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Segmentation of Multiview Data for Scene Analysis and Compression

Jesse Berent and Pier Luigi Dragotti

November 9th, 2006

Camera Sensor Networks

- Camera arrays provide samples of the plenoptic function (multiple viewpoints of a scene)
- Huge amount of data!
- The data is highly correlated and structured
- Unsupervised data analysis
 - Object or layer extraction
 - scene interpretation
 - layer based representations
 - Occlusion detection
 - innovation processes
- Applications
 - Computer vision: automatic scene interpretation
 - 3DTV



- ...

Talk Outline

- 1. Introduction to the plenoptic function
- 2. Different camera setups and the Epipolar-Plane Image (EPI)
- 3. A brief review of active contours
- 4. Derivation of 'constrained' evolution equations for the plenoptic function
- 5. Conclusion and future work

The Plenoptic Function

• 7D function that describes the intensity of each light ray that reaches a point in space [AdelsonB:91]

$$P_7 = I(V_x, V_y, V_z, \phi, \theta, \tau, \lambda)$$

- Assumptions can be made to reduce the high number of dimensions
 - 3 channels for RGB or 1 channel for grayscale
 - Static scenes
 - Viewing position constraints





Different camera setups





4D



[Stanford multi-camera array]

3D



5D



[Imperial College multi-camera array]

Plenoptic Functions

• [Images courtesy of Yizhou Wang]



The Epipolar-Plane Image (EPI) Volume

- First introduced in [BollesBM:87]
- Cameras are constrained to a line
- Points in space are mapped on to lines
- The slope of the line \propto 1/depth
- Objects correspond to 3D tubes









Object Tubes

• EPI is made of a collection of tubes













Occlusions

- A line with a larger slope will always occlude a line with a smaller one
- Occlusions occur at line intersections
- Occlusions are explicit
- Object tubes can be 'orthogonalized'

$$\mathcal{V}_1^\perp = \mathcal{V}_1 \qquad \qquad \mathcal{V}_2^\perp = \mathcal{V}_2 \cap \overline{\mathcal{V}_1^\perp}$$



Some of the Related Work

- Extracting layers using EPI analysis: Criminisi et al. 2002
- Image Cube Trajectories (ICT): Feldmann et al. 2003
- Space-time video analysis, object and occlusion volumes: Konrad and Ristivojevic 2006
- Layered stereo with occlusions: Tomasi-Lin-Birchfield 1999

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Object Tube Extraction

- We assume Lambertian and opaque surfaces
- Minimize a global cost function

$$E_{tot} = \sum_{n=1}^{N} E_n = \sum_{n=1}^{N} \iiint_{\mathcal{V}_n^{\perp}} f_n(\vec{x}) d\vec{x}$$



 $f_n(\vec{x})$ is a measure of consistency with tube n

- Separated in 2 sub-problems:
 - Estimation of contours given the slopes of the lines
 - Estimation of the slopes given the contour

Estimation of the Region Boarders: Active Contours

• Consider a cost function of the type:

$$E(\Gamma) = \iint_{\Omega(\Gamma)} f(x,y) dx dy + \iint_{\overline{\Omega}(\Gamma)} g(x,y) dx dy + \int_{\Gamma} \lambda ds$$

 Gradient [KassWT:88, CasellesKS:97, ChanV:01, Jehan-BessonBA:01]

$$rac{dE(au)}{d au} = \int_{\partial\Omega} [f(x,y) - g(x,y) + \lambda\kappa] (ec{V}\cdotec{N}) ds$$

Steepest descent

$$\frac{\partial \vec{\Gamma}(\tau)}{\partial \tau} = [f(x,y) - g(x,y) + \lambda \kappa] \vec{N} = F \vec{N}$$



Estimation of EPI Tube Contours

- We assume the slope of the lines are known
- In the case where there are 2 layers (i.e. 1 layer and the background)

$$\begin{split} E_{tot}(\tau) &= \iiint_{\mathcal{V}_{1}^{\perp}(\tau)} f_{1}(\vec{x}) d\vec{x} + \iiint_{\mathcal{V}_{2}^{\perp}(\tau)} f_{2}(\vec{x}) d\vec{x} \\ \frac{dE_{tot}(\tau)}{d\tau} &= \iiint_{\partial\mathcal{V}_{1}^{\perp}} (f_{1}(\vec{x}) - f_{2}(\vec{x})) (\vec{W} \cdot \vec{M}) d\vec{\sigma} \\ \end{split}$$



y'

From 3D to 2D Using Epipolar Geometry

- The positions of the cameras are known
- The shape of the tubes are constrained
- Leads to 'constrained' surface evolution that can be implemented in a 2D subspace



The speed function

• The gradient becomes (α =1 for fronto-parallel planes)

$$rac{dE(au)}{d au} = \int_{\partial\Omega} (ec{V}\cdotec{N}) \Big[\underbrace{\int_t f_1(ec{x})dt}_{F_1(s)} - \underbrace{\int_t f_2(ec{x})dt}_{F_2(s)} \Big] ds$$

 The functional is set to be the normalized squared difference between the intensity and the mean of the line the layer belongs to

$$f_n(ec{x}) = rac{[I(ec{x}) - \mu_n(ec{x})]^2}{L_n(ec{x})}$$

$$ec{V} = [F_1(x,y) - F_2(x,y)]ec{N}$$

Occlusion/Disocclusion

- Its not that simple...
- For an occluded layer, the functional depends also on τ and the evolution equation has additional terms that are extremely complex.
- We alleviate the problem by separating tubes into 'to be occluded' and 'disoccluded' regions (similar to [KonradR] for video)



Dealing with Multiple Tubes

- Evolve one tube at a time
- By construction, tubes compete only with the other tubes they are occluding or disoccluding
- For example:

$$egin{array}{rcl} \mathcal{V}_1^ot &=& \mathcal{V}_1 \ \mathcal{V}_2^ot(au) &=& \mathcal{V}_2(au) \cap \overline{\mathcal{V}_1^ot} \ \mathcal{V}_3^ot(au) &=& \mathbb{R}^3 \cap (\overline{\mathcal{V}_1^ot} \cap \overline{\mathcal{V}_2^ot}(au)) \end{array}$$

$$E_{tot}(\tau) = \iiint_{\mathcal{V}_{2}^{\perp}(\tau)} f_{2}(\vec{x}) d\vec{x} + \iiint_{\mathcal{V}_{1}^{\perp}} f_{1}(\vec{x}) d\vec{x} + \iiint_{\mathcal{V}_{3}^{\perp}(\tau)} f_{3}(\vec{x}) d\vec{x}$$
$$\iiint_{\mathcal{V}_{2}^{\perp}(\tau)} f(\vec{x}) d\vec{x}$$

Estimation of Line Slopes (i.e. Disparity)

- Contours are fixed
- Find the slopes of the lines
- Done jointly over all the images
- Takes into account occlusions

$$f_n(\vec{x}) = \frac{[I(\vec{x}) - \mu_n(\vec{x})]^2}{L_n(\vec{x})}$$

$$\mu_n(\vec{x}) = \mu(x, y, t, d_n(x, y))$$

• Non-linear optimization problem

$$E_{tot} = \sum_{n=1}^{N} E_n = \sum_{n=1}^{N} \iiint_{\mathcal{V}_n^{\perp}} f_n(\vec{x}) d\vec{x}$$

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Layer Disparity Model

• Disparity map can be modeled as a bicubic spline [LinT:03]

$$d_n(x,y) = \sum_{i,j} D_n(i,j) b(x-i,y-j)$$





Overall optimization

- Initialize
- Iteratively alternate
 - Segmentation given layer depth maps
 - Evolve each contour iteratively with the level set method
 - Estimation of depth maps given segmentation
 - using classical optimization methods
- End when there is no significant decrease in energy

Simulation Results

- Slanted planes can be a problem in classical stereo since there is not a 1 to 1 mapping.
- Not a problem here since slanted planes are taken into account in the model









Preliminary Experimental Results

• Tiger image sequence (15 images covering 5 degrees)



Conclusions

- The plenoptic function provides a nice framework for multiview image analysis!
- New segmentation scheme for the Epipolar-Plane Image volume
 - Constrained surface evolution (uses knowledge of camera setup for added robustness)
 - Takes into account all the images simultaneously
 - Handles occlusions
 - Is scalable to higher dimensions

Ongoing and Future Research

- Extension to the 4D and 5D cases: More degrees of freedom to the camera locations
 - Segmentation of hyper-volumes
- Scene interpretation
 - What can we learn about the scene from the shape of the tubes?
- Compression
 - Layer based representations and/or linear transforms taking into occlusions and disparities (along the EPI lines)

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Questions?