3D Scene Modeling Using Multiple Cameras

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Context

- Computer Vision: to give visual analysis capabilities to computers.
- 3D Scene Modeling: analysis of 3D scenes composed of real objects, possibly moving and deforming
 - 1. Shape Modeling.
 - 2. Motion Modeling.
 - 3. Motion Semantic Modeling (e.g. actions, activities).

• Applications :

- 3D Contents Production: TV3D, Virtual Reality, Virtualization, Interactions.
- Intelligent Environments: smart rooms,
- Surveillance.



Example : shape modeling





Example : motion modeling





Tony Hawk

Example : interactions





Outline

- Shape Modeling:
 - Acquisition Systems.
 - Geometric Approaches.
 - Photometric and geometric Approaches.
- Motion Modeling:
 - Marker Based Approaches.
 - Marker less Approaches.
- Action Modeling:
 - Model Based Approaches.
 - Holistic Approaches.



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Shape Modeling





Acquisition Systems







HD Cameras

- + quality
- cost, complexity

USB Cameras

+ cost, complexity

- quality

Firewire Cameras

+ cost, complexity

-+ quality

Issues:

- Synchronisation: improve significantly the modeling.
- Calibration: complex with lots of cameras.
- Environnement: a controled environnement is frequently required.
- Computation: important power computation required for complex models or real time applications.



2D + Depth Systems



Time of flight cameras or Stereo cameras \neq 3D Modeling



Structured Light Systems







Jean-Yves Bouguet and Pietro Perona Caltech

Technical University of Braunschweig Institute for Robotics and Process Control



Turn Tables



- Single Camera
- Calibration simplified
- Static System



Acquisition System



Stanford Spherical Gantry http://www-graphics.stanford.edu/projects/gantry/ -> Middlebury Univ. 3D Datasets



Grimage Platform



• **Acquisition** : 6-10 caméras FireWire (SONY DFW-VL500, DFW-X700, AVT Marlin). Synchronisation hardware (trigger signal).

- Visualisation : wall screen (2mx2.7m, 4096x3072 pixels) 16 projectors Sanyo LCD.
- **Computation:** PC cluster, 11 dual-xeon et 16 dual opteron connected through a double Ethernet gigabit.
- **Mobile Version:** 4-6 Firewire cameras, 6 mini-PCs and 1 laptop.



Geometric Modeling

- No *a priori* knowledge otherwise matching between model and observations.
- 2D primitive extraction:
 - points, lines -> 3D point clouds





- Régions (silhouettes) -> surfaces, volumes





Silhouette Based Approaches



[B.W. Baumgart 1975, Stanford]















































Geometric Modeling Visual Hull

Formal Definition [Laurentini'94] : The visual Hull is the maximal volume coherent with (that entirely projects onto) a set of silhouettes.

- The visual hull is the intersection of viewing cones.
- In the case of infinite viewpoints outside the convex hull of the observed objects, the visual hull corresponds to the observed surfaces without their concavities.
- In the discrete case (polygonal silhouettes, finite nb of viewpoints), the visual hull is a polyhedron.



Geometric Modeling Visual Hull



[Grimage 2005]



MINRIA

Geometric Modeling Silhouette Extraction





- Silhouettes are regions in the images where object of interest project.
- Silhouettes are estimated using low-level processes.
- Silhouettes give information on the observed surfaces.



Geometric Modeling Silhouette Extraction

- Segmentation :
 - Manual.
 - Automatic :
 - Contour tracking
 - Chroma keying (blue or green background != skin color)
 - Background subtraction (static background)



Geometric Modeling Silhouette Extraction

- Background subtraction:
 - Statistical background model
 - Gaussian
 - Gausian mixtures
 - Non parametric: histograms.
- Issues:
 - Image digitalization (noise);
 - Color ambiguities between background and foregroud objects;
 - Luminosities changes, etc.





Geometric Modeling Estimating the visual hull



Volumetric Approaches

– voxels





Surface based Approaches





Image based Approaches





Geometric Modeling-VH Voxel based approaches



- Elimination criterion for voxels
 - -> project outside a silhouette
 - Tradeoff precision/complexity



Geometric Modeling-VH Voxel based approaches



Adaptive data structures : octrees [Szeliski 93]



Geometric Modeling Estimating the visual hull



Volumetric Approaches

– voxels





Surface based Approaches





Image based Approaches




Geometric Modeling-VH Surface based approaches



- B.W. Baumgart 74:
 - 1st approach, CSGtype (constructive solid geometry);
 - polygonal contours;
 - visual hull computed as the intersection of the viewing cones.



Geometric Modeling-VH Surface based approaches



- Franco-Boyer 03:
 - The identification of the exact structure of the discrete visual hull allows exact computations.



Geometric Modeling-VH Surface based approaches



The case of a sphere observed from 4 viewpoints (4 colors)

- 3 computation steps:
- 1. Viewing segments(contributions of a silhouette vertex to the visual hull)
- 2. The mesh connecting viewing segments.
- 3. Facets by going along the oriented mesh.



Geometric Modeling-VH Computation complexity

- Polyedron intersection computation is expensive.
- Efficient algorithms:
 - Discretisation;
 - Compute in 2D.



Geometric Modeling-VH Image based approaches



[Matusik&al. Siggraph 00]



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visual hull.

Geometric Modeling-VH Image based approaches



[Matusik&al. Siggraph 00]



Geometric Modeling Visual Shapes

 \mathbf{C}_{i}



• The Visual hull is an extended bounding box.

• The idea is to search for a surface inside that bounding box and tangent to it. Tehs surfaces define the visual shapes.

• The visual shapes are, in general, a better geometric approximation of the observed than the visual hull.





Geometric Modeling Visual Shapes





• The approaches mentioned before assume exact calibration and exact silhouettes (deterministic approaches). Such data are not exact in practice (i.e. noisy).



2 images of background probabilities (white prob = 0, black prob = 1)







[Franco and Boyer 2005]

- Idea: we wish to find the content of the scene from images, as a probability grid
- Modeling the forward problem explaining image observations given the grid state is easy. It can be accounted for in a *sensor model*.
- Bayesian inference enables the formulation of our initial inverse problem from the sensor model
- Simplification for tractability: independent analysis and processing of voxels



- Unreliable silhouettes: do not make decision about their location
- Do sensor fusion: use all image information simultaneously







- I: color information in images
- B: background color model
- F: silhouette detection variable (0 or 1): hidden
- Ox: occupancy at voxel X (0 ou 1)

Sensor model:

$$P(I \mid O_X) = \sum_F P(I \mid F, B) P(F \mid O_X)$$































Silhouette consistency constraint:

$$\left[\bigcup_{\mathrm{Im}} \left[S_i - P_i(VH(\mathrm{Im}))\right] = \varnothing\right]$$

- Si is the silhouette in image i;
- Pi is the projection in image i;
- VH(Im) is the visual hull of the set of images Im.

• Using geometric information only (e.g. locations, orientations) does not allow to model precisely from images. To this purpose photometric information, or appearance, must be considered.

 $_{\odot}$ The objective is therefore to search for the geometry and the photometry that are consistent with the observations (the images).

Hypothesis :

- The observed surfaces reflect light with respect to a simplified model.
- Diffuse model (lambertian surfaces) and specular.
- constant BRDF:

[image from Wikipedia] Bidirectionnal reflectance distribution function (BRDF)

A Lambertian surface appears equally bright in any directions ——— photometric criterion:

Projections of a 3D surface point must exhibit similar photometric properties in any view where the point is visible.

[Kutulakos & Seitz IJCV00]

Approaches :

• Voxel coloring/Space carving [Seitz & Dyer'97] : voxels not photoconsistent are carved.

Approaches:

• Multi-View Stereo

[Hernandez'04,Furukawa'06,Pons'07] : an initial surface estimation is optimizeddeformed with respect to a photometric criterion.

- Lagragian methods: mesh deformations
- Eulerian methods: discrete grids
 -> level sets

Voxel Coloring

Straightforward situations for visibility: the observed scene is outside the camera center convex hull.

------ voxels are eliminated in order of increasing disctances to the camera convex hull

[Seitz] : http://www.cs.cmu.edu/~seitz/ vcolor.html

Voxel Coloring

S={} /* initial set of colored voxels is empty */
for i = 1 to r do /* traverse each of r layers */
foreach V in the ith layer of voxels do
project V into all images where V is visible
if sufficient correlation of the pixel colors
then add V to S

[Seitz] : http://www.cs.cmu.edu/~seitz/ vcolor.html

Voxel Coloring

[Seitz] : http://www.cs.cmu.edu/~seitz/ vcolor.html

Voxel Coloring

General case (Space carving) :

Voxels= full grid

Do until convergence

foreach V in Voxels do

if V on the surface

project V into all images where V is visible

 $\boldsymbol{\mathsf{if}}$ not sufficient correlation of the pixel colors

then carve V

Complexity !

Voxel Coloring

Sweeping planes

MultiView Stereo

An initial surface estimation is optimized-deformed with respect to a photometric criterion:

- Eulerian methods: values of an implicit function are evolved over a discrete-cartesian grid and with respect to a cost function (e.g. level set approaches)
 - no surface parametrization required
 - discretization issues (tradeoff precision/complexity)
 - better than space carving since voxels are not handled independently
- Lagragian methods: surface deformations in order to decrease a cost function
 - requires a surface parametrization (e.g. a mesh)

Photometric and geometric modeling

MultiView Stereo





[A. Hilton & J. Starck'06, Surry, UK]



Photometric and geometric modeling







Modeling: temporal aspects

The idea is not to model indenpedently over time sequences but to evolve instead a single model



[B. Bickel M. Botsch R. Angst W. Matusik M. Otaduy H. Pfister M. Gross Siggraph'07]

Interests :

- 1. Constraint model with nice geometric properties (e.g. correct topology)
- 2. Temporal correspondences as a byproduct (-> tells about motion)



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- Issues:
 - Precision, robustness.
 - a priori knowledge (motion model) required.
 - Markers required or not.





- Optical methods with markers:
 - Observations: markers are tracked in calibrated images
 - Passive markers: retroreflective markers
 - Active markers: markers emit their own lights for identification (e.g. different pulses per markers).
 - Motion recovery: a motion model (e.g. a skeleton) is matched with the marker locations.
 - Applications: "Motion capture" of bodies (with articuled models), faces (with deformable meshes), etc.
- Non optical methods with markers also exist: inertial or magnetic systems for instance.



- Marker based approaches, the Vicon system:
 - High precision cameras (up to 16 Mpixels) and high frequencies
 (> 400Hz).





- Markerless approaches:
 - Observations: 2D (contours, silhouettes), 3D (points, shapes) in a single or multiple calibrated images.
 - Motion recovery: the motion model is matched to the observations:
 - parametric motion models: find the best model parameters (e.g. joint angles) such that the model explains the observations.
 - non parametric models: find the best model (in the database) that explains the observations.
 - Applications: motion capture of bodies, faces, etc.



- Parametric motion model and 3D observations (locations and surface normals)
- Articulated model composed of ellipsoids.











Parametric model approach:

- Skeleton model: focus on the motion not the shape.
- 3 step approach:
 - 1. Estimate shape.
 - 2. Compute the medial axis of the estimated shape (discrete version of the medial axis).
 - 3. Find skeleton parameters such that distances between the medial axis and the skeleton are minimum.



3D Marker-less Skeleton-based Motion Tracking

Paper No.: 107



• Learning based approaches: observations (silhouettes for instance) are matched to pre-learned models with known motion.

→ Allow for non parametric motion models, e.g. motions are represented by a set of key pose transitions.



[Aggarwal & Triggs'04]



- Generative approaches (e.g. kinematic models): poses are determined by inference using a parametric model (i.e. continuous).
 - Pros: allow to represent large spaces of motions.
 - Cons: difficult to find a pose without a good initialization
- Non generative approaches: poses are recognized in a learned database.
 - Pros: initialization is not an issue.
 - Cons: limited spaces only can be represented.



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Action Modeling: Applications



Entertainment



HCI



Surveillance



Ambient Intelligence





Sports



Group Actions



• Issues:

- What is an action ? Posture and Dynamics
- Action space dimensions.
- Viewpoint dependence .
- Occlusions.
- Computation time.





- Strategies :
 - Model Based approaches: Actions are modeled in the parameter space associated to an a priori model.

- Holistic Approaches: (Holism [def.]: the tendency in nature to form wholes that are greater than the sum of the parts through creative evolution) : actions are described with respect to observations without the help of an intermediate model.



Action Modeling: Posture





Action Modeling: Dynamics







Dynamics E. Boyer - UFRIMA

Action Modeling: View-Independence

• Use 3D information







Time (keyframes)

Use View-invariance



Geometrical Invariants computed from 5 points in plane [Parameswaran05]





Rank constraints on point correspondences [Rao03]



Model based approaches:

Intermediate representation, e.g. kinematic model



Action Modeling: Holistic approaches



Motion History Images [Bobick96]



Space-Time Volumes [Blank05]



Optical Flow [Efros03]



Space-Time Interest Points [Laptev05]



3D Exemplars [Weinland07]



²D synthetic exemplars [Lv07]





[A.A. Efros, A.C. Berg, G. Mori and J. Malik'03]



Match

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3D Exemplar based approach
→ The kinematic model is replaced by learned exemplars.
→ Generative approach.

3D Exemplar

[Weinland & al.'07]



2D Observations



Exemplars associated with a database composed of 11 actions-10 acteurs



Distance Function

- Silhouette-to-Silhouette Matching
 - Assumes background subtraction
 - d = Euclidean distance between silhouettes
- Silhouette-to-Edge Matching
 - No background subtraction
 - d = Chamfer Distance between silhouette and edge image







[Weinland & al.'07]

Action Modeling: Dynamics

• Traditional representations explicitly or implicitly model dynamics.



Weizmann-Dataset [Blank05]

Is dynamic required to identify an action ?



Exemplar-based Embedding



• Intuition: similar sequences will yield similar proximities to discriminative exemplars. Hence their point representation in Rⁿ should be close.



Results

Weizmann dataset : 10 actions, 9 actors, 2D silhouettes



Silhouettes



edges



Conclusion

- Shape Modeling :
 - Industrial step (numerous start-ups and industrial projects such as 3D TV)
 - Difficulties still to be solved: appearance, temporal coherence are challenging.
- Motion Modeling:
 - Markerless approaches not really robust.
 - Difficulties: adequation model->observations hence holistic approaches ?!



Conclusion

- Action Modeling:
 - Simple actions only yet, applications to games.
 - Difficulties: the semantic interpretation is complicated due to the dimension of the space considered. To reduce such dimension, use context ?

