

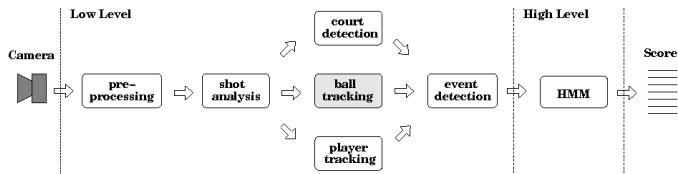
Tennis Ball Tracking

Fei Yan

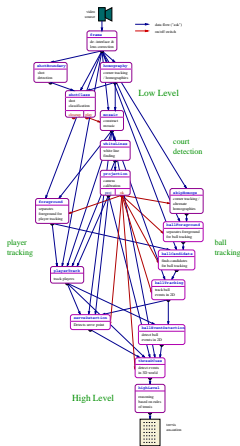
ACASVA Project Meeting
June 24, 2010
University of Surrey, Guildford

Tennis understanding system

- A system that “understands” tennis
- High level reasoning relies on low level processing:
 - court line detection / court model fitting
 - ball tracking
 - player tracking / player action recognition



Tennis understanding system



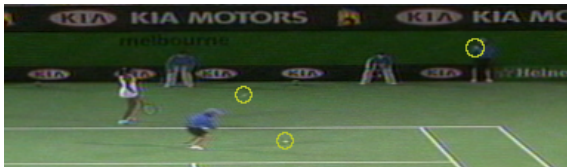
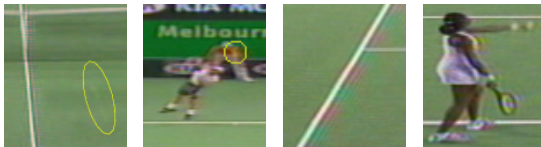
Shots		Play	What res
No	Start	End	Class
0	00:00:00	00:00:20	closeup
1	00:00:28	00:00:16	play
2	00:00:32	00:00:02	closeup
3	00:00:34	00:00:14	closeup
4	00:00:40	00:00:14	play
5	00:00:17	00:00:20	closeup
6	00:00:20	00:00:21	closeup
7	00:00:14	00:00:25	closeup
8	00:00:23	00:00:27	closeup
9	00:00:24	00:00:30	play
10	00:00:32	00:00:32	closeup
11	00:00:30	00:00:35	play
12	00:00:31	00:00:37	play
13	00:00:37	00:00:40	play
14	00:00:40	00:00:41	closeup
15	00:00:42	00:00:43	play
16	00:00:48	00:00:48	closeup

Difficulties of tennis ball tracking

- Small size, high velocity
- Motion deformation, motion blur
- Occlusion in successive frames
- Abrupt change of motion direction
- Camera pan, tilt and zoom (PTZ)
- Monocular sequences

Tennis annotation system
Difficulties of tennis ball tracking
Overall strategy
Ball candidate detection
Layered data association
Future work

Difficulties of tennis ball tracking



Overall strategy

- Track after detection (TAD) and track before detection (TBD)
 - TAD: small objects with simple representations
 - TBD: large objects with complex representations
- A TAD approach can be decomposed into:
 - candidate detection
 - data association
- Data association is the key in tennis ball tracking

Ball candidate detection

- De-interlace, geometric distortion correction
- Homography computation and motion compensation
- Motion-compensated frame differencing for motion segmentation
- An example sequence
- Motion segmentation results on the example sequence

Layered data association - candidate level association

- Candidate level association: an example
- Fitting a constant acceleration dynamic model:

$$\mathbf{v}_1 = \frac{\mathbf{z}_2 - \mathbf{z}_1}{\Delta k_{21}} - \frac{\Delta k_{21} \times \mathbf{a}}{2} \quad (1)$$

$$\mathbf{a} = 2 \times \frac{\Delta k_{21} \times (\mathbf{z}_3 - \mathbf{z}_2) - \Delta k_{32} \times (\mathbf{z}_2 - \mathbf{z}_1)}{\Delta k_{21} \times \Delta k_{32} \times (\Delta k_{21} + \Delta k_{32})} \quad (2)$$

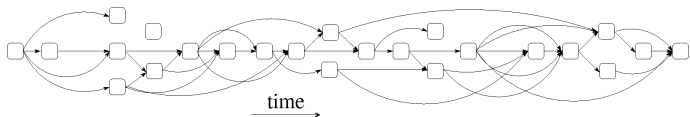
- Evaluating the quality of a model

$$C = \sum_{k=i-V}^{i+V} \sum_j \rho(\mathbf{z}_k^j) \quad (3)$$

$$\rho(\mathbf{z}_k^j) = \begin{cases} d^2(\hat{\mathbf{z}}_k, \mathbf{z}_k^j) & \text{if } d(\hat{\mathbf{z}}_k, \mathbf{z}_k^j) < d_{th} \\ d_{th}^2 & \text{if } d(\hat{\mathbf{z}}_k, \mathbf{z}_k^j) \geq d_{th} \end{cases} \quad (4)$$

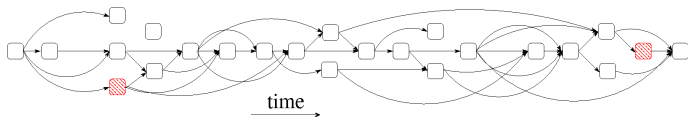
Layered data association - tracklet level association

- All tracklets in the example sequence
- Tracklet level association: shortest path problem in a directed and edge-weighted graph
 - Each node is a tracklet
 - Edge weight reflects the “compatibility” of the two nodes
 - Desired data association contained in the shortest path
- Final tracking results after event detection



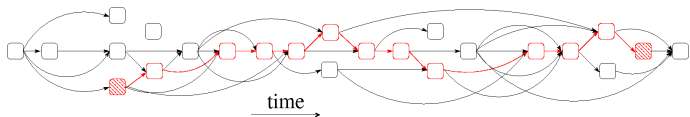
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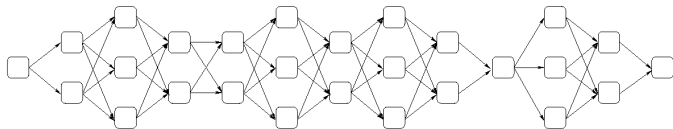
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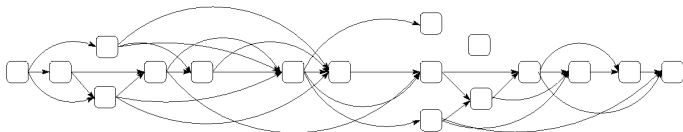
Layered data association - an efficient APSP algorithm

- General-purpose APSP algorithms:
 - Applying Dijkstra's algorithm repeatedly: $O(N^3)$
 - The Floyd-Warshall algorithm: $O(N^3)$
 - Johnson's algorithm: $O(N^2 \log N + NE)$
- Trellis graph: $\forall e_{u,v}^{p,q} \in \mathcal{E}, u + 1 = v$
- \mathcal{G} has a special topological property: $\forall e_{u,v}^{p,q} \in \mathcal{E}, u < v$

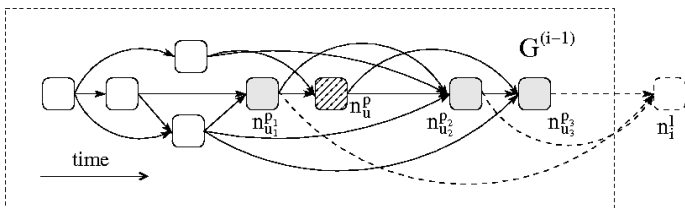


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Layered data association - an efficient APSP algorithm



- Based on dynamic programming
 - Assume APSP in $\mathcal{G}^{(i-1)}$ is solved
 - Solve APSP in $\mathcal{G}^{(i)}$ using the APSP results in $\mathcal{G}^{(i-1)}$
- Complexity: $O(N^2)$

Future work

- Doubles game?
 - More abrupt motion change and more clutter
 - Improve the robustness of the current tracker
- Other ball games?
 - Badminton? Table tennis?