

- 1 Lries to jointly solve motion segmentation and MOT problems by creating a graph with nodes representing both point trajectories and detections and finding minimum cost multi cuts for this graph
- 2 Since there are two kinds of nodes, there are three kinds of edges that can connect these together $\rightarrow c_l, c_h$ and c_{lh} respectively join $v_e - v_l, v_h - v_h$ and $v_l - v_h; v_e \rightarrow$ point trajectories and $v_h \rightarrow$ detections
- 3 The potentials or costs of these edges are mostly given by heuristics
- 3.1 A point sampling rate is used instead of feature detectors and all points with "significant underlying image structure" are tracked using large displacement OF
- 3.2 The familiar FBF is used to end trajectories that are inconsistent and whenever this happens new trajectories are inserted to maintain the sampling rate (?)
- 3.3 c_l is computed by combining motion, spatial and color distances between the trajectories \rightarrow motion and color are computed only for those with common lifetimes while spatial is also computed for these without temporal overlap
- 3.4 c_h is computed from simple features based on IOU
- 3.5 c_{lh} is computed by combining the spatio-temporal distance between the point and the center of the detection with the value of the segmentation template at the point
4. The solution to the MC-MC problem is APX hard so heuristics (47 years old) are used to get approx. solvers that are apparently good enough in practice.
- 5 A lot of heuristics are used for computing the edge costs for each of the datasets where different ones seem to be needed depending on the exact nature of the detections \rightarrow too many detections need to be grouped into tracklets while too sparse ones, for instance those obtained after non maximum suppression require a crude empirical normalizing factor.
6. A general performance observation is that the false negatives decrease while false positives rise