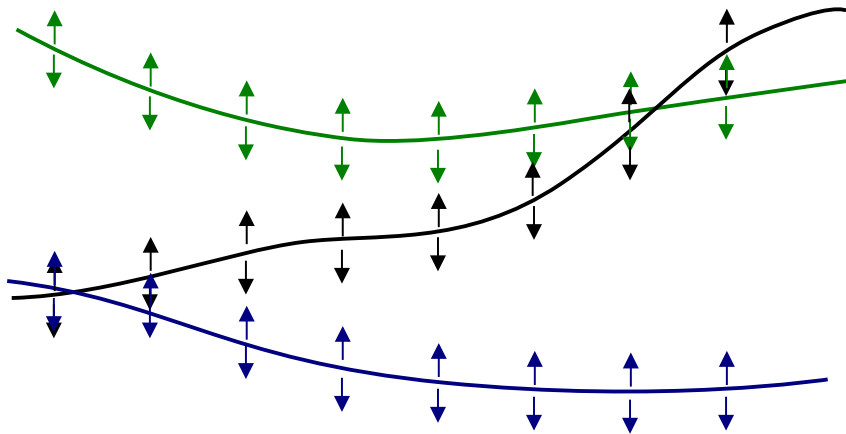
➔ Multiple measurements by one person



# Continuous Energy Minimization

$$E(\mathbf{X}) = E_{\text{obs}} + E_{\text{dyn}} + E_{\text{exc}} + E_{\text{per}} + E_{\text{reg}}$$

State vector:  $X, Y$ -locations of **all** targets at **all** frames  $\mathbf{X} \in \mathbb{R}^d, d \approx 2000$



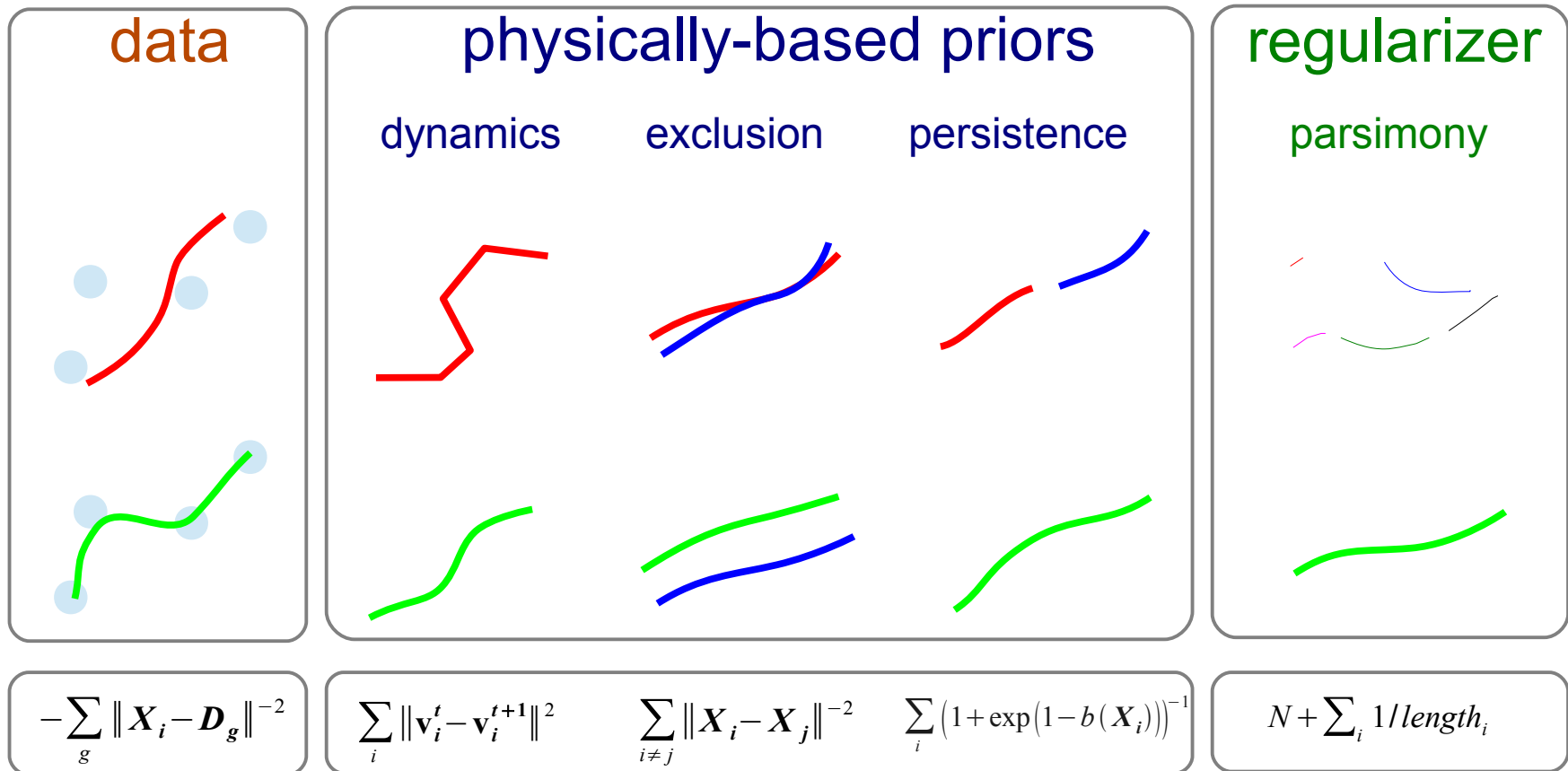
[Milan et al., PAMI 2014]

# Why Continuous Energy?

- Continuous trajectories
  - low sensor resolution not an issue
- No implicit data association
  - multiple assignments possible
- Provides best results
  - Measured by standard tracking metrics

# The Energy

$$E = E_{\text{obs}} + aE_{\text{dyn}} + bE_{\text{exc}} + cE_{\text{per}} + dE_{\text{reg}}$$

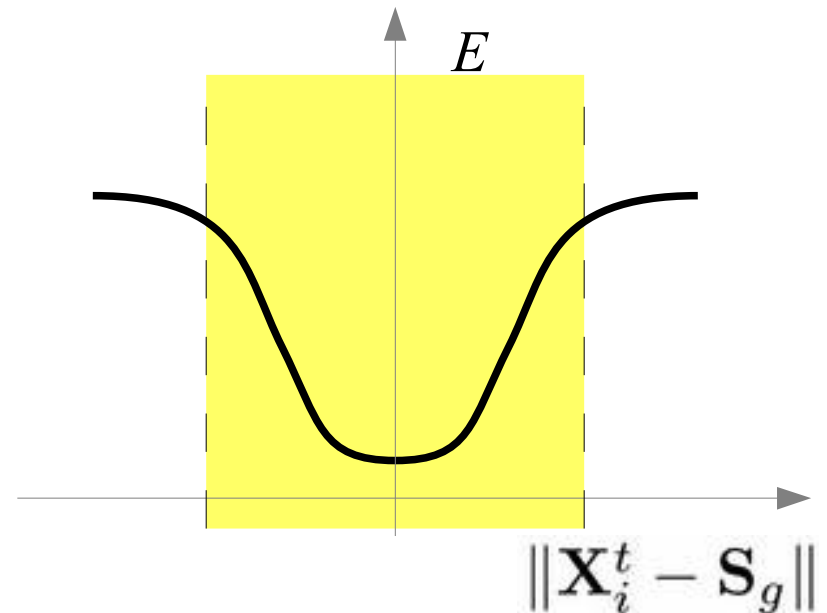
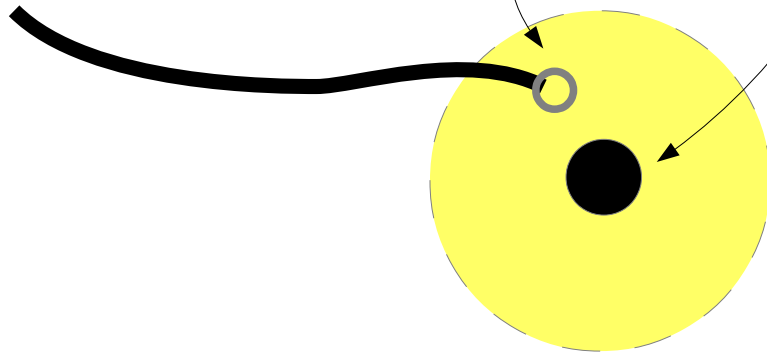




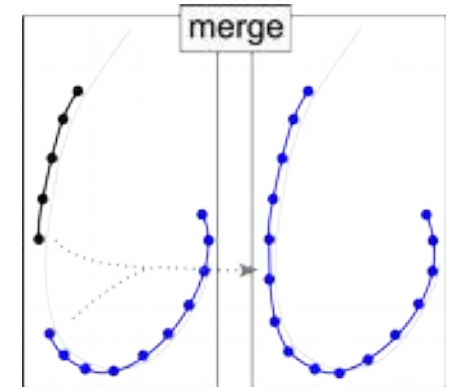
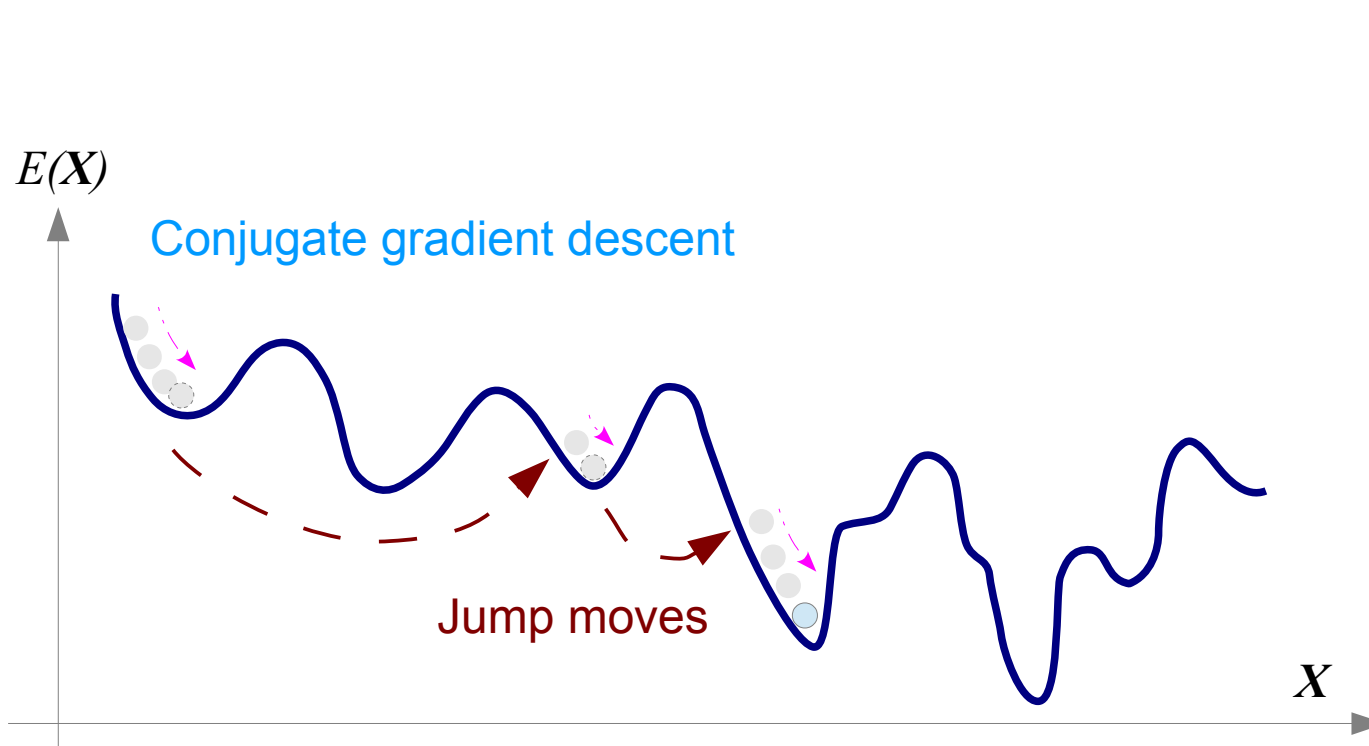
# Data Term

$$E_{\text{det}}(\mathbf{X}) = \sum_{i=1}^N \sum_{t=s_i}^{e_i} \left[ \lambda - \sum_{g \in G(t)} \frac{s^2}{\|\mathbf{X}_i^t - \mathbf{S}_g\|^2 + s^2} \right]$$

lobe size



# Optimization



Merge – Split  
Grow – Shrink  
Add – Remove

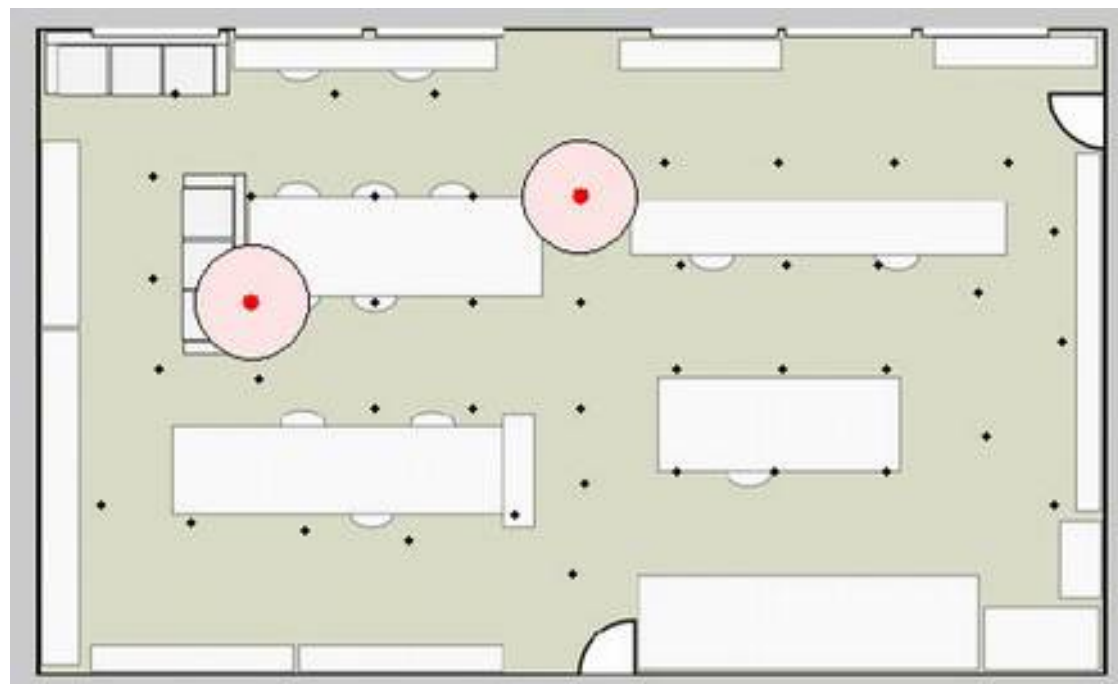
- conjugate gradient descent for **local optimization**
- discontinuous jumps to determine **dimensionality (number of targets)**

# Experiments

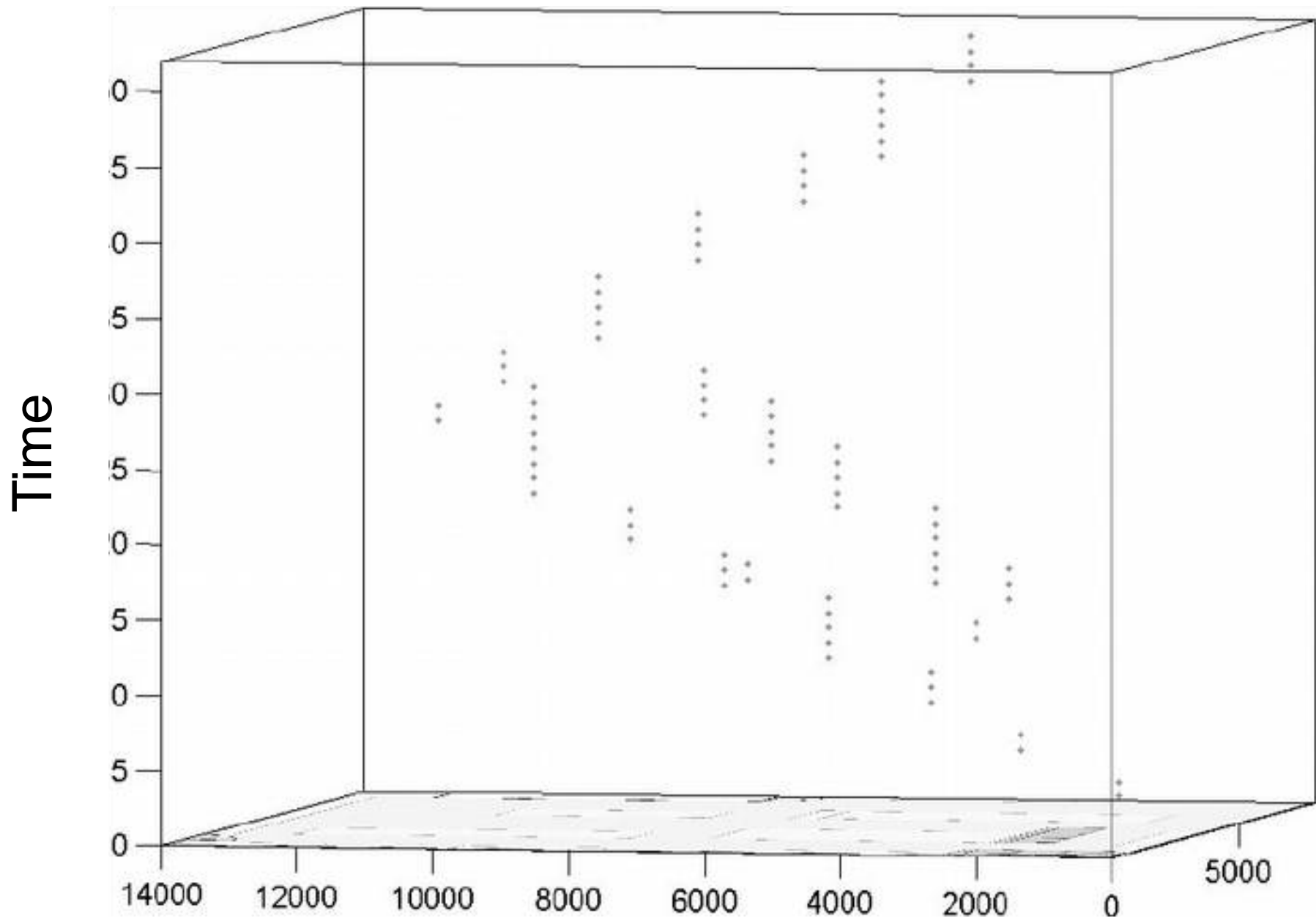
## Synthetic Data

- Manual assignment of keyframes
- Interpolation and sensor simulation

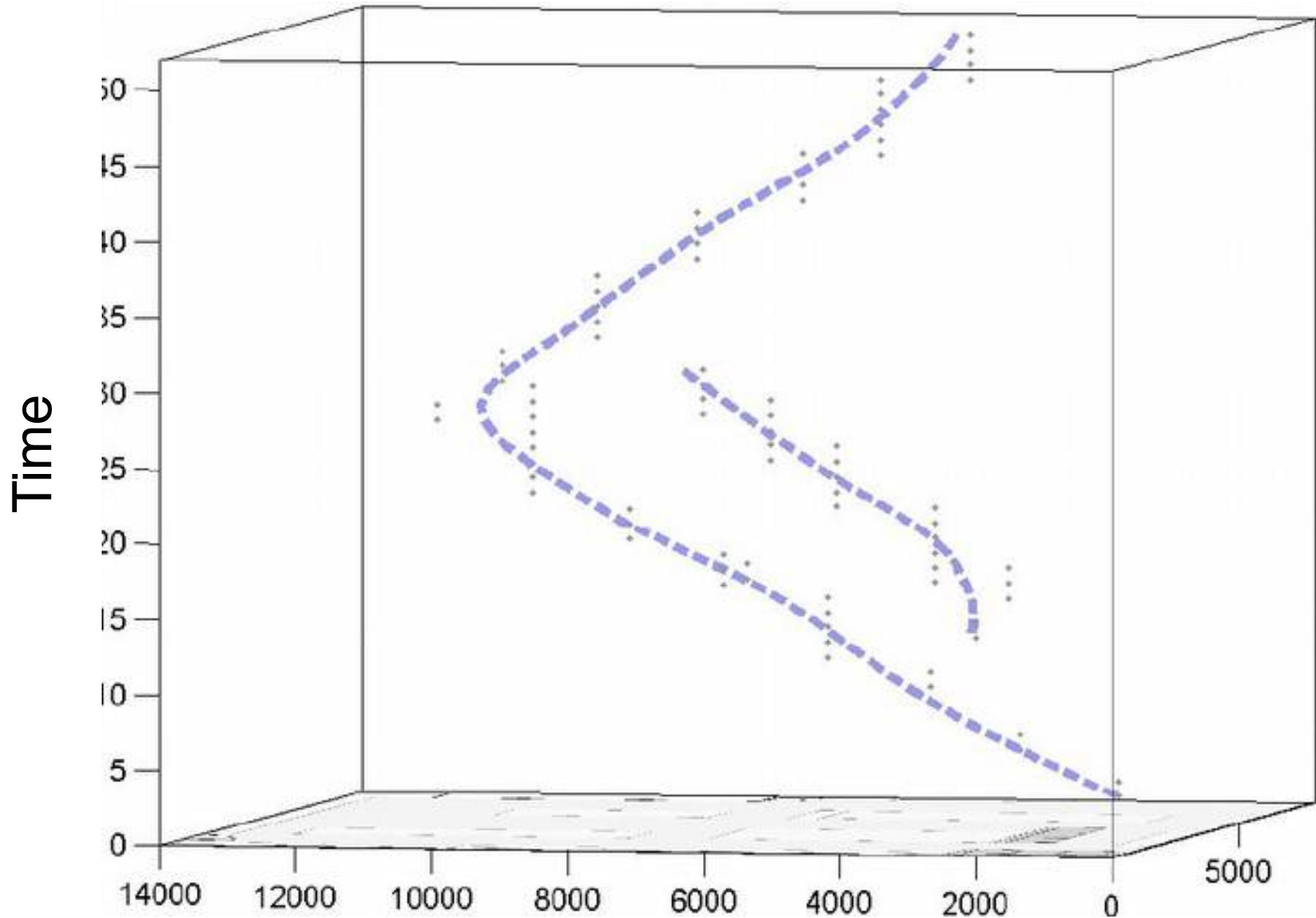
# Measurements



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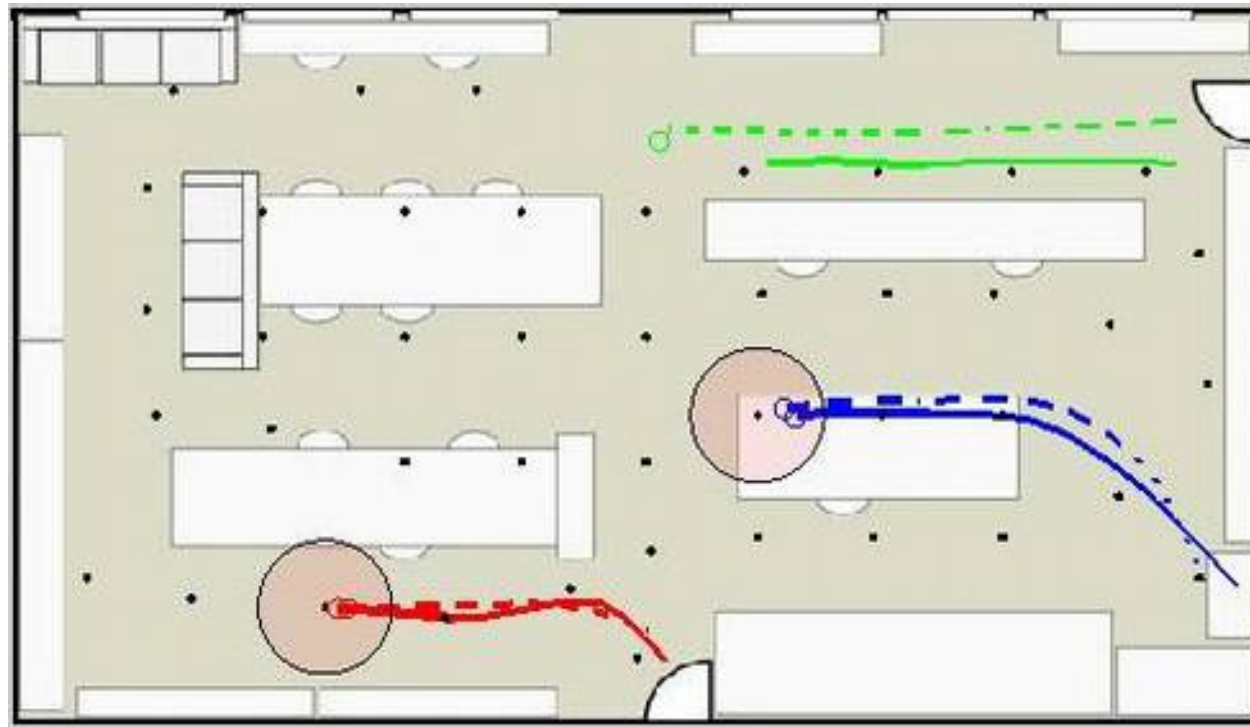
# Ground Truth



# Experiments

## Synthetic Data

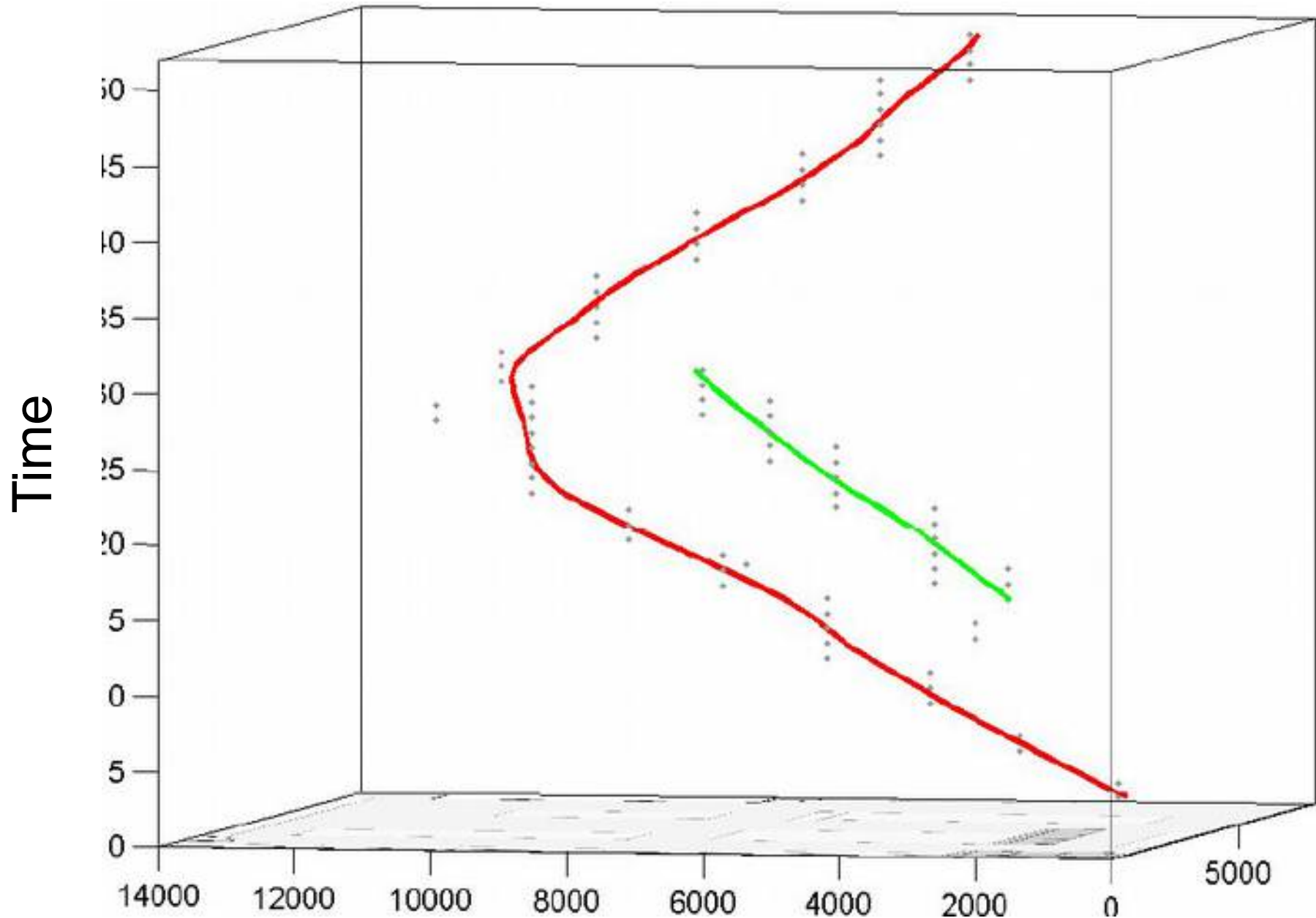
- Manual assignment of keyframes
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————— **Result**

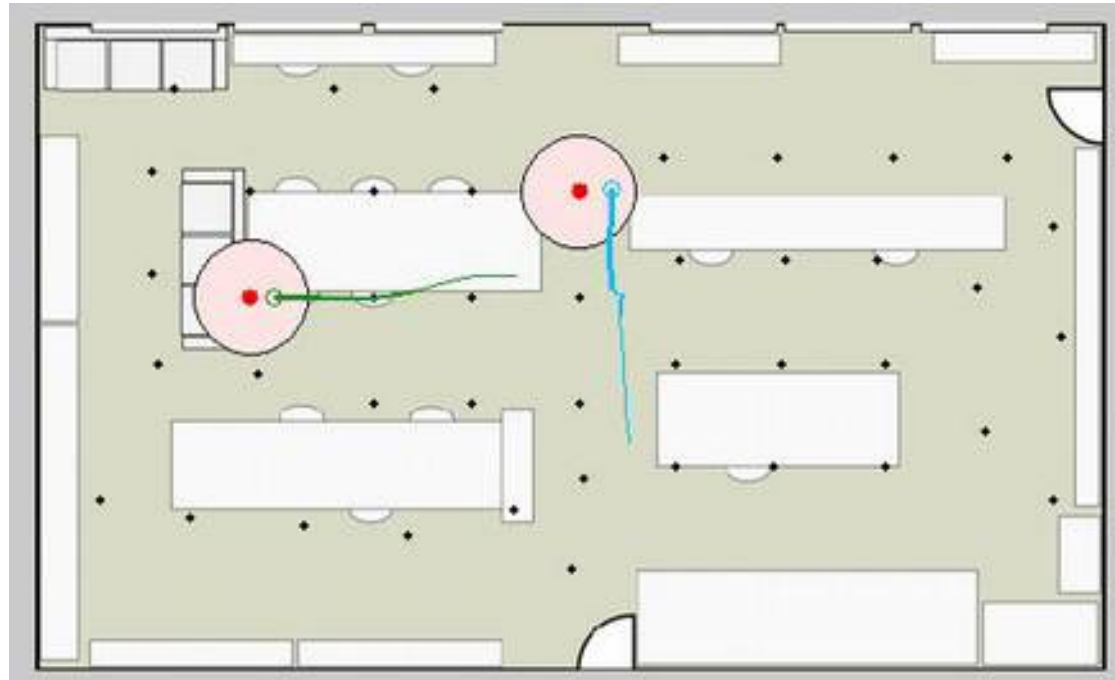
..... **GT**

# Result



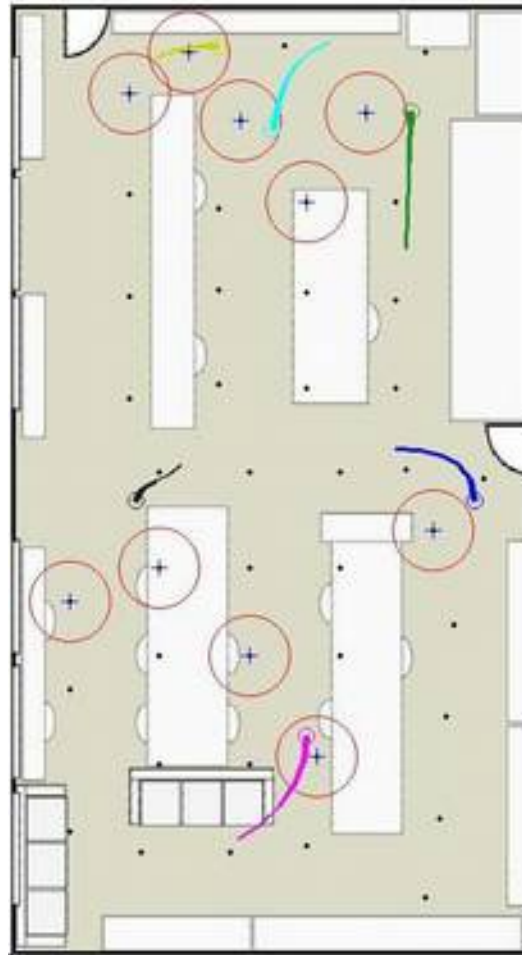


# Result



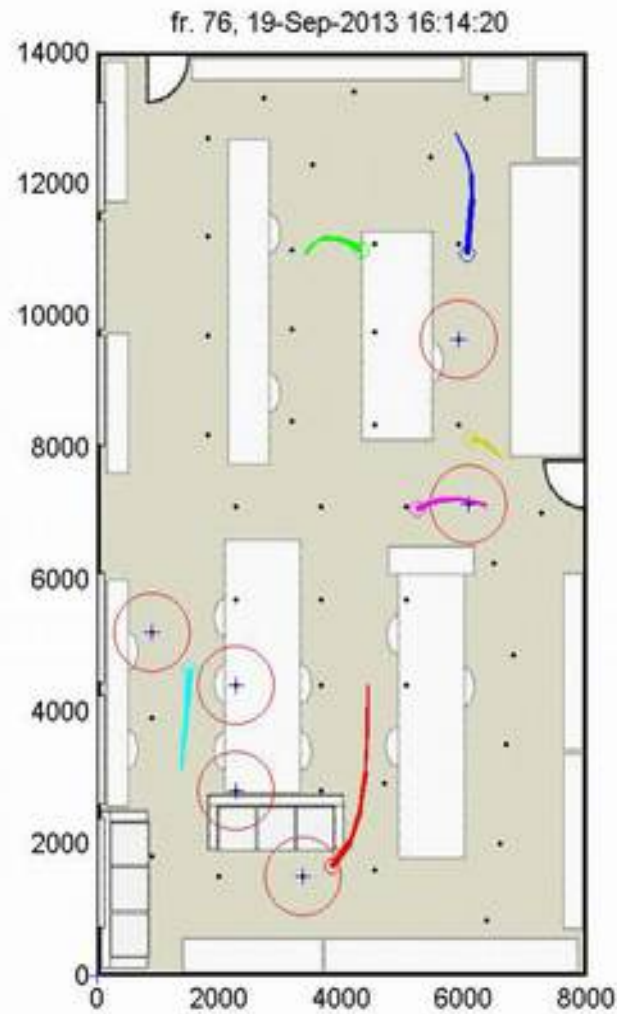
# Real Data

- Up to six people in a large lab
- Two cameras (2 Hz)
- Temporal alignment
- Annotation of key frames (very approximate)

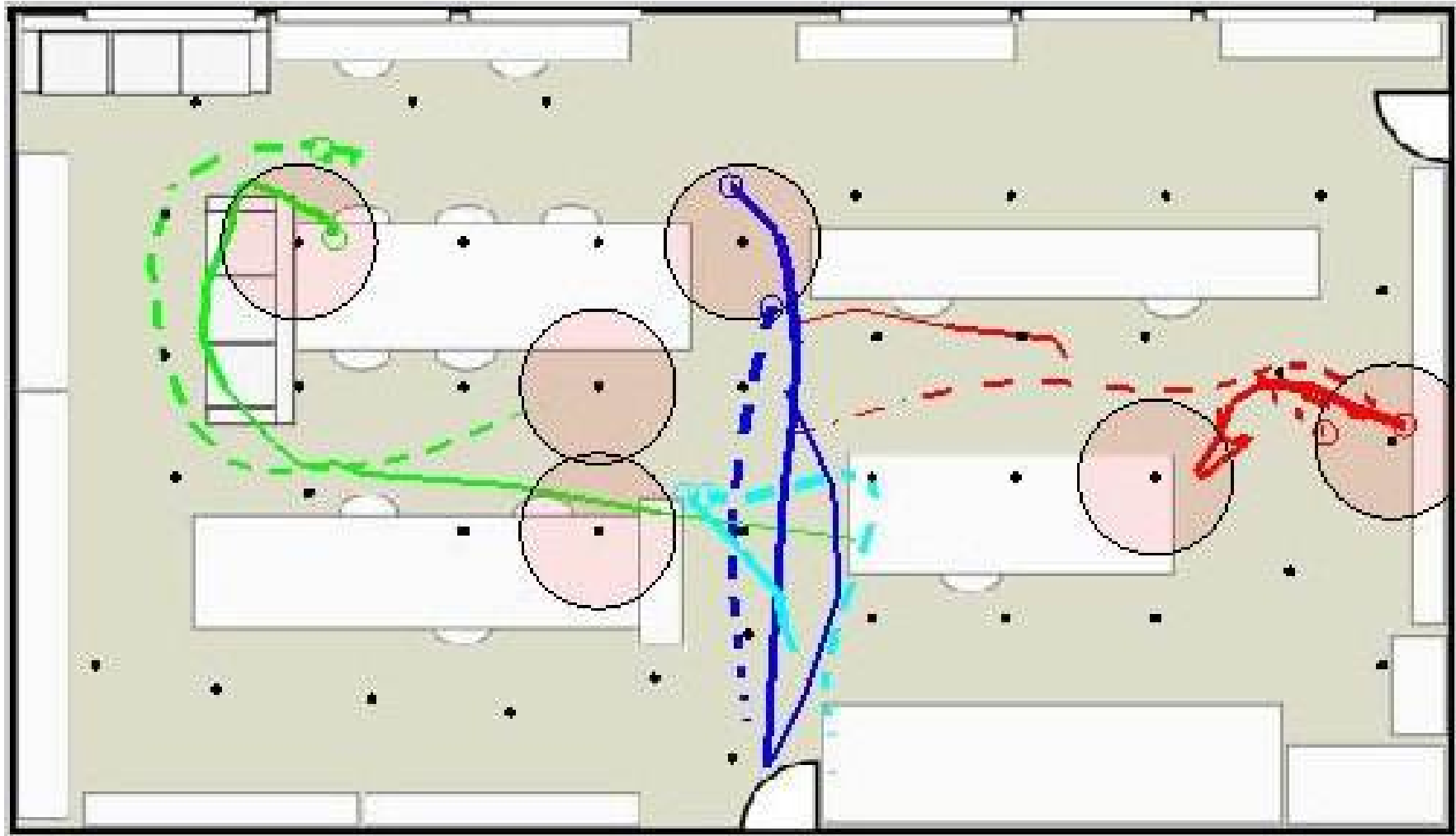


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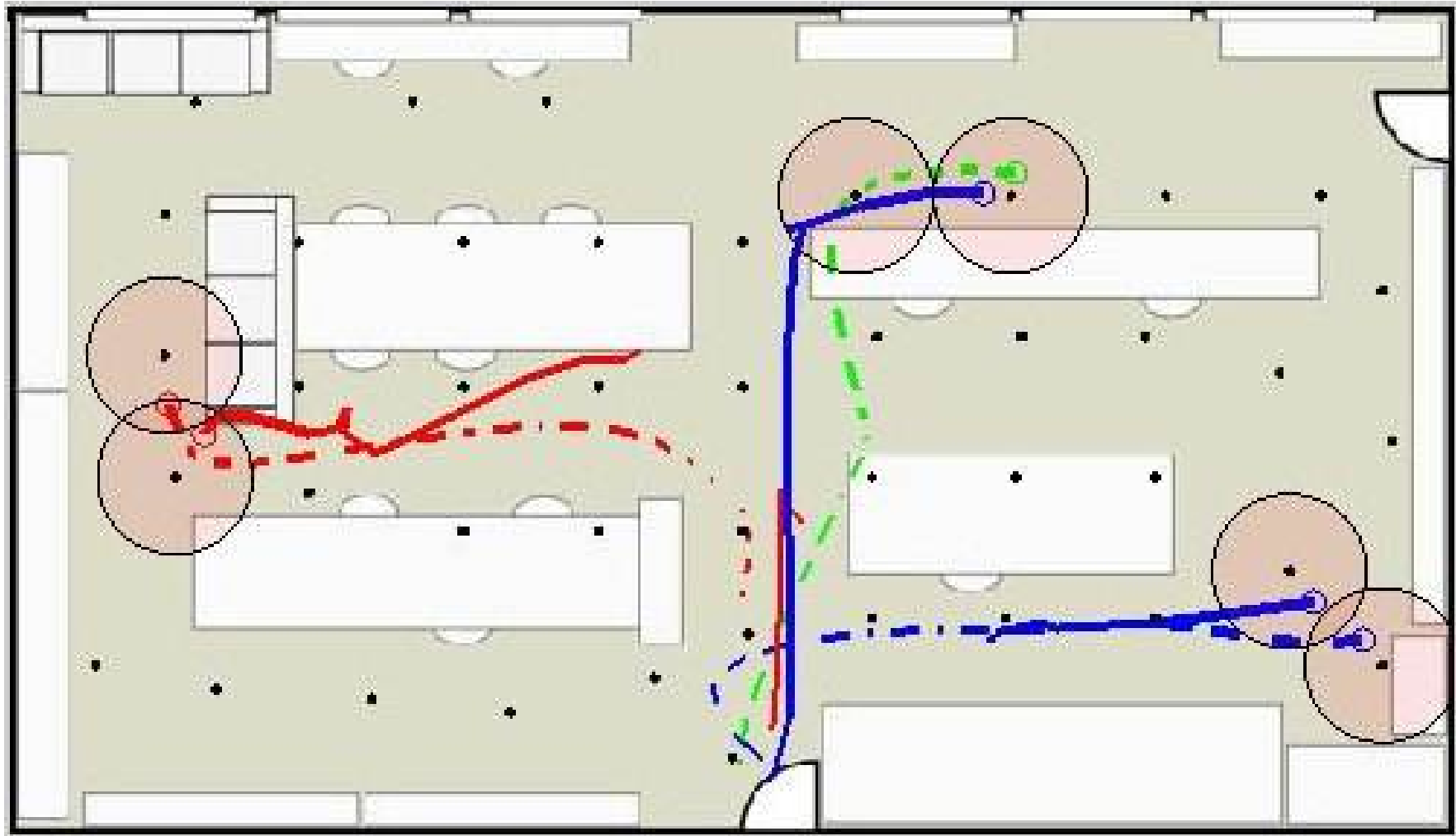
# Results (real)



————— **Result**

..... **GT**

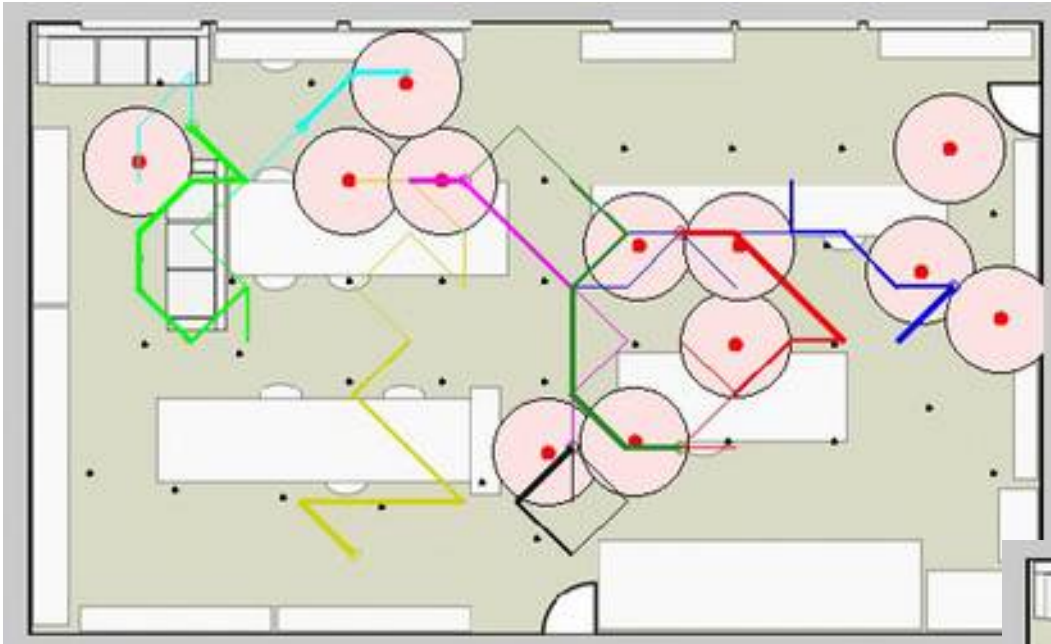
# Results (real)



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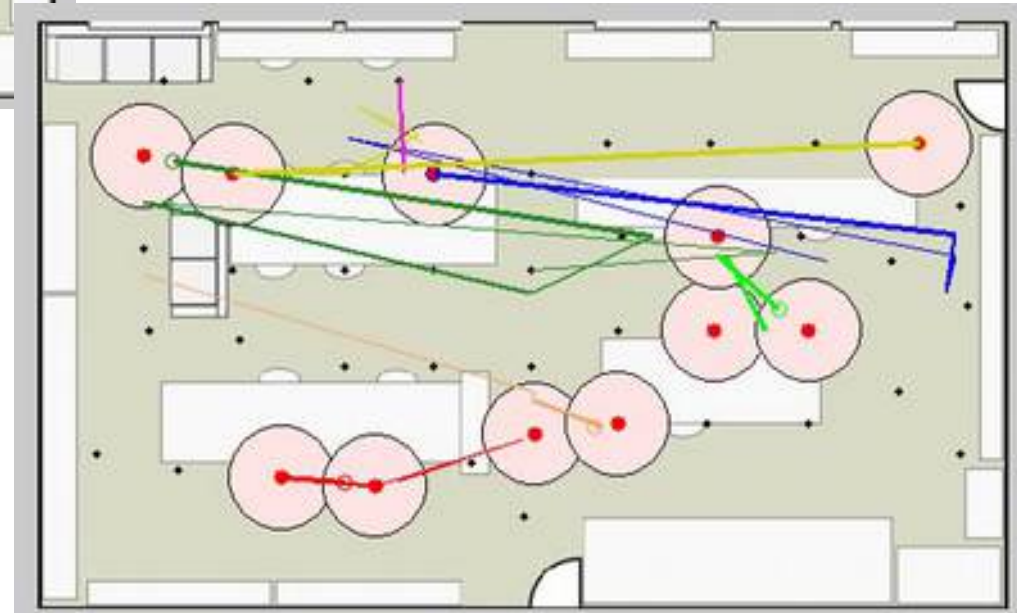
..... **GT**

# Other Approaches



[Berclaz et al., PAMI 2011]

[Tao et al., Sensors 2012]



# Quantitative Results

MOTA = normalized error count

Dataset	Method	MOTA [%]	MOTP [%]	ID sw	#Targets (MAE)
synthetic	Ours	76.0	73.6	13	0.54

MOTP = localization error (73%  $\approx$  35 cm)

Real data	Ours	55.3	54.6	43	0.76
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# Quantitative Results

Dataset	Method	MOTA [%]	MOTP [%]	ID sw	#Targets (MAE)
synthetic	<b>Ours</b>	<b>76.0</b>	<b>73.6</b>	13	<b>0.54</b>
	Linear DA [1]	66.6	64.6	58	0.57
	DP [2]	55.9	65.3	57	0.62
	KSP [3]	75.5	67.5	<b>6</b>	1.52
Real data	<b>Ours</b>	<b>55.3</b>	<b>54.6</b>	<b>43</b>	<b>0.76</b>
	Linear DA [1]	9.3	50.1	252	1.00
	DP [2]	9.6	47.3	128	1.25
	KSP [3]	31.1	48.3	48	1.52

[1] Tao et al., *Sensors* 2012

[2] Pirsiavsah et al., *CVPR* 2011

[3] Berclaz et al., *PAMI* 2014



# Advertisement



## Multiple Object Tracking Benchmark

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## 2D MOT 2015

This benchmark contains video sequences in unconstrained environments filmed with both static and moving cameras. Tracking and evaluation are done in image coordinates.

### Training Set

	Sample	Name ↑↓	FPS ↑↓	Resolution	Length ↑↓	Tracks ↑↓	Boxes ↑↓	Description	Source	Ref.
1		TUD-Stadtmitte	25	640x480	179 (00:07)	10	1156	A static camera at about 2 meters height shows walking people on the street.	<a href="#">link</a>	[1]
2		TUD-Campus	25	640x480	71 (00:03)	8	359	A short sequence with side-view pedestrians	<a href="#">link</a>	[2]
3		PETS09-S2L1	7	768x576	795 (01:54)	19	4652	A widely used sequence showing up to 8 walking pedestrians, partly in unusual patterns.	<a href="#">link</a>	[3]

# Advertisement



## 2D MOT 2015

This benchmark contains

evaluation are done in image

- 22 Sequences (old + new)
- > 1300 Trajectories
- > 100,000 Bounding boxes
- Live online evaluation

	Sample	Name	Description	Source	Ref.
1		TUO- Stadtmitte	night shows	<a href="#">link</a>	[1]
2		TUO- Campus	A short sequence with side-view pedestrians	<a href="#">link</a>	[2]
3		PETS09- S2L1	A widely used sequence showing up to 8 walking pedestrians, partly in unusual patterns.	<a href="#">link</a>	[3]

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## Multiple Object Tracking Benchmark

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### 2D MOT 2015 Results

MOTA and MOTP are part of the CLEAR protocol [1], MT, ML and Frag were introduced in [2].

Tracker ↓↑	MOTA ↓↑	MOTP ↓↑	FAR ↓↑	MT ↓↑	ML ↓↑	FP ↓↑	FN ↓↑	ID Sw. ↓↑	Frag ↓↑	Hz ↓↑	Specifications	Detector	Ref.
TBD	<b>31.6</b>	<b>72.8</b>	1.4 %	<b>13.7 %</b>	48.2 %	9,929	30,735	1,801	1,583	1.4	2.6 GHz, 1 Core	No	[3,4]
TC_ODAL	31.0	72.0	2.4 %	23.1 %	<b>31.1 %</b>	11,649	<b>18,856</b>	<b>394</b>	<b>976</b>	8.3	2.6 GHz, 1 Core	No	[5]
CEM	19.3	70.8	2.5 %	8.6 %	46.6 %	14,169	34,580	814	1,017	1.0	2.6 GHz, 1 Core	No	[6]
SMOT	17.4	71.4	<b>1.4 %</b>	2.5 %	55.6 %	<b>8,070</b>	41,346	1,323	1,881	3.8	2.6 GHz, 1 Core	No	[7]
LP2D	11.5	71.0	1.8 %	2.4 %	64.5 %	10,156	41,723	2,495	2,564	112.3	2.6 GHz, 1 Core	No	[8]
DP_NMS	3.9	70.4	3.5 %	8.5 %	33.7 %	20,000	32,180	6,885	4,291	<b>288.0</b>	2.6 GHz, 1 Core	No	[9]

[view detailed results](#) 

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## Multiple Object Tracking Benchmark

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### 2D MOT 2015 Results

MOTA and MOTP are part of the CLEAR protocol [1], MT, ML and Frag were introduced in [2]

<http://motchallenge.net>

Track	Detector	Ref.
TC_0	No	[3,4]
CF	No	[5]
SMOT	No	[6]
LP2D	No	[7]
DP_NMS	No	[8]

Track	MOTA	MOTP	MT	ML	Frag	Time	Time	Time	Time	Time	Time	Time	Time	Time
SMOT	17.4	71.4	1.4 %	2.5 %	55.6 %	8,070	41,346	1,323	1,881	3.8	2.6 GHz, 1 Core	No	[7]	
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view detailed results

# Summary

- A principled alternative to preserve privacy
- Continuous energy with soft assignments
- Still a very challenging problem
- Data + Code online

<http://research.milanton.net/irtracking/>