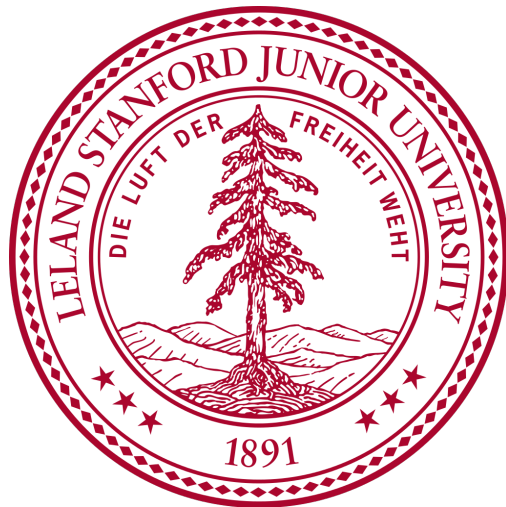


# Joint Embeddings of Shapes and Images via CNN Image Purification

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Daniel Cohen-Or   Leonidas J. Guibas  
(\* Joint First Authors)



# Joint Embeddings of Shapes and Images via CNN Image Purification



Deep learning is so cool for so many problems...

# Deep learning, yay or nay?

A piece of cake,  
elementary math...

$$Y = f(X)$$

What the hell is  
the  $f$ ?



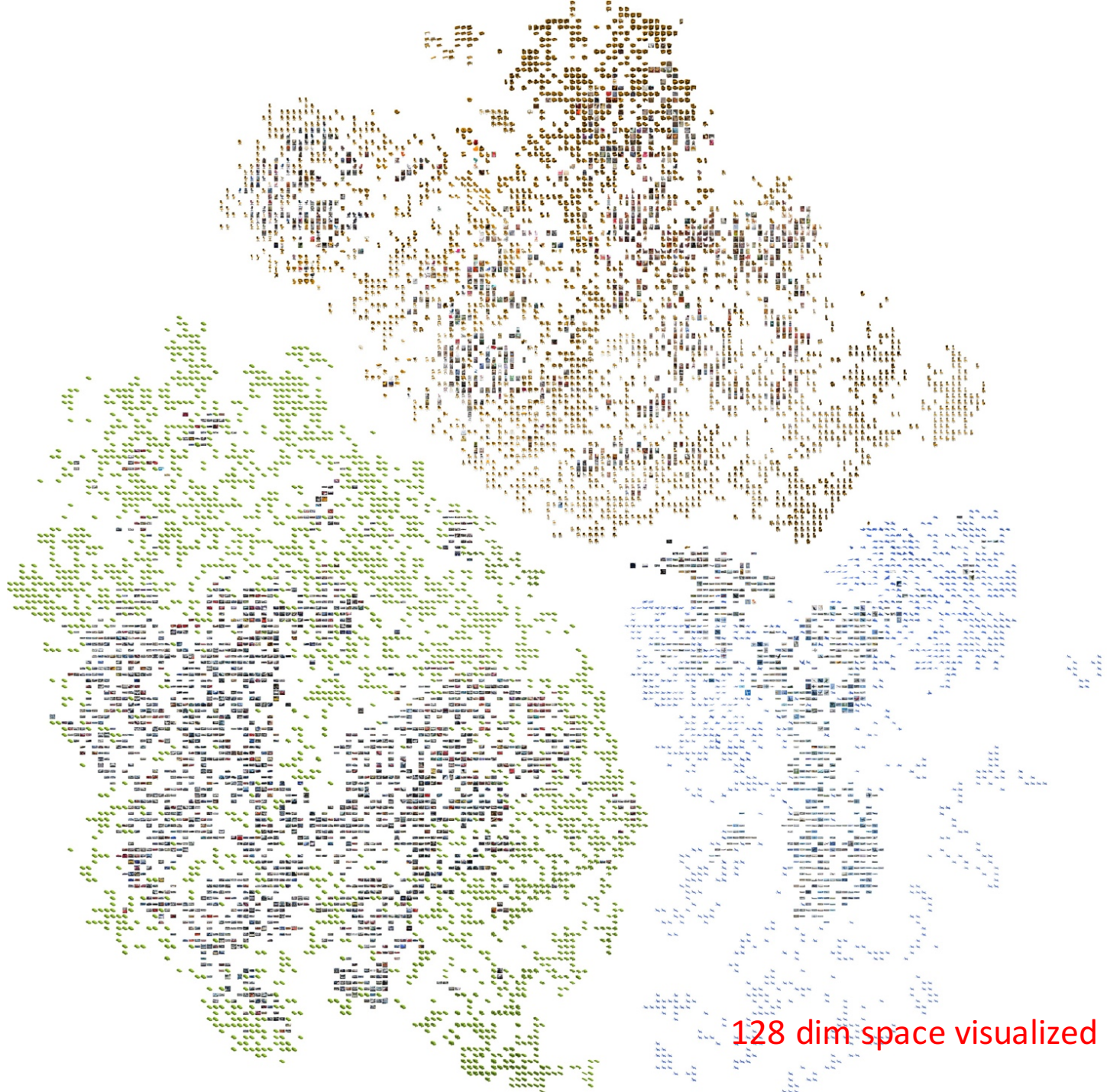
It eats, a lot!





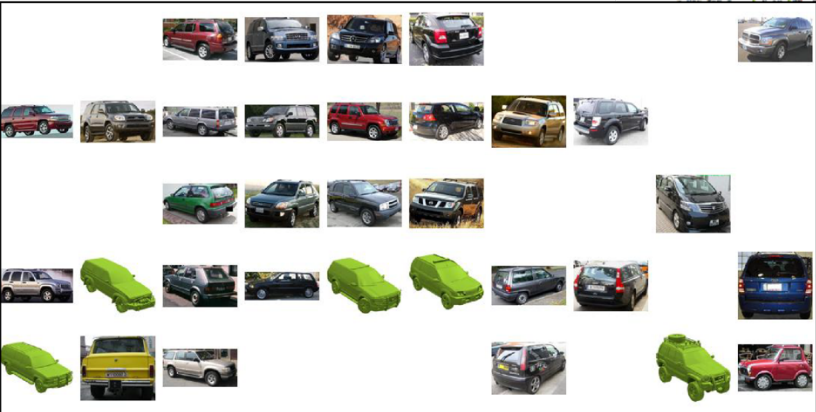
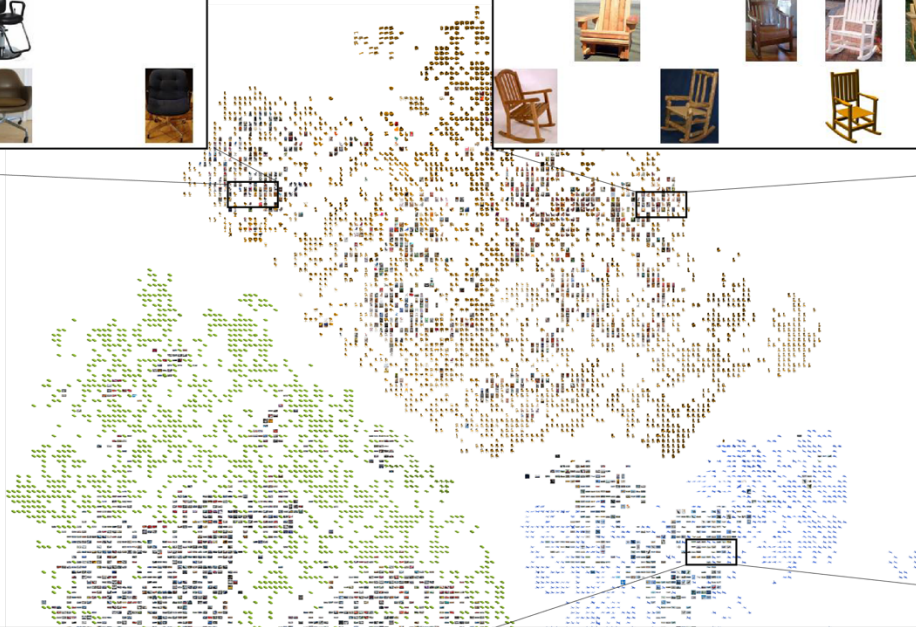
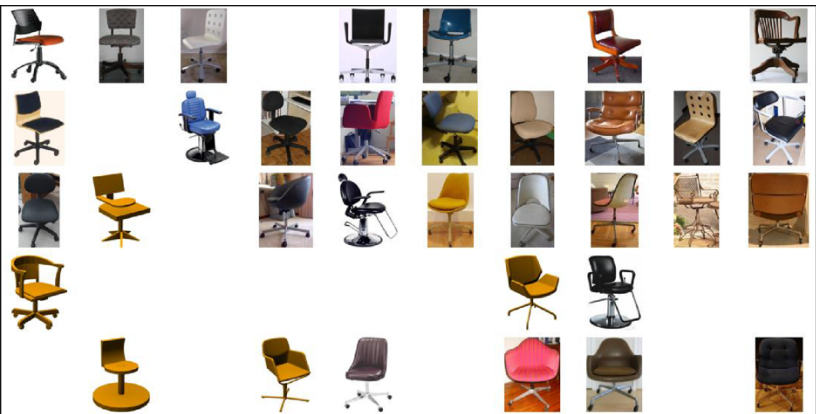
A “FoodTech” for Deep Learning

# Joint Embeddings of Shapes and Images via CNN Image Purification



128 dim space visualized by t-SNE





# Image based Shape Retrieval

Query



Top 5 Neighbors

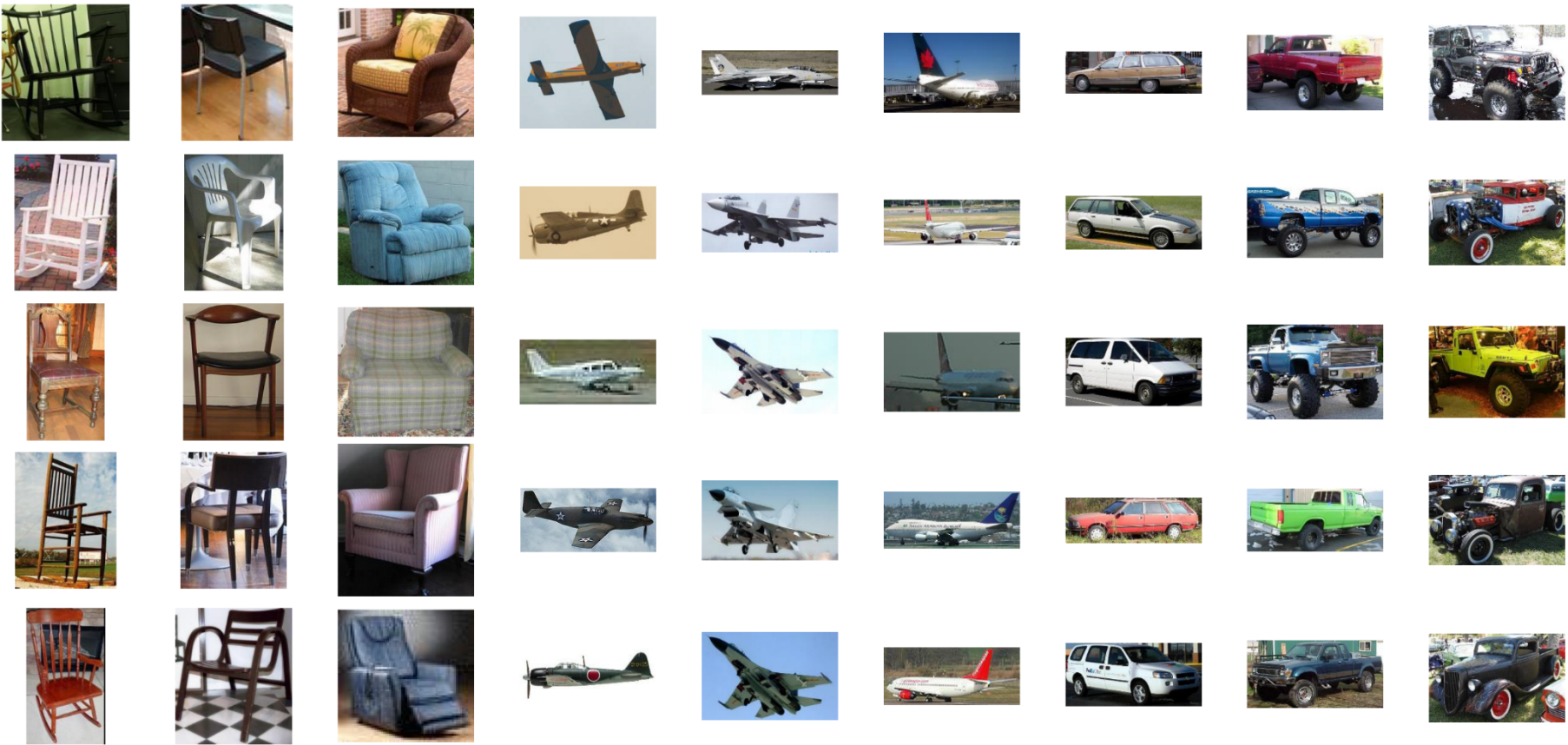


# Shape based Image Retrieval

Query

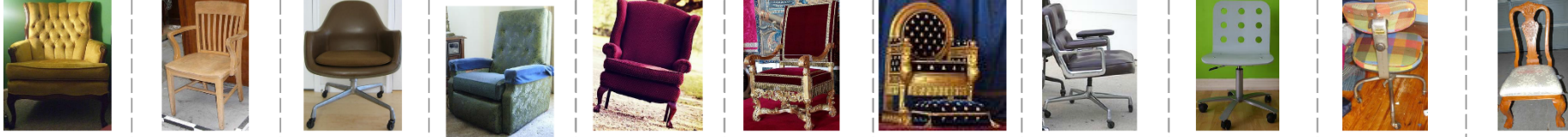


Top 5 Neighbors



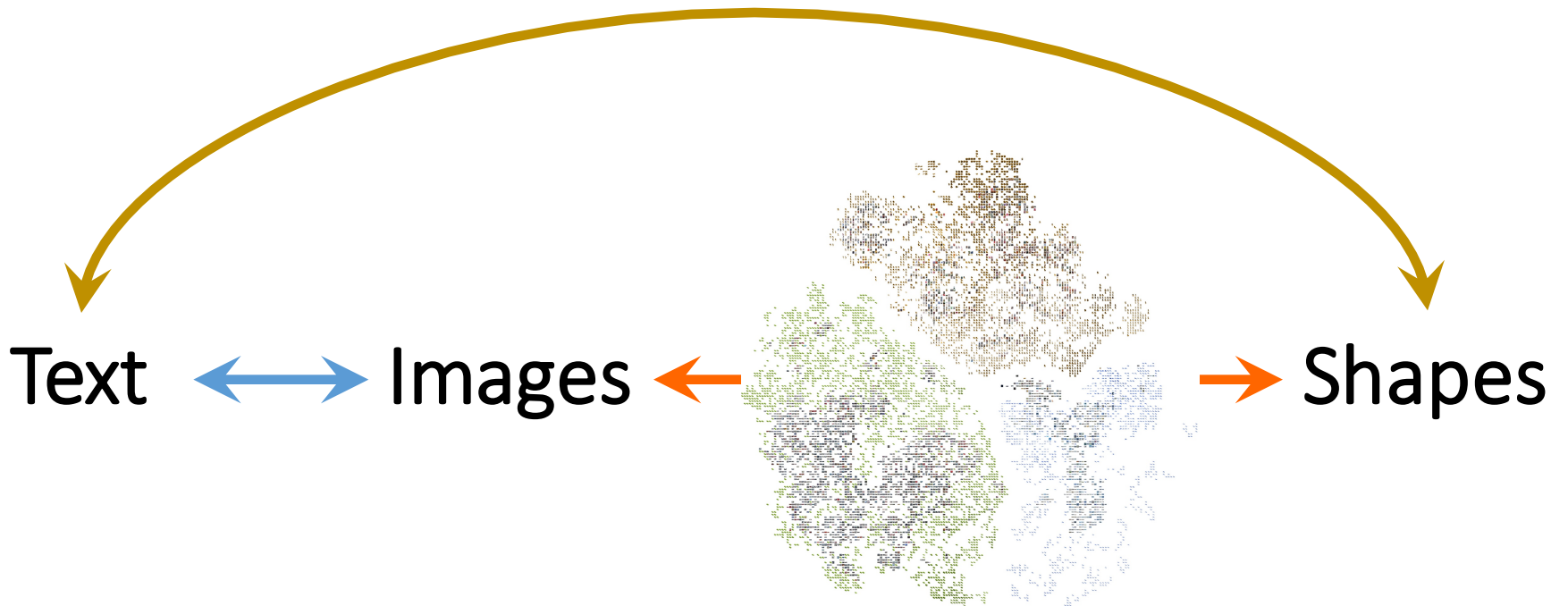
# Cross-View Image Retrieval

Query

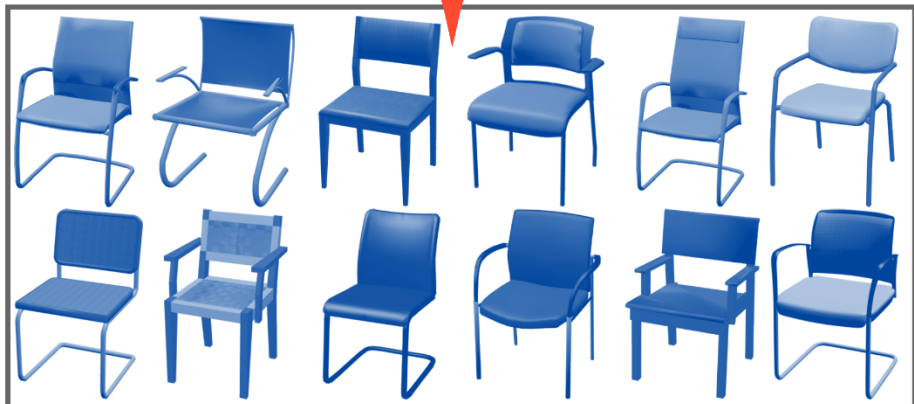
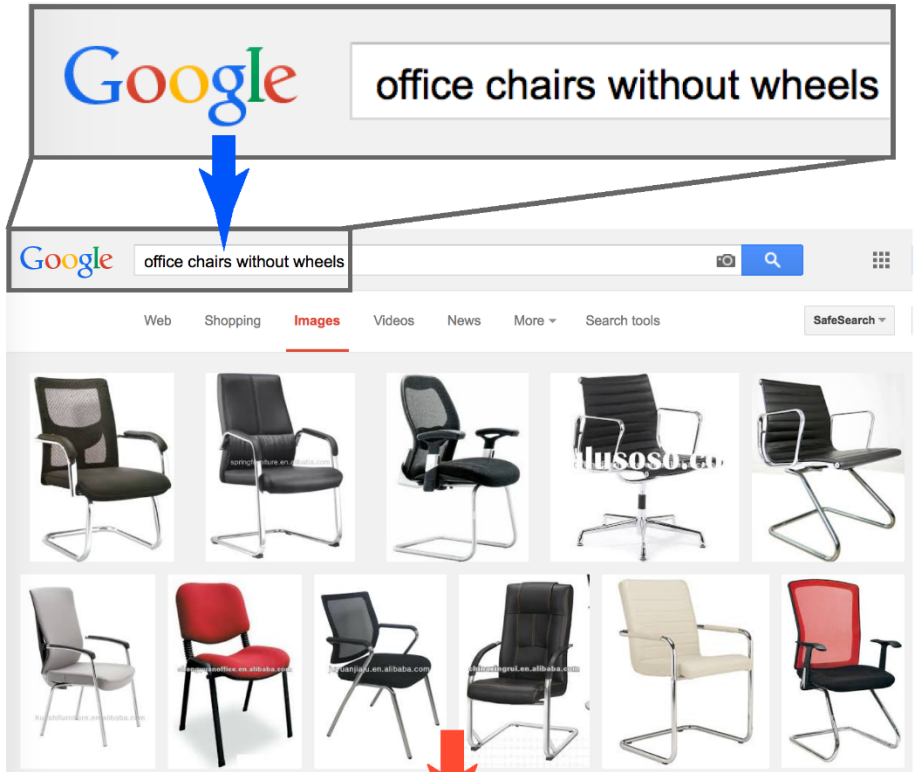
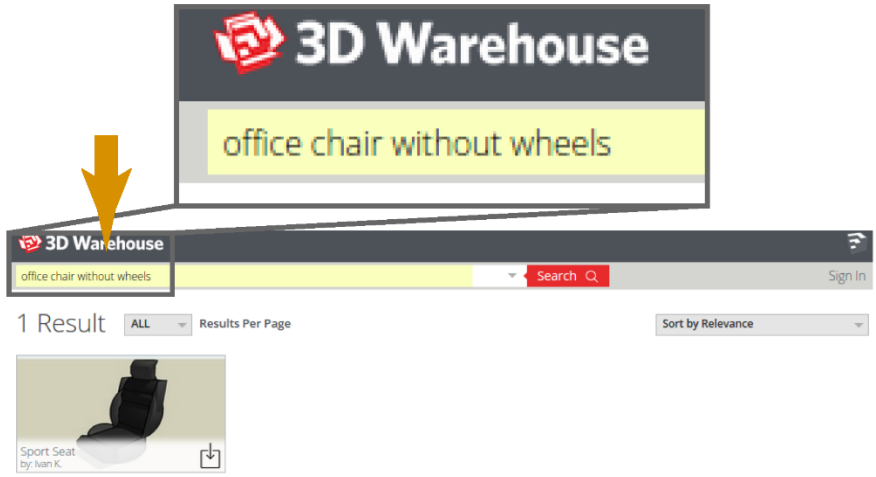





Top 3 Neighbors





# Text based Shape Retrieval

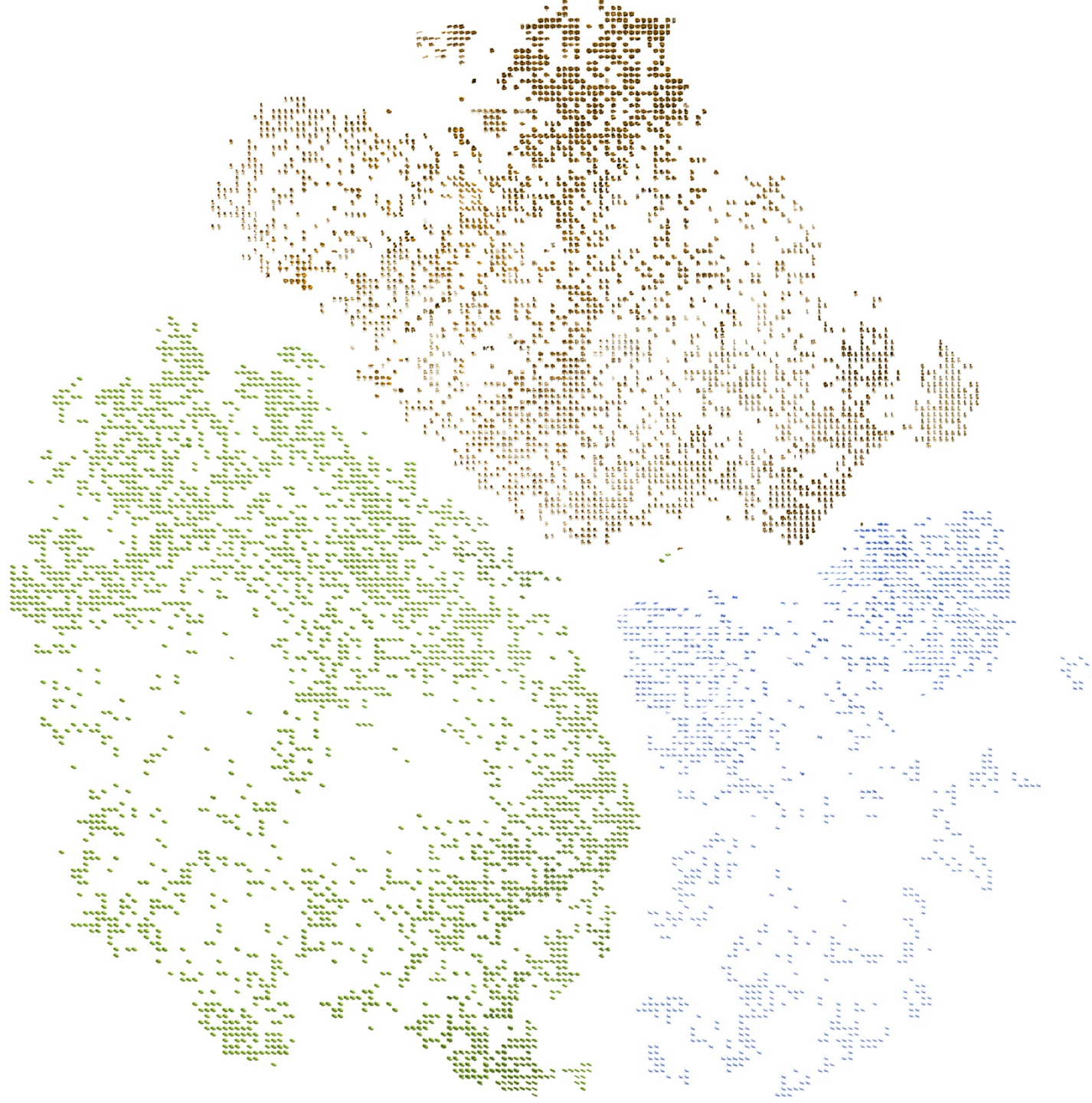


-  Text to Shape Query (TS)
-  Text to Image Query (TI)
-  Image to Shape Query (IS)

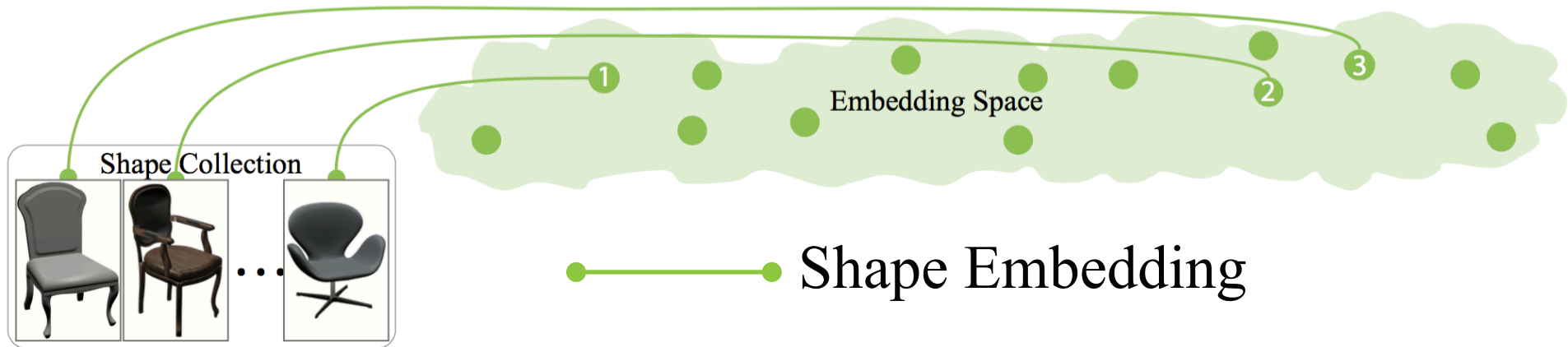
# Text based Shape Retrieval

The image shows a screenshot of the 3D Warehouse website. At the top, there is a search bar with the text "chairs with round back" and a search button. Below the search bar, there are 45 search results displayed in a grid. Each result includes a small image of a chair and its name and designer. The results are sorted by relevance. The chairs shown include various styles such as dining chairs, lounge chairs, and armchairs.

The image shows a screenshot of a Google search for "chairs with round back". The search results are displayed in the "Images" tab. A red arrow points from the search results to a grid of blue chairs, illustrating the concept of text-based shape retrieval. The grid contains 14 blue chairs of various styles, all rendered in a uniform blue color, representing the retrieval of shapes based on the text query.







$$\textit{Similarity}_{(s_i, s_j)} = \|\mathcal{P}_i - \mathcal{P}_j\|$$

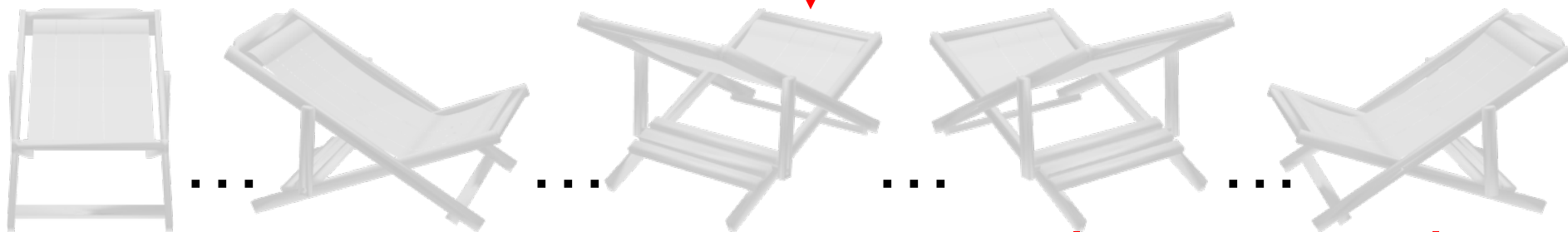
Many choices for  $\mathcal{P}_i$ :

Shape Histograms, Spin Images, Spherical Harmonics, Shape Distributions, etc.

LFD-HoG  
Very Strong!



Light Field Rendering



HoG

HoG

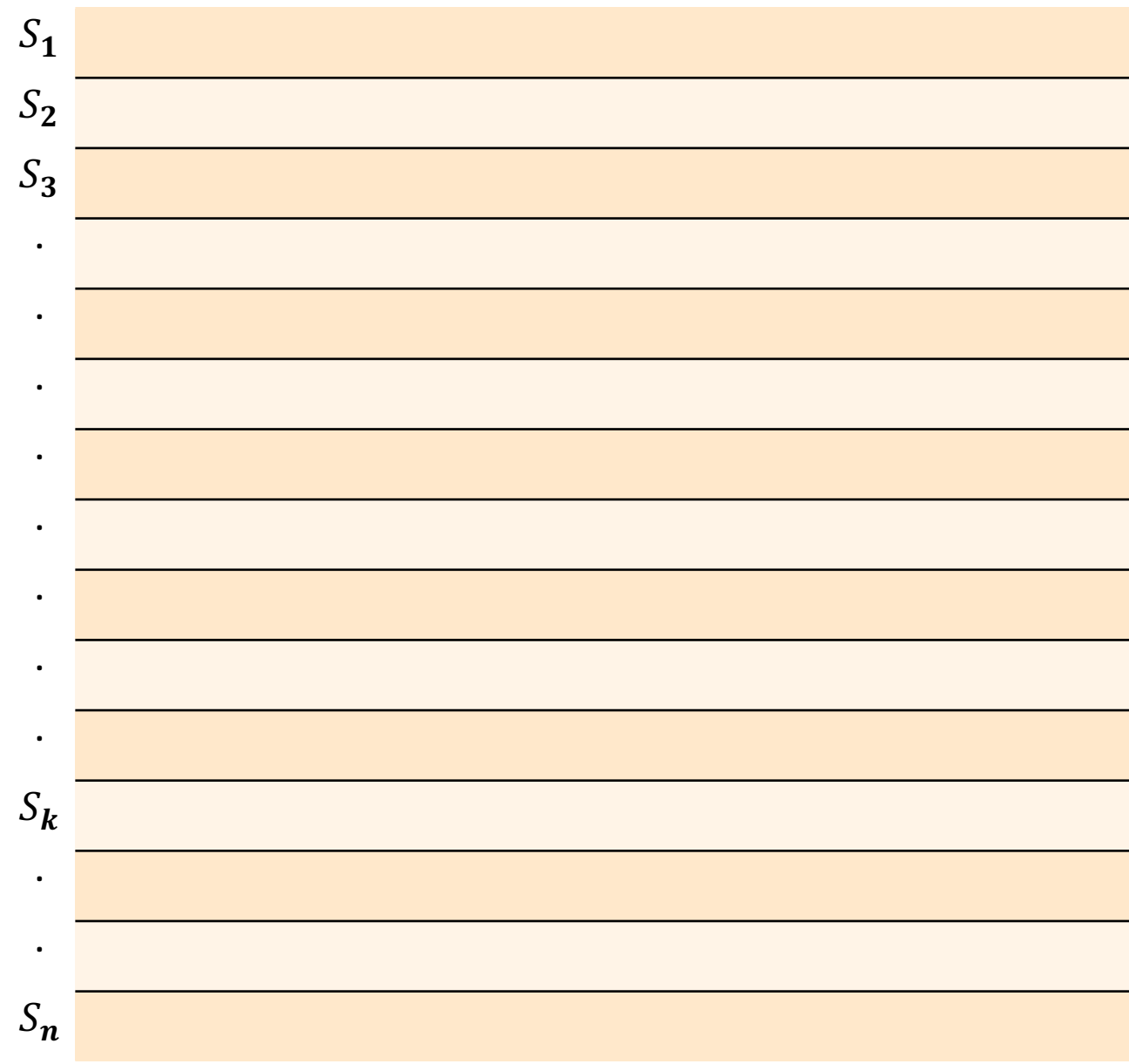
HoG

HoG

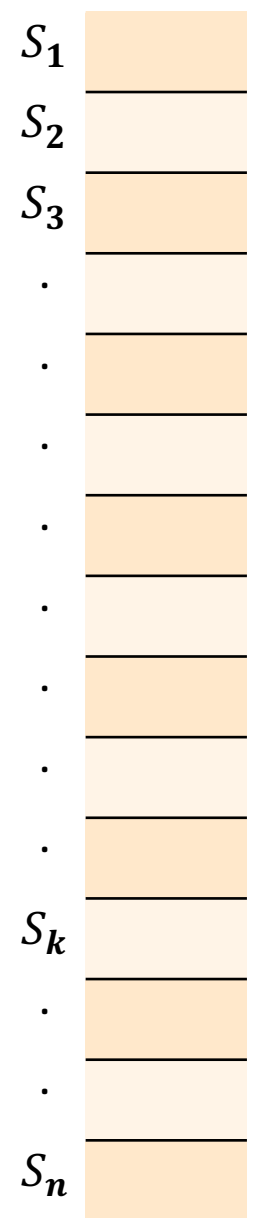
HoG

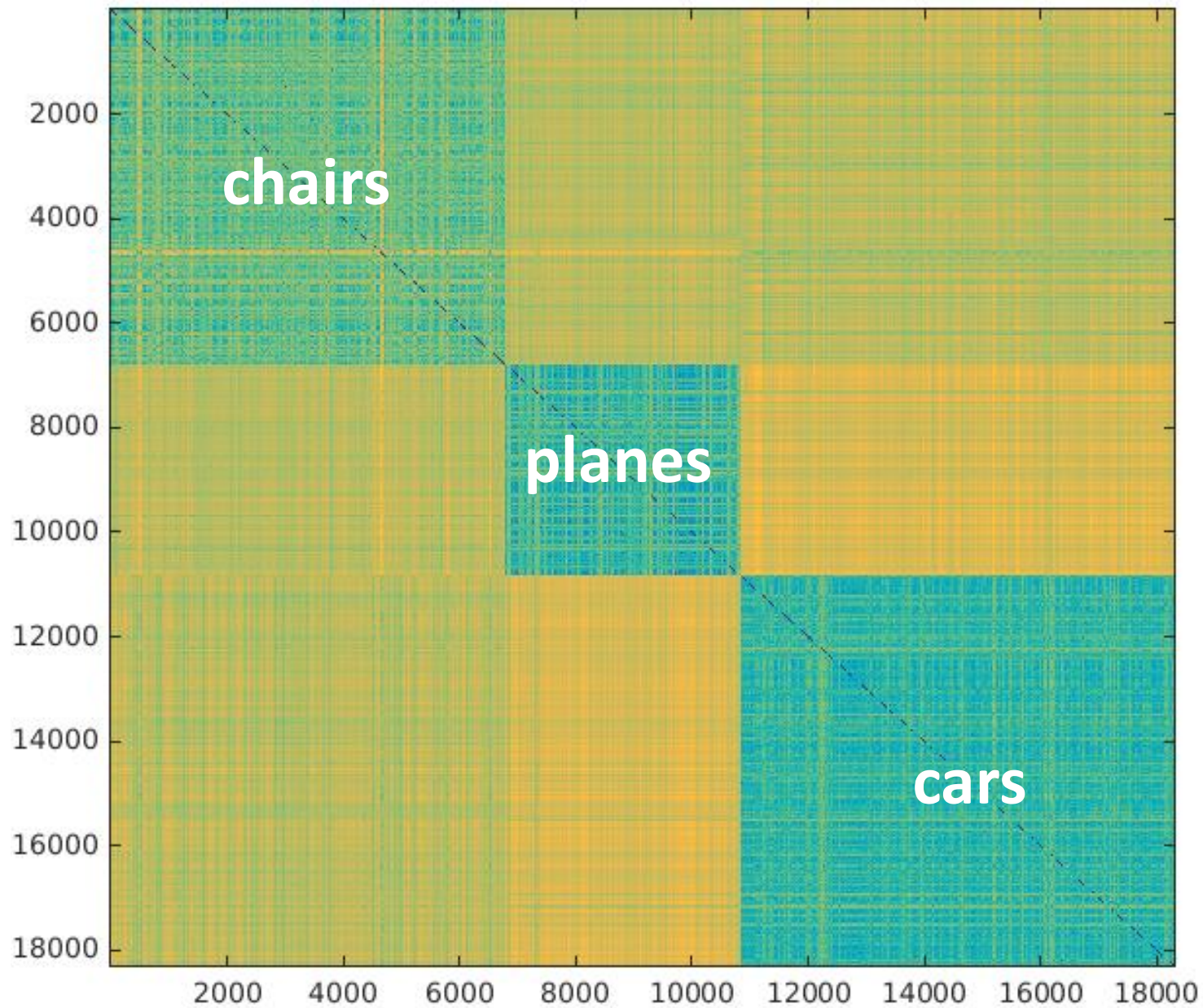
Concatenate





PCA →





Distance Matrix:  $d(S_i, S_j)$  in the  $(i, j)$  –  $th$  element

	$S_1$	$S_2$	$S_3$	·	·	·	·	·	·	·	·	·	·	$S_n$
$S_1$														
$S_2$														
$S_3$														
·														
·														
·														
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·														
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·														
·														
·														
·														
·														
·														
$S_k$														
·														
·														
$S_n$														

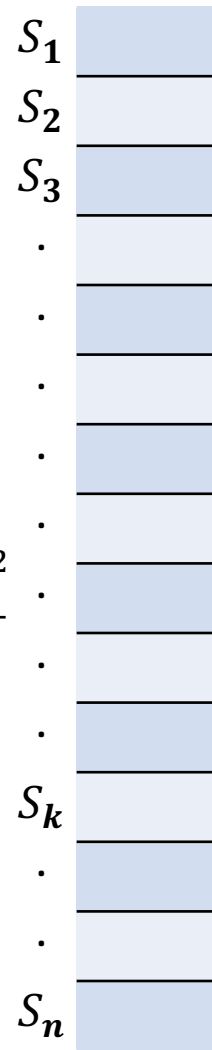
Distance Matrix:  $d(S_i, S_j)$  in the  $(i, j)$  - th element

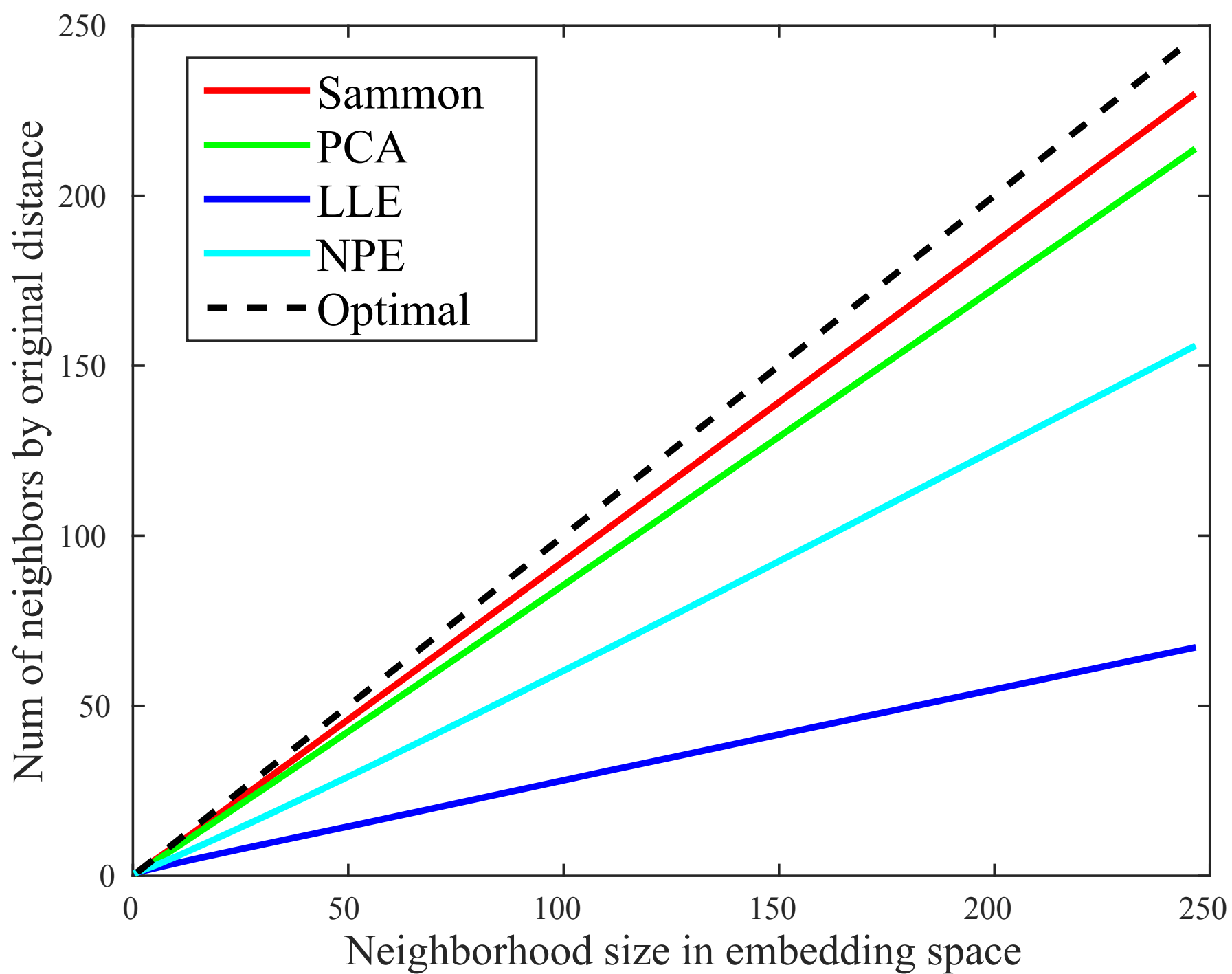
Each row can serve as the embedding point

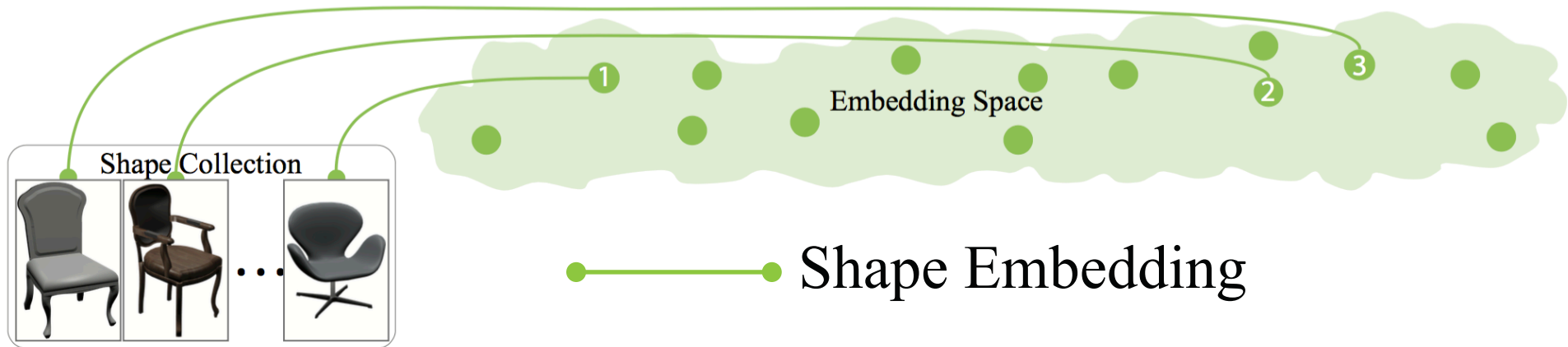


Sammon's Error

$$E = \frac{1}{\sum_{i < j} d_{ij}^*} \sum_{i < j} \frac{(d_{ij}^* - d_{ij})^2}{d_{ij}^*}$$



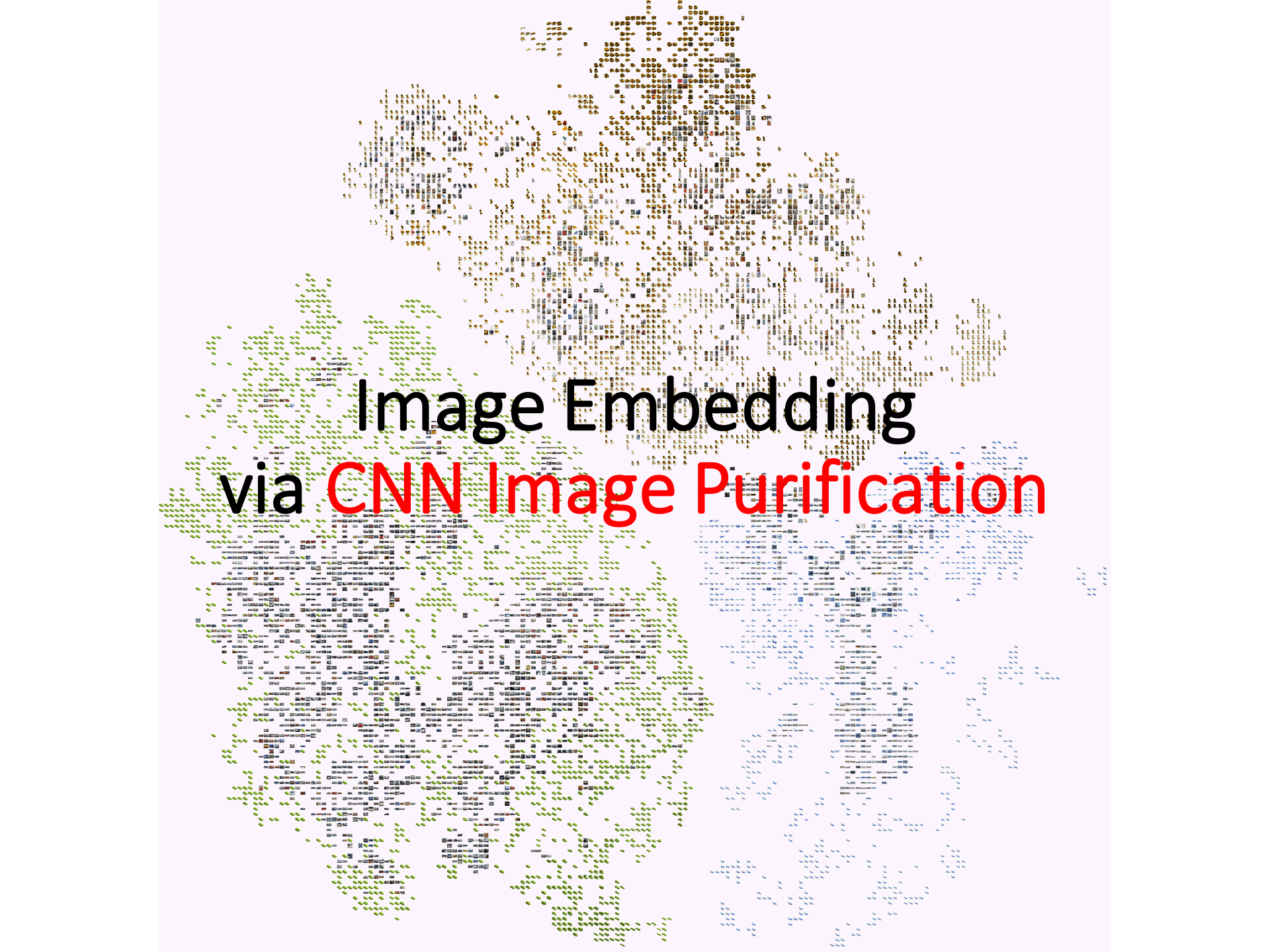




$$\textit{Similarity}_{(s_i, s_j)} = \|\mathcal{P}_i - \mathcal{P}_j\|$$

Our choice of embedding point  $\mathcal{P}_i$ :

1. Extract Light Field HoG Descriptors
2. Compute Distance Matrix
3. MDS with Sammon's Error



# Image Embedding via CNN Image Purification



# Deep learning, yay or nay?



$I_1$



$I_2$



$I_3$

A piece of cake,  
elementary math...

$$\mathcal{P}_i = f(I_i)$$
$$\|\mathcal{P}_2 - \mathcal{P}_3\| < \|\mathcal{P}_1 - \mathcal{P}_2\|$$

What the hell is  
the  $f$ ?



## Chair

A seat for one person, with a support for the back; "he put his coat over the back of the chair and sat down"

1460  
pictures

94.26%  
Popularity  
Percentile



Numbers in brackets: (the number of synsets in the subtree)

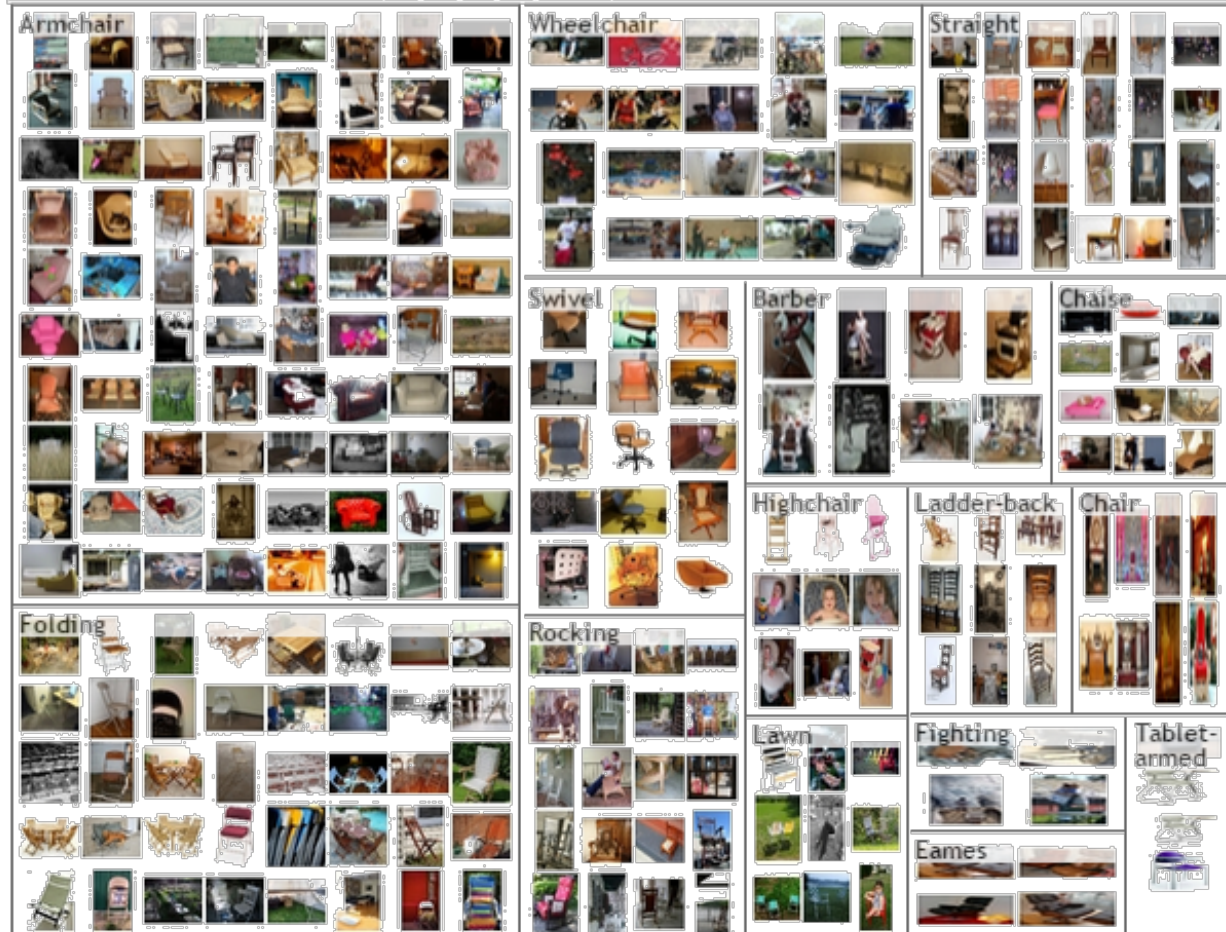
- ImageNet 2011 Fall Release (32326)
  - plant, flora, plant life (4486)
  - geological formation, formation (17)
  - natural object (1112)
  - sport, athletics (176)
  - artifact, artefact (10504)
    - instrumentality, instrumentation
      - device (2760)
      - implement (726)
      - container (744)
      - hardware, ironware (0)
      - equipment (479)
      - ceramic (6)
      - means (0)
      - toiletry, toilet articles (57)
      - conveyance, transport (566)
      - connection, connexion, connr
      - weaponry, arms, implement
      - furnishing (222)
        - furniture, piece of furnitu
          - baby bed, baby's bed
          - bedroom furniture (2)
          - bedstead, bedframe
          - bookcase (0)
          - buffet, counter, sideb
          - cabinet (3)
          - chest of drawers, che
          - dining-room furniture
          - etagere (0)

### Treemap Visualization

### Images of the Synset

### Downloads

ImageNet 2011 Fall Release > A > I > F > Seat > Chair



# S H A P E N E T



SHAPENET

Search  Q Options ▾

About Download Stats Publications

Choose a taxonomy:

ShapeNetCore ▾

- bowl(1,186)
- bus,autobus,coach,charabanc,double-
- cabinet(9,1571)

Synset Models TreeMap Stats

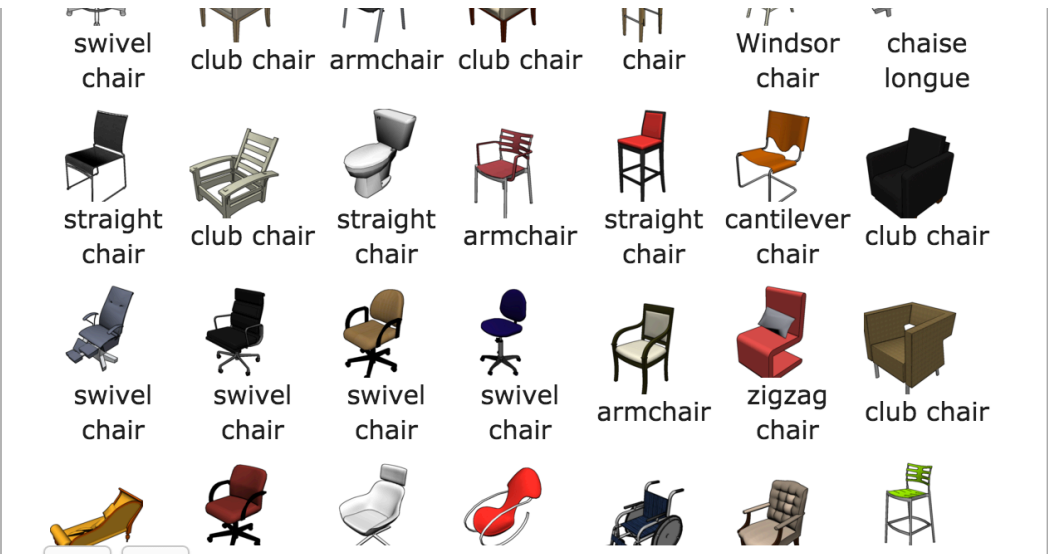
Displaying 1 to 160 of 6827

< 1 2 3 4 5 6 7 8 9 10 ... 43 >

armchair armchair chair Windsor chair swivel chair armchair straight chair

<http://shapenet.org>

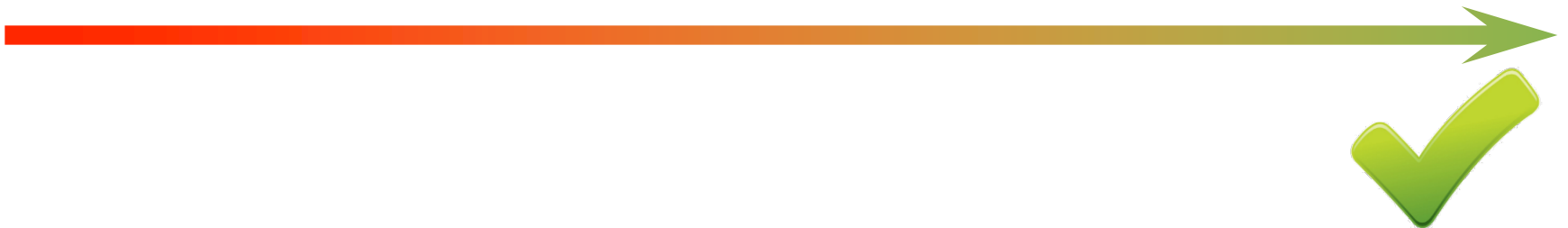
- dishwasher,dish washer,dishwashing m
- display,video display(5,1093)
- earphone,earpiece,headphone,phone(0
- faucet,spigot(2,744)
- file,file cabinet,filing cabinet(1,298)
- guitar(1,797)
- helmet(3,162)
- jar(1,596)
- knife(2,424)
- lamp(2,2318)
- laptop,laptop computer(0,460)
- loudspeaker,speaker,speaker unit,loud:

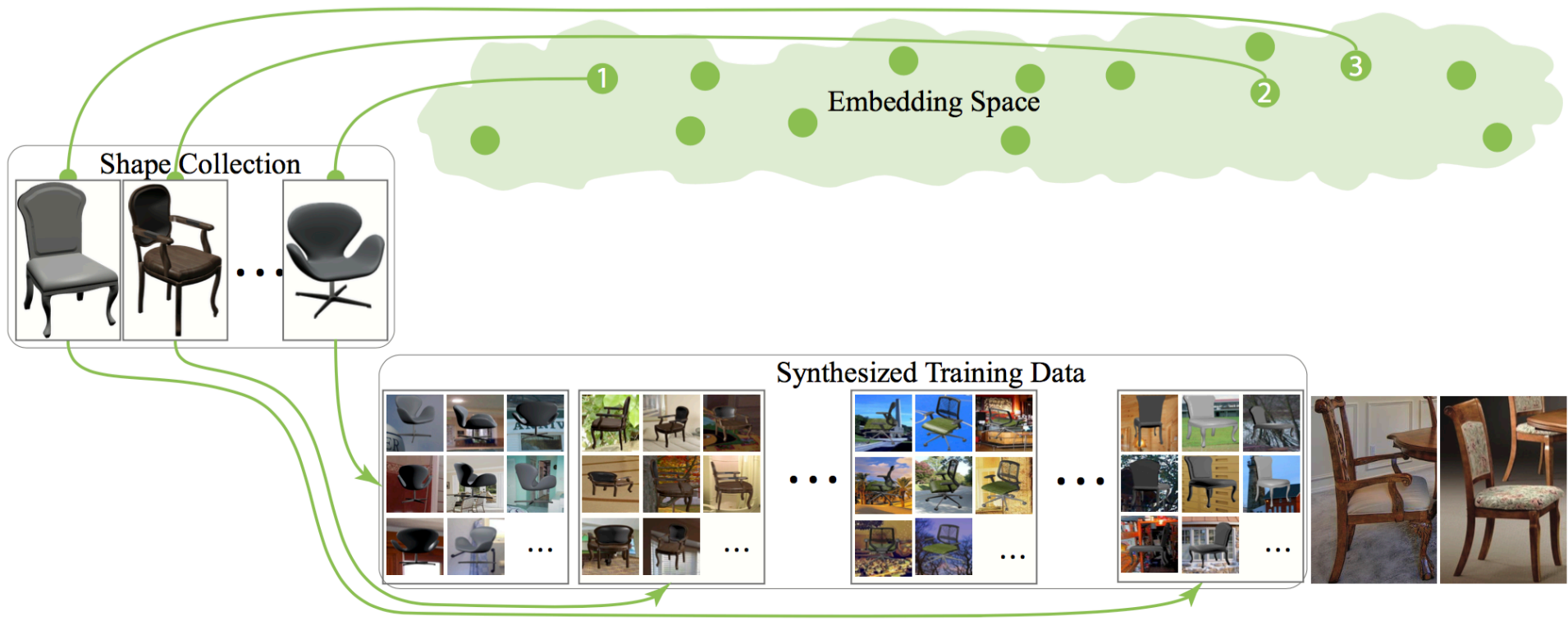


- ✓ A model is worth a thousand images!
- ✓ Rendering:  $Image = f(Properties)$
- ✓ Computer Vision:  $Properties = f^{-1}(Image)$



It eats, a lot!





● —● Shape Embedding      → Image Synthesis

Many image-point pairs  $(I_{S_i}, \mathcal{P}_i)$

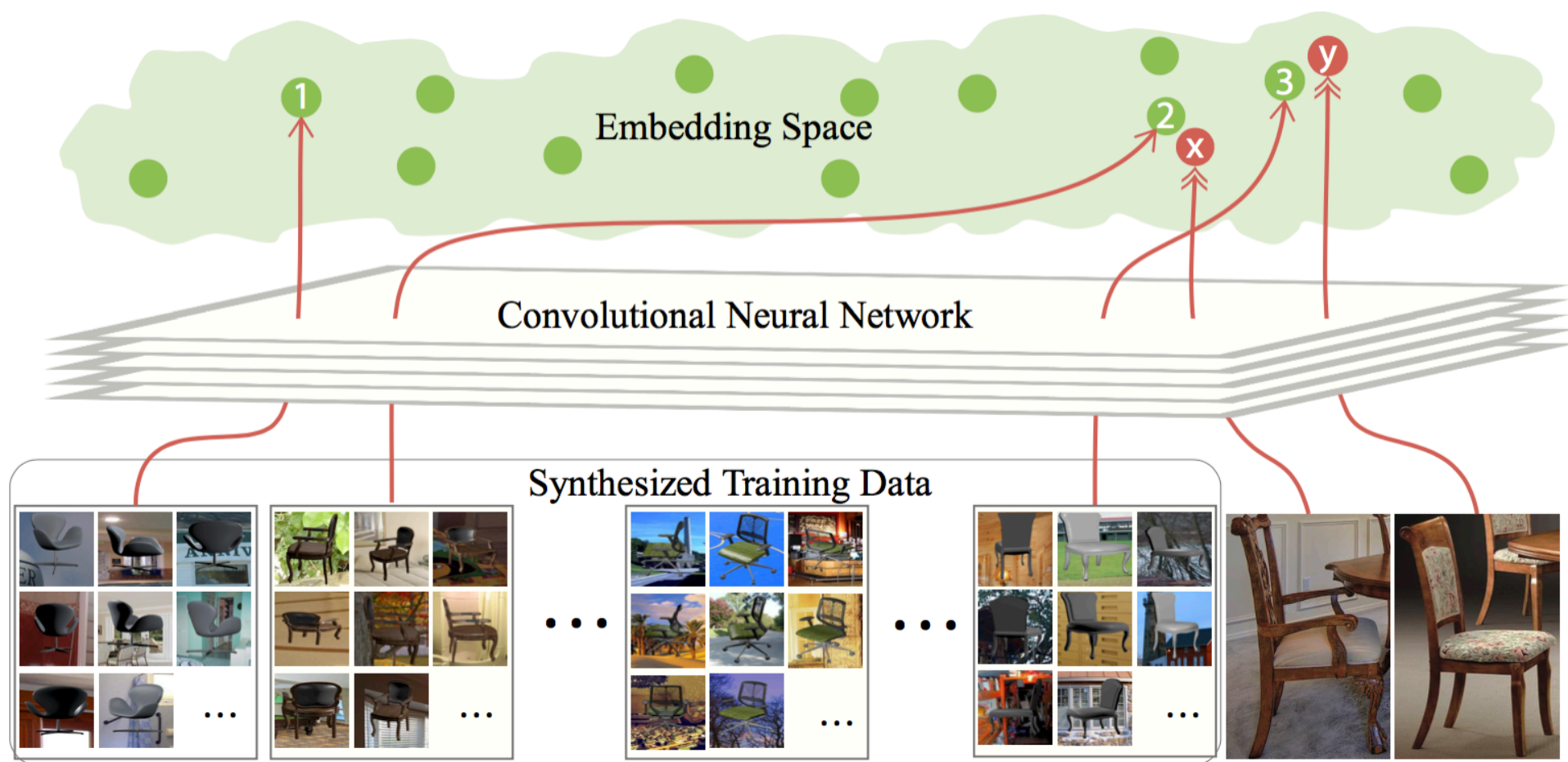


$\neq 10^{14} *$



It's not only the number...





Input: many image-point pairs  $(I_{S_i}, \mathcal{P}_i)$

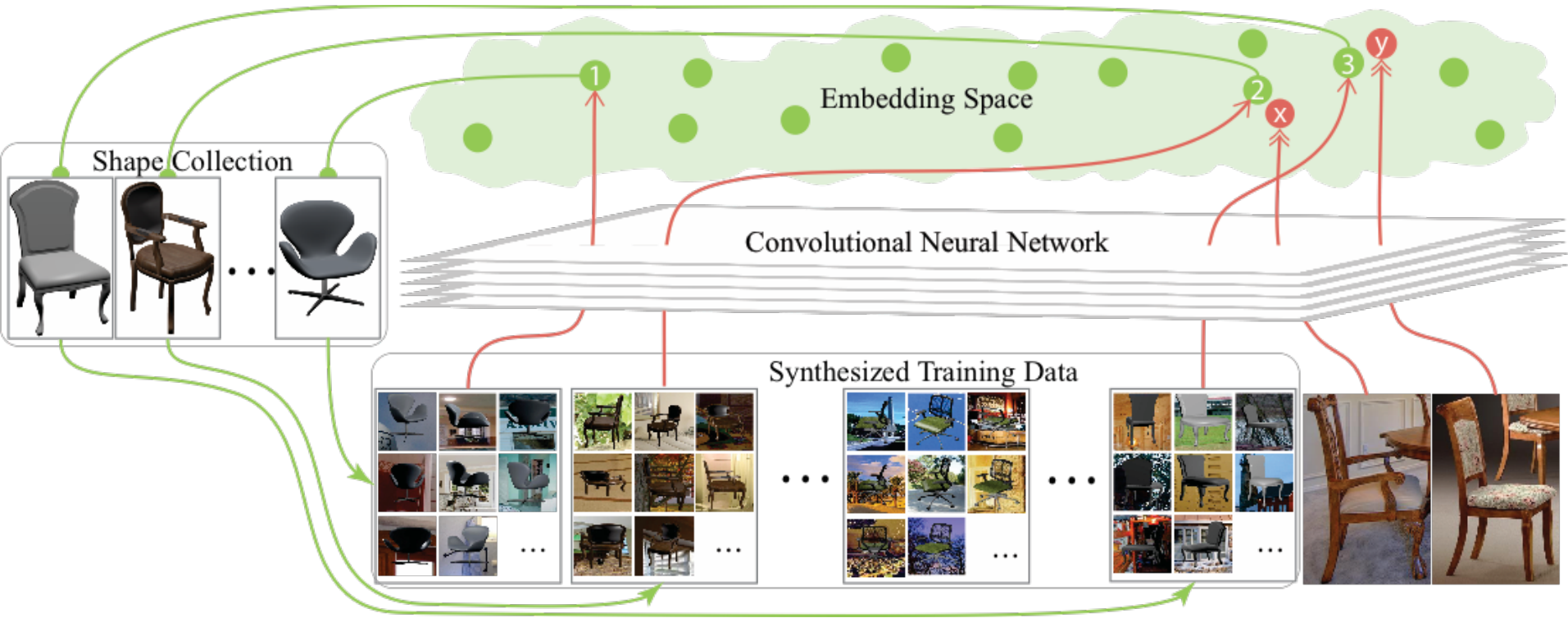
Task: learn the function  $\mathcal{P}_i = f(I_{S_i})$



Hey, wake up!

Here comes the most important slide!

● Shape Embedding → Precious High Quality Supervision



→ Image Synthesis → Messy but Nutritional Training Data

→ Training Phase

→ Testing Phase →  $\mathcal{P}_i = f(I_{S_i})$ , the hell function

# Quantitative Evaluation

	HoG	BoW	LLC	FisherVector	AlexNet fc7 (ImageNet)	AlexNet fc7 (fine tune)	Siamese (64 neighbors)	Siamese (0 neighbor)	Ours
Chair-clutter	0.698	0.681	0.690	0.665	0.706	0.724	0.691	0.701	<b>0.765</b>
Chair-clean	0.710	0.678	0.717	0.675	0.744	0.757	0.724	0.723	<b>0.801</b>
Car	0.278	0.280	0.283	0.270	0.287	0.293	0.285	0.259	<b>0.312</b>

AUC of **image to image retrieval** precision-recall curve

Median rank of	HoG	AlexNet fc7 (ImageNet)	AlexNet fc7 (fine tune)	Siamese (64 nbors)	Siamese (0 nbor)	Ours
first matched	<b>1</b>	7	5	3	3	<b>1</b>
last matched	32	84	71	94	49	<b>5</b>

First and last image match rankings in **shape to image retrieval**

# Quantitative Evaluation

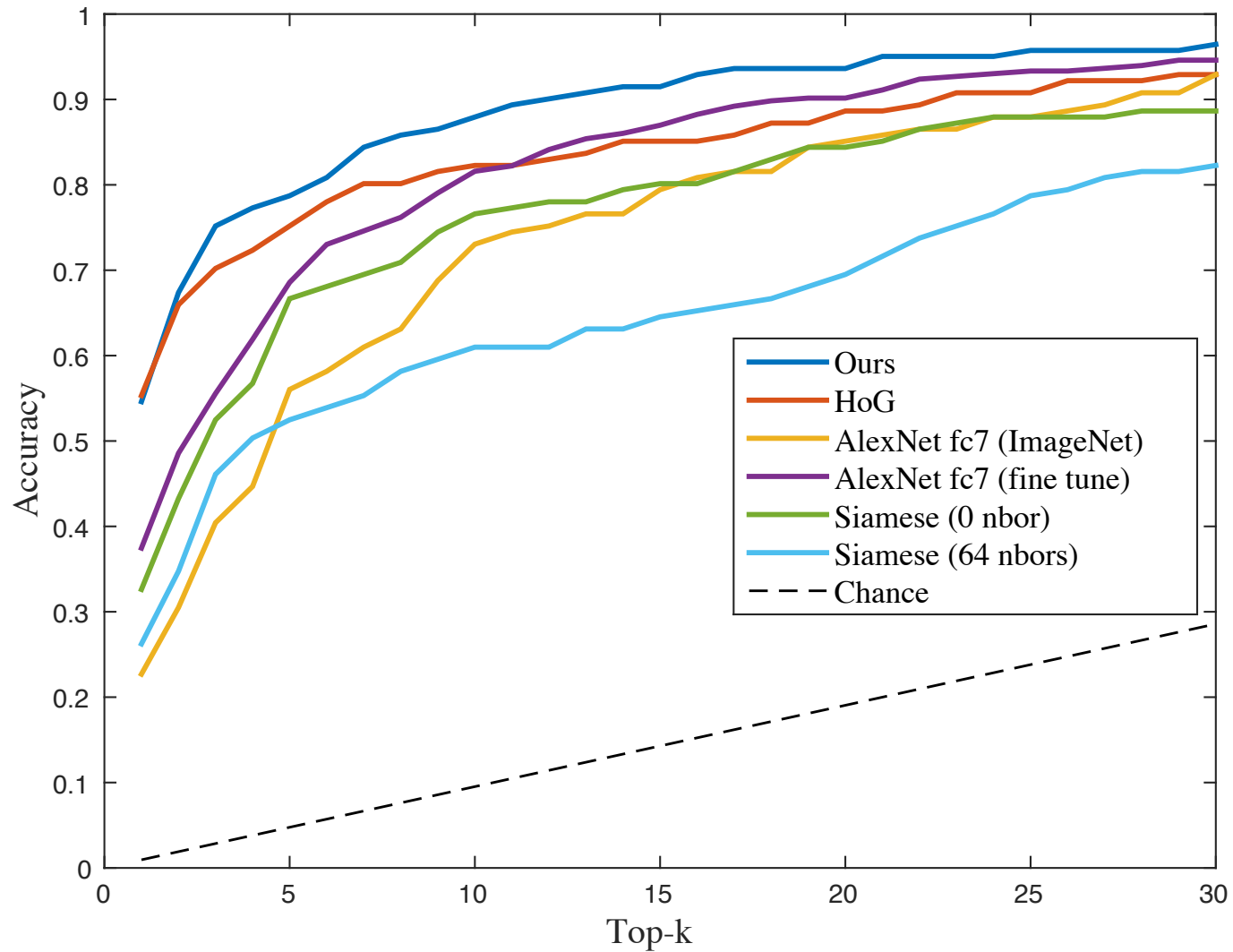
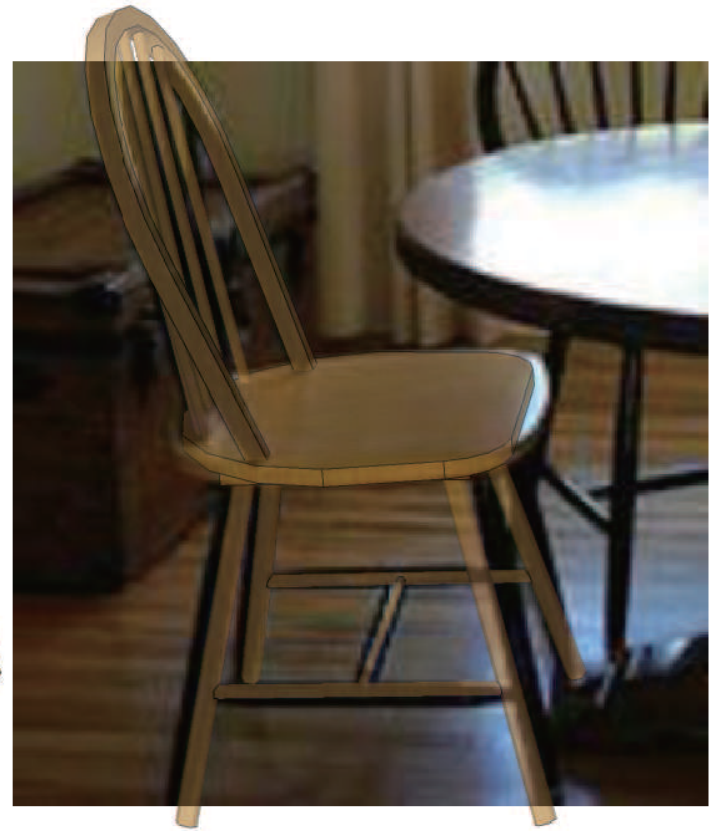


Image to shape retrieval

# Key Steps towards 3D Reconstruction



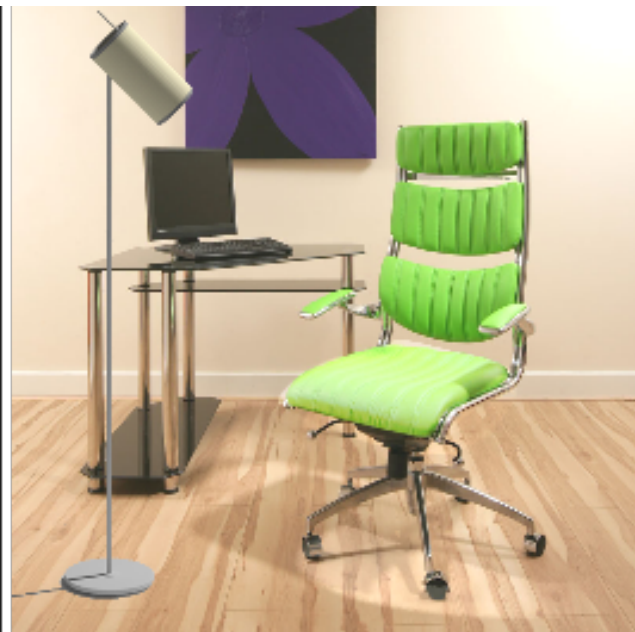
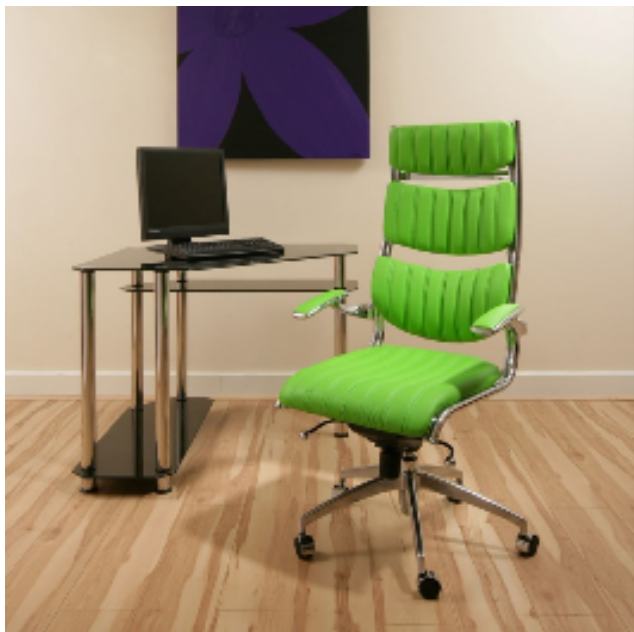
**Similar Shape Retrieval**

**+**

**Viewpoint estimation**

Render for CNN: Viewpoint Estimation in Images Using CNNs  
Trained with Rendered 3D Model Views, ICCV 2015 Oral







Stay Cool with

<http://shapenet.github.io/JointEmbedding/>  
<http://shapenet.github.io/RenderForCNN/>



# Take Home Messages

- Train with synthetic, and act on real
- Asymmetry between synthesis and learning
  - Analogy to encoding/decoding in cryptology
- Promising directions
  - Occlusion patterns
  - Contextual information (depth images)

Thank you!



