



# Object Detection and Trajectory Clustering from Low-Level Point Tracks

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March 18, 2009



# Principal Investigator



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## Research Areas

- Computer vision in large-scale camera networks: calibration and tracking
- Integrating digital images and LiDAR
- Image and video change analysis
- Image processing and machine learning for radiation therapy applications

## Potential Military Applications

- SIGINT / IMINT / GEOINT / MASINT analysis and understanding
- Large-scale image and range sensor fusion; activity recognition; change detection and understanding

## Contact Information

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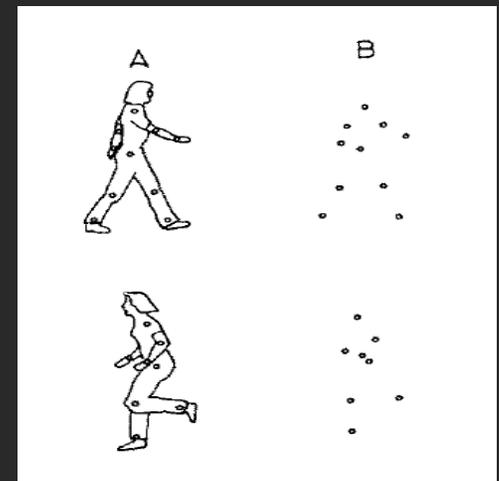
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# Perceptual Organization



- A swarm of moving green dots represents tracked features
- Human observers can still identify/count moving objects (Johansson 75)





# Automatically Extracting Dominant Motions





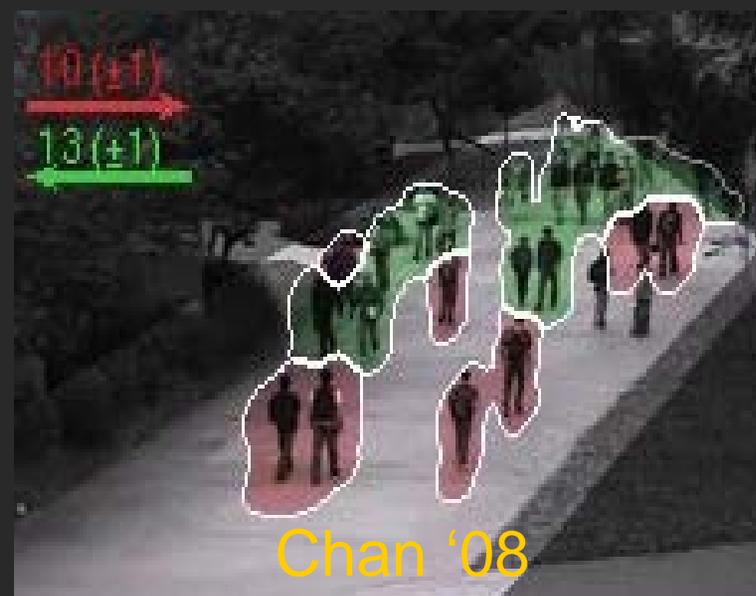
# Motivation: Crowd Behavior Analysis



- **Challenges: inter- and intra-object occlusions, shadows, illumination changes, difficulty in model fitting**
- **Applications: surveillance, search and rescue, crisis management**

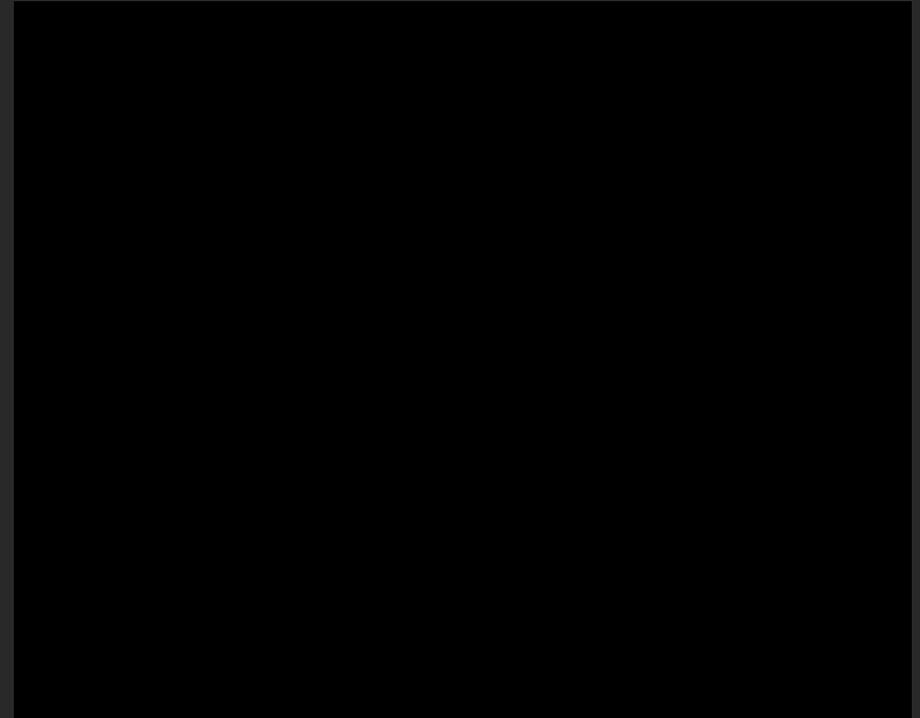


# Related Work





# Low-Level Feature Point Tracking

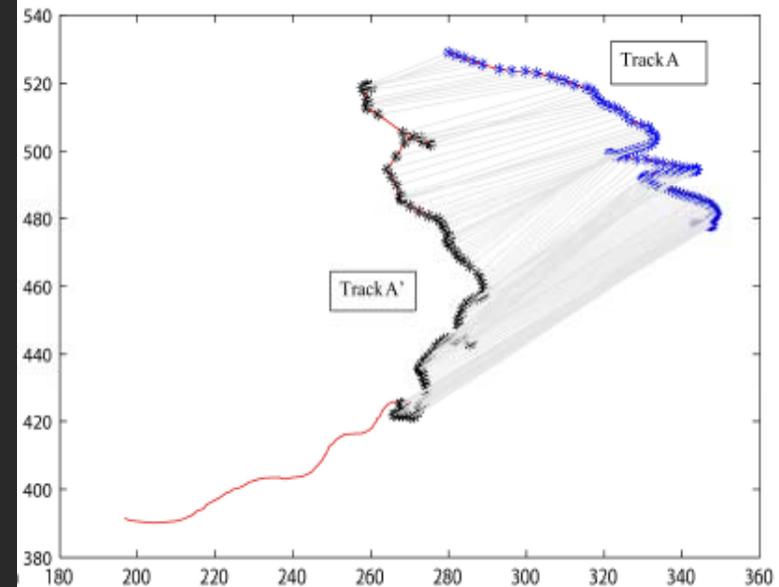
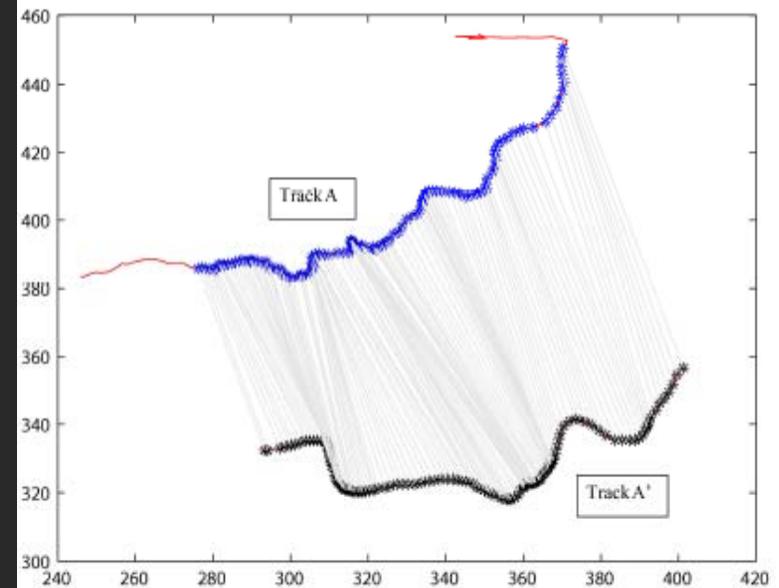


- Tomasi-Kanade and Rosten-Drummond features
- Tracked with hierarchical KLT algorithm
- Many feature tracks are partial or inaccurate



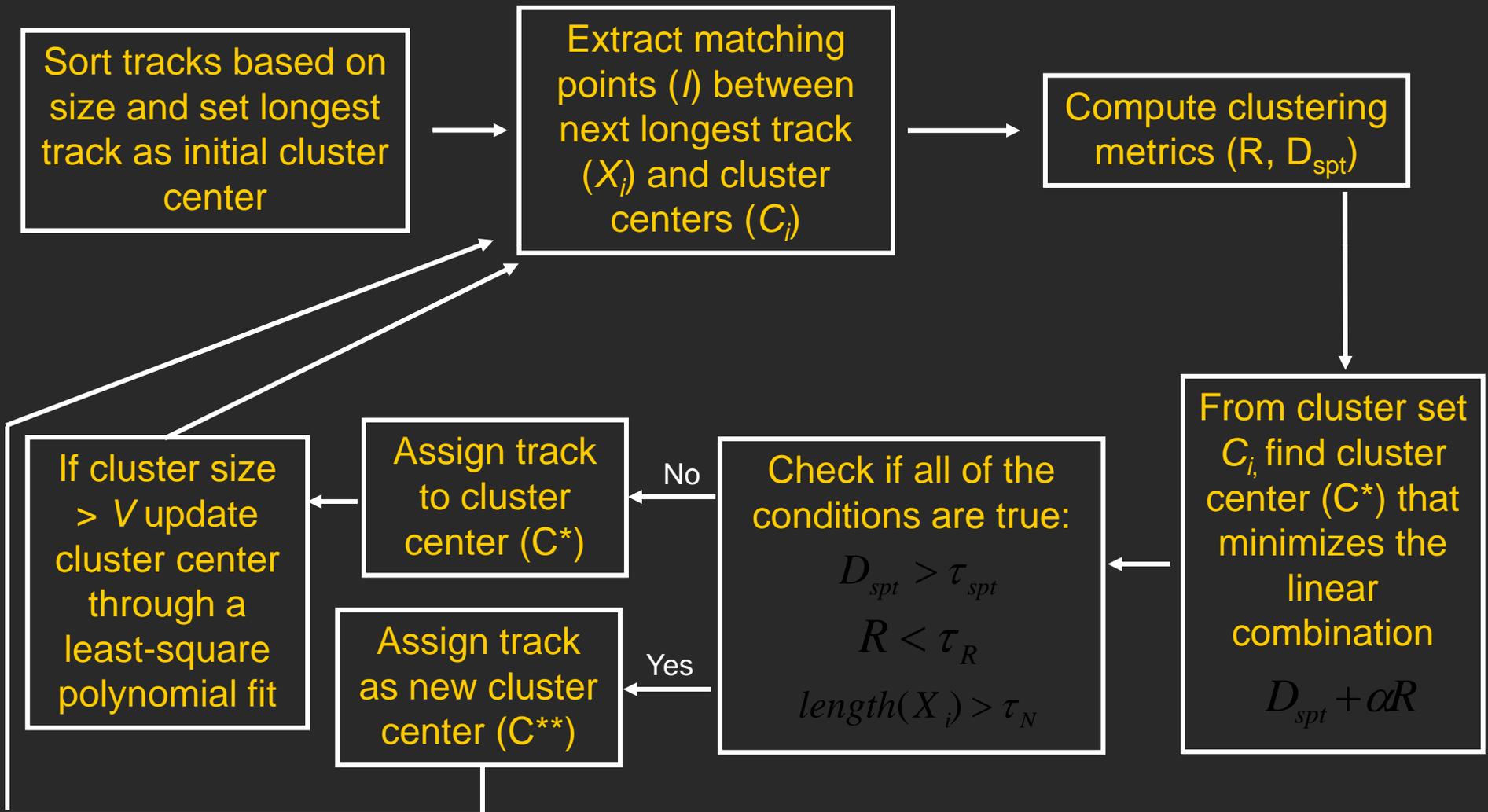
# Track Matching Metric

- Distance between tracks measured using longest common subsequences
- Fast point matching algorithm based on dynamic programming
- Parameters to control matching flexibility in space and time





# Dominant Motion Clustering Algorithm



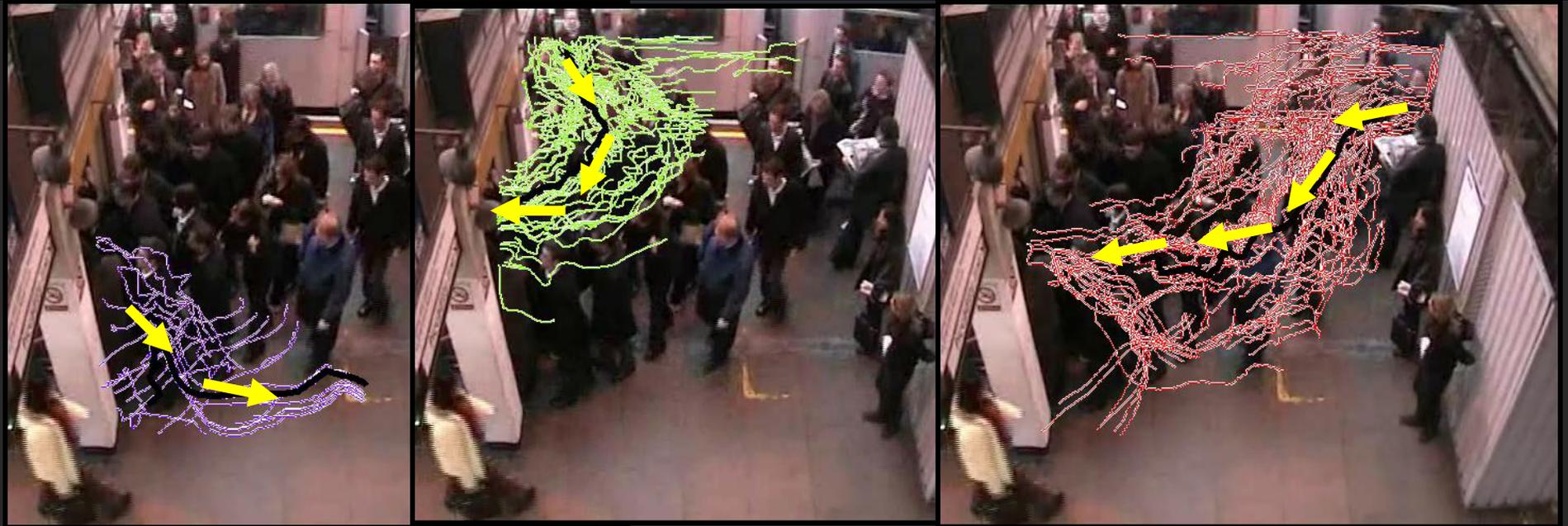


# Station Sequence





# Station Sequence



## Automatically extracted clusters:

- Entering platform
- Exiting platform from left
- Exiting platform from right



# Escalator Sequence





# Airport Sequence



- Left to right, right to left, entering office, loitering



# Campus Sequence



- Two dominant motions
- One anomalous motion: pedestrian doing a U-turn (detected as a small cluster)



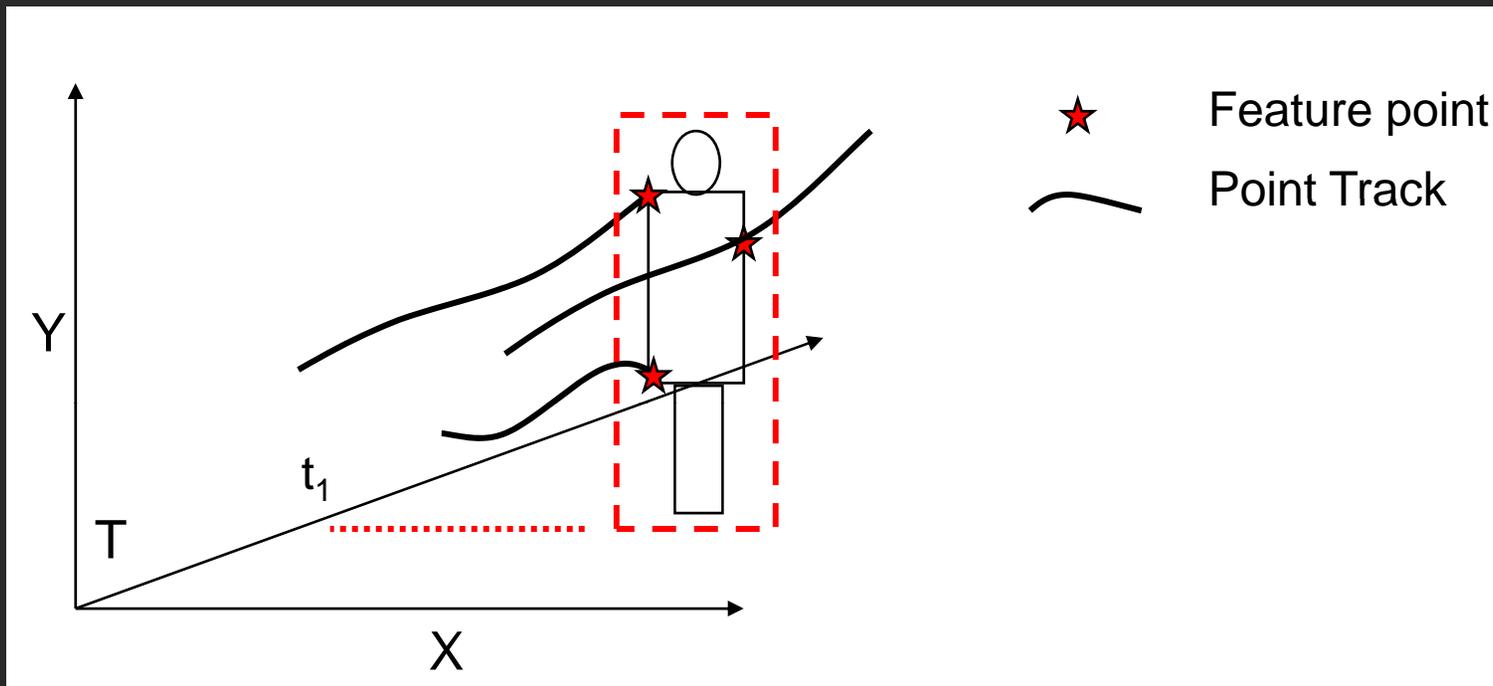
# Coherent Motion Regions



**We collect point tracks into spatio-temporal coherent motion regions (i.e. potential objects), using a greedy algorithm to select the best set of CMRs.**



# Coherent Motion Regions

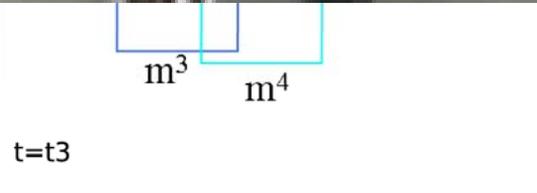
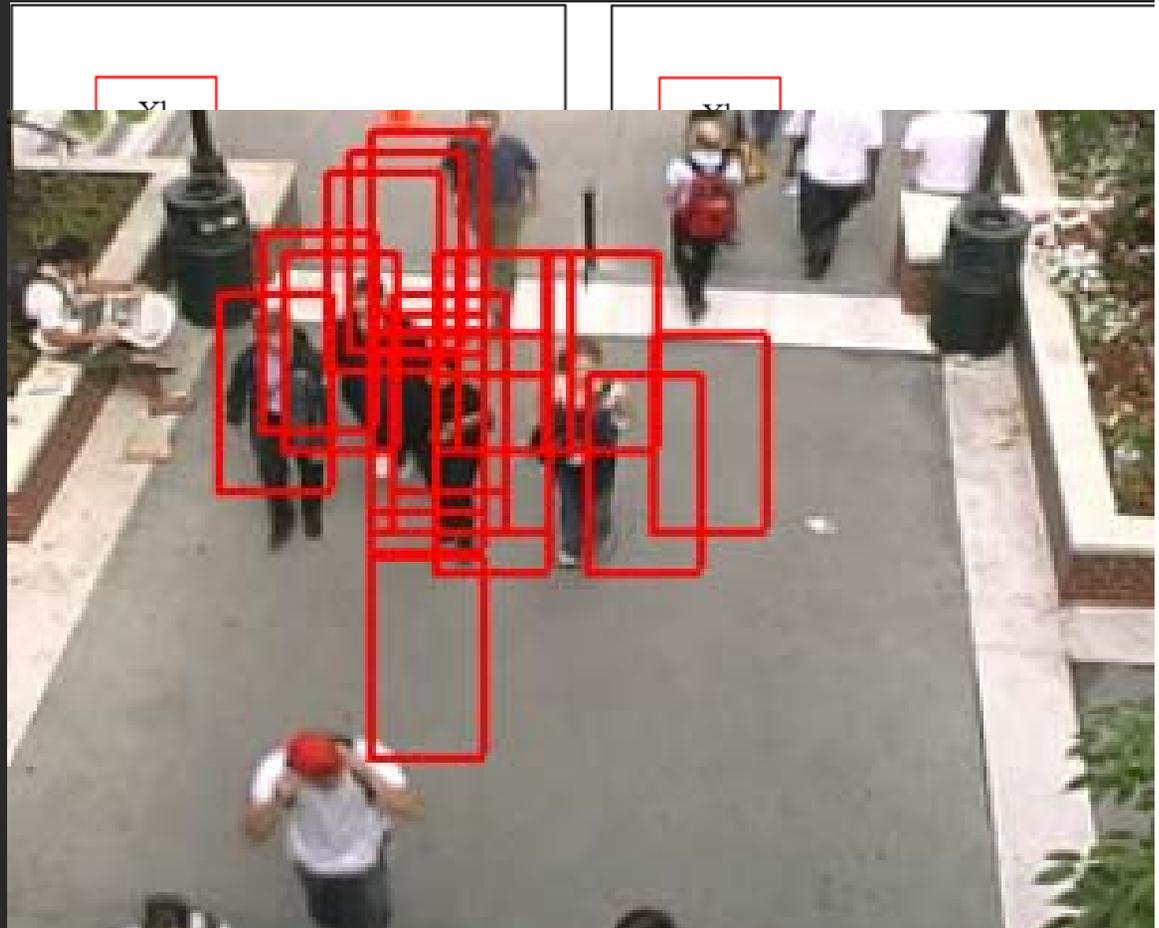


- **CMR: a contiguous chunk of  $(x,y,t)$  space that completely contains a set of point tracks**
- **May contain tracks with little to no spatial or temporal overlap**
- **We didn't consider temporal aspects previously**



# CMR Candidates

- User supplies rough CMR dimensions
- Slide spatial window over each frame
- Generate columns of a binary matrix that represents potential CMRs



(c)

$X^6$	0	0	1	1
$X^7$	1	0	1	0
$X^8$	0	0	0	1

(d)



# CMR Strength

- We define a track similarity measure  $S(i, j)$
- Basic idea: point tracks from the same object should have low maximum distance and low variance in distance

$$S(i, j) = \exp\{-\alpha * (\max(D(i, j) + \text{var}(D(i, j))))\}$$

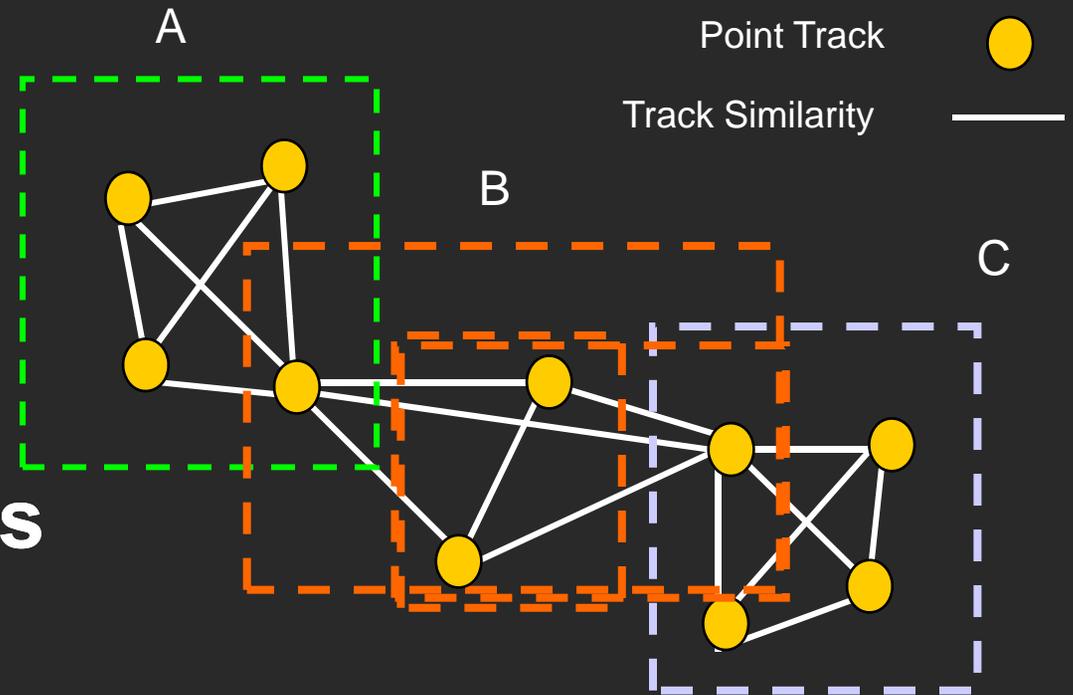
- We define the strength of a CMR as:

$$L(j) = A(j)^T SA(j)$$



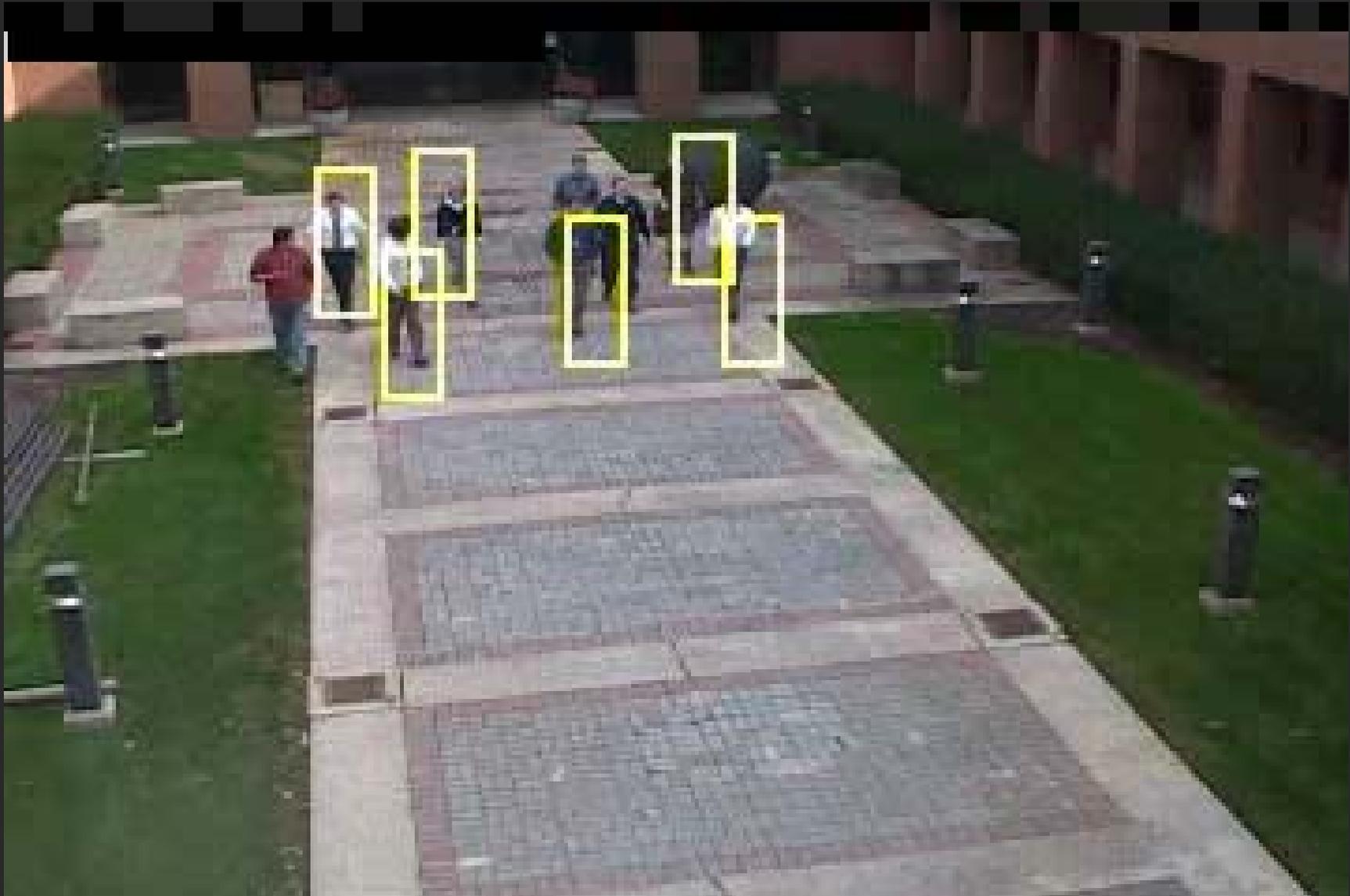
# Optimal CMR Subset

- We use a greedy algorithm
- Select strongest CMR
- Break connections to remaining candidates and iterate
- Explicitly enforces  $\leq 1$  CMR per track





# GE Sequence



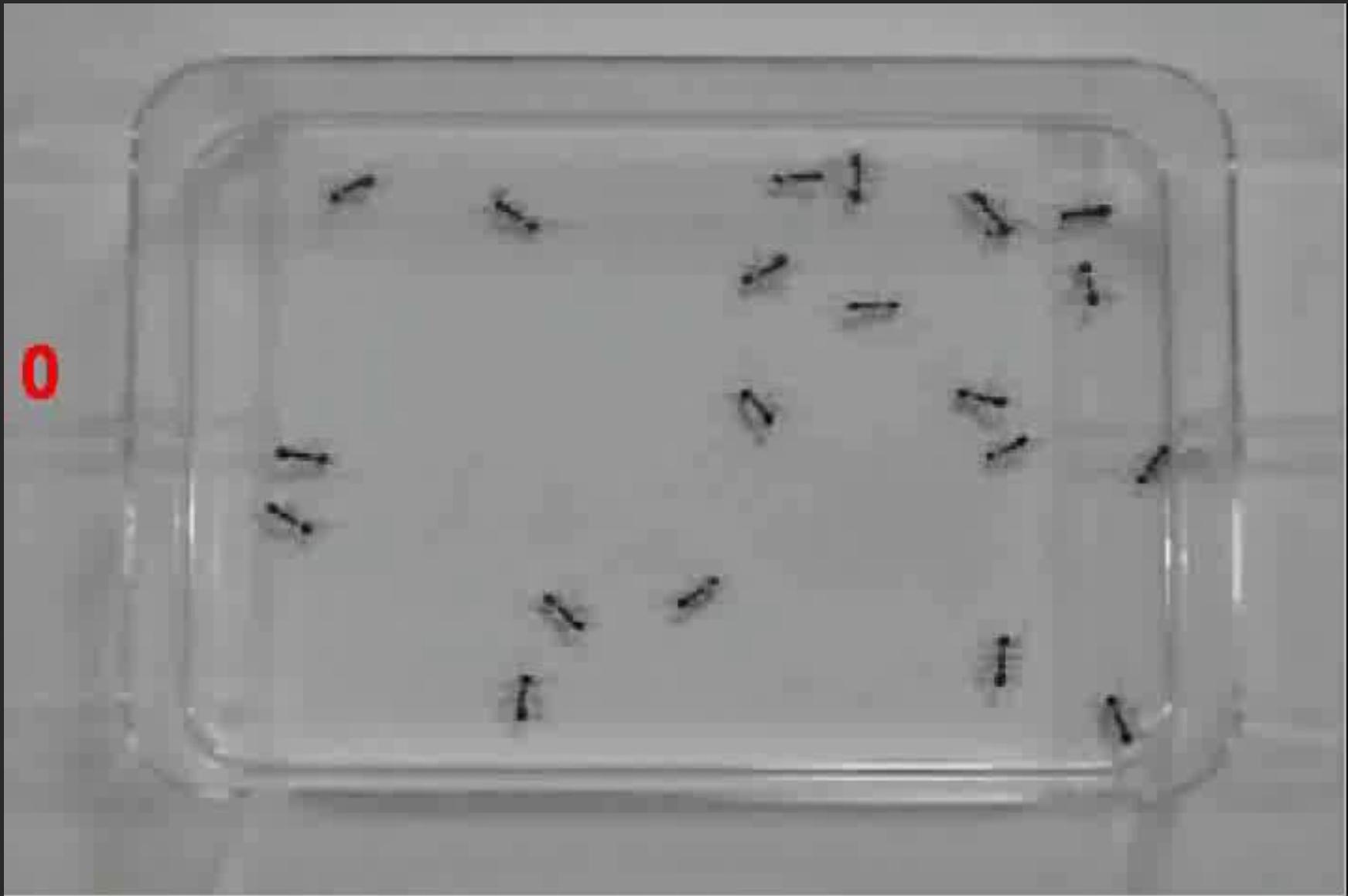


# CalTrans Sequence





# Ants Sequence



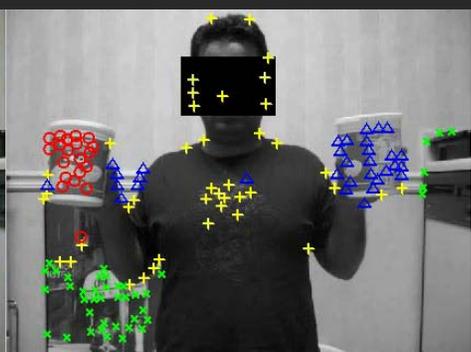
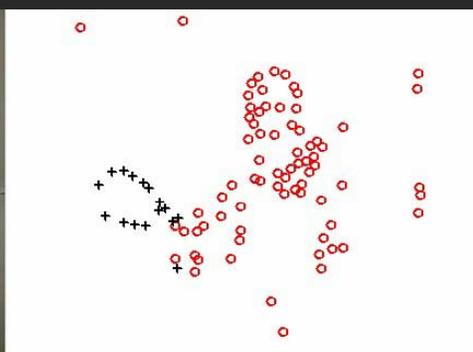
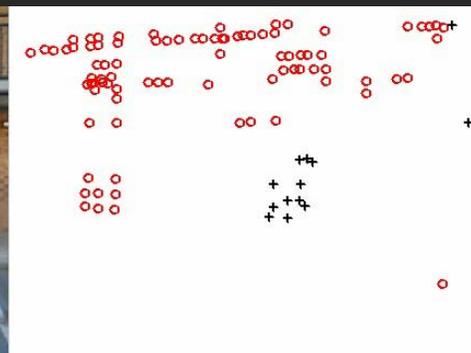
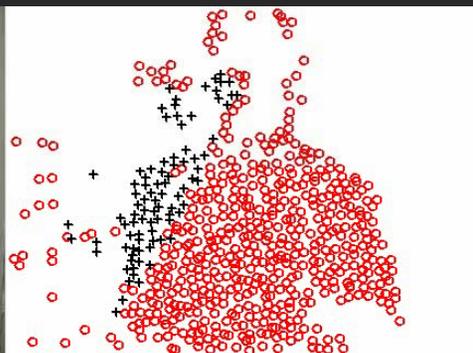


# Night Highway Sequence





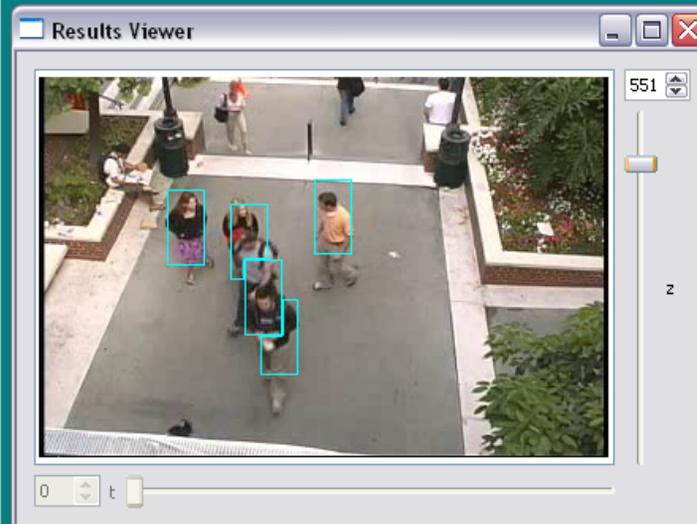
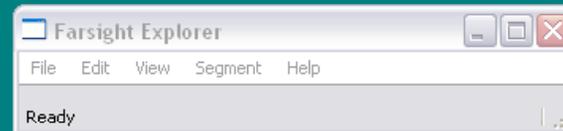
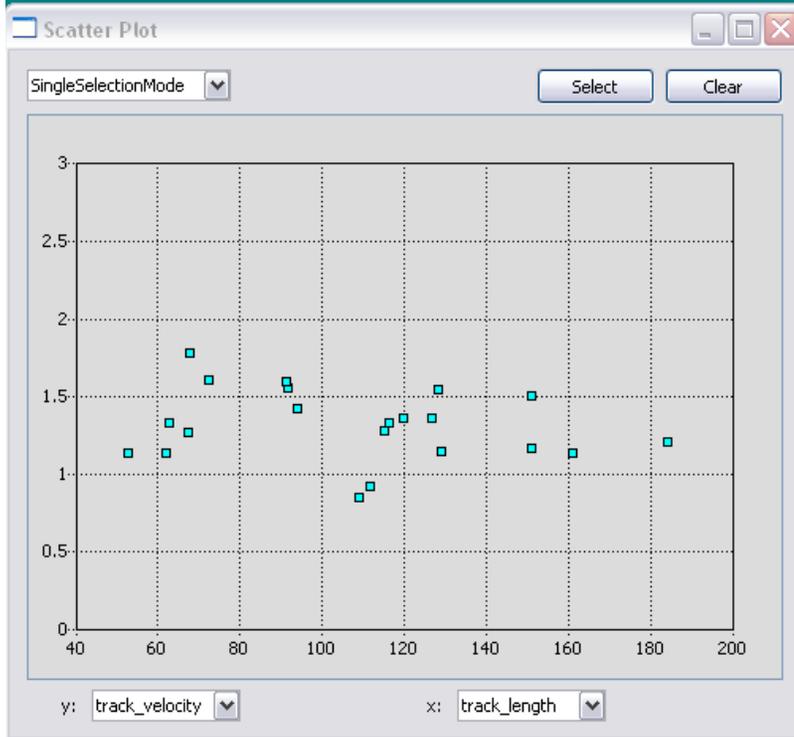
# Independent Motion Segmentation



We apply non-negative factorization of a matrix of point track velocities to obtain fast, accurate segmentation of independent motions.



# Integration into FARSIGHT



Table

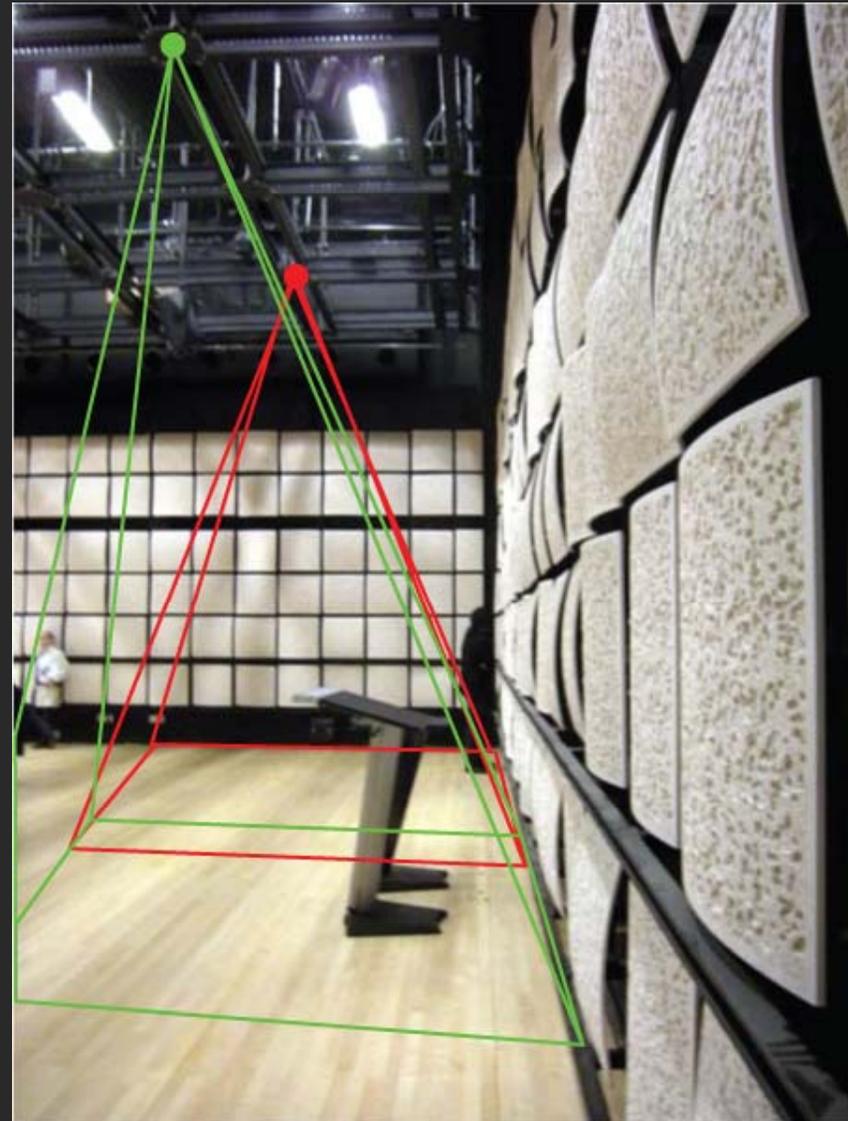
	id	track_length	track_velocity
1	1	111.42	0.928499
2	2	61.6493	1.14165
3	3	93.8204	1.42152
4	4	119.752	1.36082
5	5	150.736	1.50736
6	6	91.6099	1.55271
7	7	184.101	1.21119
8	8	126.551	1.36077
9	9	114.856	1.27618
10	10	116.173	1.33532
11	11	127.939	1.54143
12	12	108.763	0.849711
13	13	91.2176	1.60031
14	14	160.875	1.13292
15	15	52.4493	1.1402
16	16	150.723	1.16839
17	17	72.3394	1.60754
18	18	67.7477	1.78283
19	19	128.756	1.14961
20	20	67.2791	1.26942
21	21	62.6369	1.3327

- A powerful interface for interactive visualization and editing of results developed at RPI for biological applications



# EMPAC Video Analytics Testbed

- 50x40x30 ft<sup>3</sup> black-box studio at EMPAC, RPI
- 12 ceiling-mounted fixed-focus network cameras, 4 wall-mounted PTZ network cameras
- Testbed for camera network, tracking, activity recognition, crowd motion research
- Platform for URPs





# Future Work

- **Tuning “significant” dominant motions**
- **Variable-size CMRs**
- **Application to portal monitoring environment**
- **Multi-source fusion**
- **Independent motion segmentation**



# Thanks

- Collaborators: Anil Cheriyadat, Bhudendra Bhaduri (ORNL)

- Support: DHS, NSF, DARPA



- Papers:

<http://www.ecse.rpi.edu/~rjradke/pubs.htm>

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