

# Lecture 14: Shape Google: Rigid Shape Statistics

COMPSCI/MATH 290-04

Chris Tralie, Duke University

3/1/2016

# Announcements

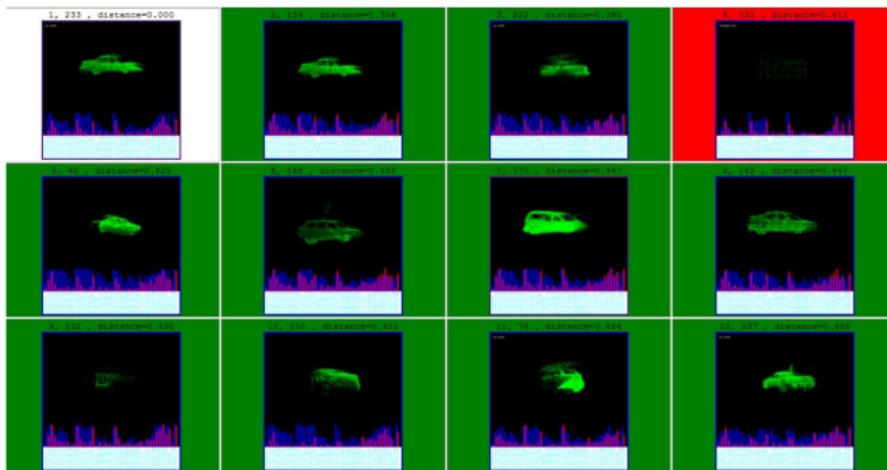
- ▷ Group Assignment 1 Full Submission Due Tomorrow (Wednesday) 11:55 PM
- ▷ Rank Top 3 Final Project Choices By Friday 3/5 (Groups of 3-4)
- ▷ Attendance Policy Clarification
- ▷ Midterm Next Thursday 3/10

# Table of Contents

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

# Random Sampling By Area

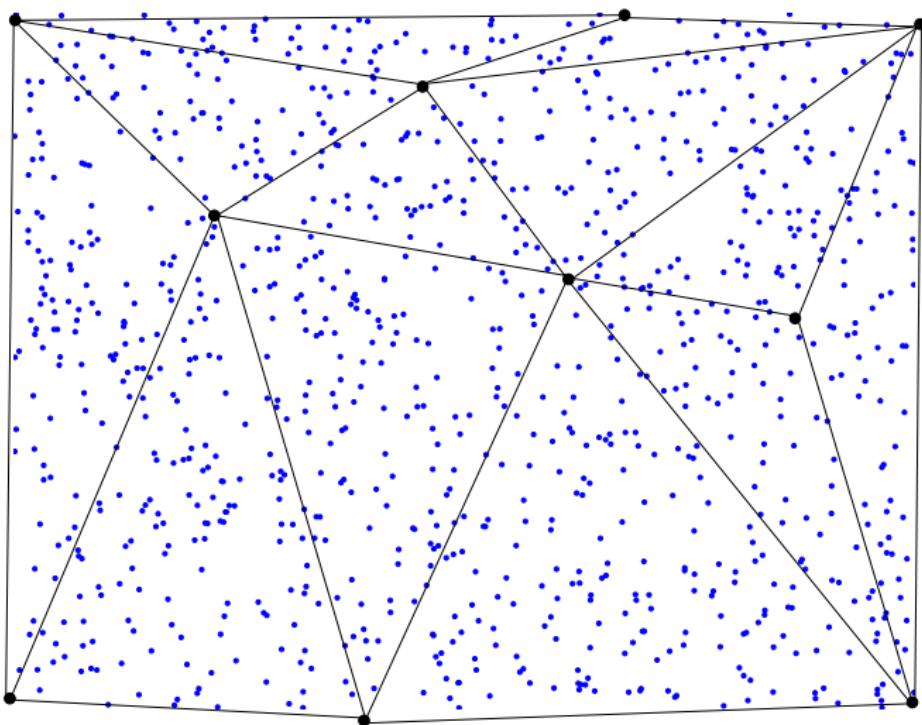
Goal: Given a shape, return similar shapes in a database



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  $\{\vec{x}_i\}_{i=1}^N$

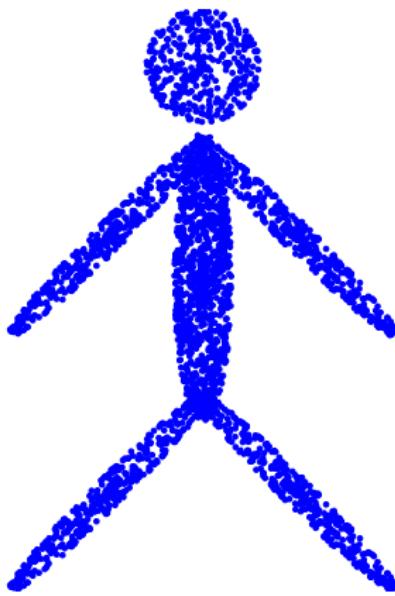
- ▷ Subtract off centroid
- ▷ Root mean square scale. Want

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

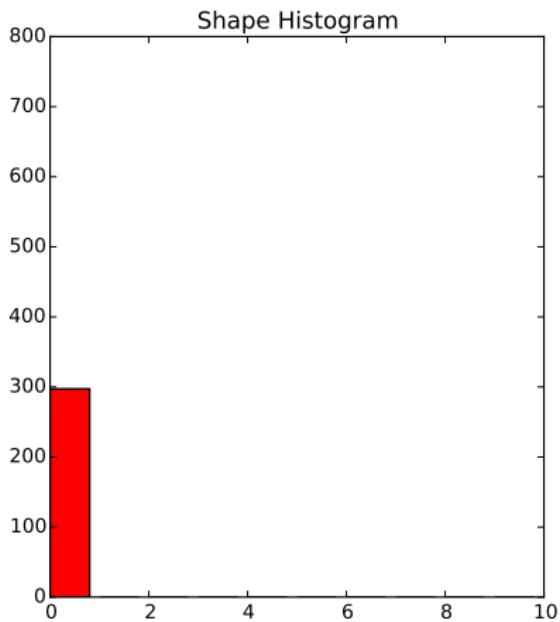
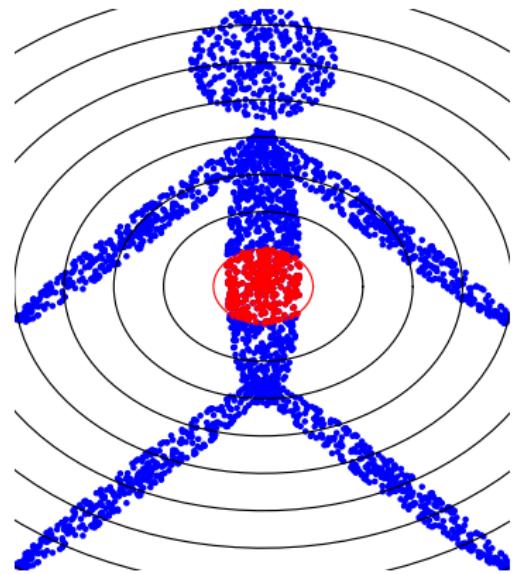
# Shape Matching Criteria

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

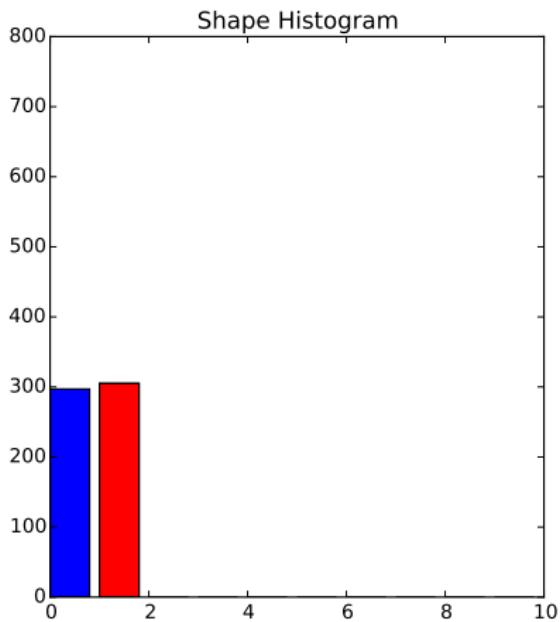
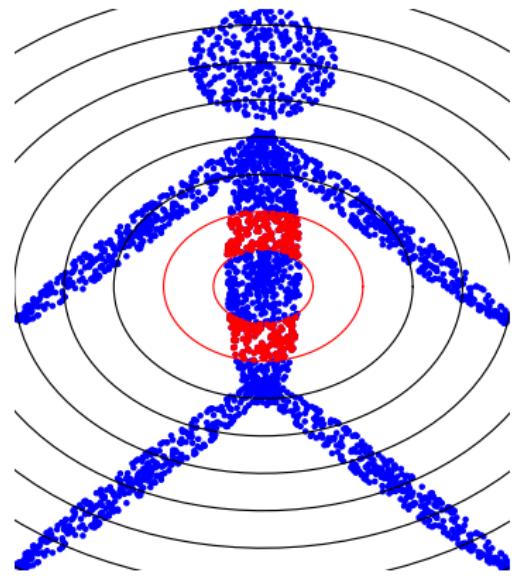
# Shape Histogram: Shells



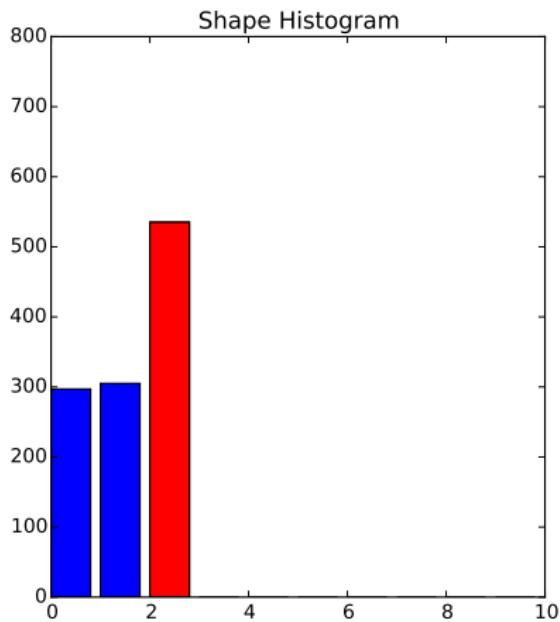
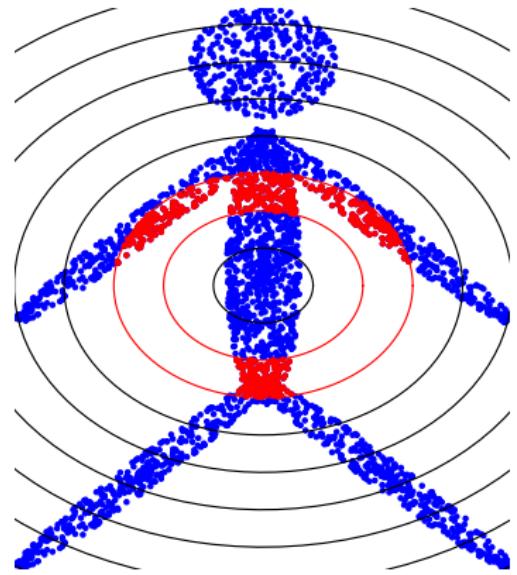
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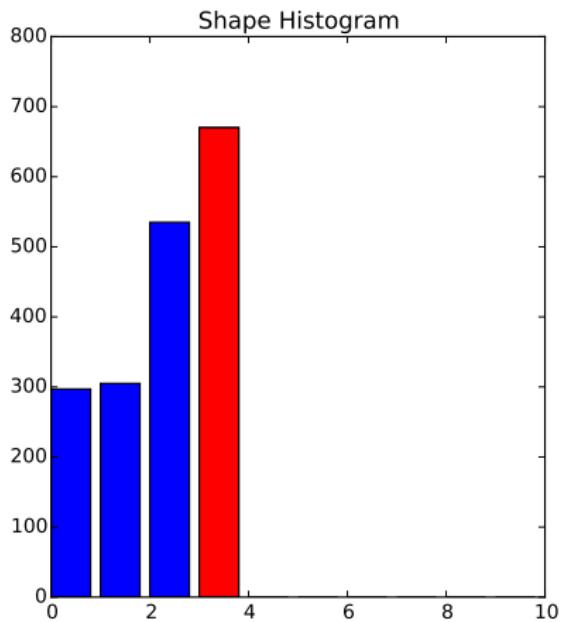
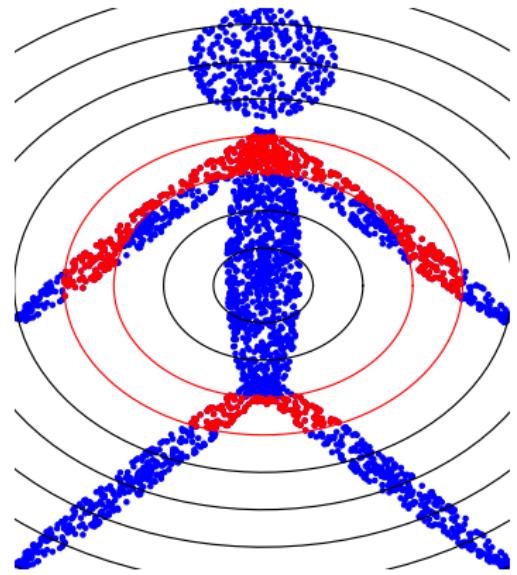
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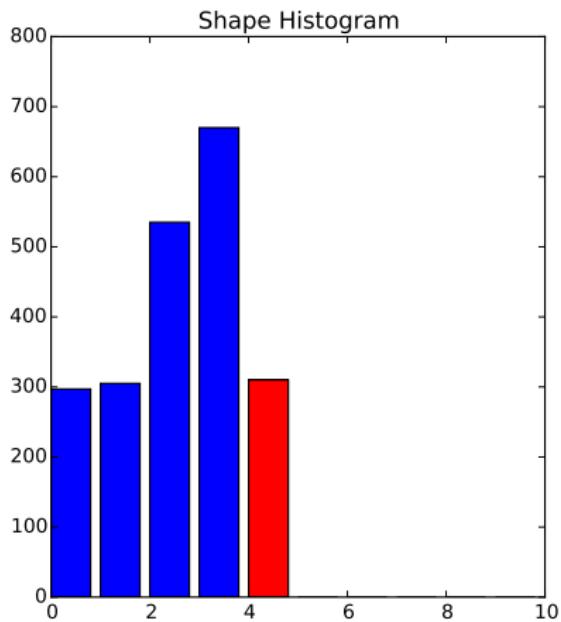
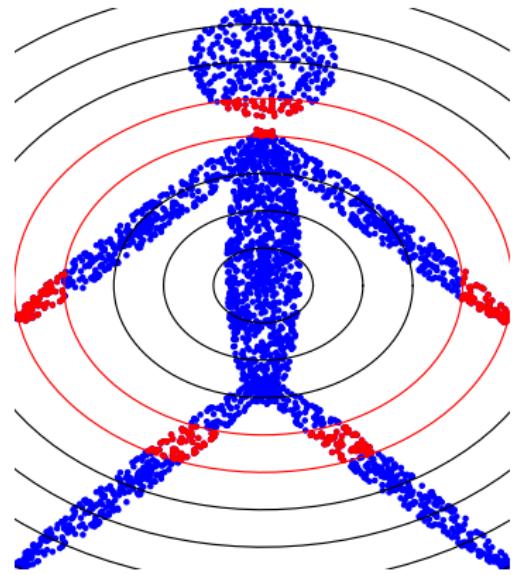
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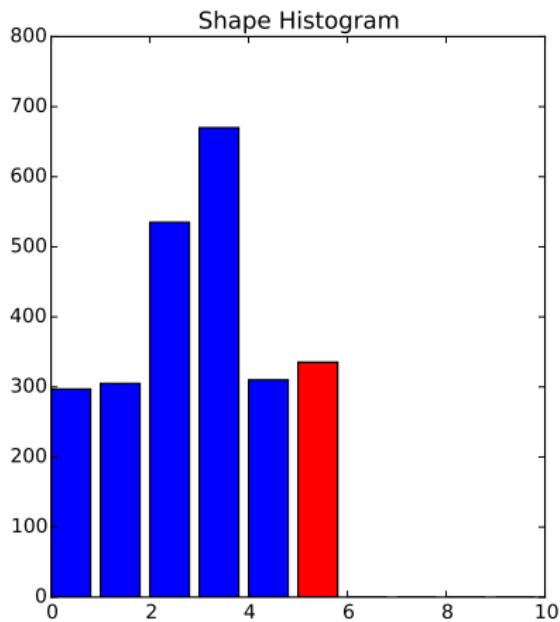
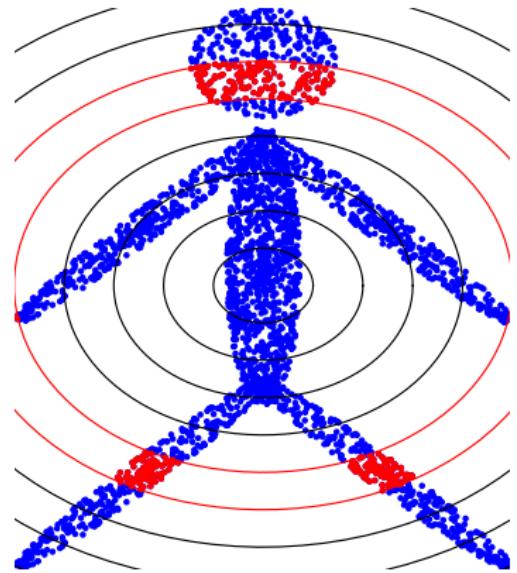
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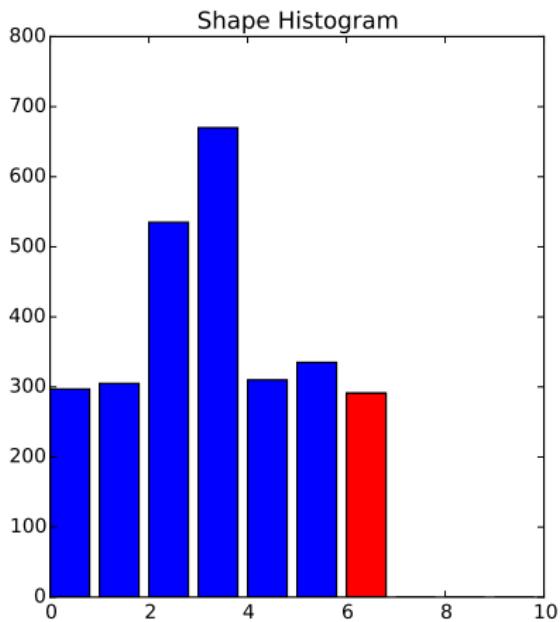
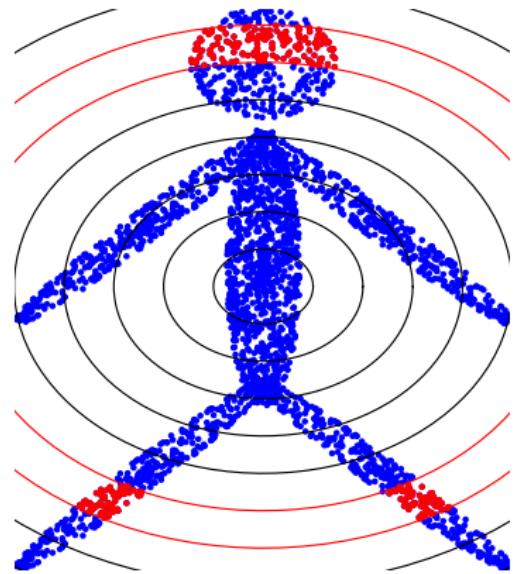
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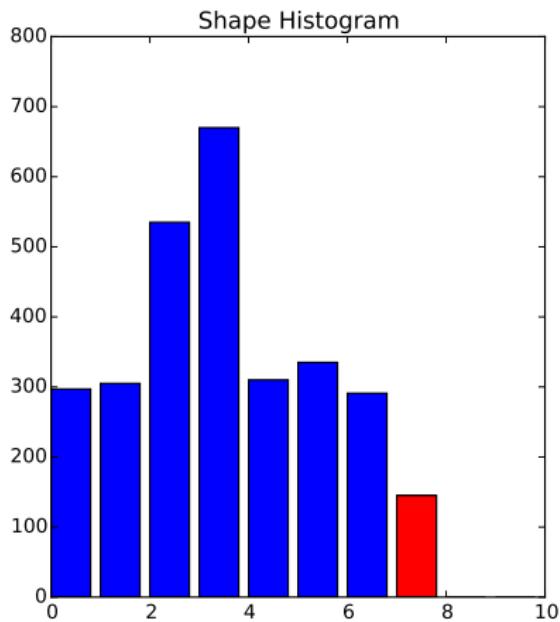
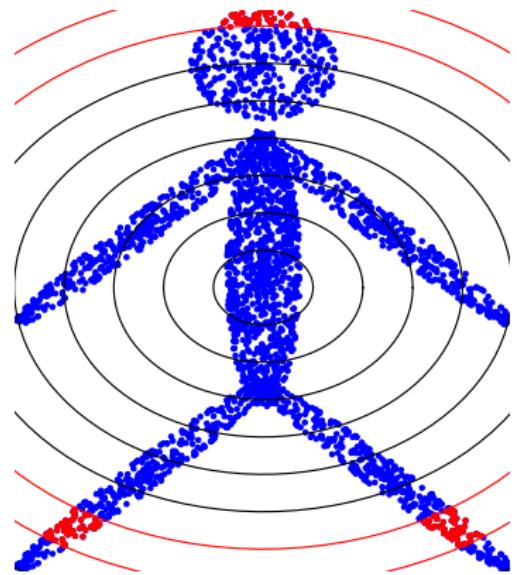
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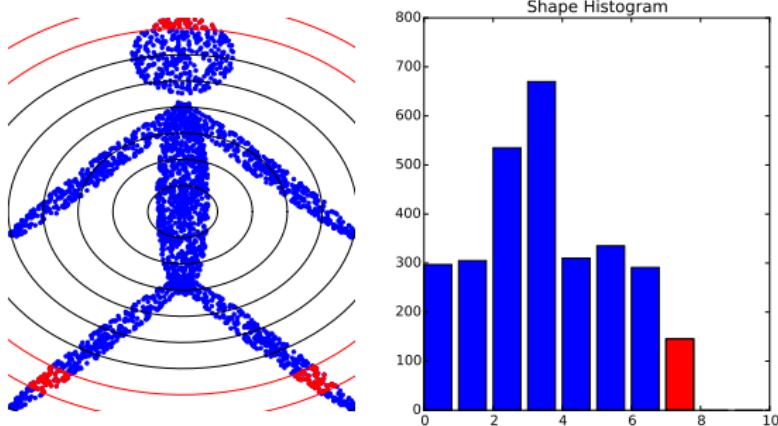
# Shape Histogram: Shells



# Shape Histogram: Shells

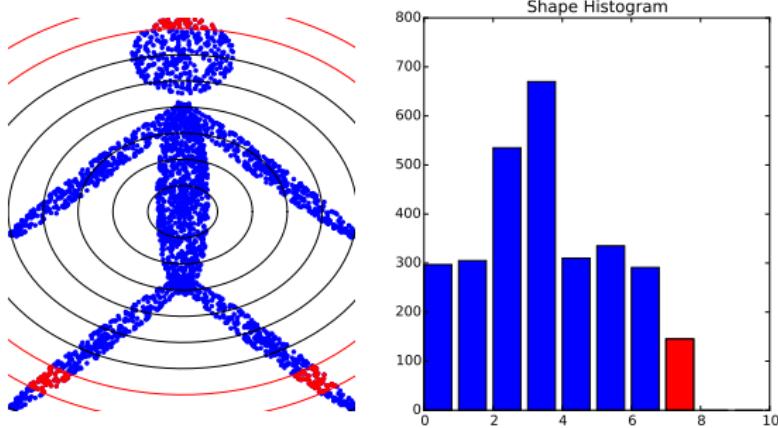


# Shape Histogram: Shells



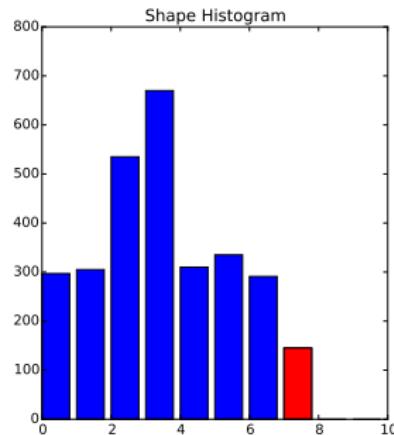
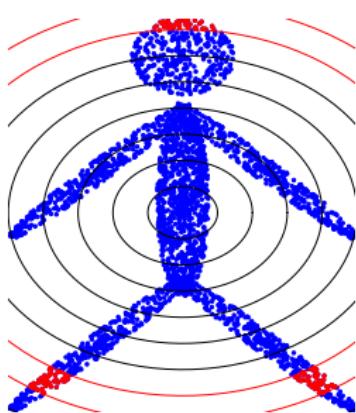
- ▷ Quick To Compute

# Shape Histogram: Shells



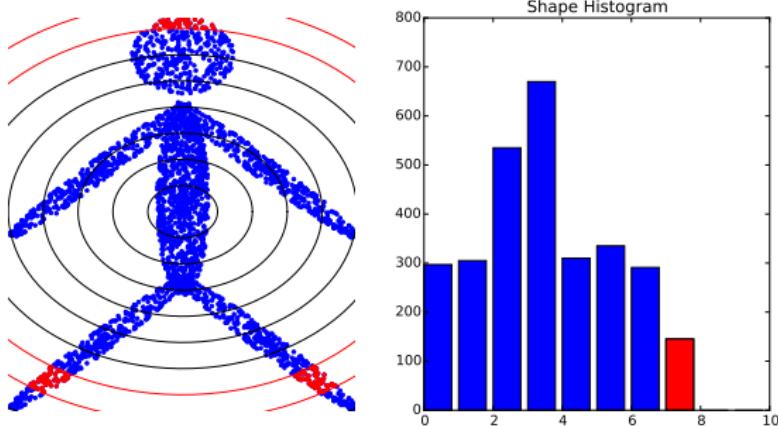
▷ Quick To Compute

# Shape Histogram: Shells



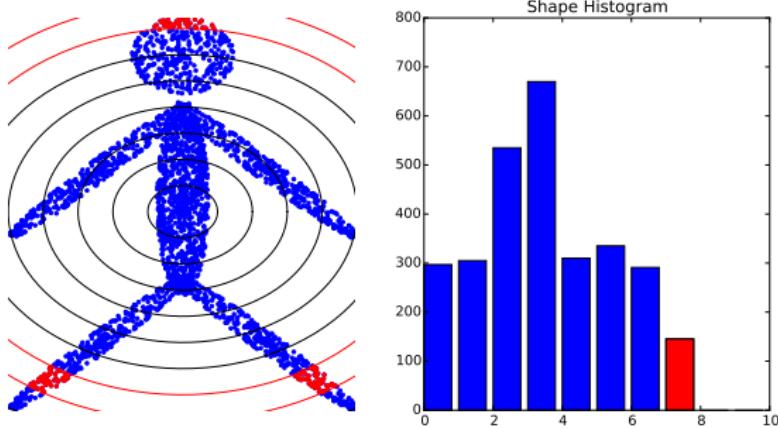
- ▷ Quick To Compute
- ▷ Concise To Store

# Shape Histogram: Shells



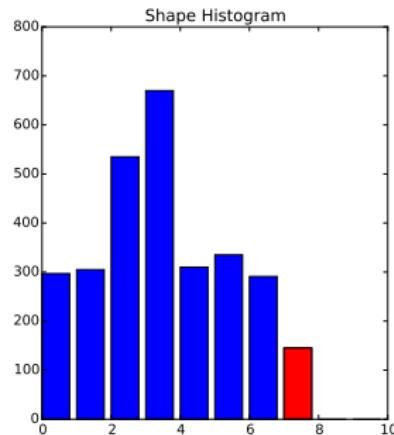
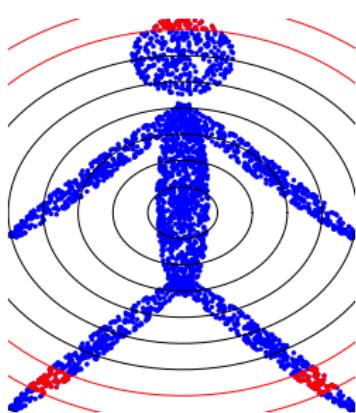
- ▷ Quick To Compute
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# Shape Histogram: Shells



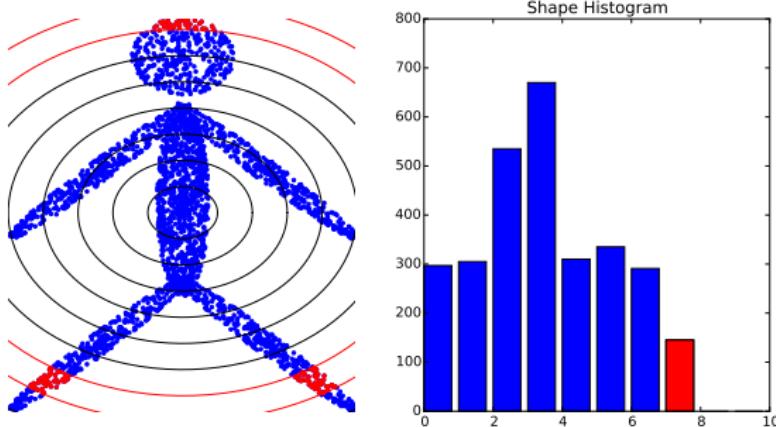
- ▷ Quick To Compute
- ▷ Concise To Store
- ▷ Rotation Invariant

# Shape Histogram: Shells



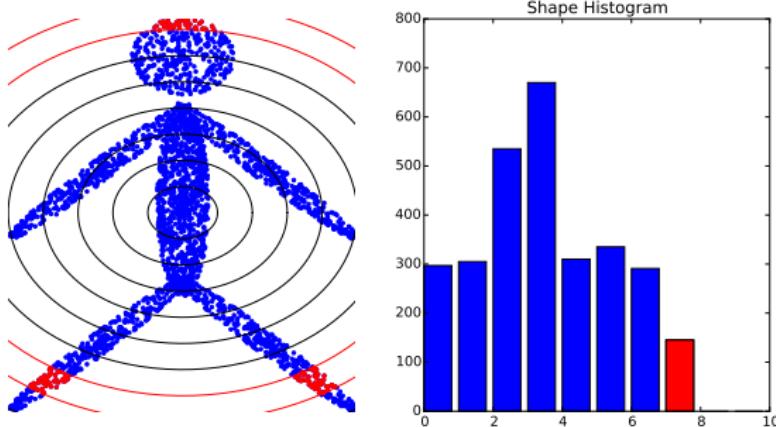
- ▷ Quick To Compute
- ▷ Concise To Store
- ▷ Rotation Invariant

# Shape Histogram: Shells



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

# Shape Histogram: Shells



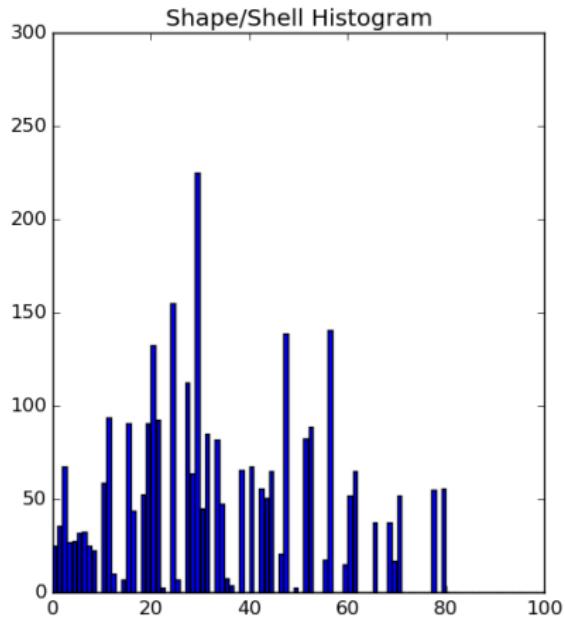
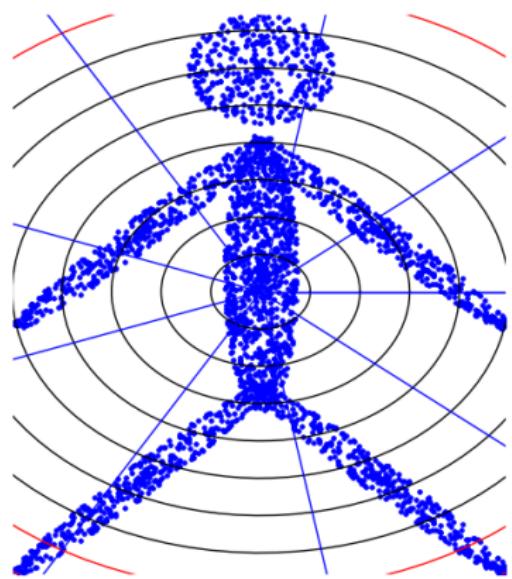
- ▷ Quick To Compute
- ▷ Concise To Store
- ▷ Rotation Invariant
- ▷ Discerning

# Shape Histogram: Shells

What can't it tell apart?

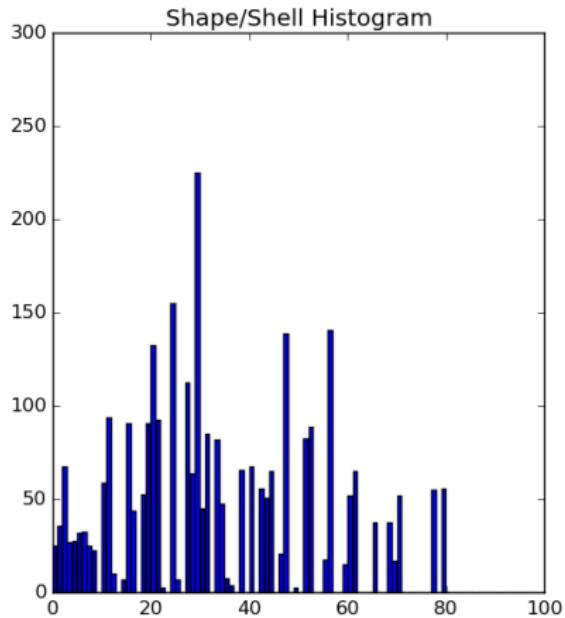
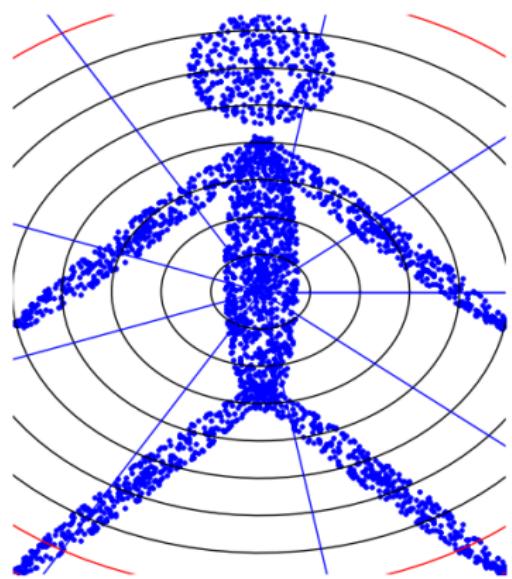
SHOW VIDEO

# Shape Histogram: Shells And Sectors



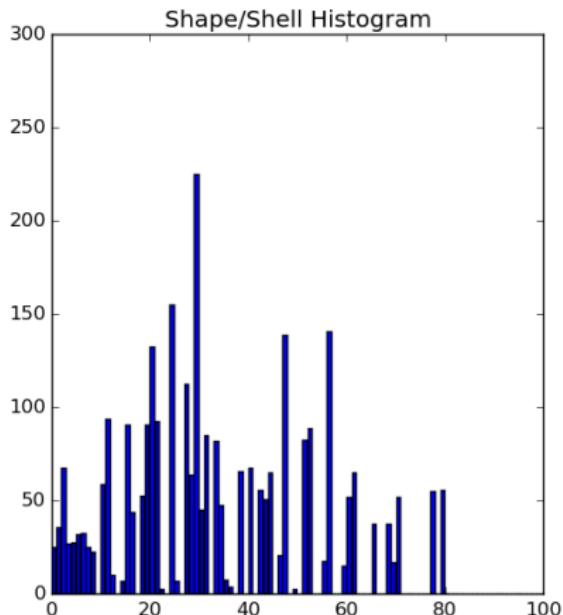
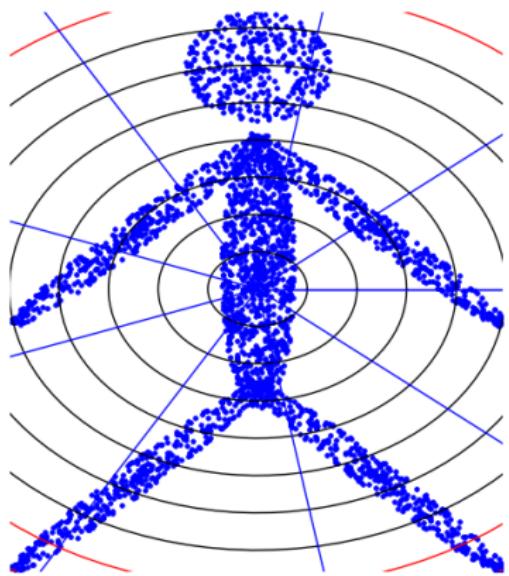
SHOW VIDEO

# Shape Histogram: Shells And Sectors



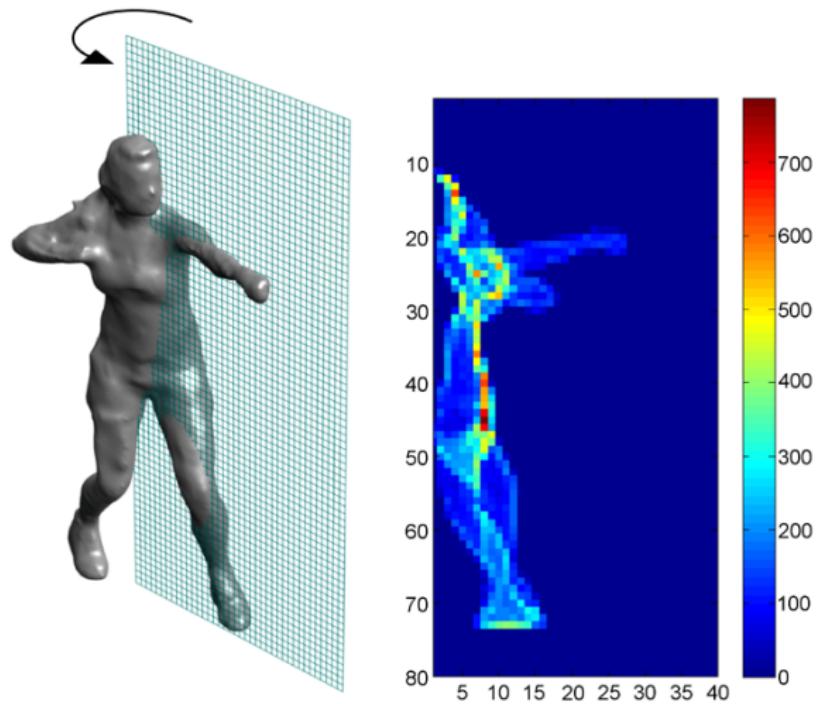
Still Rotation Invariant?

# Shape Histogram: Shells And Sectors



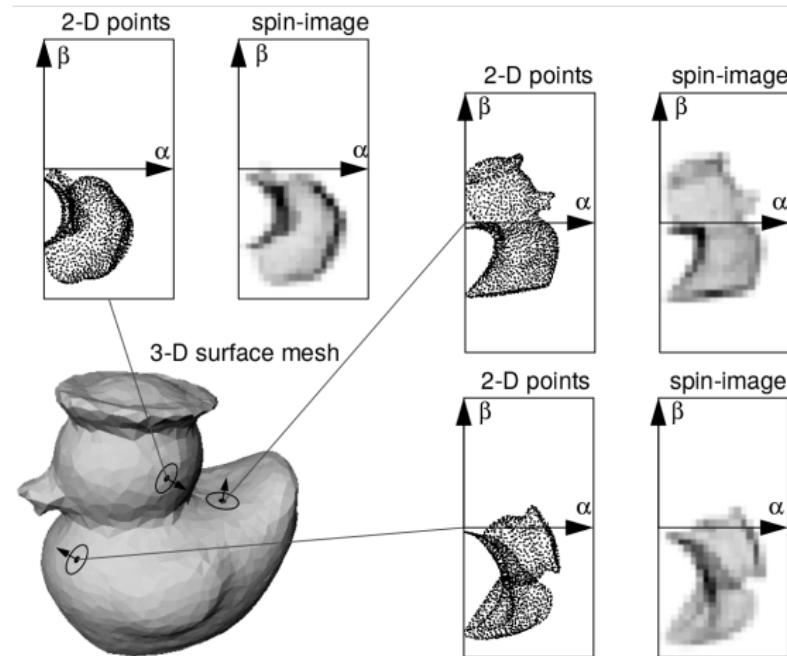
- ▷ Sort sectors within each shell
- ▷ Record PCA eigenvalues within each shell

# Spin Images



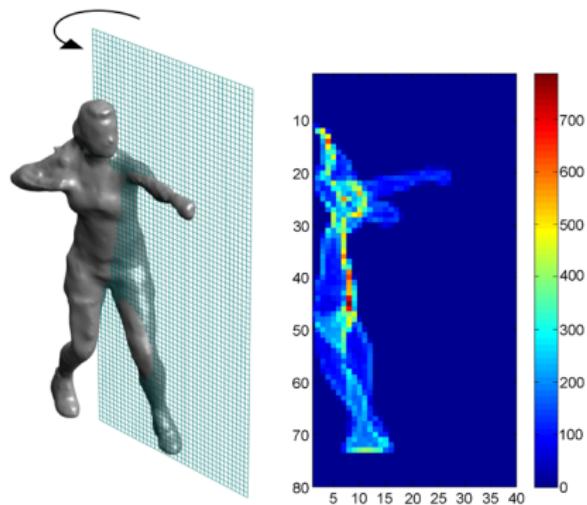
Johnson/Herbert 1999, Figure Huang 2010

# Spin Images: Rubber Duck



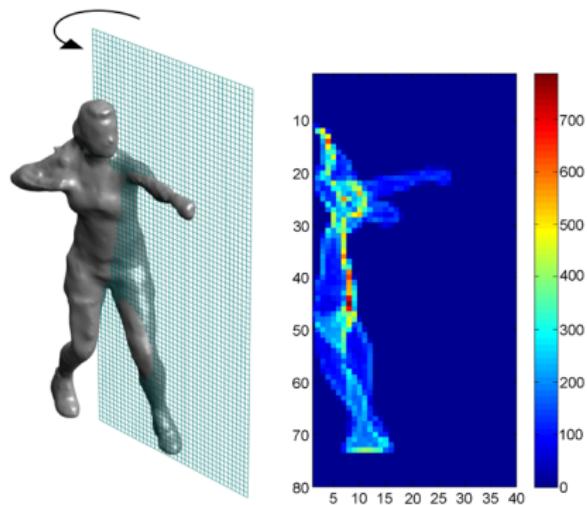
Johnson/Herbert 1999

# Spin Images



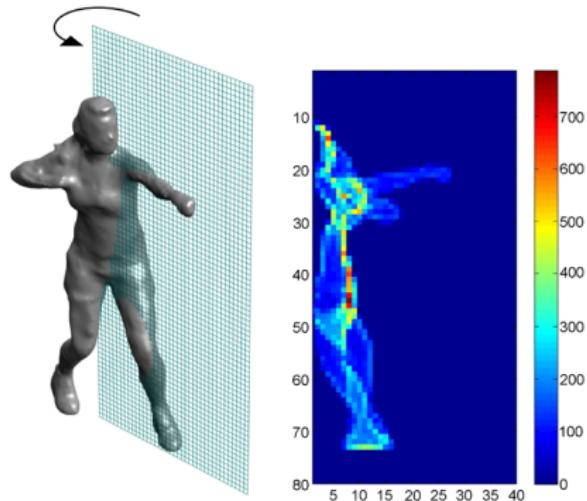
- ▷ Quick To Compute

# Spin Images



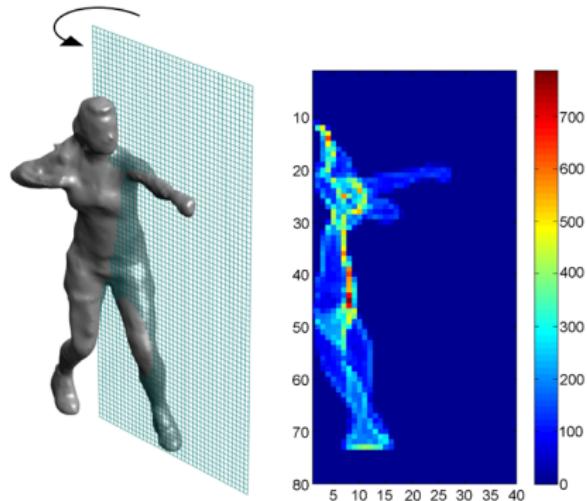
- ▷ Quick To Compute

# Spin Images



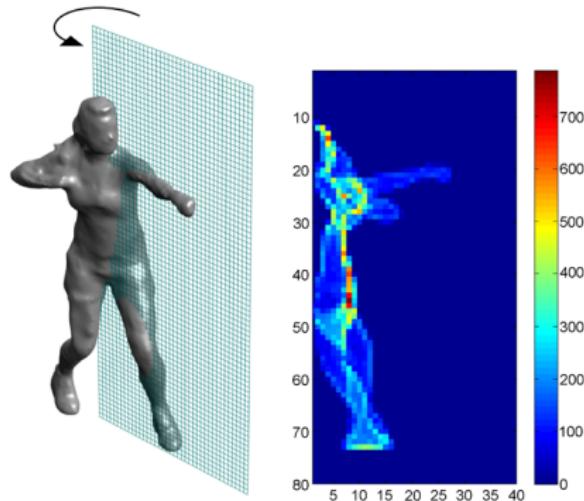
- ▷ Quick To Compute
- ▷ Concise To Store

# Spin Images



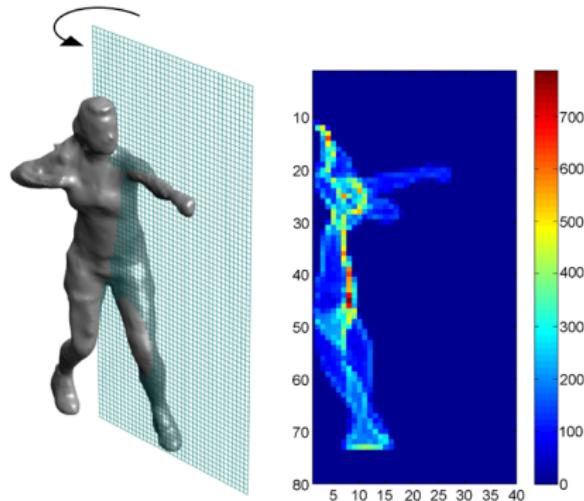
- ▷ Quick To Compute
- ▷ Concise To Store (Can compress images)

# Spin Images



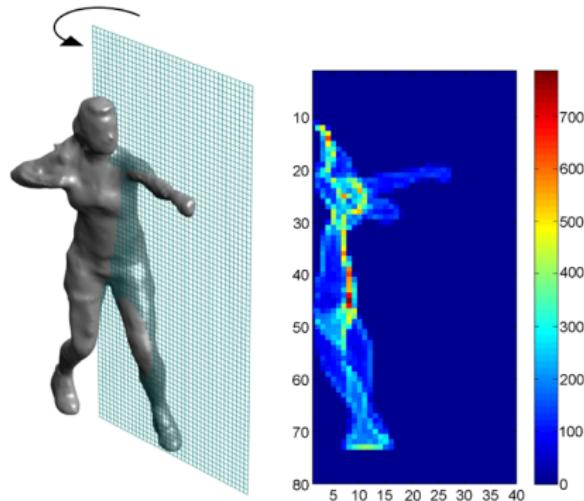
- ▷ Quick To Compute
- ▷ Concise To Store (Can compress images)
- ▷ Rotation Invariant

# Spin Images



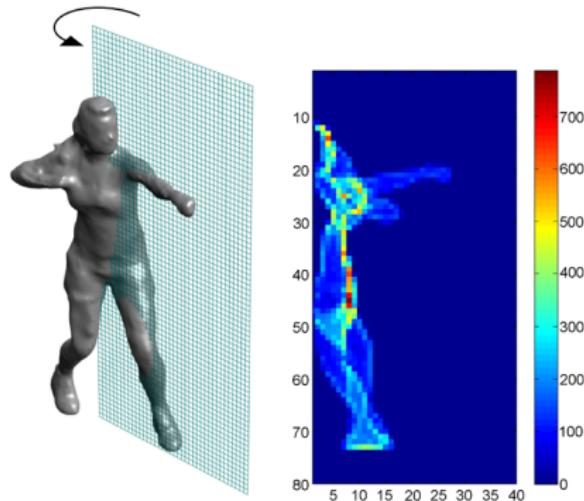
- ▷ Quick To Compute
- ▷ Concise To Store (Can compress images)
- ▷ Rotation Invariant (Careful with principal axis stability)

# Spin Images



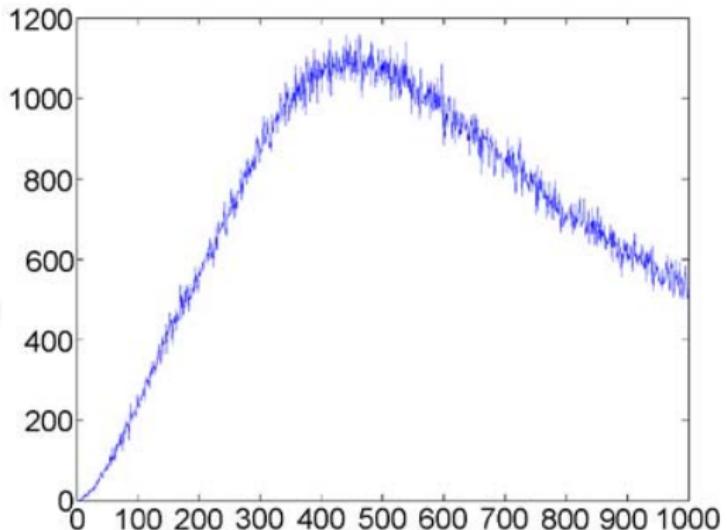
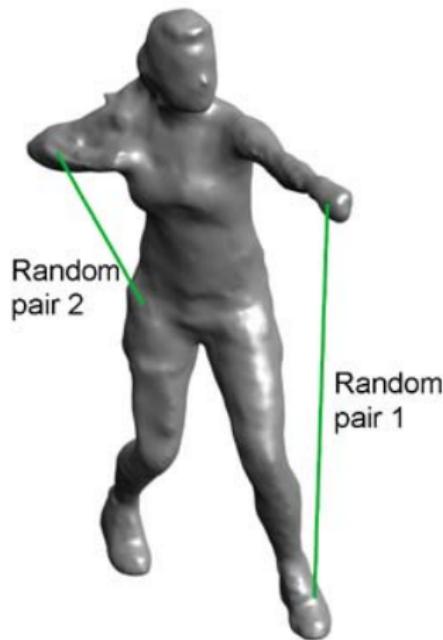
- ▷ Quick To Compute
- ▷ Concise To Store (Can compress images)
- ▷ Rotation Invariant (Careful with principal axis stability)
- ▷ Discerning

# Spin Images



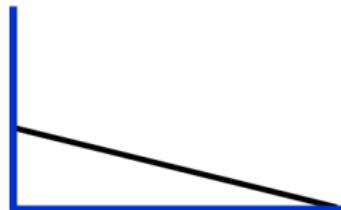
- ▷ Quick To Compute
- ▷ Concise To Store (Can compress images)
- ▷ Rotation Invariant (Careful with principal axis stability)
- ▷ Discerning

## D2: Distance Histograms

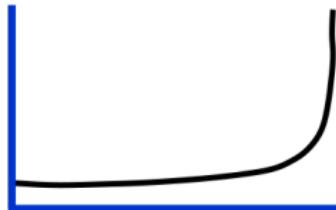


Osada 2003, Figure from Huang 2010

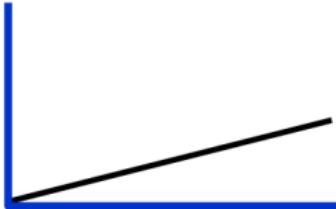
## D2: Primitive Examples



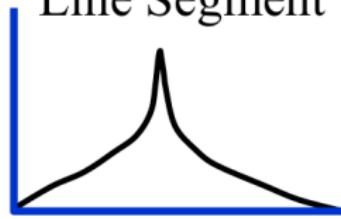
Line Segment



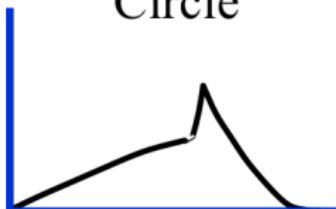
Circle



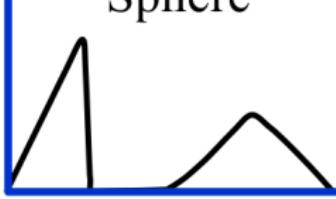
Sphere



Cylinder



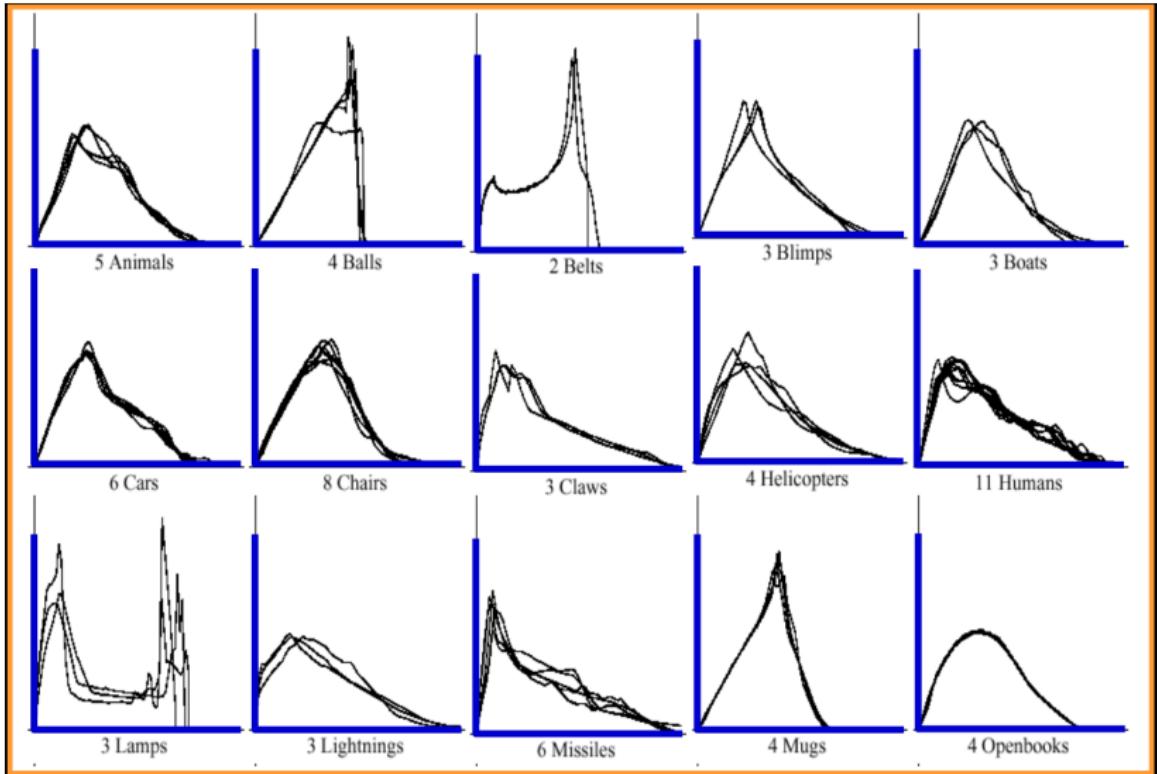
Cube



Two Spheres

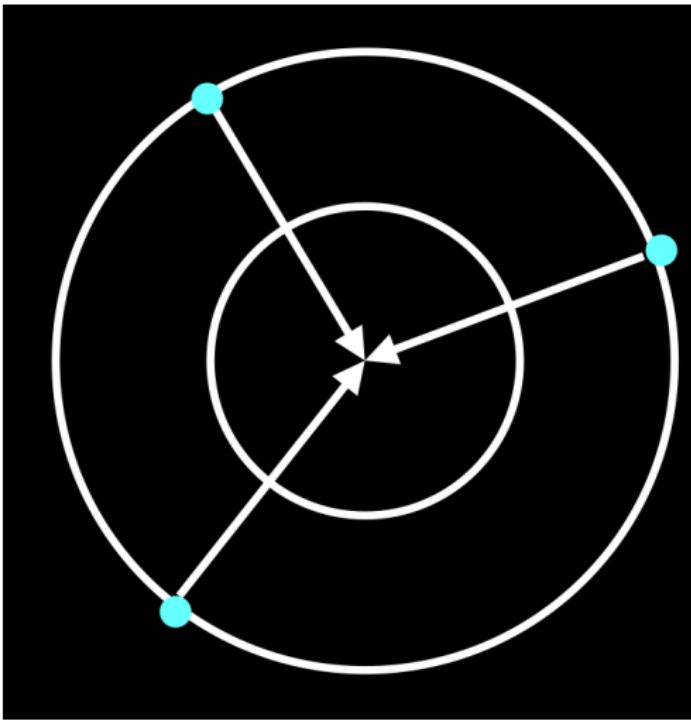
Osada 2003

## D2: Real Examples



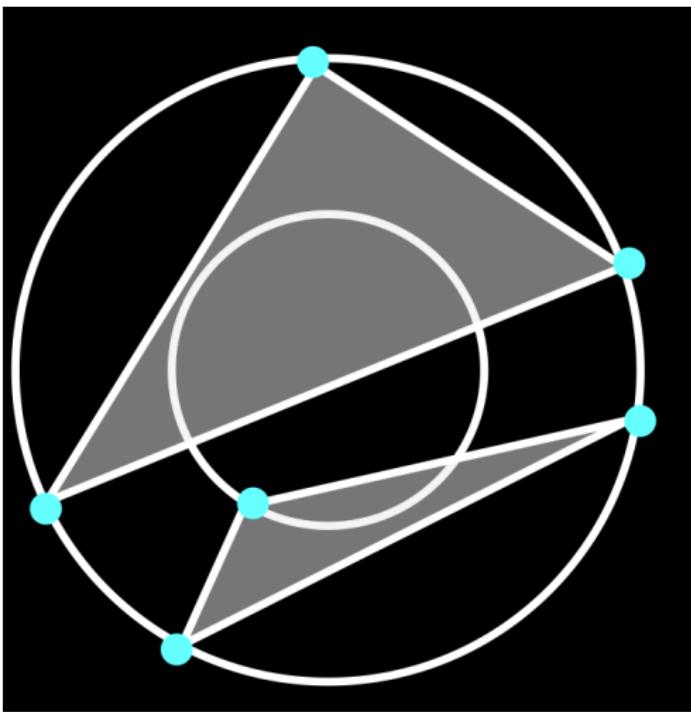
Osada 2003

# D1: Randomly Sample Points



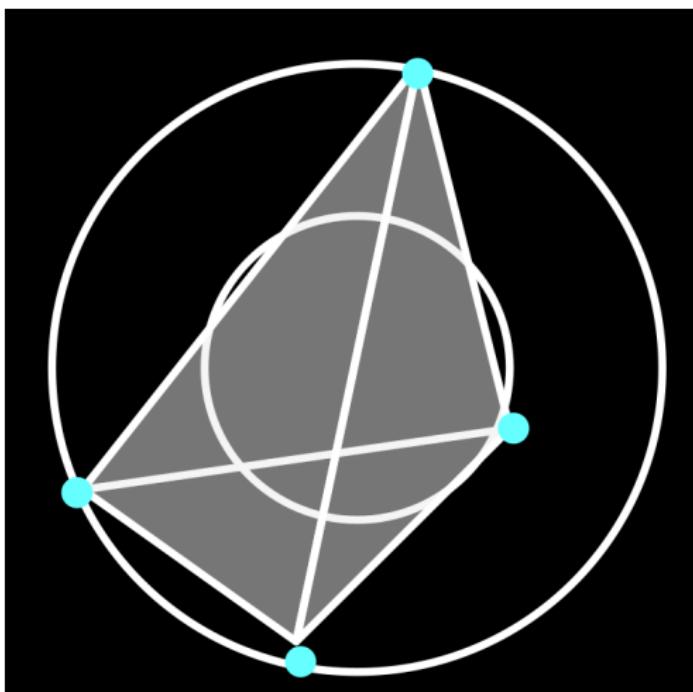
Osada 2003

## D3: Randomly Sample Areas



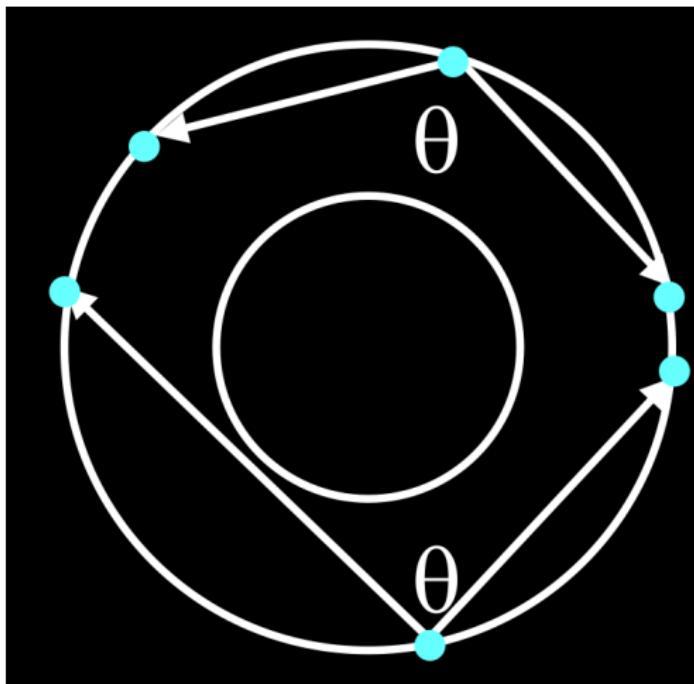
Osada 2003

## D4: Randomly Sample Volumes



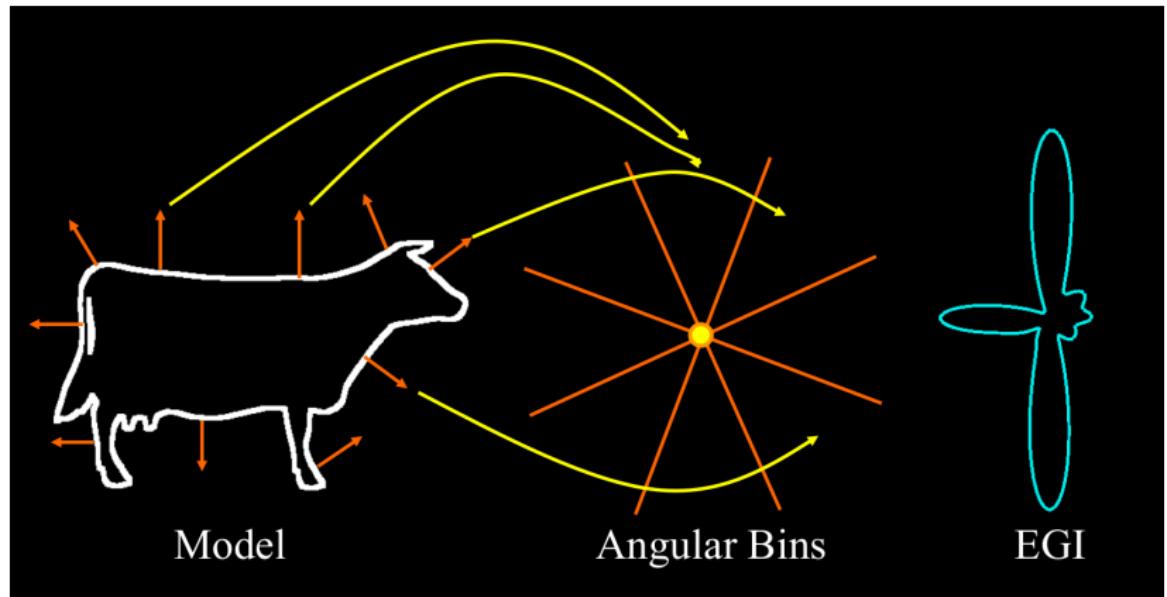
Osada 2003

## A3: Randomly Sample Angles



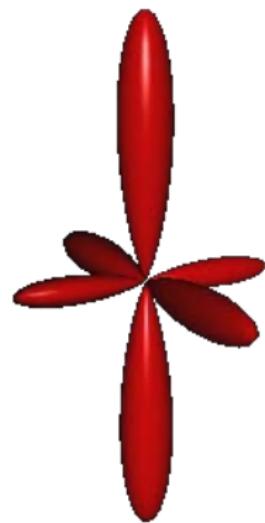
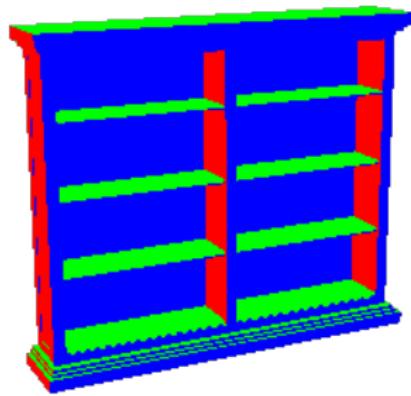
Osada 2003

# Extended Gaussian Image



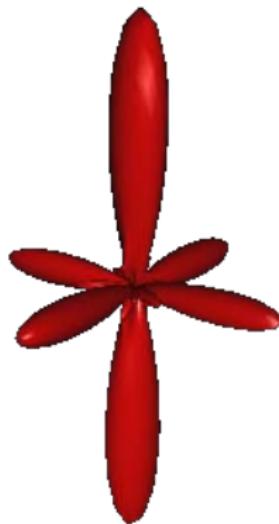
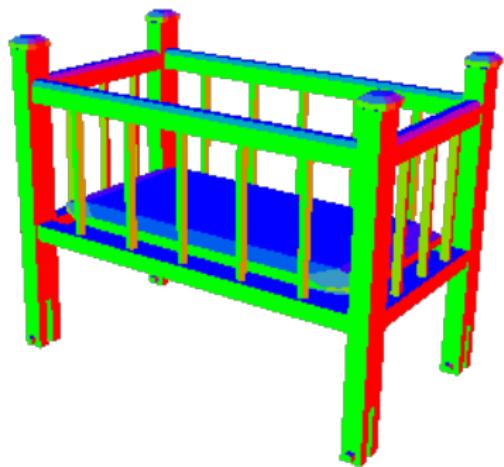
Funkhouser 2004

# Extended Gaussian Image



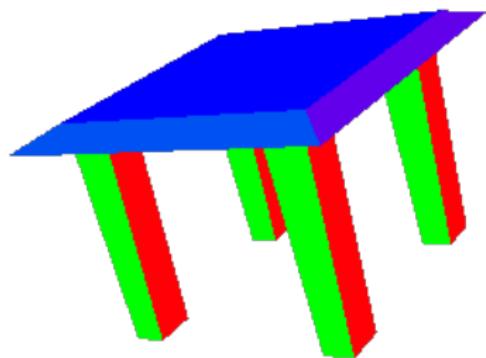
Funkhouser 2004

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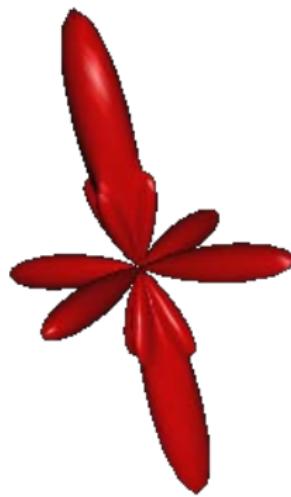
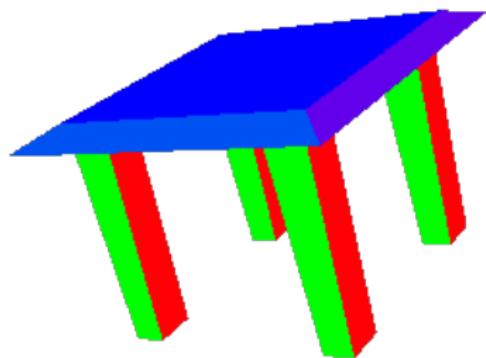
Funkhouser 2004

# Extended Gaussian Image



Funkhouser 2004

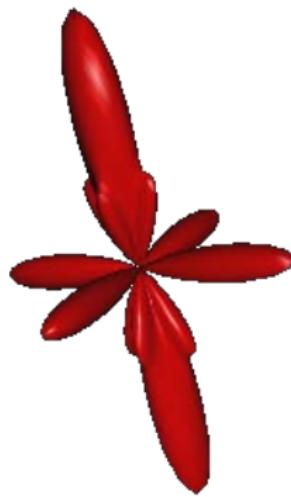
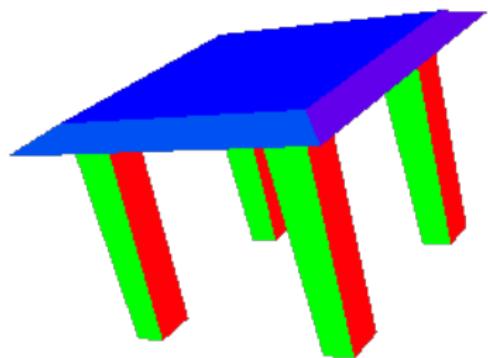
# Extended Gaussian Image



- ▷ Efficient To Compute / Concise To Store

Funkhouser 2004

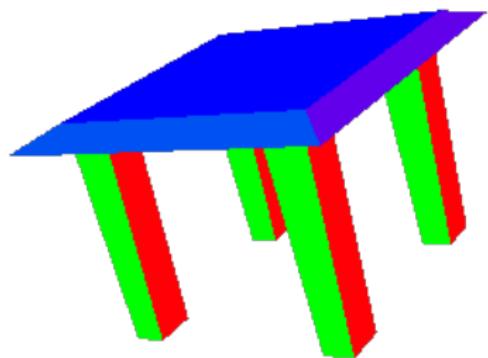
# Extended Gaussian Image



- ▷ Efficient To Compute / Concise To Store

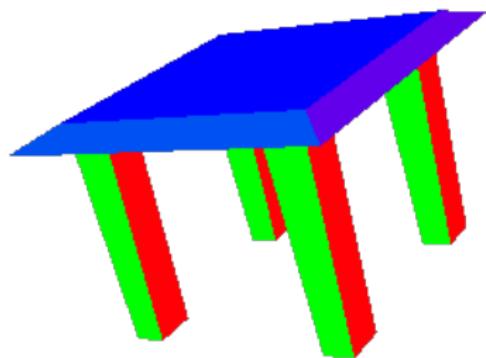
Funkhouser 2004

# Extended Gaussian Image



▷ Discerning  
Funkhouser 2004

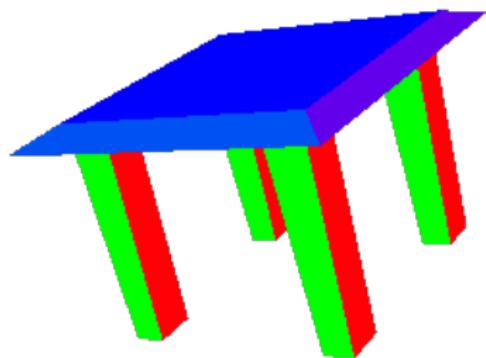
# Extended Gaussian Image



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

Funkhouser 2004

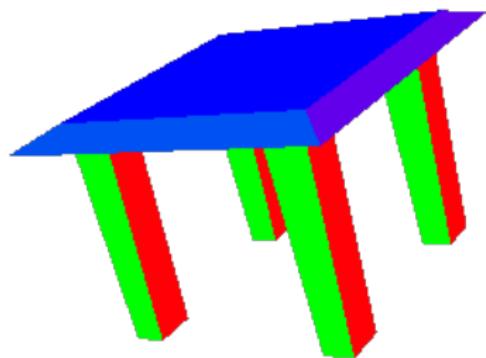
# Extended Gaussian Image



▷ Rotation Invariant

Funkhouser 2004

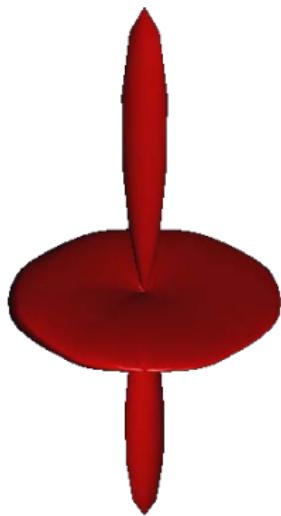
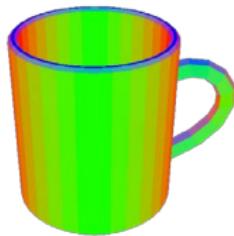
# Extended Gaussian Image



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

Funkhouser 2004

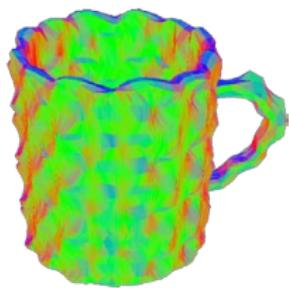
# Extended Gaussian Image



▷ Robust To Noise?

Funkhouser 2004

# Extended Gaussian Image



- ▷ Not Robust To Noise!

Funkhouser 2004

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# 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  $h_2$

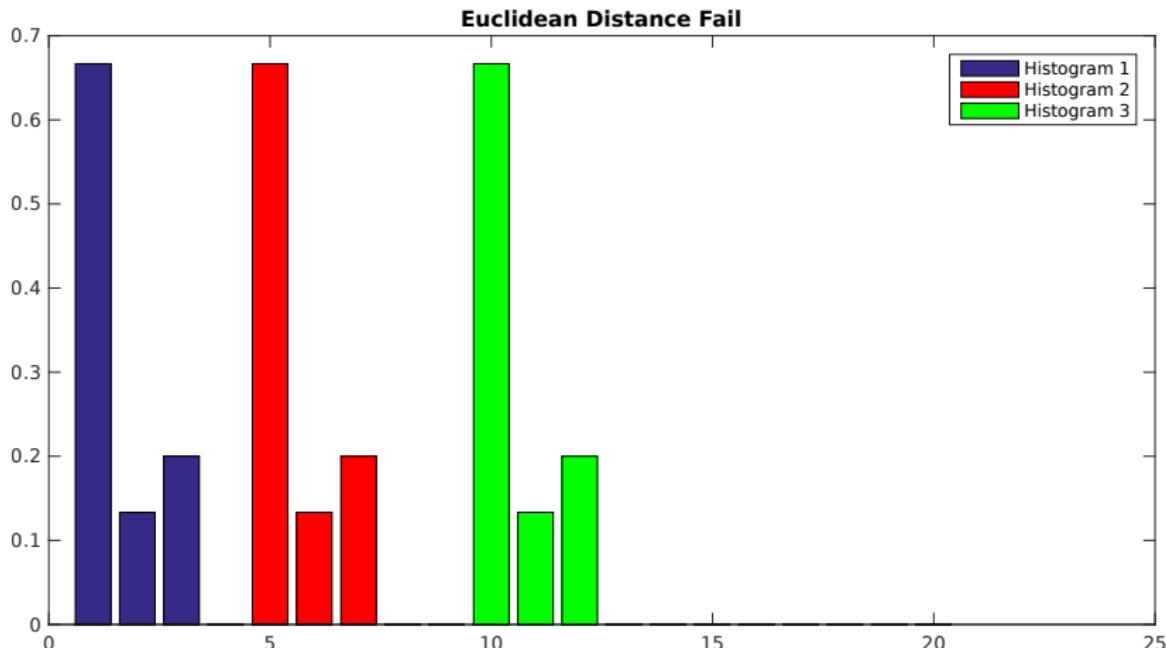
$$d_E(h_1, h_2) = \sqrt{\sum_{i=1}^N (h_1[i] - h_2[i])^2}$$

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

# Histogram Cosine Distance

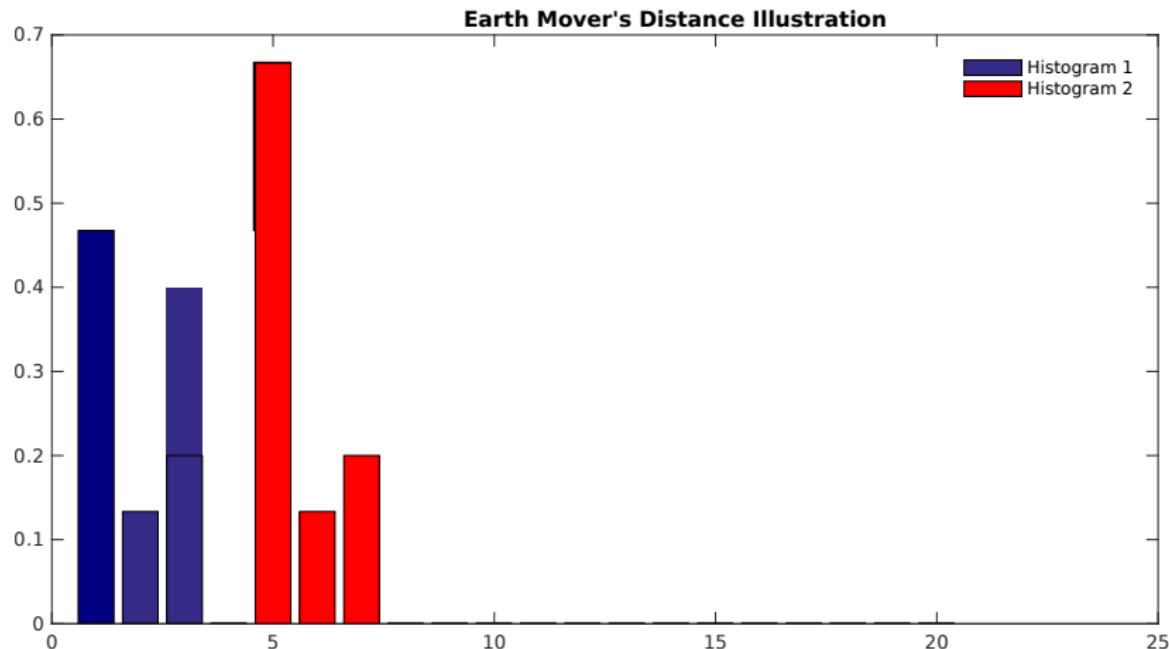
$$d_C(h_1, h_2) = \cos^{-1} \left( \frac{\vec{h}_1 \cdot \vec{h}_2}{\|\vec{h}_1\| \|\vec{h}_2\|} \right)$$

# Euclidean Distance Shortcomings



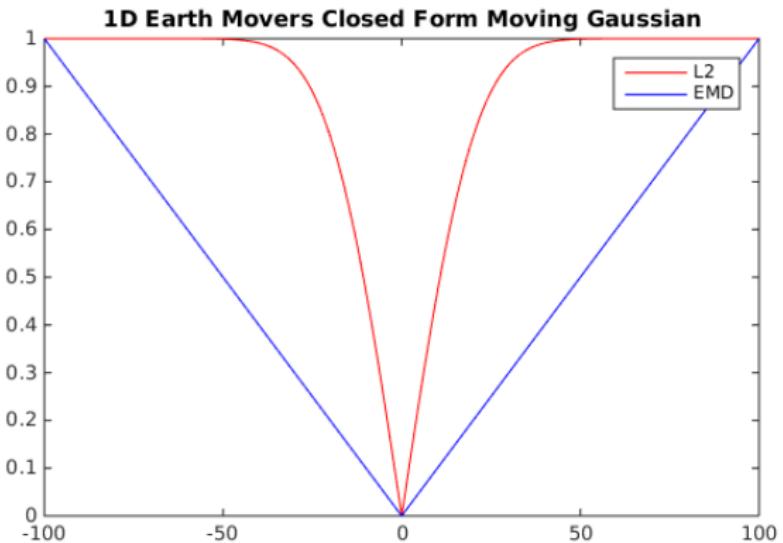
They all have the same distance!

# 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

# Evaluation Strategy

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



Rusinkiewiz/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