



Lecture 15: Motion

Motion segmentation

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CS131 Computer Vision: Foundations and Applications



What will we learn today?

- Motion segmentation
 - Formulation
 - Method
 - Results





Recap

- Key assumptions
 - **Small motion:** points do not move very far
 - **Brightness constancy:** projection of the same point looks the same in every frame
 - **Spatial coherence:** points move like their neighbors

Reminder: Gestalt – common fate



Common Fate



Motion segmentation

- How do we represent the motion in this scene?



Source: Silvio Savarese



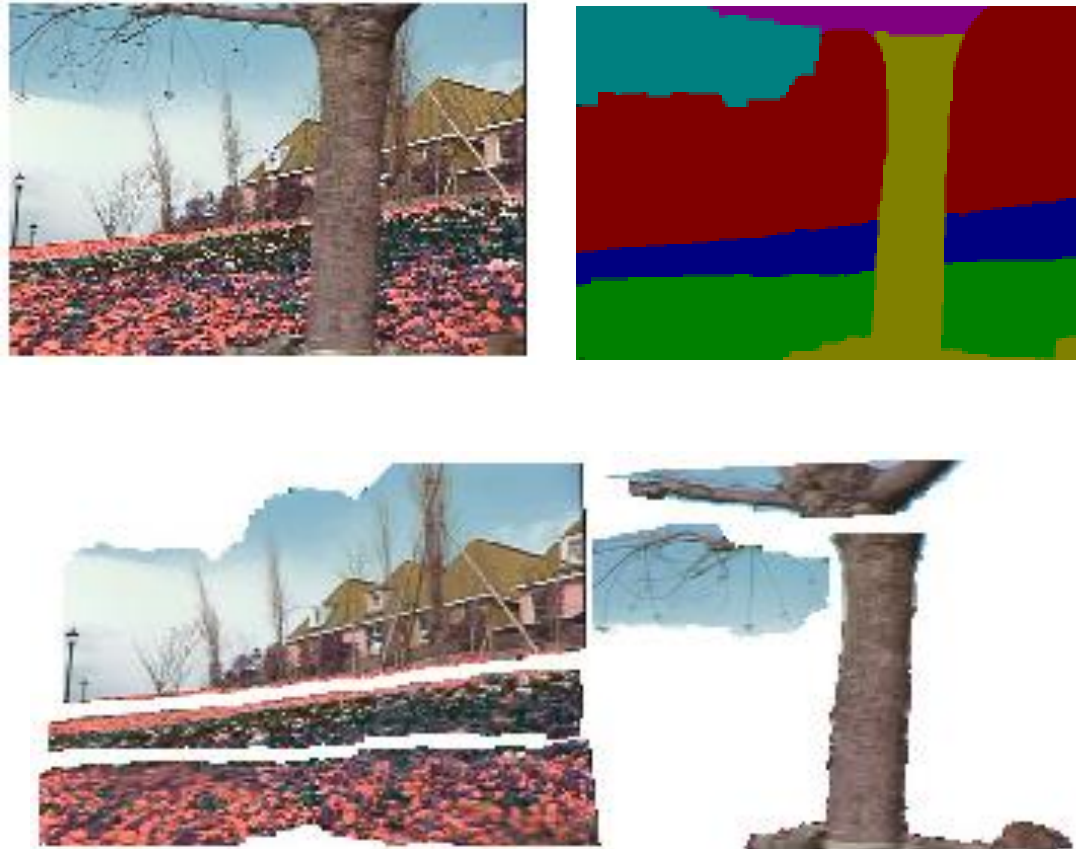
Motion

Common fate

5

Motion segmentation

- Break image sequence into “layers” each of which has a coherent (affine) motion



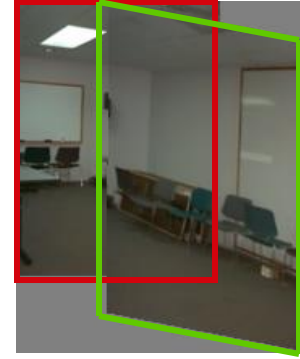
Affine motion

$$u(x, y) = a_1 + a_2x + a_3y$$

$$v(x, y) = a_4 + a_5x + a_6y$$

- Substituting into the brightness constancy equation:

$$I_x \cdot u + I_y \cdot v + I_t \approx 0$$

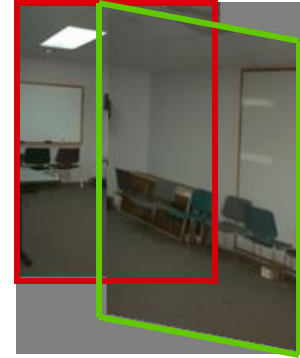


Affine motion

$$u(x, y) = a_1 + a_2x + a_3y$$

$$v(x, y) = a_4 + a_5x + a_6y$$

- Substituting into the brightness constancy equation:



$$I_x(a_1 + a_2x + a_3y) + I_y(a_4 + a_5x + a_6y) + I_t \approx 0$$

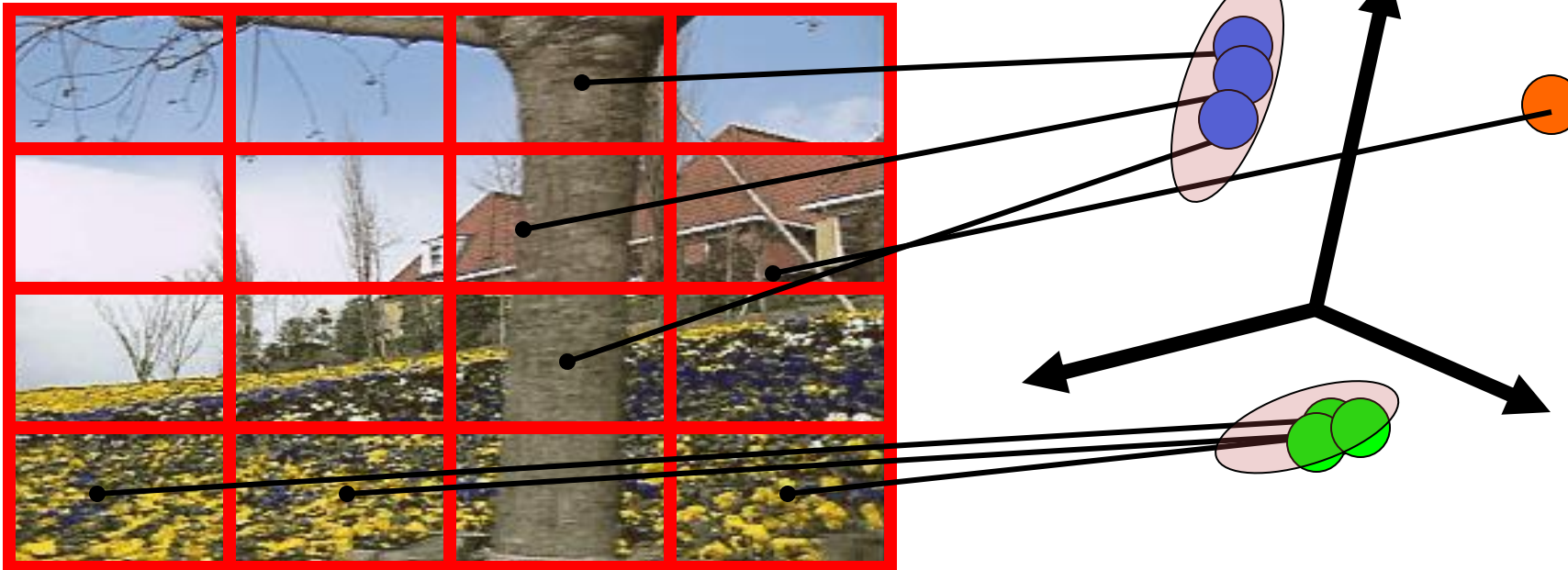
- Each pixel provides 1 linear constraint in 6 unknowns
- Least squares minimization:

$$Err(\vec{a}) = \sum \left[I_x(a_1 + a_2x + a_3y) + I_y(a_4 + a_5x + a_6y) + I_t \right]^2$$

How do we estimate the layers?

1. Obtain a set of initial affine motion hypotheses

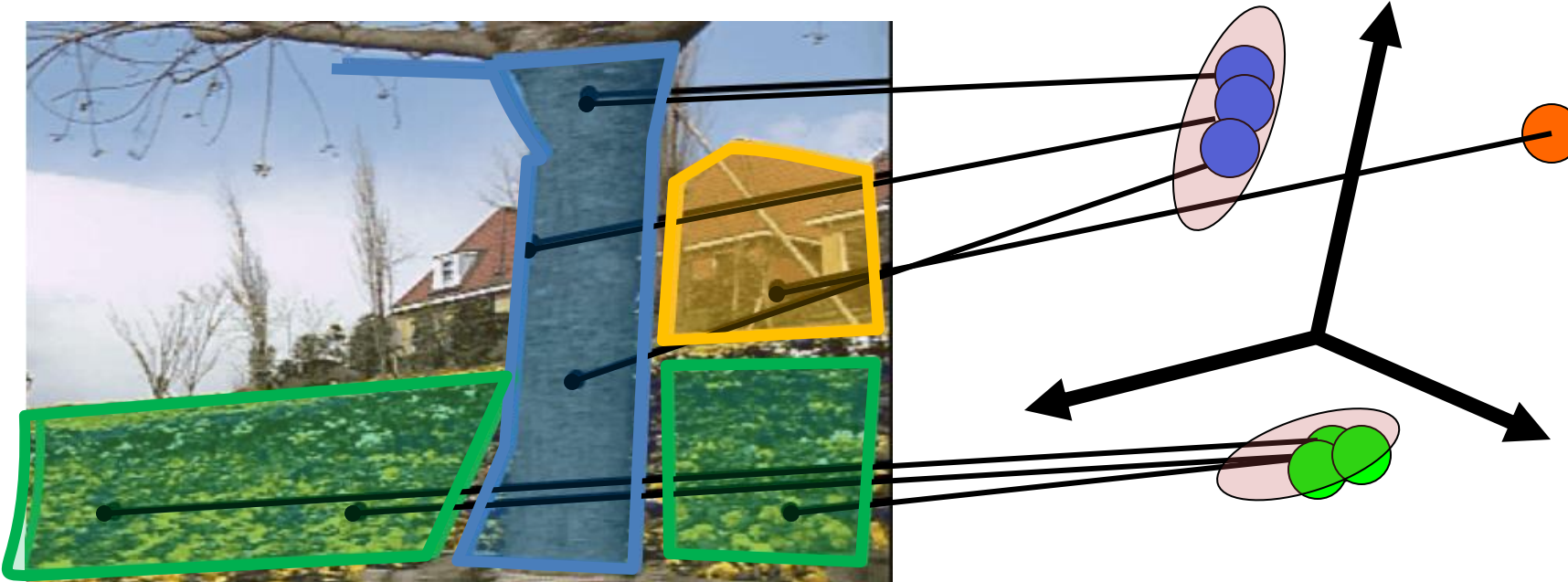
- Divide the image into blocks and estimate affine motion parameters in each block by least squares
 - Eliminate hypotheses with high residual error
- Map into motion parameter space
- Perform k-means clustering on affine motion parameters
 - Merge clusters that are close and retain the largest clusters to obtain a smaller set of hypotheses to describe all the motions in the scene



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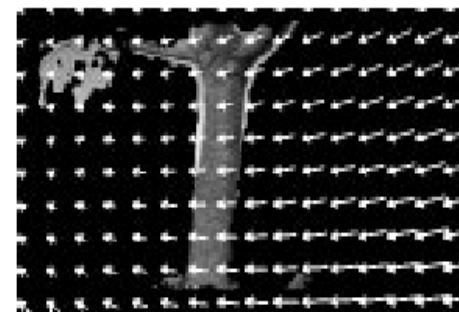
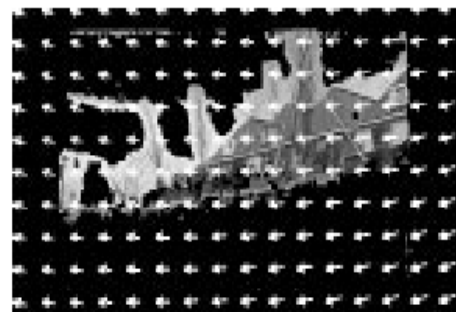
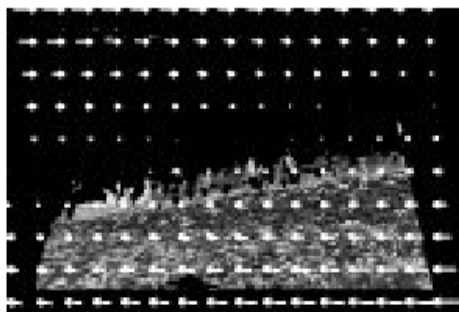
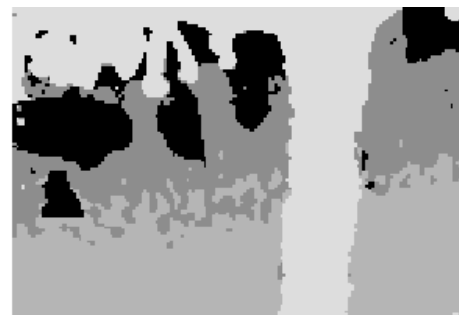
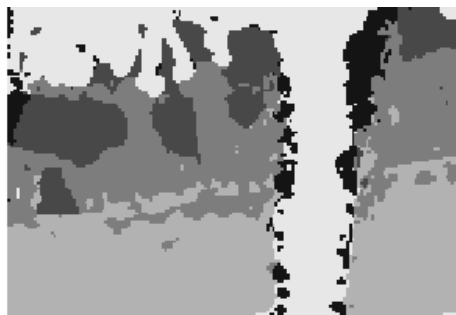
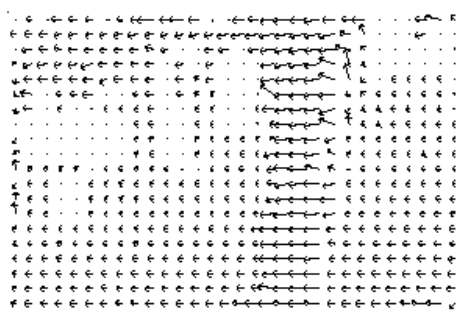
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2. Iterate until convergence:

- Assign each pixel to best hypothesis
 - Pixels with high residual error remain unassigned
- Perform region filtering to enforce spatial constraints
- Re-estimate affine motions in each region

Example result



J. Wang and E. Adelson. [Layered Representation for Motion Analysis](#). CVPR 1993.

Source: Silvio Savarese



Summary

- Motion segmentation
 - Formulation
 - Method
 - Results

