



V³D² Workshop 2003 Frankfurt

3D Model Retrieval

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DFG Teilprojekt „Ähnlichkeitssuche durch Gestaltcharakterisierung auf 3D Datenbanken”

DFG SPP V3D2 Phase 3

Joint project of

Multimedia Signal Processing Group

Prof. Dietmar Saupe

Dejan V. Vranić (starting date: April 2002)

Databases, Data Mining and Visualization Group

Prof. Daniel Keim

Tobias Schreck (starting date: November 2002)

University of Konstanz, Dept. of Comp. Science

Overview

Part A: 3D Model Retrieval (Dejan V. Vranić)

- Problem of 3D Model Retrieval
- Requirements
- Approach
- Demo
- Exposition of a small group of feature vectors

Part B: Kombination von Feature-Vektoren (Tobias Schreck)

- Problemstellung
- Untersuchungsansatz
- Effektivitätsresultate
- Entwicklungsziel: Query Processor

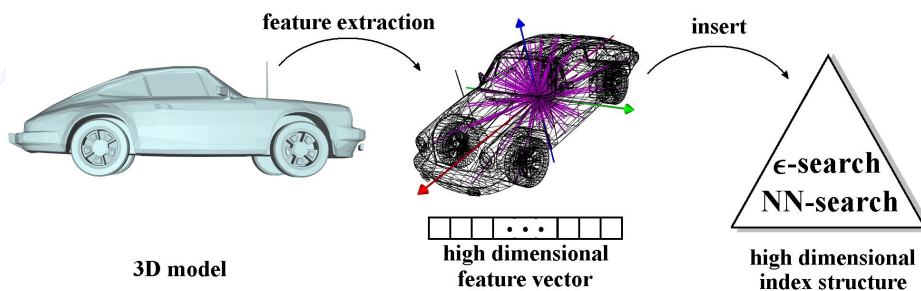
Content-based retrieval

♦ Goal

- Use a multimedia object (image, movie clip, audio file, or 3D-model) as a key;

♦ Approach

- Extract appropriate features (descriptors (D)) from objects automatically and use them as points in a search space;



Typical retrieval algorithm

◆ Normalization

- Problem: 3D objects are defined in arbitrary coordinates and units;

◆ Feature Extraction

- Problem: describe object in few variables yet with sufficient discriminant power;

◆ Similarity search

- Problem: for a given query object efficiently find nearest neighbors (curse of dimensionality).

3D-shape descriptors criteria

- ◆ Invariance with respect to translation, rotation, scaling, and reflections;
- ◆ Robustness with respect to level-of-detail (different tessellations);
- ◆ Robustness with respect to noise and outliers;
- ◆ Efficient feature extraction and retrieval;
- ◆ Multiresolution feature representation;
- ◆ Discriminating of shape similarities.

Demo

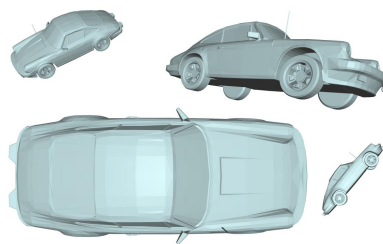
Invariance w.r.t. similarity transforms

Three approaches:

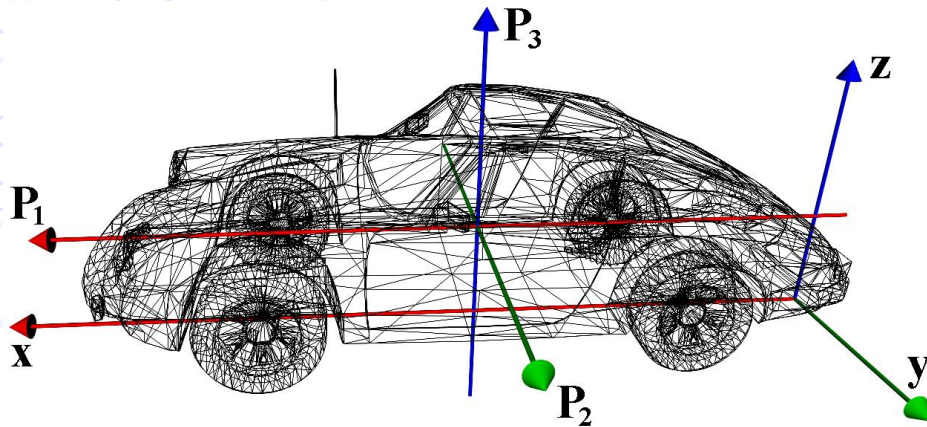
- ◆ **Finding canonical coordinate frame**
 - Principle Component Analysis (PCA);
- ◆ **Aligning objects pairwise**
 - Fitting objects (time-consuming);
- ◆ **Defining Ds that possess the invariance inherently**
 - Considering some relative features of sets of points or triangles (curvature spectrum, “shape distributions”, representation of topology);
 - By summing up (or averaging) features over a group of transformations (e.g., FFT on a sphere).

Normalization (pose estimation)

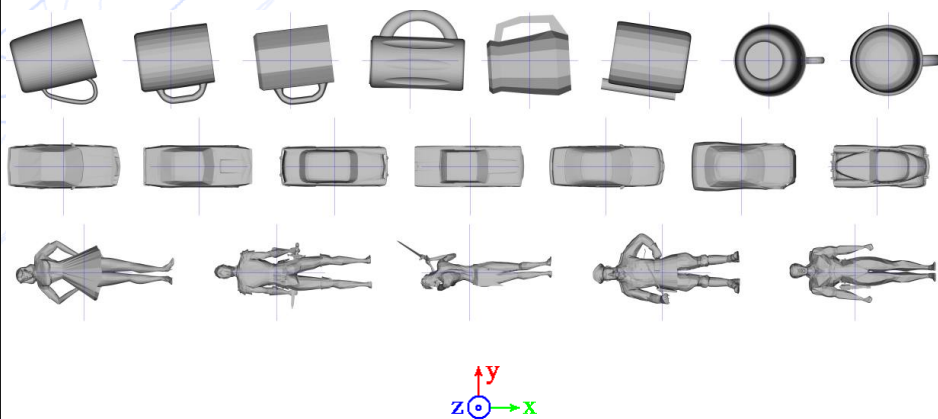
- ◆ **Purpose**
 - Finding a canonical frame (i.e., position and orientation);
- ◆ **Result**
 - Translation, rotation, scaling, and reflection invariance of descriptors;
- ◆ **Tools**
 - Center of mass,
 - “Continuous” PCA (CPCA) (rotation),
 - Test based on moments (reflections), and
 - Formula for calculating the scale factor.



The outcome of the CPCA



Examples of pose estimation using the CPCA



3D-shape descriptors

Classification of our 3D-shape descriptors:

- ◆ **Geometry-based**
 - Volume,
 - Voxel, and
 - Moment-based Ds;
- ◆ **Image-based**
 - Ray-based,
 - Depth-buffer,
 - Silhouette,
 - Shading, and
 - “Complex” Ds.

Representation: spatial or frequency (spectral) domain.

Ray-based D (spatial domain)

◆ Idea

- Sample a 3D-model in regularly spaced directions and treat these samples as components of a feature vector.

◆ Realization

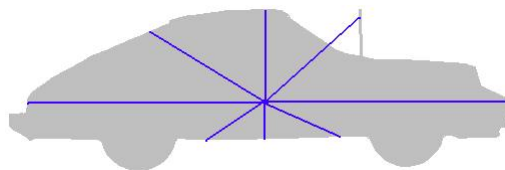
- Measure extent of a model

$$r: S^2 \rightarrow R$$

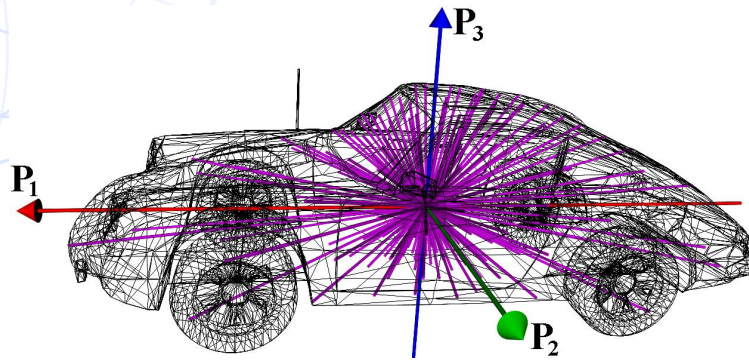
$$\mathbf{u} \mapsto \max\{r \geq 0 \mid r\mathbf{u} \in I \cup \{\mathbf{0}\}\},$$

\mathbf{u} – a directional unit vector

I – the point set of the model



Ray-based feature vector



Features as functions on a sphere

[Saupe, Vranić, MMSP 2001, DAGM 2001]

◆ Spherical harmonics

- Use Fourier coefficients of a function on the 2-sphere to form a feature vector with an embedded multiresolution representation;

◆ Realization

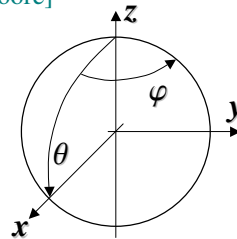
- Sample a 3D-model in appropriate radial directions \mathbf{u}_{ij} ($i, j = 0, \dots, n-1$) and treat these samples as values of a function on the 2-sphere.

Spherical harmonics

[Heally, Rockmore, Kostelec, Moore]

◆ Functions on the 2-sphere

$$\begin{aligned} f: S^2 \rightarrow \mathbb{C} \\ (\theta, \varphi) \rightarrow f(\theta, \varphi), \\ 0 \leq \theta \leq \pi, \\ 0 \leq \varphi < 2\pi; \end{aligned}$$

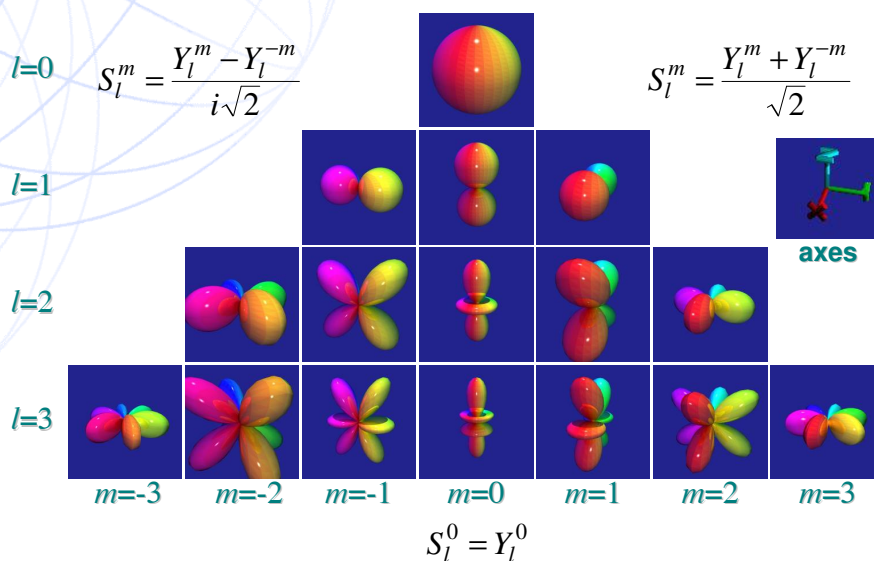


◆ Spherical harmonics $Y_l^m: S^2 \rightarrow \mathbb{C}$, $|m| \leq l$ provide orthonormal basis of $L^2(S^2)$;

◆ Subspace X_l spanned by $\{ Y_l^m \mid -l \leq m \leq l \}$ is invariant w.r.t. rotations of the sphere;

Spherical harmonics visualization

(Courtesy of Quantum Chemistry Group, University of Oviedo)



Fourier transform on the sphere

- ◆ Expansion of the function $f(\theta, \varphi)$

$$f \approx \sum_{l \geq 0} \sum_{|m| \leq l} c(l, m) \cdot Y_l^m;$$

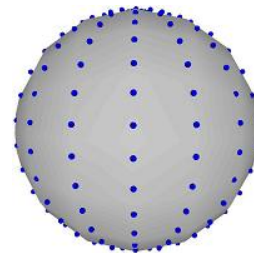
- ◆ $c(l, m) = \langle f, Y_l^m \rangle$ is (l, m) - Fourier coefficient;
- ◆ Use special FFT algorithms (Healy et al), public domain source code.

Frequency domain representation

I = point set of (filled-in) triangles ,

$$r : S^2 \rightarrow R$$

$$\mathbf{u} \mapsto \max \{r \geq 0 \mid r\mathbf{u} \in I \cup \{\mathbf{0}\}\},$$

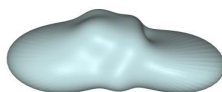


$$\mathbf{u}_{ij} = (x_{ij}, y_{ij}, z_{ij}) = (\cos \varphi_i \sin \theta_j, \sin \varphi_i \sin \theta_j, \cos \theta_j),$$

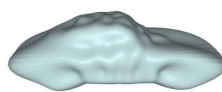
$$\varphi_i = 2i\pi / n, \quad \theta_j = (2j+1)\pi / 2n, \quad i, j = 0, \dots, n-1.$$



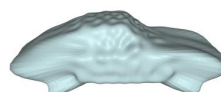
Original



8^2 harmonics



16^2 harmonics



24^2 harmonics

Absolute values of coefficients

$$c(l, m) = \overline{c(l, -m)} \Rightarrow |c(l, m)| = |c(l, -m)|$$

$$l = 0$$

$$1.161$$

$$l = 1$$

$$0.064 \quad 0.163 \quad 0.064$$

$$l = 2$$

$$0.213 \quad 0.037 \quad 0.373 \quad 0.037 \quad 0.213$$

$$l = 3$$

$$0.017 \quad 0.008 \quad 0.010 \quad 0.008 \quad 0.010 \quad 0.008 \quad 0.017$$

$$m=-3 \quad m=-2 \quad m=-1 \quad m=0 \quad m=1 \quad m=2 \quad m=3$$

Dimension: $n(n+1)/2, l < n$.

“Complex” feature vector

[Vranić, Saupe ICME 2002]

I = point set of (filled-in) triangles,

$$\mathbf{u}_{ij} = (x_{ij}, y_{ij}, z_{ij}) = (\cos \varphi_i \sin \theta_j, \sin \varphi_i \sin \theta_j, \cos \theta_j),$$

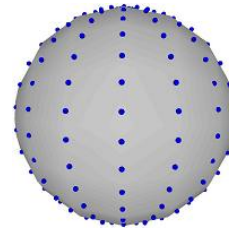
$$\varphi_i = 2i\pi/n, \quad \theta_j = (2j+1)\pi/2n, \quad i, j = 0, \dots, n-1,$$

$$r : S^2 \rightarrow \mathbb{C}$$

$$r(\mathbf{u}) = x(\mathbf{u}) + i y(\mathbf{u}),$$

$$x : S^2 \rightarrow [0, +\infty) \in \mathbb{R}, \quad y : S^2 \rightarrow [0, 1] \in \mathbb{R},$$

$$x(\mathbf{u}) = \max \{ x \geq 0 \mid x \mathbf{u} \in I \cup \{\mathbf{0}\} \}, \quad y(\mathbf{u}) = \begin{cases} 0, & \text{if } x(\mathbf{u}) = 0 \\ \mathbf{u} \cdot \mathbf{n}(\mathbf{u}), & \text{otherwise} \end{cases}.$$



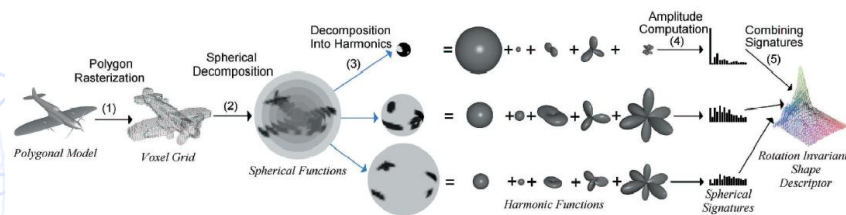
◆ Find spherical power spectrum for $r(\mathbf{u})$ and for $l < n$ use all coefficient magnitudes as components of the feature vector

Dimension: $n^2, l < n$.

Related work

- ♦ **[Paquet et al. '98]**
 - Cords-based, moments-based, and wavelet transform-based descriptors (Ds);
- ♦ **[Suzuki et al. '00]**
 - Grid-based and “rotation invariant” Ds;
- ♦ **[MPEG-7]**
 - 3D shape spectrum and 2D-3D multiple view;
- ♦ **[Hilaga et al. '01]**
 - Topology matching;
- ♦ **[Funkhouser et al. '01, '02, '03]**
 - Reflective symmetry, shape distribution, spherical harmonics, skeletal graphs.

Harmonic 3D descriptor (H3D)



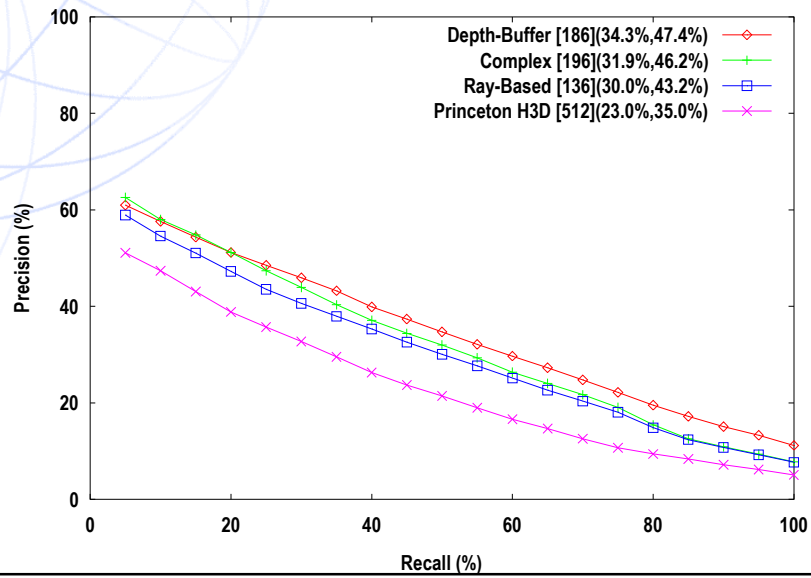
- Spherical harmonics Y_l^m ($S^2 \rightarrow \mathbb{C}$, $|m| \leq l$) provide orthonormal basis of $L^2(S^2)$;

- ♦ Subspace X_l spanned by $\{ Y_l^m \mid -l \leq m \leq l \}$ is invariant w.r.t. rotations of the sphere;

$$f_{i,l} = \sqrt{\sum_{|m| \leq l} |c_i(l,m)|^2}, \quad 0 \leq l < n, \quad 1 \leq i \leq R$$

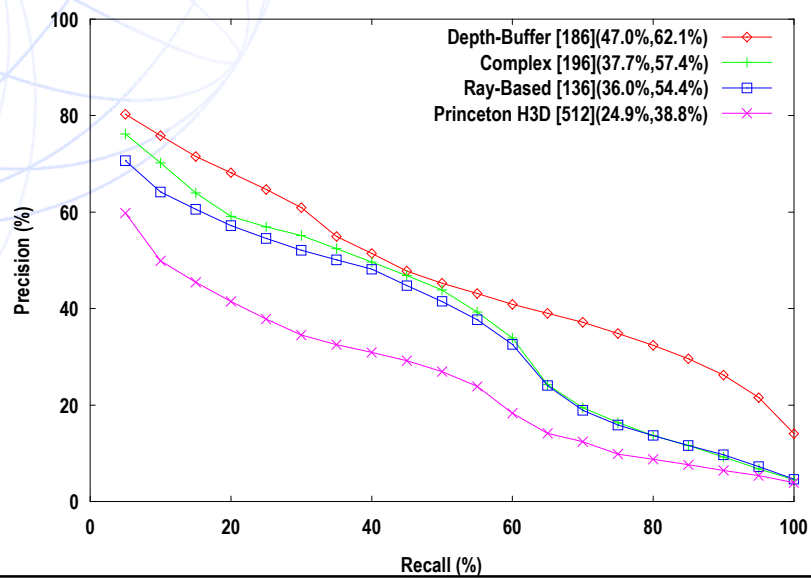
Comparison of performance

Our Database - Reclassified (472 Queries)



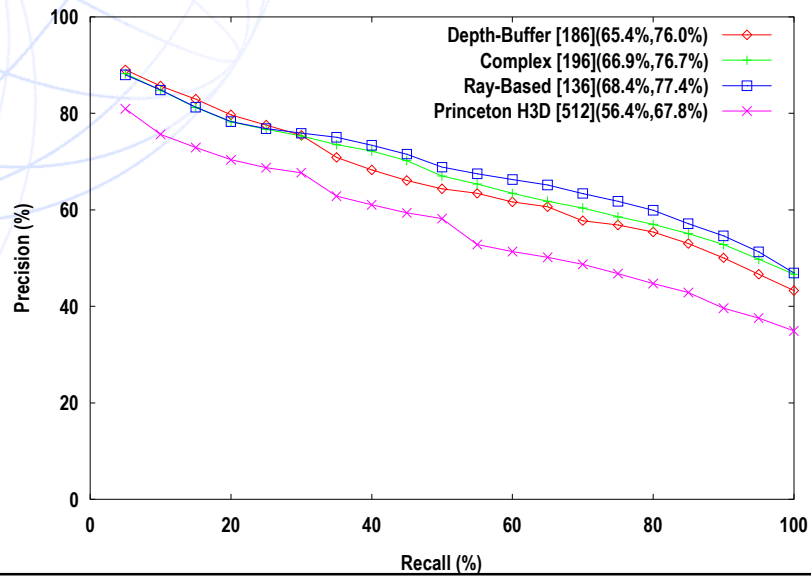
Comparison of performance

Our Database - Bottles, Cars, Missiles, Planes, and Swords (173 Queries)



Comparison of performance

MPEG-7 Database (227 Queries)



Comparison of performance

Reclassified MPEG-7 Database (222 Queries)

