

PARTIAL 3D OBJECT RETRIEVAL

Guide: Dr. Shankar G.

Presented by:

1. Apoorva Malemath
2. (USN-01FE16BCS041)

Outline

1. Motivation and Scope
2. Problem Statement
3. Objectives
4. Related Works
5. System Model
6. Methodology
7. Implementation
8. Results
9. Conclusions and Future Scope

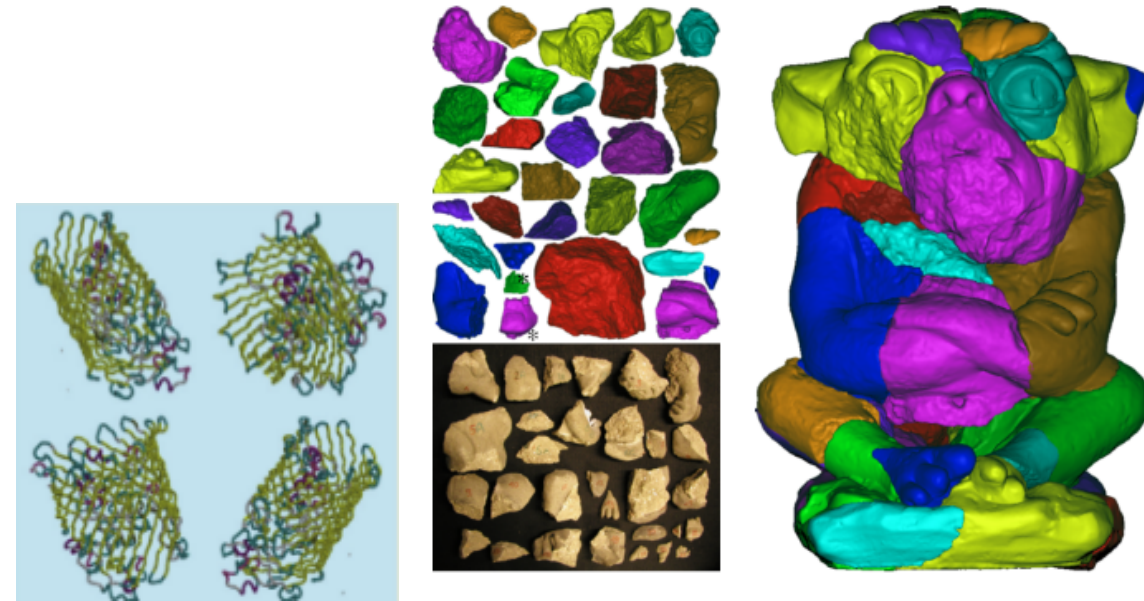
Motivation and Scope

Increasing amount of 3D data.

Ability of capture devices to **produce low-cost multimedia data.**

Applications:

1. Cultural Heritage
2. Archaeology
3. 3D protein retrieval and classification
4. 3D retrieval for museums



Problem Statement

To develop a technique to retrieve the similar objects given a partial query.

Input



Output



Partial Query

Retrieved Objects

Objectives

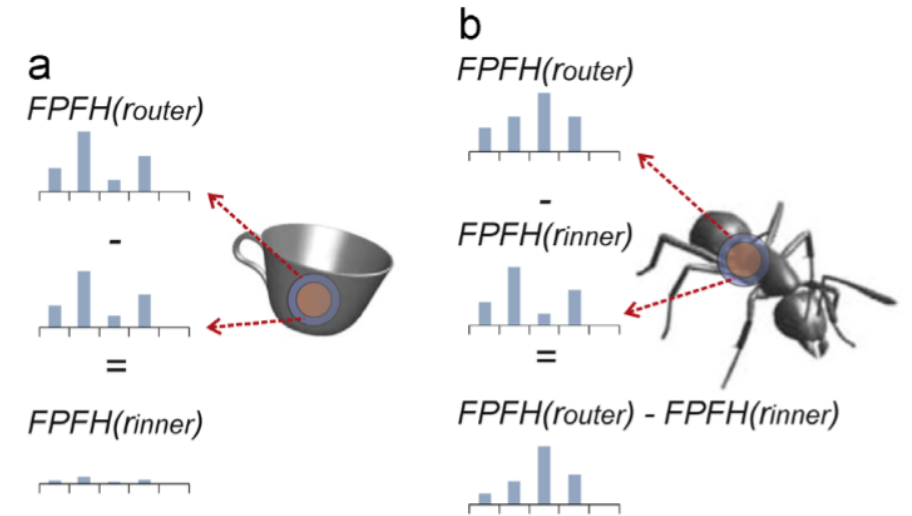
1. To extract the features from the given 3D objects.
2. To perform encoding of the extracted features.
3. To match the input query with the dataset.
4. To evaluate the performance of the proposed technique.

Related Works

Fisher encoding of Differential Fast Point Feature histograms for Partial 3D Object Retrieval, 2016 [1]

The proposed system is based on hybrid shape matching scheme so as to account for both global and local shape similarities.

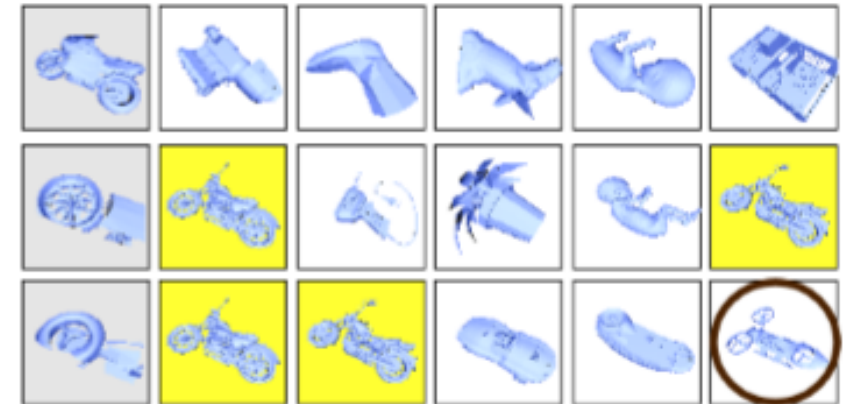
- Differential fast point feature histogram
- Fisher vector.



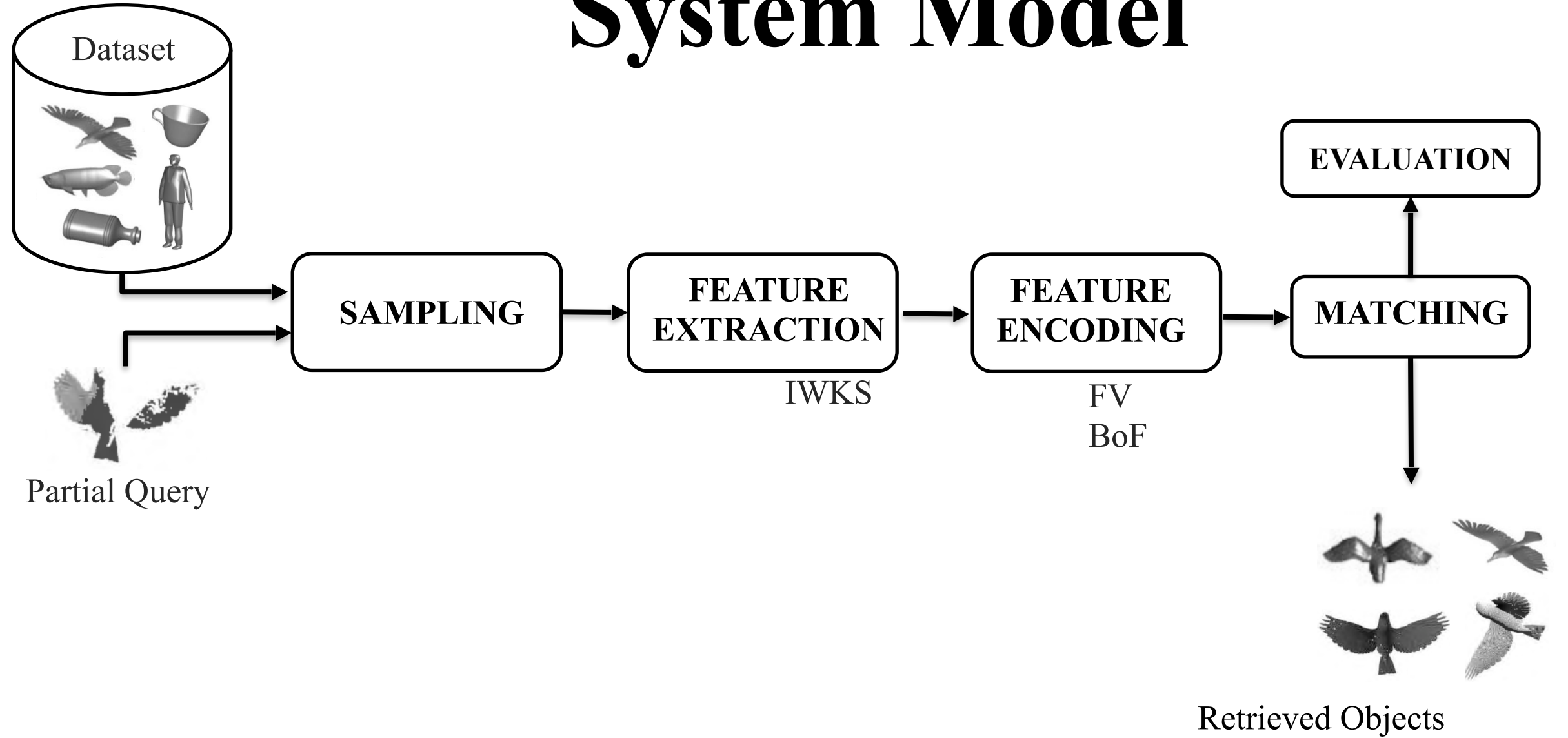
A Query-by-Example Concept and User Interface for Global and Partial 3D Object Retrieval, 2009. [2]

It is based on the idea to apply an existing global 3D descriptor on both whole models and model segments.

- Partial similarity - Local to local correspondences
- Model segmentation



System Model



Methodology

1. Feature Extraction

Improved Wave Kernel Signature (IWKS)

- Characterize a point by the average probabilities of quantum particles of different energy levels.
- Energies correspond to frequency.
- Eigen values of the LBO.
- Cube root scaling.
- Curvature aggregation.

Methodology

2. Feature Encoding

- **Bag-of-features (BOF)**
 - Convert the descriptors to codewords.
 - K-means clustering
- **Fisher Vector (FV)**
 - Gaussian Mixture Model used as visual code book.
 - Captures the average first and second order differences between descriptors and the centers of GMM.

Methodology

- For K^{th} GMM, we define,

$$u_k = \frac{1}{N\sqrt{\pi_k}} \sum_{i=1}^N q_{ik} \sum_k^{-1/2} (x_i - \mu_k)$$

$$v_k = \frac{1}{N\sqrt{2\pi_k}} \sum_{i=1}^N q_{ik} [(x_i - \mu_k) \sum_k^{-1} (x_i - \mu_k) - 1]$$

- Fisher encoding of feature vector is given by,

$$f = [u_1^T, v_1^T, \dots, u_k^T, v_k^T]$$

Methodology

3. Matching

- **Manhattan Distance (L1- norm)**

$$d = \sum_{i=1}^N |a_i - b_i|$$

- **Euclidean Distance (L2-norm)**

$$d = \sqrt{\sum_{i=1}^N (a_i - b_i)^2}$$

Methodology

- **Cosine Distance**

$$d = \frac{\sum_{i=1}^N (x_i y_i)}{\sqrt{\sum_{i=1}^N x_i^2} \sqrt{\sum_{i=1}^N y_i^2}}$$

- **Earth Mover Distance**

$$d = \frac{\sum_{i=1}^M \sum_{j=1}^N f_{i,j} d_{i,j}}{\sum_{i=1}^M x_i^2 \sum_{j=1}^N f_{i,j}}$$

Methodology

4. Performance Evaluation

Nearest Neighbor (NN)

- Ratio of objects in the query's class that are present in the top N matches. $N=1$ is considered.
- **First Tier (FT)**
- Given a query, it is the precision when C objects have been retrieved, where C is the number of relevant objects. $K=|C|-1$ is considered for first tier.

Methodology

Second Tier (ST)

- Given a query, it is the precision when $2 \times C$ objects have been retrieved, where C is the number of relevant objects in the 3D dataset. $K = 2 \times (|C| - 1)$ for the second tier.
- **E-measure**
- F-measure, which is the weighted harmonic mean of precision and recall. F-measure is defined as:

$$F = \frac{(1 + \alpha) \times \textit{precision} \times \textit{recall}}{\alpha \times \textit{precision} + \textit{recall}}$$

Where α is the weight. Let $\alpha = 1$ then,

$$F = \frac{2 \times \textit{precision} \times \textit{recall}}{\alpha \times \textit{precision} + \textit{recall}}$$

Methodology

E-measure is defined as,

$$E = 1 - F$$

Discount cumulat $E = 1 - \frac{1}{\frac{1}{precision} + \frac{1}{recall}}$

- DCG penalizes the objects which are relevant and appear lower in the result list.

$$DCG_r = relevant_1 + \sum_{i=2}^r \frac{relevant_i}{\log_2(i+1)}$$

Dataset

SHREC 2011 Range Scan
Classes : 50
Objects per class : 20



Results

Encoding	K	NN	FT	ST	E	DCG
FV	10	0.0000	0.0180	0.0350	0.0223	0.3085
	20	0.0067	0.0107	0.0303	0.0149	0.3037
	25	0.0067	0.0107	0.0310	0.0162	0.3068
	50	0.0000	0.0097	0.0310	0.0162	0.3037
	100	0.0000	0.0093	0.0293	0.0151	0.3064
	125	0.0133	0.0133	0.0347	0.0200	0.3081
	150	0.0133	0.0117	0.0303	0.0164	0.3079
	200	0.0133	0.0160	0.0317	0.0179	0.3102
BoF	10	0.0000	0.0143	0.0243	0.0164	0.3056
	50	0.0200	0.0217	0.0417	0.0246	0.3170
	100	0.0000	0.0140	0.0317	0.0182	0.3104
	150	0.0200	0.0177	0.0327	0.0205	0.3128
	200	0.0267	0.0257	0.0430	0.0285	0.3209
	300	0.0400	0.0257	0.0437	0.0282	0.3218
	500	0.0533	0.0240	0.0443	0.0290	0.3219
	1000	0.0600	0.0527	0.0527	0.0333	0.3222

Performance metrics obtained using L1-Norm for matching

Results

Encoding	K	NN	FT	ST	E	DCG
FV	10	0.0067	0.0177	0.0337	0.0221	0.3086
	20	0.0000	0.0180	0.0350	0.0223	0.3085
	25	0.0000	0.0183	0.0357	0.0218	0.3078
	50	0.0000	0.0157	0.0310	0.0195	0.3068
	100	0.0000	0.0140	0.0310	0.0187	0.3079
	125	0.0000	0.0140	0.0283	0.0182	0.3077
	150	0.0067	0.0160	0.0287	0.0190	0.3098
	200	0.0000	0.0153	0.0287	0.0192	0.3088
BoF	10	0.0067	0.0403	0.0690	0.0446	0.3362
	50	0.0200	0.0200	0.0387	0.0226	0.3171
	100	0.0000	0.0160	0.0407	0.0218	0.3119
	150	0.0067	0.0203	0.0383	0.0231	0.3114
	200	0.0333	0.0247	0.0460	0.0285	0.3187
	300	0.0400	0.0190	0.0450	0.0295	0.3198
	500	0.0333	0.0267	0.0503	0.0313	0.3212
	1000	0.0533	0.0267	0.0527	0.0333	0.3222

Performance metrics obtained using L2-Norm for matching

Results

Encoding	K	NN	FT	ST	E	DCG
FV	10	0.0133	0.0140	0.0267	0.0167	0.3064
	20	0.0133	0.0137	0.0257	0.0149	0.3073
	25	0.0133	0.0177	0.0307	0.0187	0.3103
	50	0.0133	0.0180	0.0323	0.0218	0.3108
	100	0.0133	0.0153	0.0333	0.0195	0.3065
	125	0.0133	0.0167	0.0320	0.0197	0.3086
	150	0.0133	0.0140	0.0300	0.0190	0.3090
	200	0.0000	0.0150	0.0320	0.0179	0.3086
BoF	10	0.0067	0.0493	0.0907	0.0569	0.3436
	50	0.0200	0.0243	0.0467	0.0277	0.3243
	100	0.0000	0.0217	0.0397	0.0249	0.3177
	150	0.0000	0.0140	0.0287	0.0172	0.3073
	200	0.0133	0.0273	0.0520	0.0338	0.3239
	300	0.0400	0.0247	0.0447	0.0292	0.3247
	500	0.0267	0.0247	0.0440	0.0277	0.3262
	1000	0.0400	0.0290	0.0497	0.0292	0.0292

Performance metrics obtained using Cosine Distance for matching

Results

Encoding	K	NN	FT	ST	E	DCG
FV	10	0.0067	0.0157	0.0350	0.0218	0.3096
	20	0.0000	0.0173	0.0427	0.0254	0.3126
	25	0.0000	0.0167	0.0437	0.0269	0.3119
	50	0.0000	0.0167	0.0417	0.0228	0.3117
	100	0.0067	0.0170	0.0403	0.0236	0.3139
	125	0.0000	0.0160	0.0413	0.0231	0.3134
	150	0.0067	0.0187	0.0407	0.0251	0.3163
	200	0.0000	0.0160	0.0423	0.0241	0.3142
BoF	10	0.0067	0.0167	0.0417	0.0246	0.3136
	50	0.0000	0.0197	0.0397	0.0244	0.3187
	100	0.0667	0.0317	0.0633	0.0418	0.3354
	150	0.0600	0.0323	0.0630	0.0390	0.3338
	200	0.0200	0.0153	0.0370	0.0226	0.3134
	300	0.0067	0.0180	0.0347	0.0218	0.3102
	500	0.0067	0.0157	0.0370	0.0215	0.3160
	1000	0.0000	0.0187	0.0420	0.02625	0.3144

Performance metrics obtained using EMD for matching

Results

Comparison with state-of-the art technique

Feature	NN	FT	ST
RSI	0.0892	0.0734	0.0713
PSI	0.0933	0.0812	0.0770
SC	0.0861	0.0825	0.0771
FPFH	0.1167	0.0799	0.074

Performance metrics of state-of-the art technique on SHREC'09

Encoding	Distance	K	NN	FT	ST
FV	l_1 -norm	200	0.0133	0.016	0.0317
	l_2 -norm	10	0.0067	0.0177	0.0337
	Cosine	50	0.0133	0.0180	0.0323
	EMD	150	0.0067	0.0187	0.0407
BoF	l_1 -norm	1000	0.0600	0.0527	0.0527
	l_2 -norm	1000	0.0533	0.0267	0.0527
	Cosine	1000	0.0400	0.0290	0.0497
	EMD	100	0.0667	0.0317	0.0633

Performance metrics of proposed partial 3D object retrieval technique on SHREC'11

Results

- We observed that the proposed technique using BoFs encoding outperformed FV encoding techniques.
- L1-norm distance out performed other distance measures.

Conclusions And Future Scope

We proposed a partial 3D object retrieval technique comprising five modules, namely, sampling, feature extraction, feature encoding, matching and performance evaluation.

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- We demonstrated the performance of the proposed technique on SHREC'11 dataset.
-
- We observed that the proposed technique using BoFs encoding outperformed other encoding techniques. Also, for matching L1-norm distance performed better.
-
- As future work the proposed technique can be extended by using other feature extraction methods and matching techniques.
- Also weighting of codewords can be done in BoF.

THANK-YOU

REFERENCES:

- [1] Michalis A. Savelonas, Ioannis Pratikakis, Konstantinos Sfikas. Fisher encoding of Differential Fast Point Feature histograms for Partial 3D Object Retrieval, 2016.
- [2] I.Pratikakis, M.Spagnuolo, T. Theoharis, and R.Veltkamp. A Query-by-Example Concept and User Interface for Global and Partial 3D Object Retrieval, 2009.
- [3] Mathieu Aubry, Ulrich Schlickewei, Daniel Cremers. The Wave Kernel Signature: A Quantum Mechanical Approach to Shape Analysis.

Implementation

PFF
46622 90161 0
0.345632 -0.989338 -0.226566
-0.063311 0.753159 -0.400532
1.0083 -0.183685 -0.54204
-0.725983 -0.052494 -0.085058
-0.386981 0.021914 0.190186
-0.401598 0.049609 0.176745
-0.411009 0.038168 0.181225
-0.377571 0.033354 0.185706
-0.411137 0.040663 0.183257
-0.387342 0.024409 0.192218
-0.410374 0.044023 0.184144
-0.386346 0.027768 0.193105
-0.408288 0.047346 0.183649
-0.38426 0.031092 0.19261
-0.405669 0.049743 0.181904
-0.381641 0.033489 0.190865
-0.403221 0.050571 0.179377
-0.379193 0.034317 0.188338
-0.367505 0.00965 0.199779
-0.382122 0.037345 0.186338
-0.391533 0.025905 0.190818
-0.358095 0.021091 0.195299
-0.391894 0.0284 0.19285
-0.367866 0.012146 0.201811
-0.390898 0.03176 0.193737
-0.36687 0.015505 0.202698
-0.388811 0.035083 0.193242
-0.364784 0.018829 0.202203
-0.386193 0.03748 0.191497
-0.362165 0.021225 0.200458
-0.383745 0.038308 0.18897
-0.359717 0.022054 0.197931
-0.348029 -0.002613 0.209372
-0.362646 0.025082 0.195931
-0.372056 0.013641 0.200411
-0.338618 0.008878 0.204892

Feature
Extraction

1	0.050280	0.051050	0.051972	0.053143	0.054726	0.056962	0.060149	0.064598	0.070571	0.078203	0.087451	0.098063	0.109598	0.121453	0.132929	0.143311	0.151967	0.158453	0.162594	0.164515	0.164631	0.163583
	0.162155	0.161166	0.161363	0.163312	0.167302	0.173294	0.180938	0.189663	0.198832	0.207903	0.216550	0.224720	0.232612	0.240599	0.249110	0.258491	0.268878	0.280108	0.291690	0.302850	0.312642	0.320110
	0.324443	0.325113	0.321956	0.315195	0.305407	0.293443	0.280317	0.267066	0.254630	0.243741	0.234870	0.228215	0.223730	0.221194	0.220282	0.220641	0.221963	0.224026	0.226730	0.230098	0.234247	0.239350
	0.245562	0.252953	0.261442	0.270754	0.280422	0.289827	0.298280	0.305132	0.309868	0.312186	0.312024	0.309545	0.305080	0.299066	0.291974	0.284248	0.276260	0.268288	0.260501	0.252976	0.245722	0.238733
	0.232030	0.225708	0.219944	0.214995	0.211151	0.208685	0.207803	0.208604	0.211071	0.215077	0.220413	0.102731										
2	0.042697	0.044213	0.045481	0.046398	0.046928	0.047143	0.047268	0.047709	0.049019	0.051825	0.056716	0.064150	0.074384	0.087446	0.103124	0.120979	0.140382	0.160577	0.180768	0.200208	0.218264	0.234458
	0.248453	0.260025	0.269015	0.275288	0.278722	0.279228	0.276806	0.271602	0.263963	0.254442	0.243771	0.232792	0.222370	0.213293	0.206182	0.201408	0.199046	0.198861	0.200361	0.202896	0.205802	0.208540
	0.210817	0.212627	0.214217	0.215979	0.218281	0.221315	0.224978	0.228847	0.232262	0.234468	0.234801	0.232844	0.228534	0.222201	0.214558	0.206643	0.199717	0.195140	0.194214	0.198002	0.207145	0.221708
	0.241091	0.264037	0.288754	0.313146	0.335108	0.352831	0.365053	0.371206	0.371424	0.366443	0.357410	0.345682	0.332634	0.319534	0.307462	0.297275	0.289600	0.284820	0.283083	0.284294	0.288145	0.294149
	0.301702	0.310163	0.318928	0.327500	0.335525	0.342807	0.349292	0.355036	0.360164	0.364835	0.369209	0.100928										
3	0.022735	0.024567	0.027117	0.030519	0.034885	0.040241	0.046473	0.053298	0.060292	0.066970	0.072905	0.077811	0.081584	0.084289	0.086113	0.087315	0.088171	0.088941	0.089852	0.091088	0.092804	0.095145
	0.098252	0.102269	0.107327	0.113500	0.120755	0.128901	0.137563	0.146210	0.154234	0.161067	0.166300	0.169773	0.171618	0.172243	0.172260	0.172384	0.173300	0.175545	0.179413	0.184912	0.191785	0.199576
	0.207735	0.215731	0.223143	0.229725	0.235421	0.240343	0.244716	0.248808	0.252868	0.257099	0.261647	0.266625	0.272122	0.278197	0.284839	0.291908	0.299071	0.305780	0.311293	0.314768	0.315414	0.312657
	0.306290	0.296556	0.284134	0.270045	0.255476	0.241577	0.229276	0.219139	0.211317	0.205575	0.201393	0.198111	0.195083	0.191804	0.187995	0.183622	0.178869	0.174068	0.169609	0.165846	0.163022	0.161224
	0.160371	0.160241	0.160524	0.160882	0.161015	0.160698	0.159805	0.158306	0.156249	0.153733	0.150879	0.104129										
4	0.104774	0.100940	0.096745	0.092171	0.087250	0.082113	0.077012	0.072329	0.068549	0.066199	0.065778	0.067698	0.072232	0.079466	0.089252	0.101160	0.114462	0.128156	0.141053	0.151909	0.159569	0.163117
	0.161988	0.156065	0.145724	0.131828	0.115631	0.098595	0.082140	0.067390	0.055019	0.045218	0.037801	0.032362	0.028446	0.025670	0.023796	0.022758	0.022669	0.023802	0.026580	0.031558	0.039398	0.050831
	0.066582	0.087255	0.113168	0.144151	0.179360	0.217162	0.255171	0.290474	0.320019	0.341093	0.351752	0.351113	0.339440	0.318049	0.289049	0.255009	0.218606	0.182311	0.148170	0.117684	0.091801	0.070987
	0.055357	0.044801	0.039092	0.037936	0.040964	0.047673	0.057353	0.069033	0.081486	0.093321	0.103139	0.109741	0.112310	0.110544	0.104680	0.095428	0.083815	0.071003	0.058102	0.046026	0.035414	0.026606
	0.019678	0.014507	0.010844	0.008393	0.006855	0.005972	0.005534	0.005385	0.005415	0.005549	0.005739	0.099635										
5	0.030284	0.028884	0.027447	0.026002	0.024613	0.023370	0.022385	0.021783	0.021690	0.022222	0.023477	0.025534	0.028445	0.032219	0.036805	0.042075	0.047817	0.053741	0.059511	0.064777	0.069228	0.072623
	0.074837	0.075895	0.076003	0.075579	0.075240	0.075758	0.077947	0.082526	0.089976	0.100441	0.113692	0.129159	0.146001	0.163206	0.179698	0.194449	0.206590	0.215498	0.220865	0.222719	0.221411	0.217558
	0.211963	0.205531	0.199173	0.193725	0.189864	0.188030	0.188376	0.190749	0.194722	0.199687	0.204970	0.209950	0.214145	0.217260	0.219180	0.219928	0.219612	0.218358	0.216271	0.213411	0.209789	0.205388
	0.200183	0.194178	0.187423	0.180034	0.172210	0.164227	0.156431	0.149211	0.142957	0.138010	0.134612	0.132862	0.132698	0.133890	0.136068	0.138758	0.141439	0.143603	0.144822	0.144800	0.143417	0.140747

IWKS features

Input 3D Point

Implementation

```
RQ0001
D00325 13.729901000000002
D00263 13.808893
D00665 13.821432000000001
D00568 13.936519
D00915 14.029943
D00478 14.08895
D00156 14.139003
D00468 14.160554999999999
D00873 14.166057
D00051 14.167767999999999
D00921 14.185222
D00953 14.201017000000002
D00198 14.203201000000002
D00173 14.222005
D00504 14.226023999999999
D00676 14.245052
D00577 14.260414
D00889 14.263792
D00310 14.278835
D00631 14.303379000000001
D00607 14.307772
D00781 14.312448000000002
D00946 14.314570999999999
D00860 14.318677
D00654 14.333682000000001
D00044 14.346034
D00836 14.359932
D00933 14.371594
D00045 14.372935
D00789 14.38384
D00359 14.393061999999999
D00458 14.399866000000001
D00053 14.399992999999998
D00881 14.406676999999998
D00747 14.420335000000001
D00951 14.429309
D00901 14.455884
D00323 14.459999
D00978 14.46285
D00837 14.47649
D00158 14.481909
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16.664856	15.944921	16.114256	14.346034	14.372935	15.086486	16.499138	24.813869	19.012891
16.809121	15.820039	24.461756	15.316389	15.797004	16.392833	16.034865	17.359610	17.277325
22.359112	17.747075	18.056772	19.572358	15.210226	14.640430	18.920930	17.665648	14.766238
19.969264	15.207854	15.125418	17.475769	21.244660	16.013304	19.765733	16.358264	14.853928
18.298576	16.482294	22.562756	20.953228	20.454971	15.046720	18.544733	19.286704	14.578660
16.717046	15.974663	18.495464	18.620275	17.208055	15.810090	24.928934	16.395471	17.016678
15.741292	20.848572	15.146948	14.687022	16.887125	18.724995	15.381240	15.061201	16.308775
16.776185	19.019627	18.216160	15.238976	18.612929	14.824546	14.536049	15.154107	19.327745
15.190304	14.608425	16.594414	14.798928	18.297127	16.890854	14.946721	21.212264	15.184657
14.690491	16.657235	14.916384	17.587172	16.305881	18.500377	19.341536	17.406914	16.234812
16.515398	15.709943	15.967297	15.845501	16.085257	15.536251	14.525725	17.691798	20.820275
19.251932	14.614850	13.808893	16.482635	16.989161	16.107411	15.859863	27.213714	15.675002
20.806755	16.452433	14.907622	15.254660	15.826778	14.483572	16.692820	14.872060	17.496717
16.764768	15.582310	17.201866	16.253425	15.578367	16.552320	15.090085	18.178024	16.185699
21.219833	17.216389	14.459999	16.252771	13.729901	19.106687	18.257746	17.241214	15.285468
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15.857382	16.145354	15.083883	14.721060	18.567730	19.493520	17.594716	17.117788	17.351547
16.507596	15.040100	17.348597	18.756786	16.033034	16.330388	18.068995	17.844276	15.334921
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26.175523	17.991816	29.180205	14.226024	14.973580	16.363649	18.168313	19.426279	19.565403
14.988161	18.544091	15.490701	15.985469	16.365934	16.099103	18.872161	22.159959	15.621667
14.718398	16.987271	15.385979	14.690066	16.226250	14.779066	21.589350	14.965812	16.461616
16.661609	17.732022	16.909288	18.521760	17.533926	15.437341	18.726334	13.936519	16.307409
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16.521372	18.035684	19.569748	15.667983	13.821432	15.243081	20.219066	20.681037	15.389233

Distance Matrix

Generating
Ranked List

Ranked List

Feature
Encoding and
Matching