

Lecture 14: Shape Google: Rigid Shape Statistics

COMPSCI/MATH 290-04

Chris Tralie, Duke University

3/1/2016

Announcements

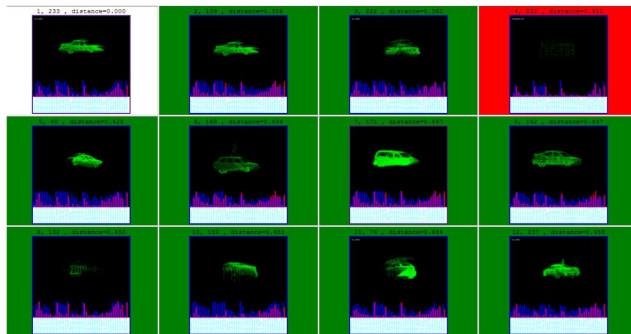
- ▷ Group Assignment 1 Full Submission Due Tomorrow (Wednesday) 11:55 PM
- ▷ Rank Top 3 Final Project Choices By Friday 3/5 (Groups of 3-4)
- ▷ Attendance Policy Clarification
- ▷ Midterm Next Thursday 3/10

Table of Contents

- ▶ Shape Statistics / Algorithms
- ▷ Comparing Shape Statistics
- ▷ Classification / Performance Evaluation

Random Sampling By Area

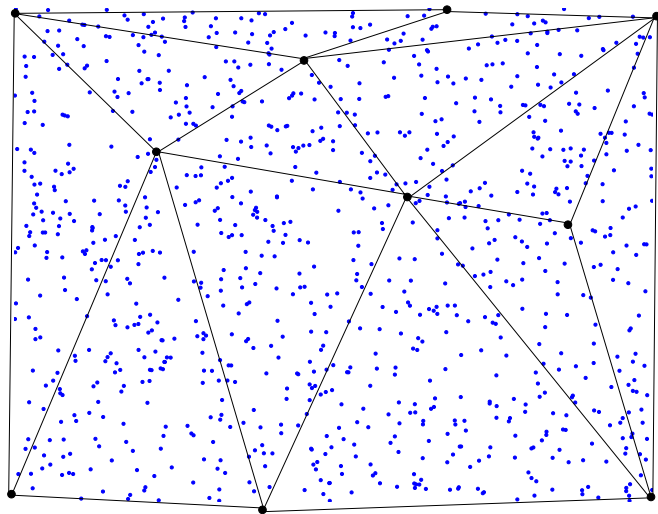
Goal: Given a shape, return similar shapes in a database



Tralie 2010

- ▷ Focus on point clouds
- ▷ Focus on shapes similar under *rigid motion*

Random Sampling By Area



Centroid Centering / RMS Scaling

For a point cloud $\{\vec{x}_i\}_{i=1}^N$

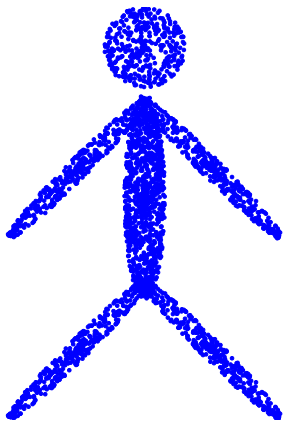
- ▷ Subtract off centroid
- ▷ Root mean square scale. Want

$$\sqrt{\frac{1}{N} \sum_{i=1}^N \|\vec{x}_i\|^2} = 1$$

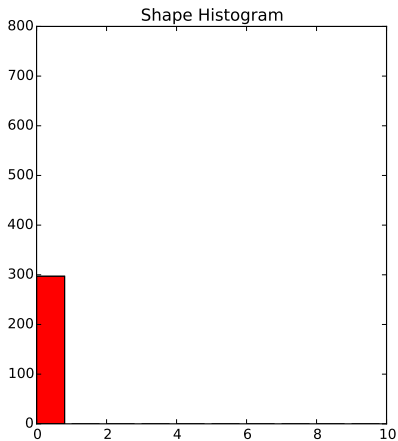
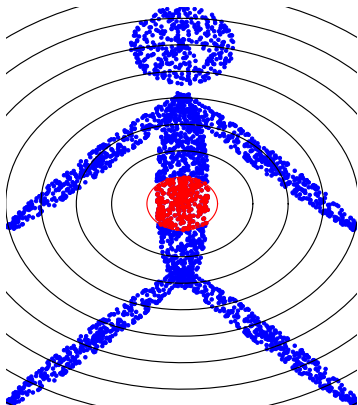
Shape Matching Criteria

- ▷ Concise To Store
- ▷ Quick to compute
- ▷ Efficient to match
- ▷ Discerning
- ▷ Noise tolerant
- ▷ *Rotation Invariant*

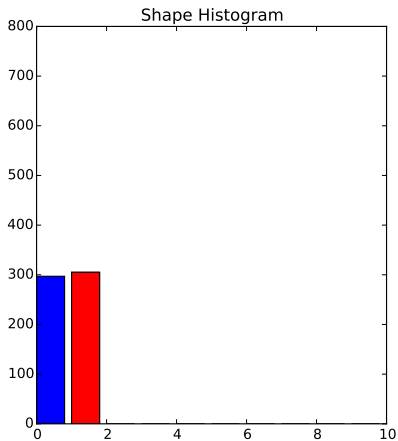
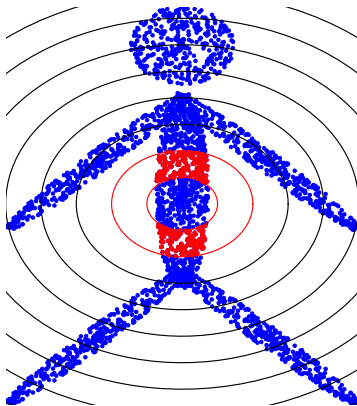
Shape Histogram: Shells



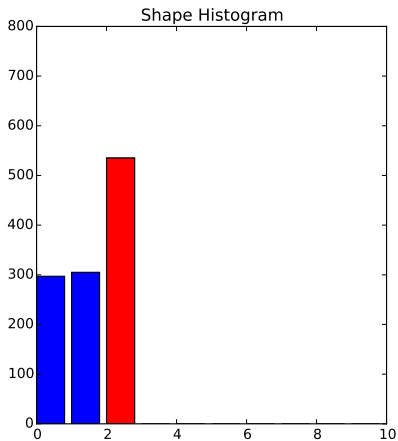
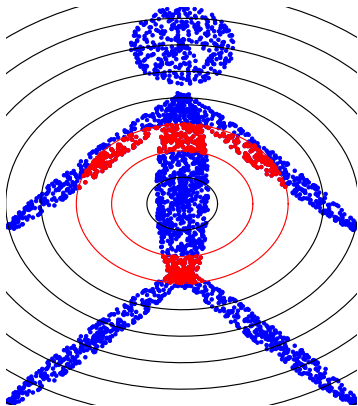
Shape Histogram: Shells



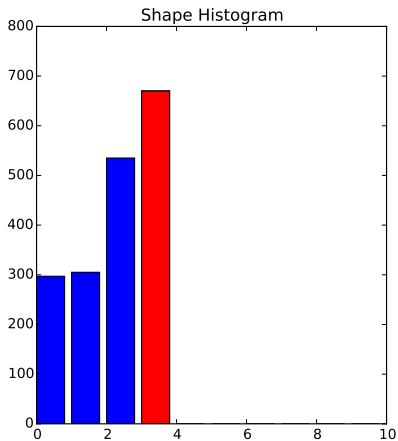
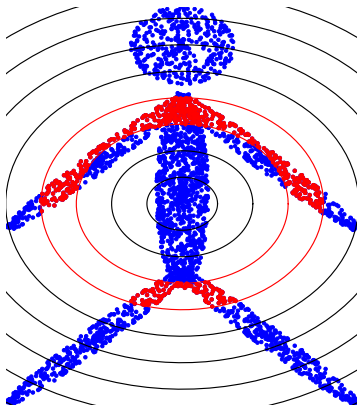
Shape Histogram: Shells



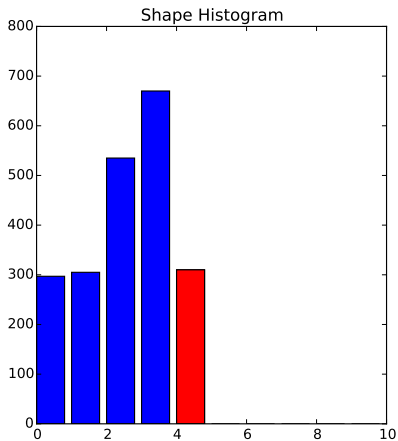
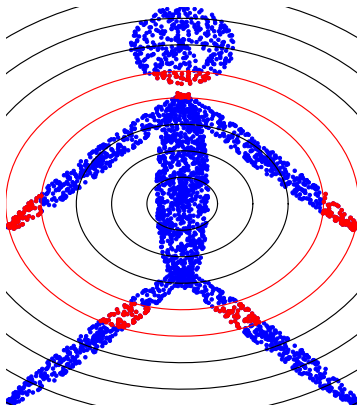
Shape Histogram: Shells



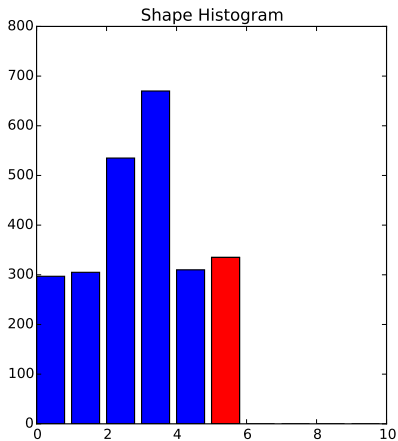
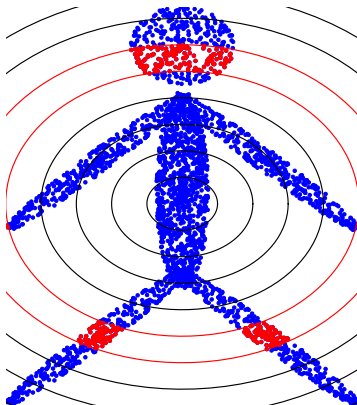
Shape Histogram: Shells



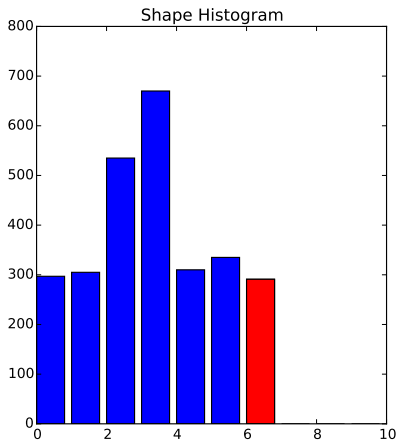
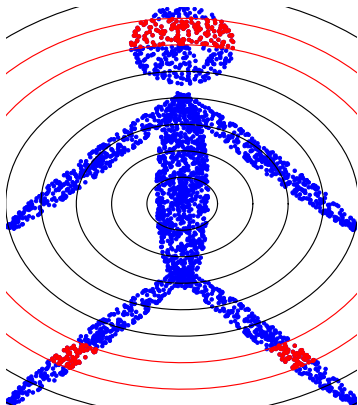
Shape Histogram: Shells



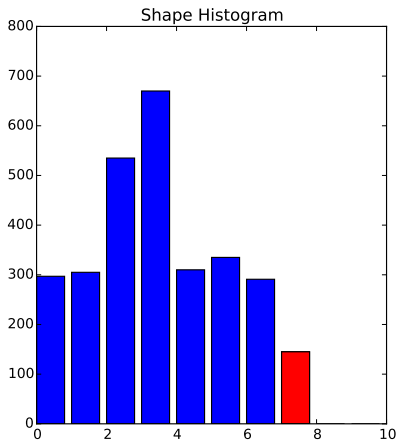
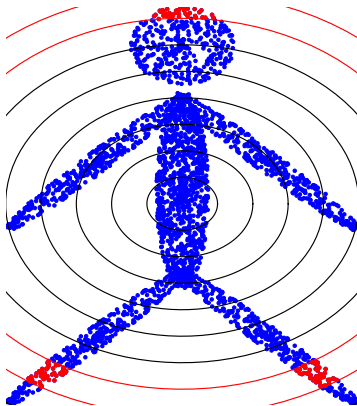
Shape Histogram: Shells



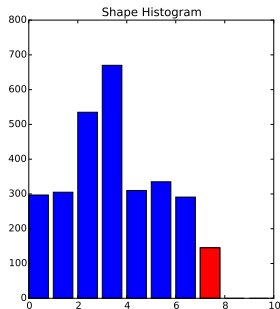
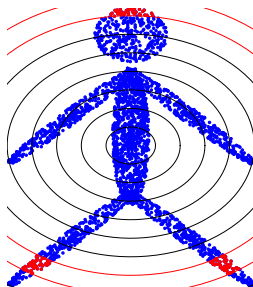
Shape Histogram: Shells



Shape Histogram: Shells

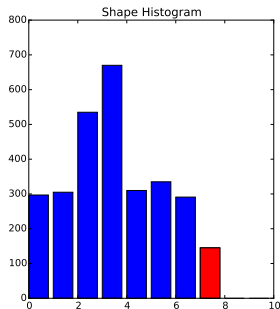
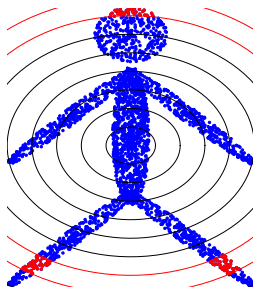


Shape Histogram: Shells



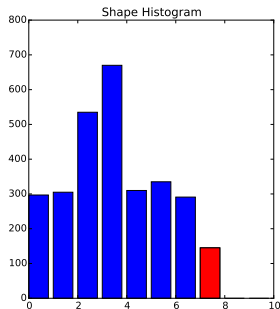
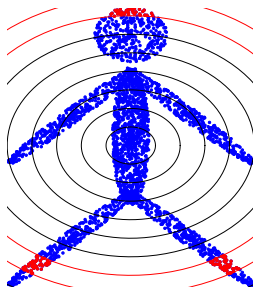
▷ Quick To Compute

Shape Histogram: Shells



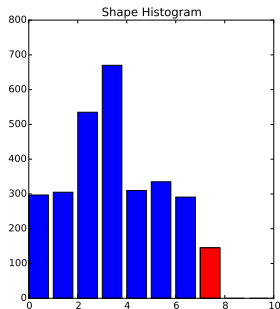
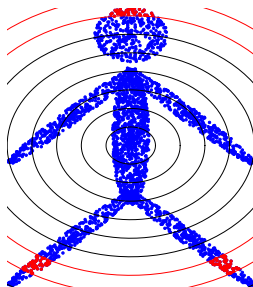
▷ Quick To Compute

Shape Histogram: Shells



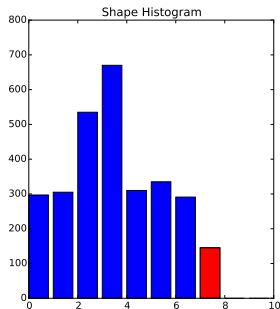
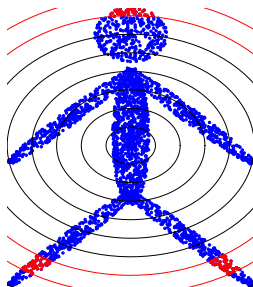
- ▷ Quick To Compute
- ▷ Concise To Store

Shape Histogram: Shells



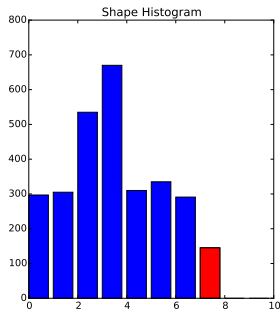
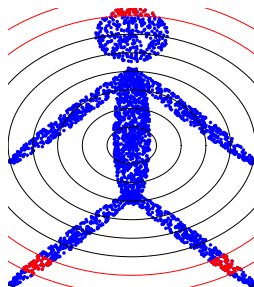
- ▷ Quick To Compute
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Shape Histogram: Shells



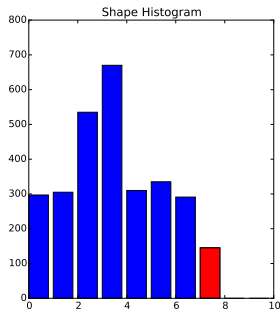
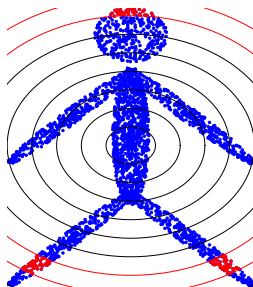
- ▷ Quick To Compute
- ▷ Concise To Store
- ▷ Rotation Invariant

Shape Histogram: Shells



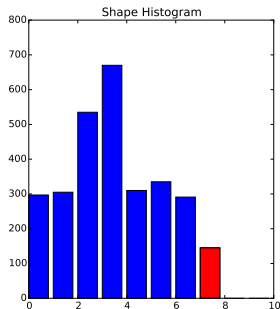
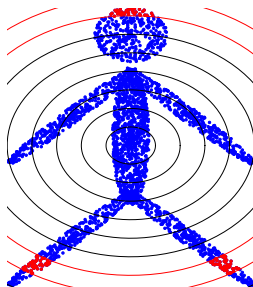
- ▷ Quick To Compute
- ▷ Concise To Store
- ▷ Rotation Invariant

Shape Histogram: Shells



- ▷ Quick To Compute
- ▷ Concise To Store
- ▷ Rotation Invariant
- ▷ Discerning

Shape Histogram: Shells

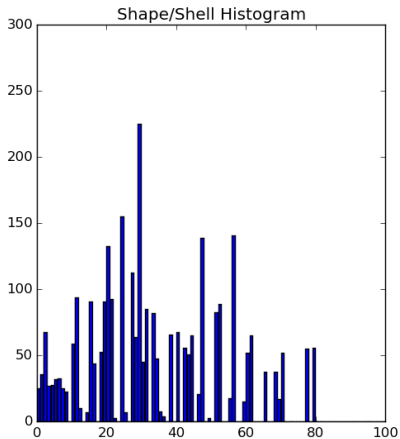
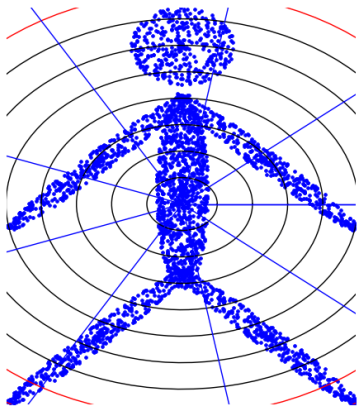


- ▷ Quick To Compute
- ▷ Concise To Store
- ▷ Rotation Invariant
- ▷ Discerning

Shape Histogram: Shells

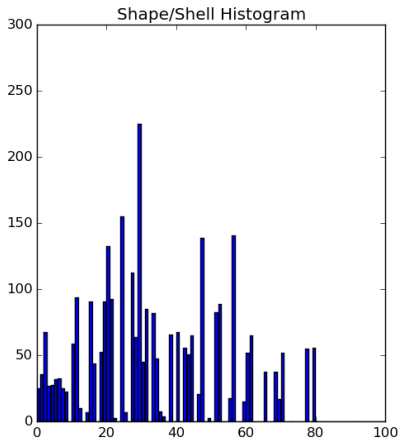
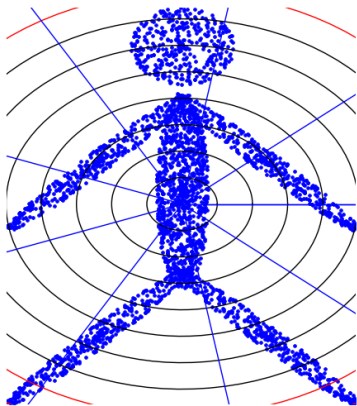
What can't it tell apart?
SHOW VIDEO

Shape Histogram: Shells And Sectors



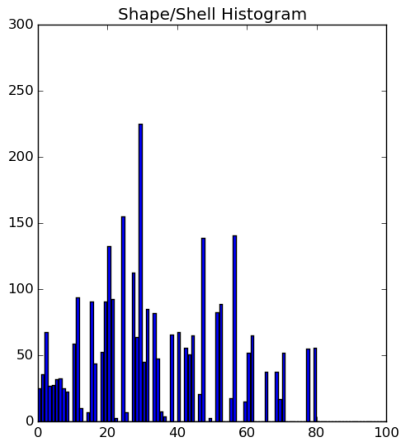
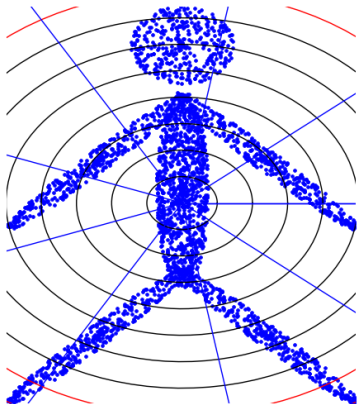
SHOW VIDEO

Shape Histogram: Shells And Sectors



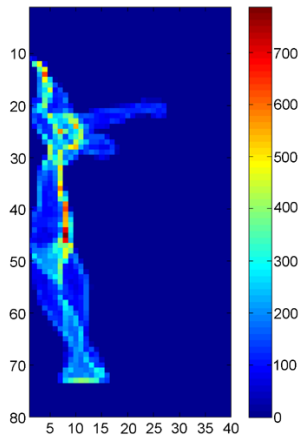
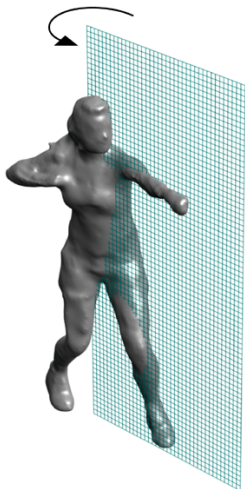
Still Rotation Invariant?

Shape Histogram: Shells And Sectors



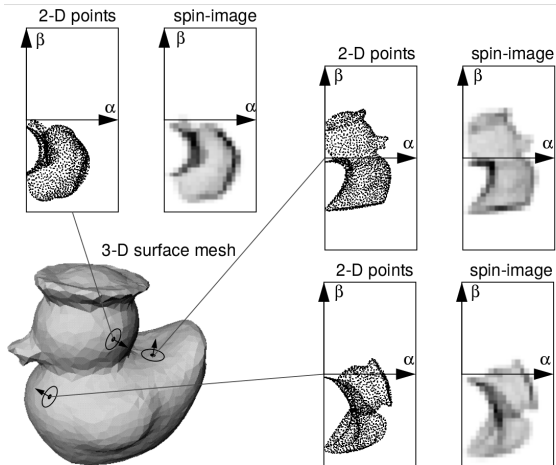
- ▷ Sort sectors within each shell
- ▷ Record PCA *eigenvalues* within each shell

Spin Images



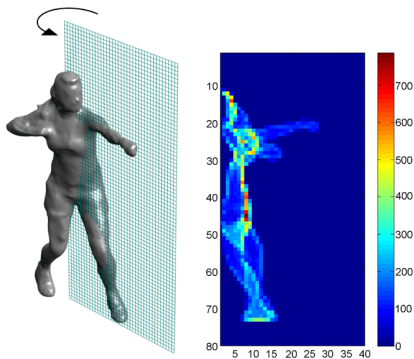
Johnson/Herbert 1999, Figure Huang 2010

Spin Images: Rubber Duck



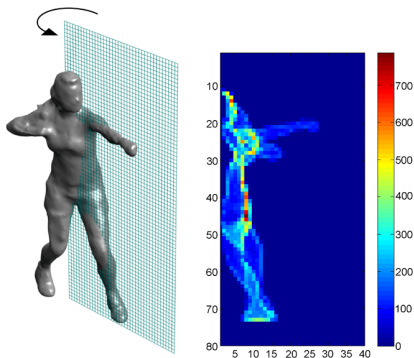
Johnson/Herbert 1999

Spin Images



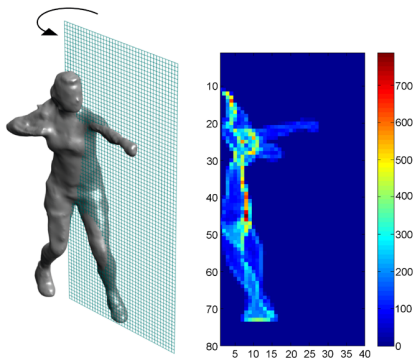
▷ Quick To Compute

Spin Images



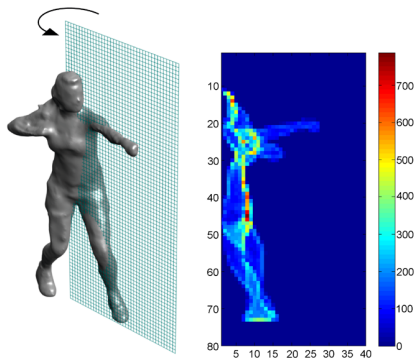
▷ Quick To Compute

Spin Images



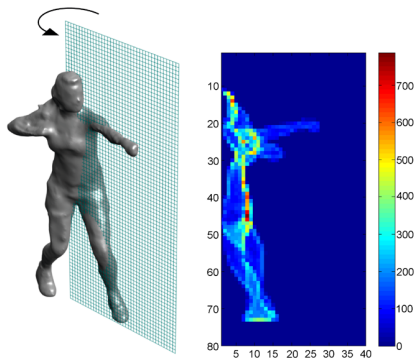
- ▷ Quick To Compute
- ▷ Concise To Store

Spin Images



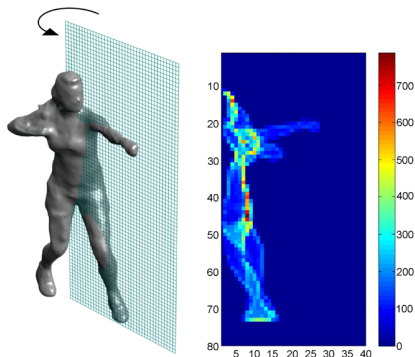
- ▷ Quick To Compute
- ▷ Concise To Store (Can compress images)

Spin Images



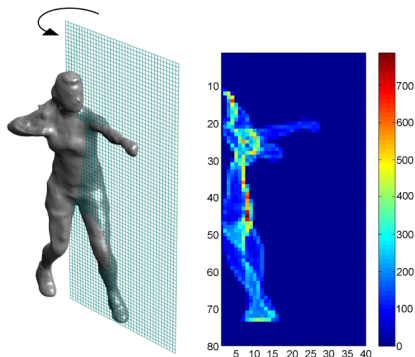
- ▷ Quick To Compute
- ▷ Concise To Store (Can compress images)
- ▷ Rotation Invariant

Spin Images



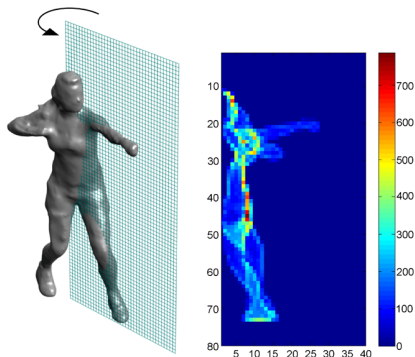
- ▷ Quick To Compute
- ▷ Concise To Store (Can compress images)
- ▷ Rotation Invariant (Careful with principal axis stability)

Spin Images



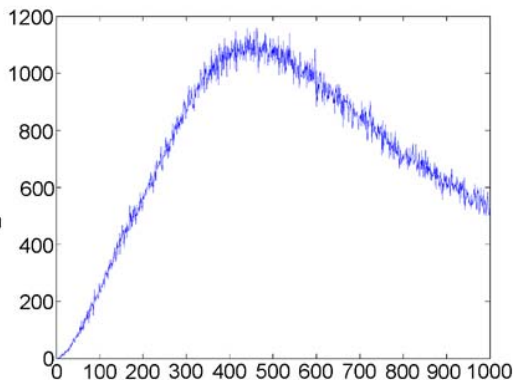
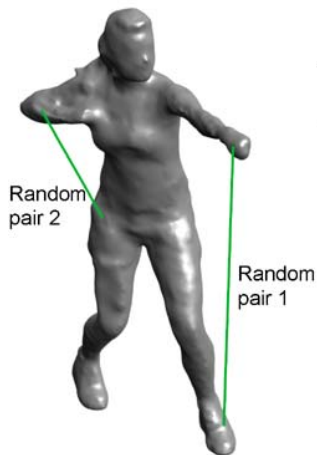
- ▷ Quick To Compute
- ▷ Concise To Store (Can compress images)
- ▷ Rotation Invariant (Careful with principal axis stability)
- ▷ Discerning

Spin Images



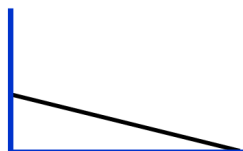
- ▷ Quick To Compute
- ▷ Concise To Store (Can compress images)
- ▷ Rotation Invariant (Careful with principal axis stability)
- ▷ Discerning

D2: Distance Histograms

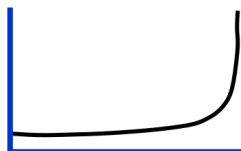


Osada 2003, Figure from Huang 2010

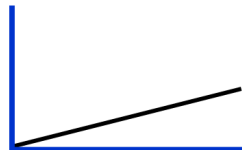
D2: Primitive Examples



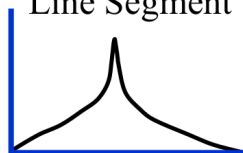
Line Segment



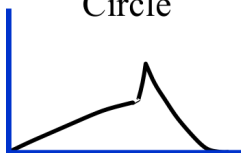
Circle



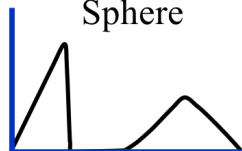
Sphere



Cylinder



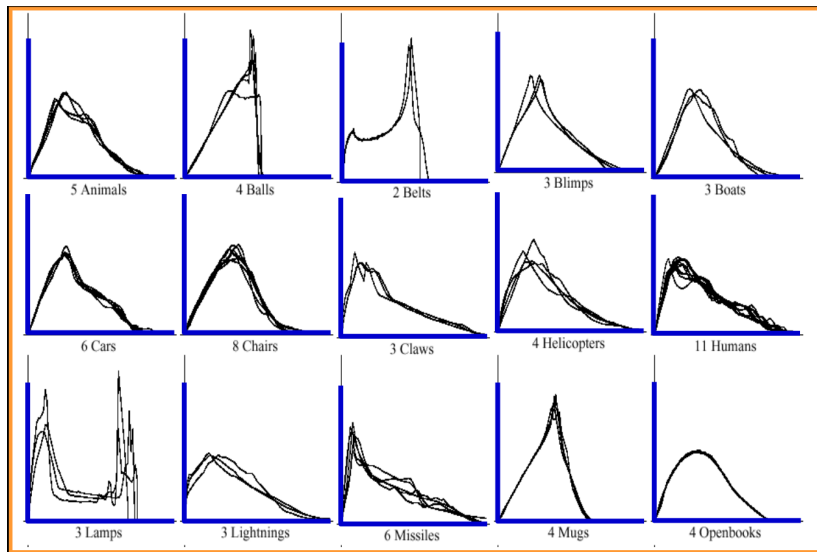
Cube



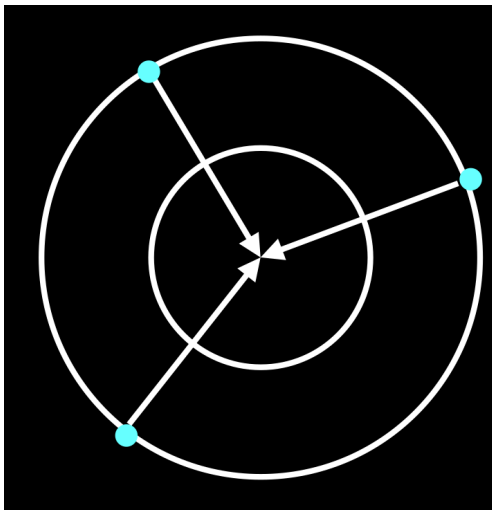
Two Spheres

Osada 2003

D2: Real Examples

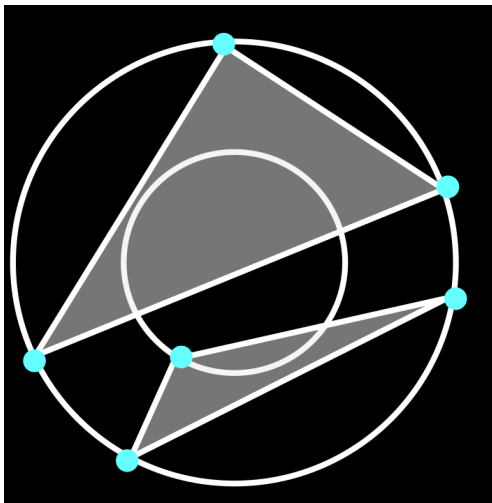


D1: Randomly Sample Points



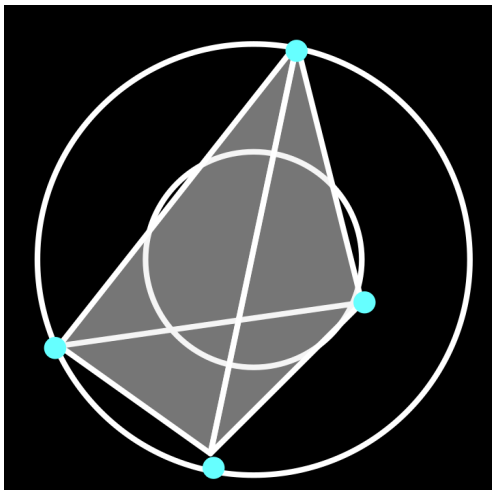
Osada 2003

D3: Randomly Sample Areas



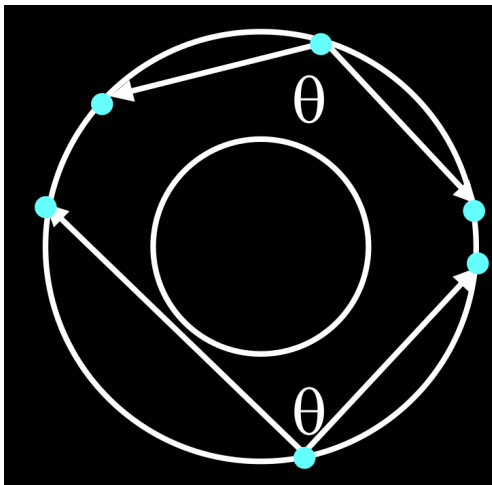
Osada 2003

D4: Randomly Sample Volumes



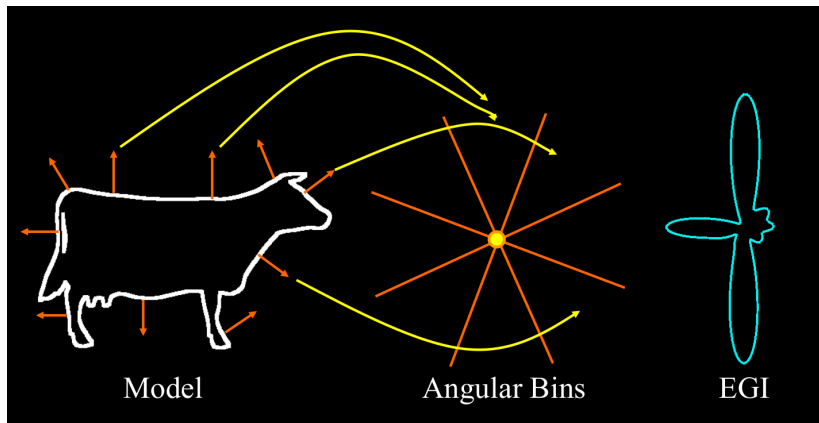
Osada 2003

A3: Randomly Sample Angles



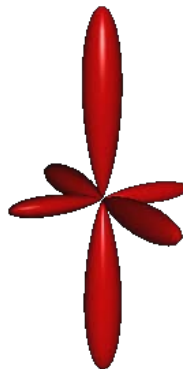
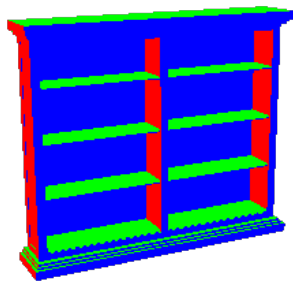
Osada 2003

Extended Gaussian Image



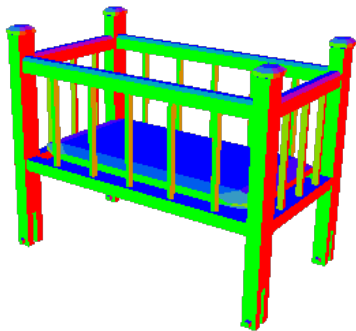
Funkhouser 2004

Extended Gaussian Image

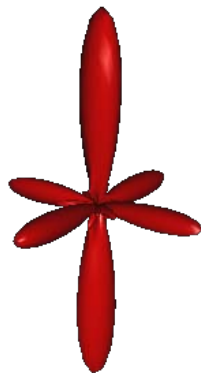


Funkhouser 2004

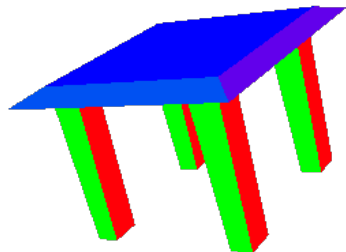
Extended Gaussian Image



Funkhouser 2004

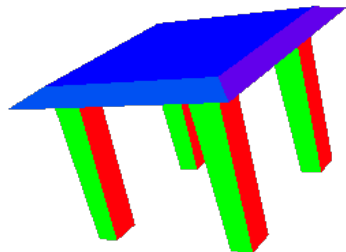


Extended Gaussian Image



Funkhouser 2004

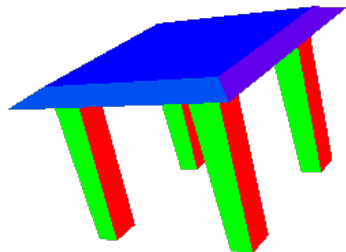
Extended Gaussian Image



▷ Efficient To Compute / Concise To Store

Funkhouser 2004

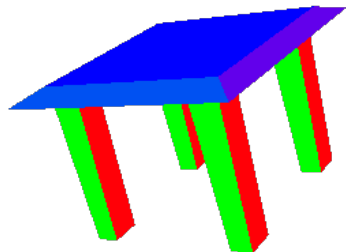
Extended Gaussian Image



▷ Efficient To Compute / Concise To Store

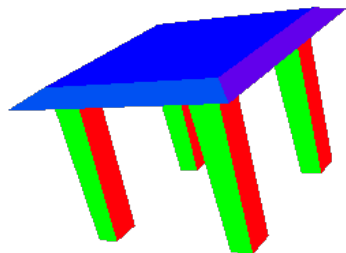
Funkhouser 2004

Extended Gaussian Image



▷ Discerning
Funkhouser 2004

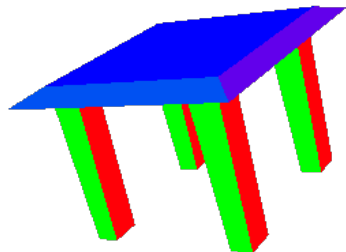
Extended Gaussian Image



▷ **Discerning** (Only fully describes convex objects)

Funkhouser 2004

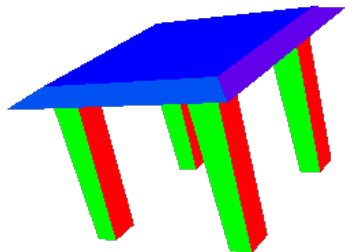
Extended Gaussian Image



▷ Rotation Invariant

Funkhouser 2004

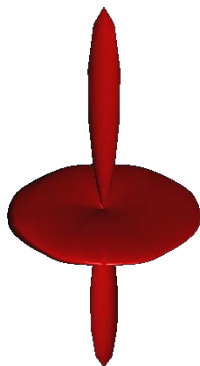
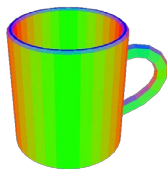
Extended Gaussian Image



- ▷ **Rotation Invariant** (Rotate To Align With PCA Axes)

Funkhouser 2004

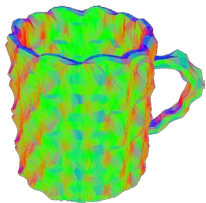
Extended Gaussian Image



▷ Robust To Noise?

Funkhouser 2004

Extended Gaussian Image



▷ **Not Robust To Noise!**

Funkhouser 2004

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- ▷ Classification / Performance Evaluation

Normalize Histograms By Mass

$$h'[i] = \frac{h[i]}{\sum_{k=1}^N h[k]}$$

In other words, all bins should sum to 1

Histogram Euclidean Distance

For histograms h_1 and h_2

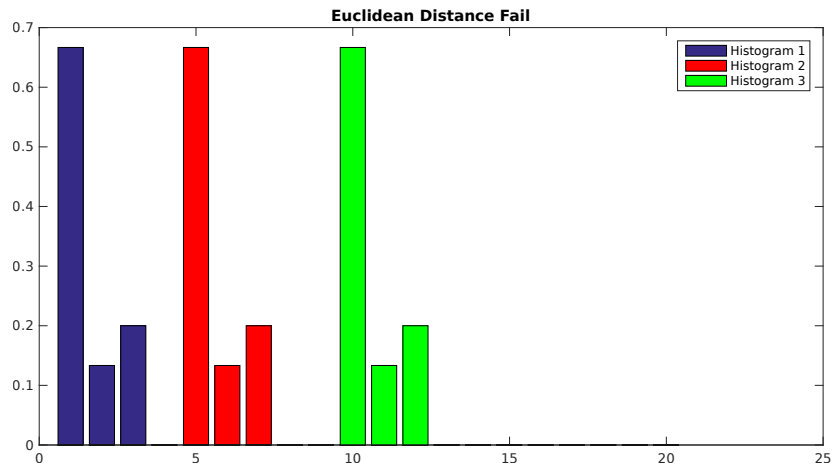
$$d_E(h_1, h_2) = \sqrt{\sum_{i=1}^N (h_1[i] - h_2[i])^2}$$

Just thinking of h_1 and h_2 as high dimensional Euclidean vectors! Each histogram bin is a dimension

Histogram Cosine Distance

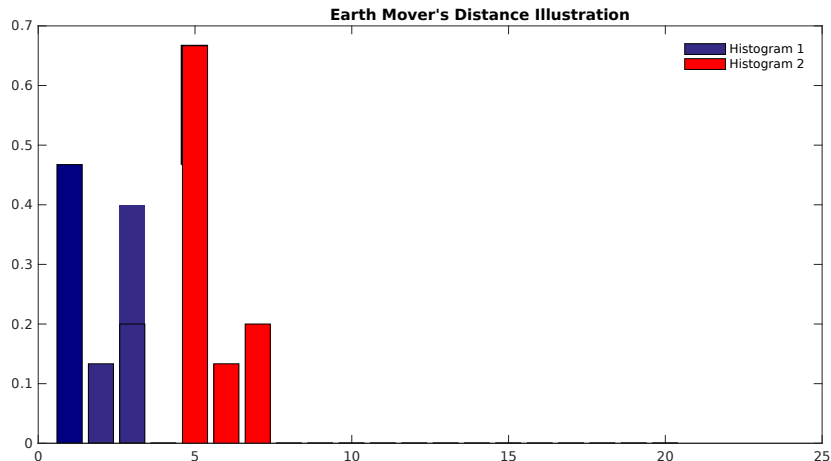
$$d_C(h_1, h_2) = \cos^{-1} \left(\frac{\vec{h}_1 \cdot \vec{h}_2}{\|\vec{h}_1\| \|\vec{h}_2\|} \right)$$

Euclidean Distance Shortcomings



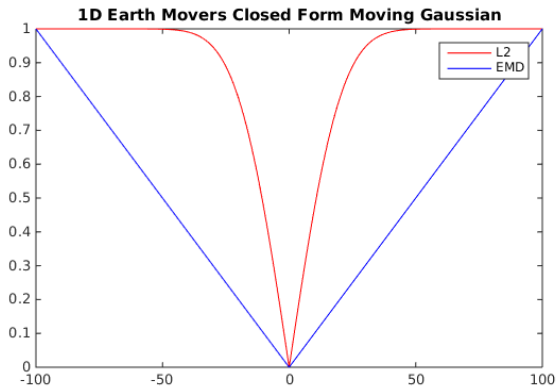
They all have the same distance!

Euclidean Distance Shortcomings



Move earth from blue to red

Earth Mover's Distance



Chi Squared Distance

$$d_{\chi}(h_1, h_2) = \frac{1}{2} \sum_{i=1}^N \frac{(h_1[i] - h_2[i])^2}{h_1[i] + h_2[i]}$$

Exclude values for which $h_1[i] = h_2[i] = 0$

Table of Contents

- ▷ Shape Statistics / Algorithms
- ▷ Comparing Shape Statistics
- ▶ Classification / Performance Evaluation

Do *leave one out* technique

Use each item as test item in turn, compare to database

- ▶ Summarize evaluation statistics over entire database by *averaging them*

Precision / Recall



Rusinkewicz/Funkhouser 2009

Other Evaluation Metrics

- ▷ Average Precision (Area Under Precision/Recall Curve)
- ▷ Mean Reciprocal Rank (1/rank of first correct item)
- ▷ Median Reciprocal Rank

1 is perfect score