



SIGGRAPH2010

The People Behind the Pixels



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SKETCH-BASED 3D SHAPE RETRIEVAL

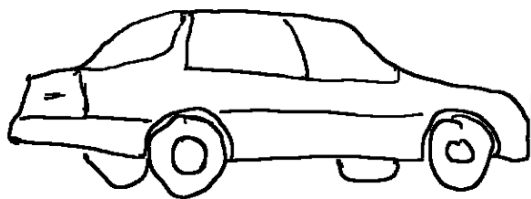
Goal: sketch-based shape retrieval



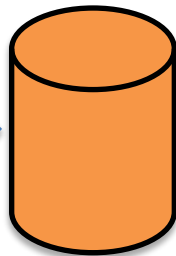
query result

input sketch

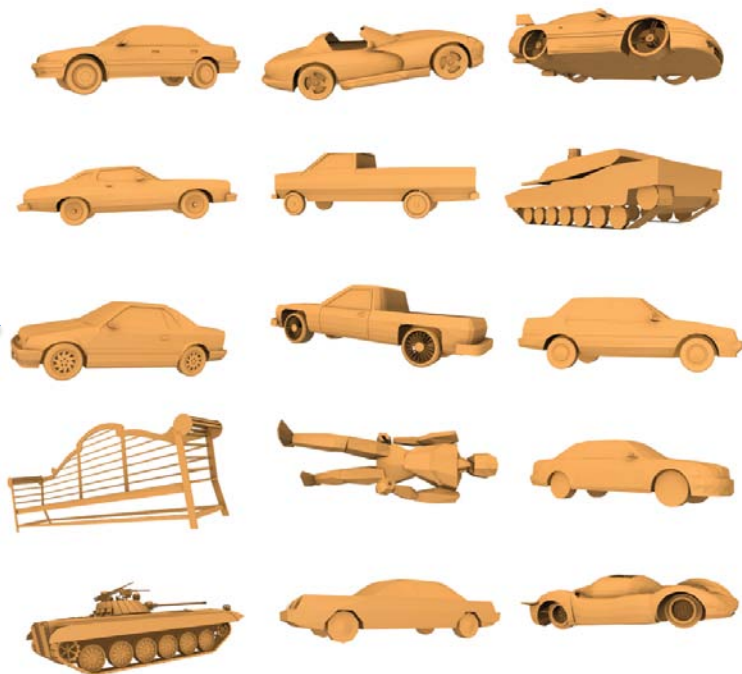
3D model database



query




retrieve



Browser

Comparison View Sketch View Clustering View

pen size: 3 pen eraser



Settings Query Statistics

search type: global bot

descriptor: sift_line_25

query: num results: 50

clustering: num clusters: 5
num iterations: 20
min changes frac: 0.01

vis query:
vis comparison:

Outline



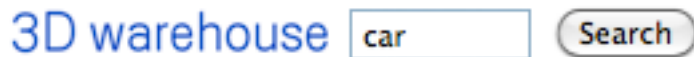
- **Background**
 - **Sketches as input**
- Overview
- Framework
- Results

Why sketches as input?



- 3 common strategies for input

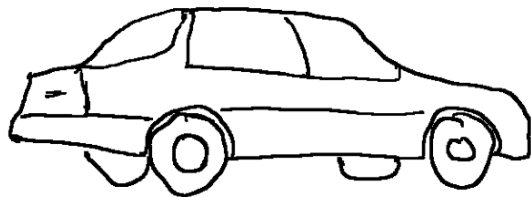
keywords



quick, simple, semantics

no/incorrect tags

sketch (2D)



rather simple, independent of external data

requires drawing skills

model (3D)



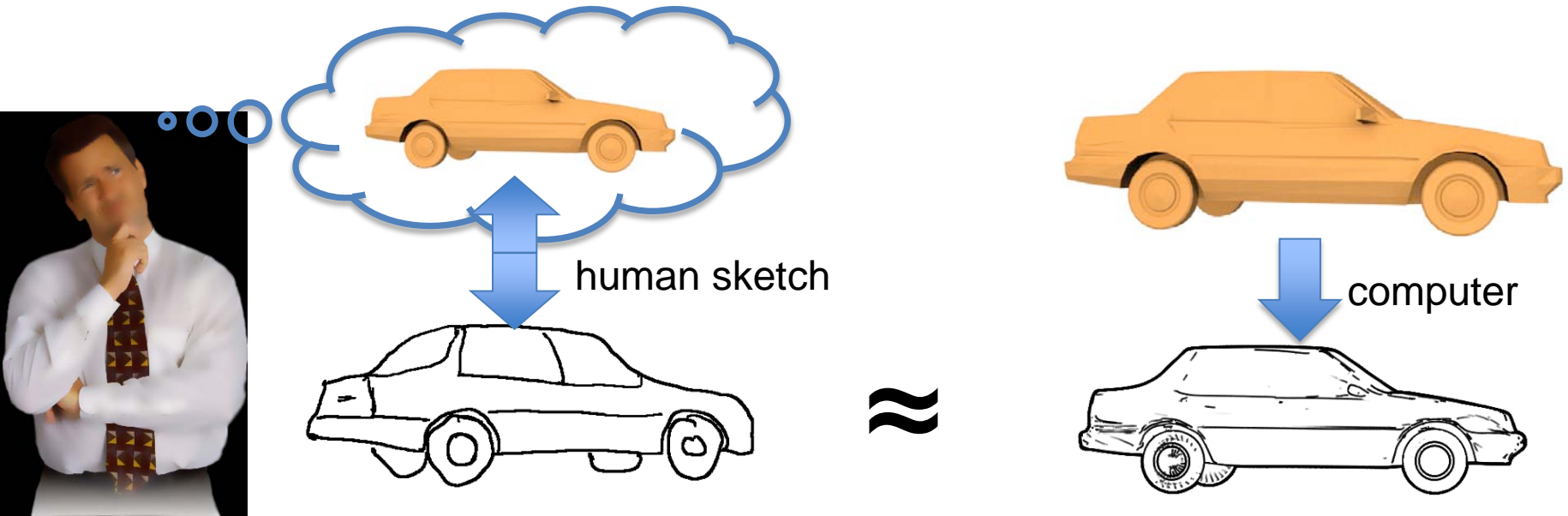
rich input leads to good results

example often not available

Why sketches as input?



- Shape parts index into human memory [Hoffman'97]
- 80-90% of lines explained by known definitions [Cole'08]



Outline

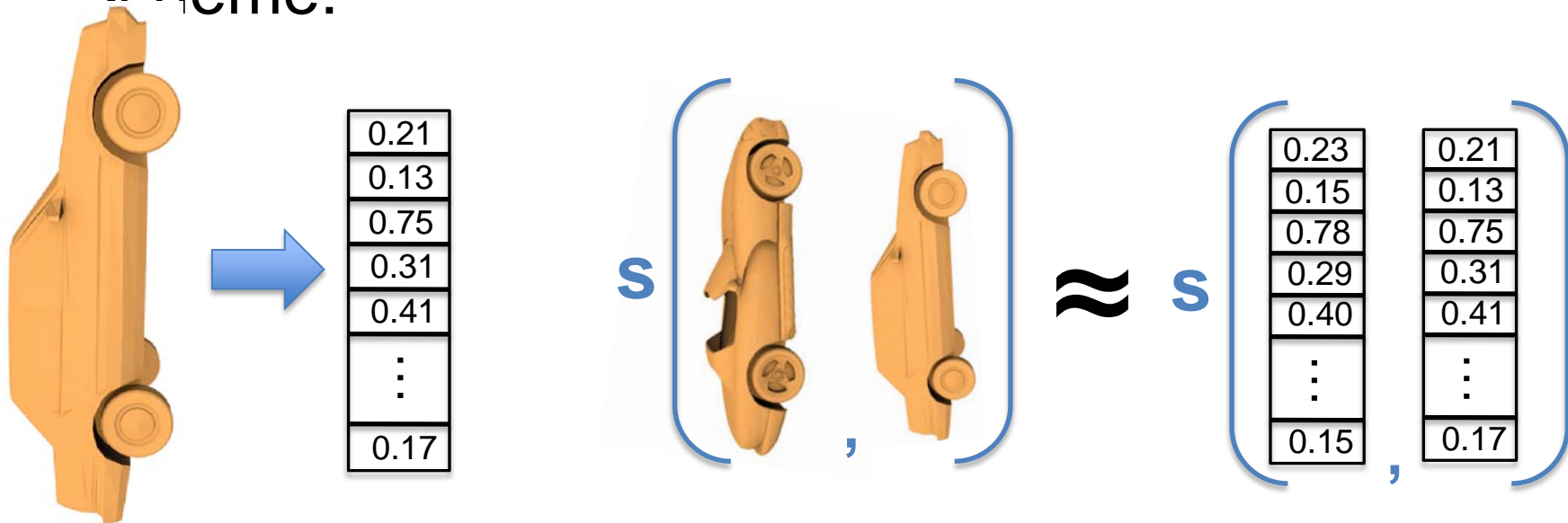


- Background
- **Overview**
 - **Previous work**
 - **Comparison with our approach**
- Framework
- Results



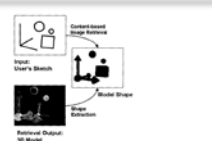





Overview

- Current retrieval systems rely on common scheme:



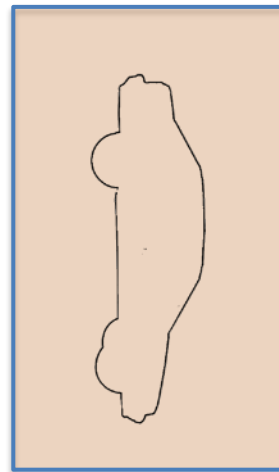
Previous work: global features



	[Löffler, 2000]
	[Funkhouser'03]
	[Pu'05]
	[Hou'07]
	[Shin'07]
	[Napoleon'09]



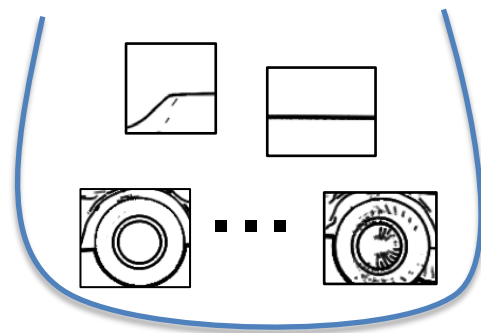
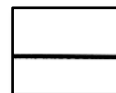
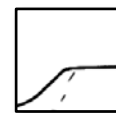
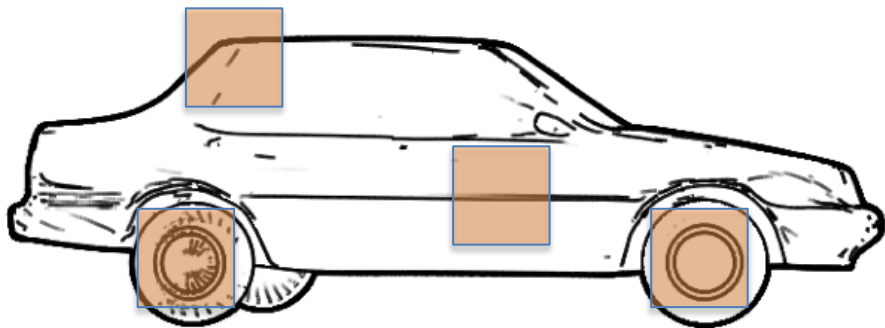
global analysis global descriptor



0.21
0.13
0.75
0.31
0.41
⋮
0.17

Our approach: local features

- Independent local features allow for:
 - translation invariance
 - partial matching
 - standard search data structures



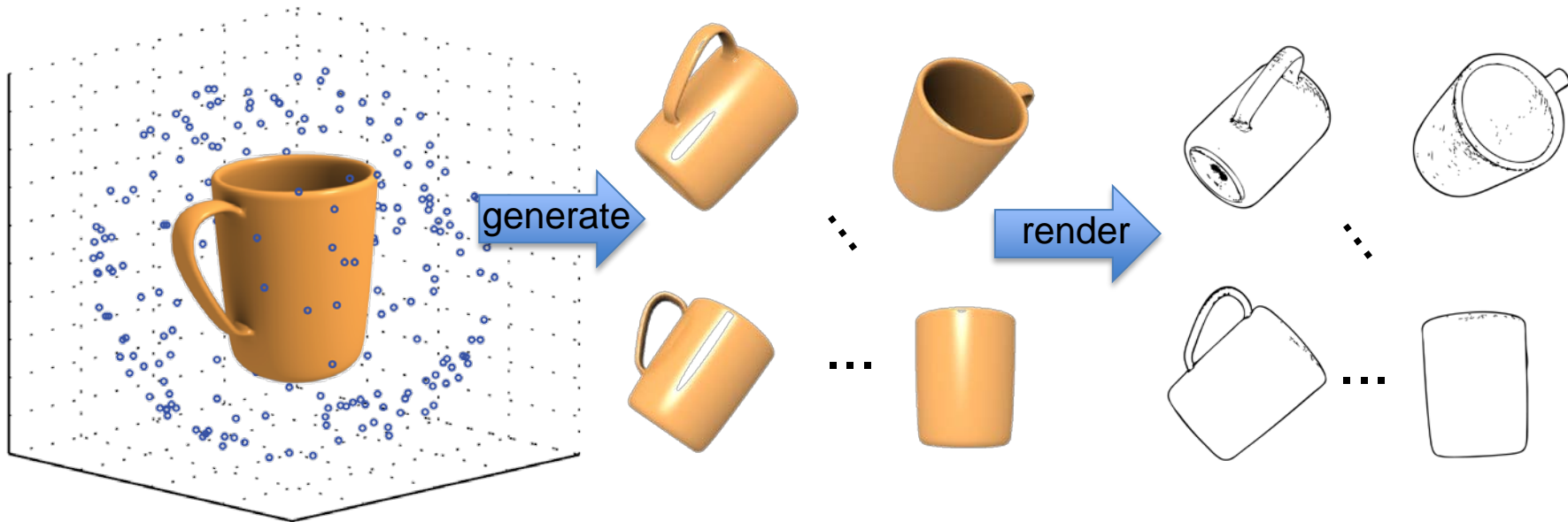
Bag-of-features [Sivic'03]

Outline

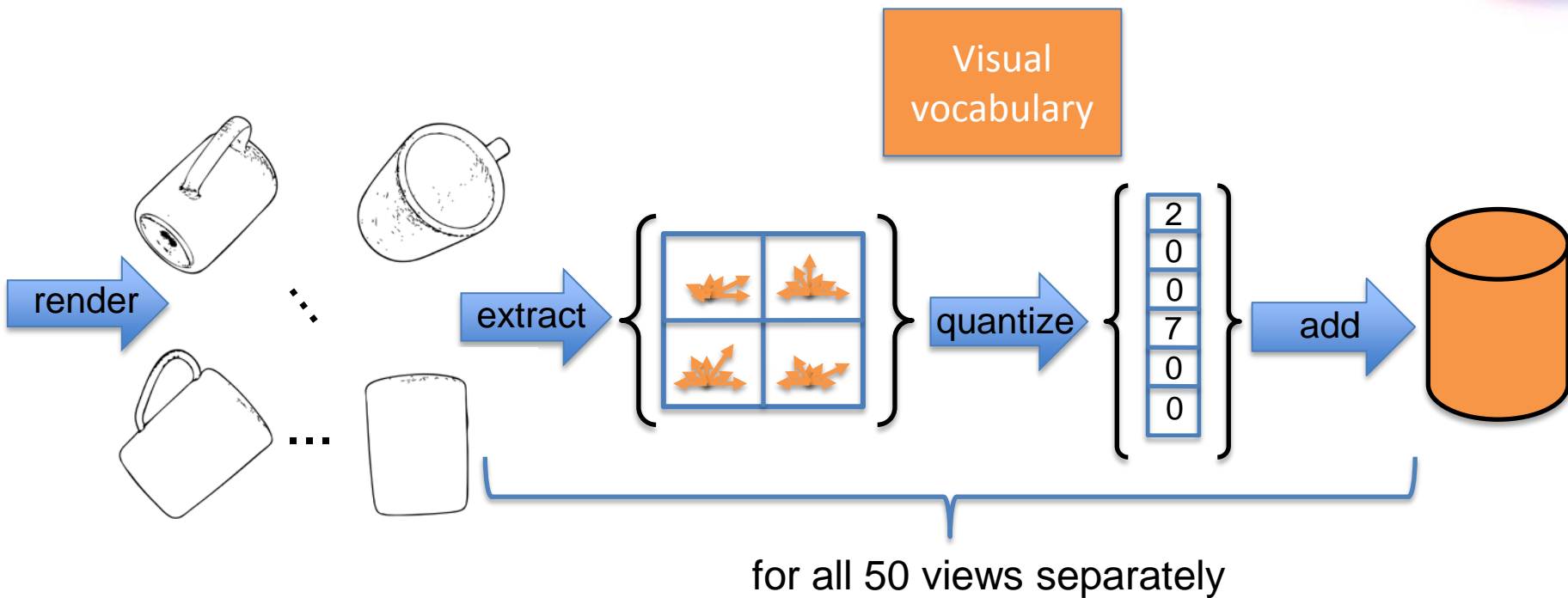


- Background
- Overview
- **Framework**
 - **Offline indexing**
 - **Learning visual vocabulary**
 - **Online search**
- Results

Offline indexing



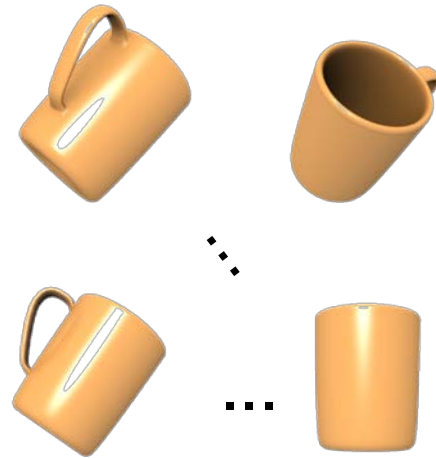
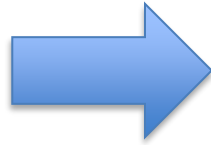
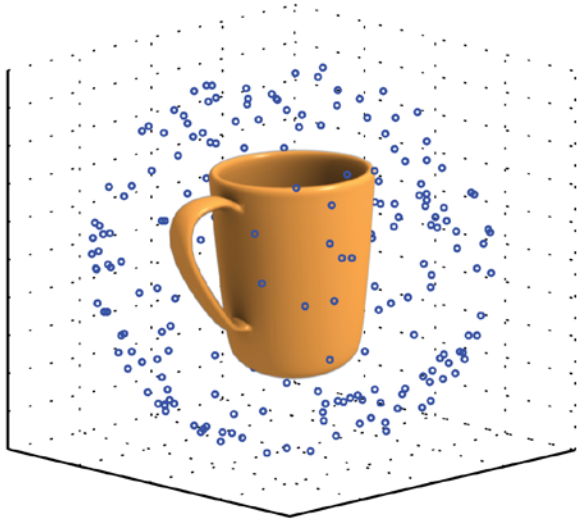
Offline indexing



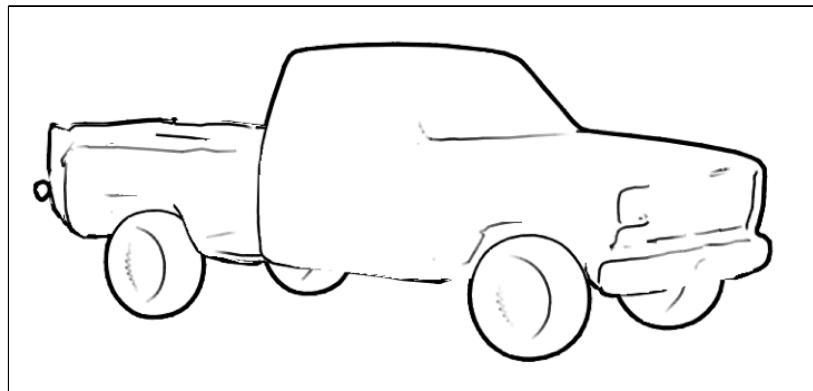
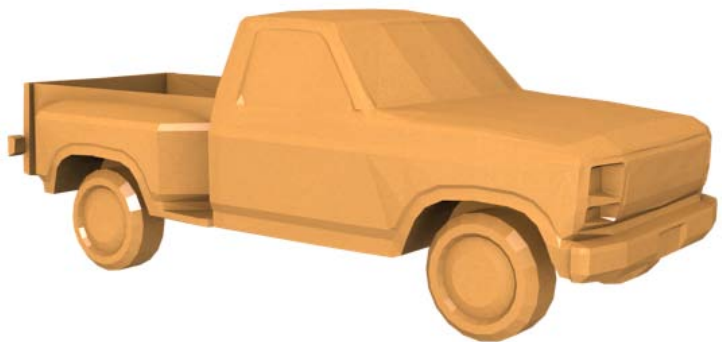
Offline indexing: view generation



- Uniformly sample bounding sphere: 50 samples



Offline indexing: NPR lines

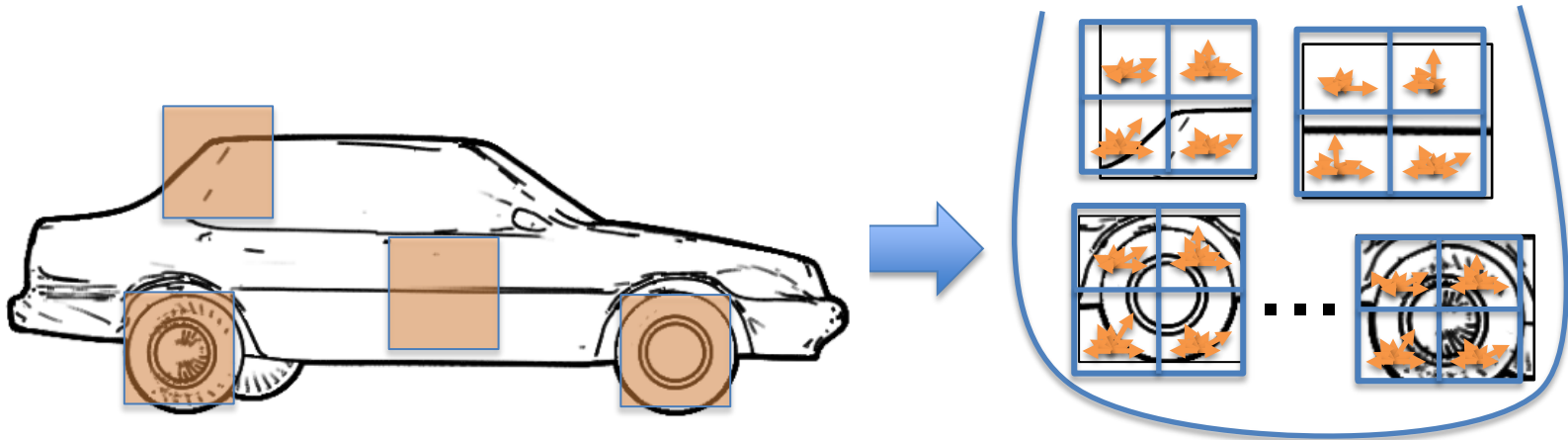


- Occluding contours
- Suggestive contours [DeCarlo'03]

Offline indexing: sampling & features



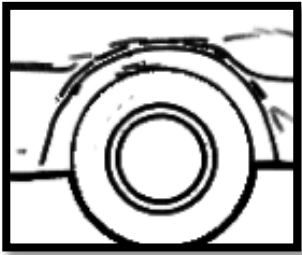
- Sampling: 500 random samples on lines
- Representation: should be concise & robust
 - local image statistics



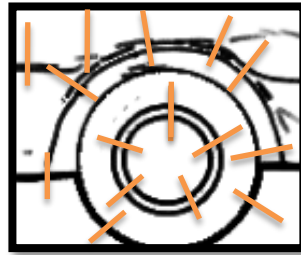
Offline indexing: features



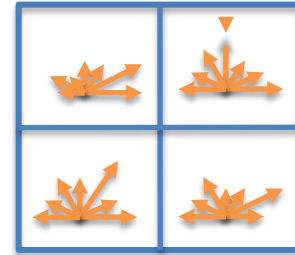
(1) Extract **local region**



(2) estimate **orientations**



(3) **distribution** of orientations



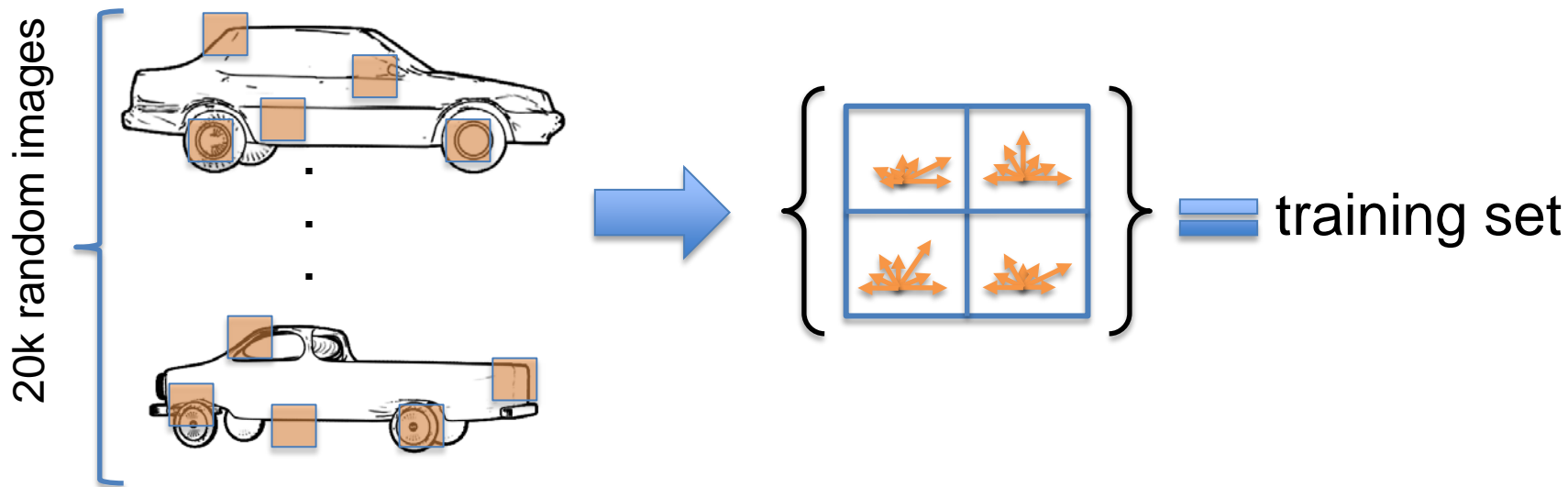
4x4 spatial, 8 radial bins

- No directionality information in gradients
- Binned distribution invariant to small deformations

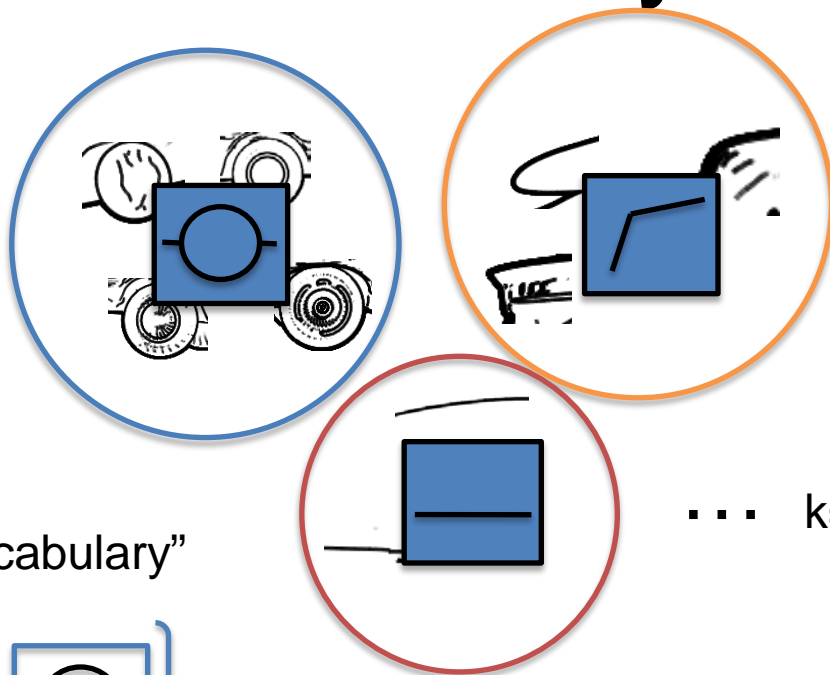
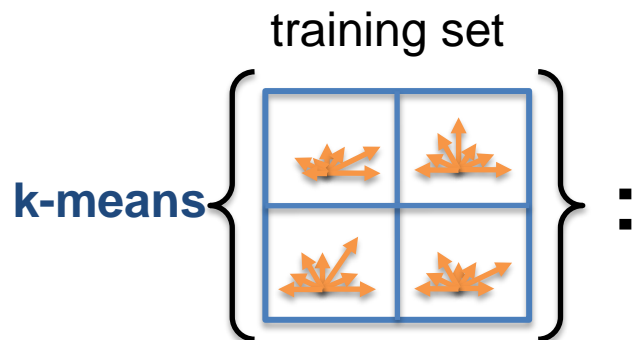
Offline indexing: visual vocabulary



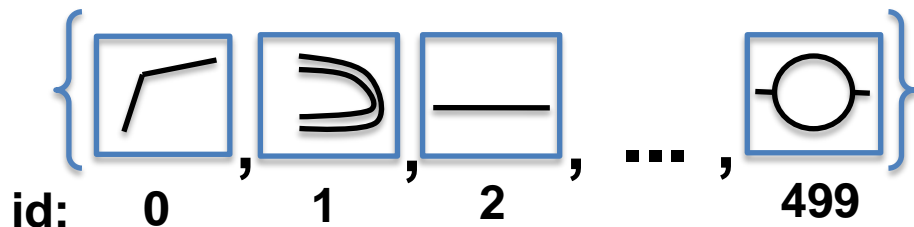
- 20k images (sampled from 50 views each of 2k models)
- 500 local features each
 - Training set size: 10 million local features



Offline indexing: visual vocabulary



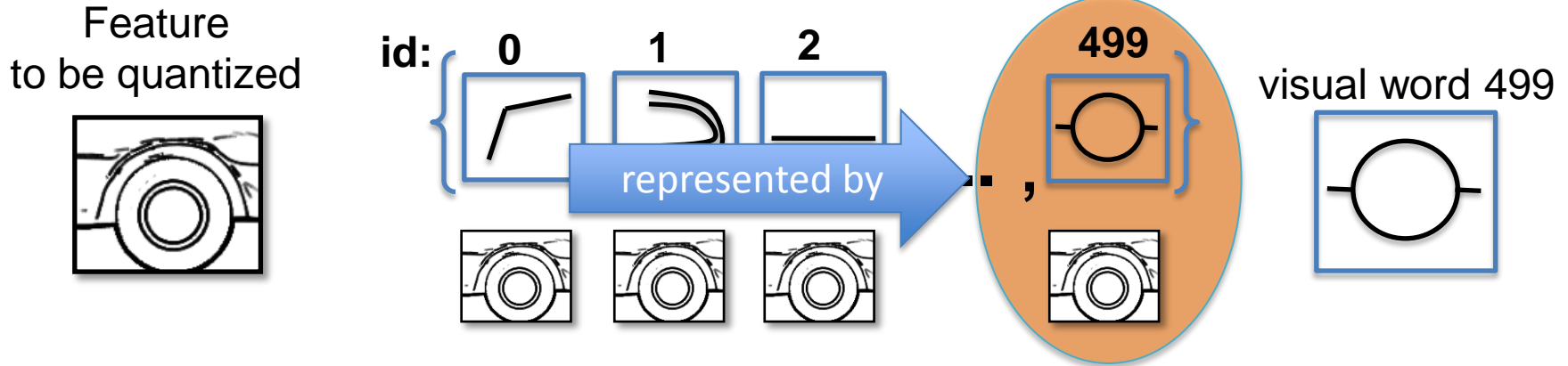
Cluster centers form “visual vocabulary”



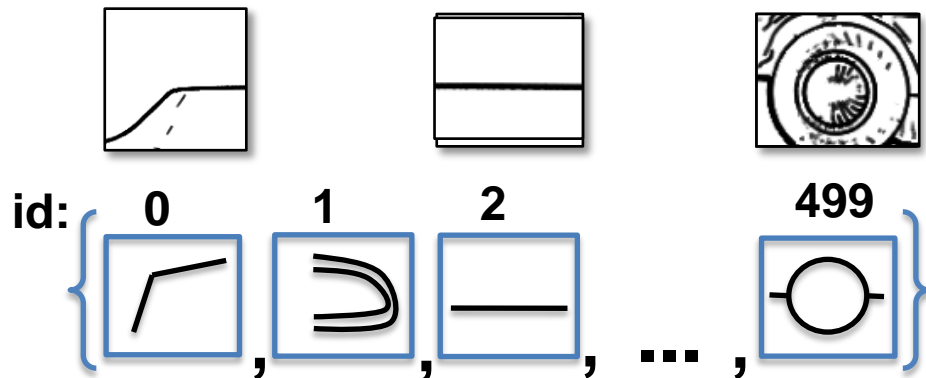
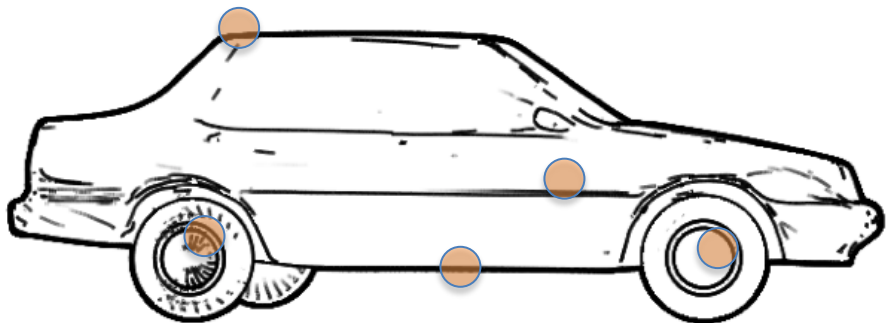
Offline indexing: quantization



- Quantization allows for
 - More compact representation
 - Grouping of perceptually similar features

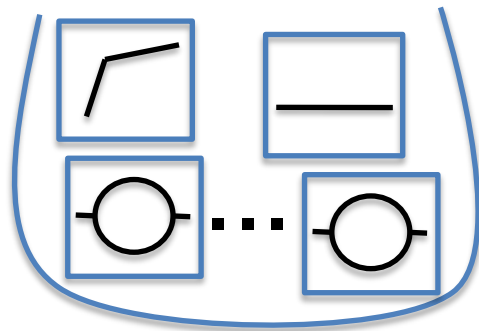
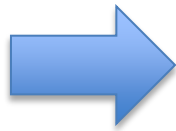
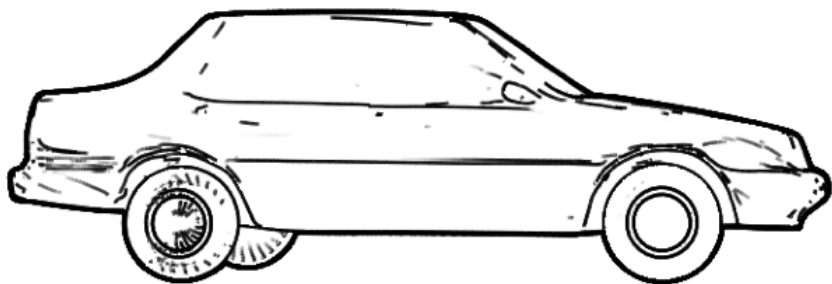


Offline indexing: representation



histogram of "visual words"

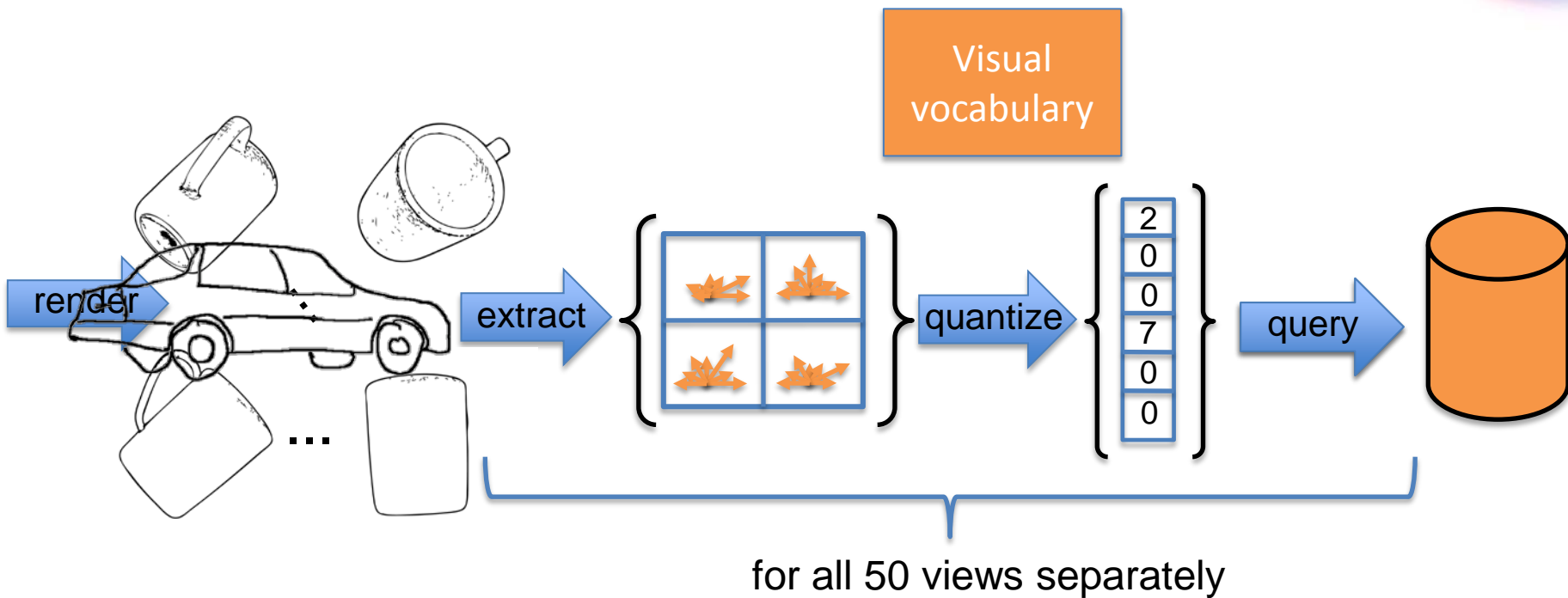
Offline indexing: representation



=

1
0
2
⋮
2

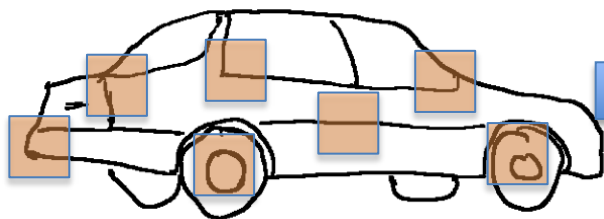
Online search



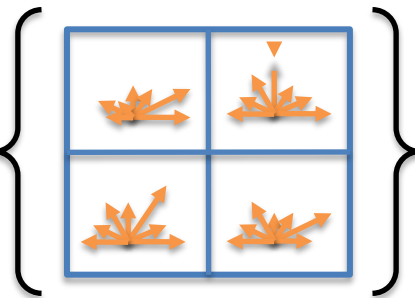
Online Search



sample 500 locations



500 feature vectors



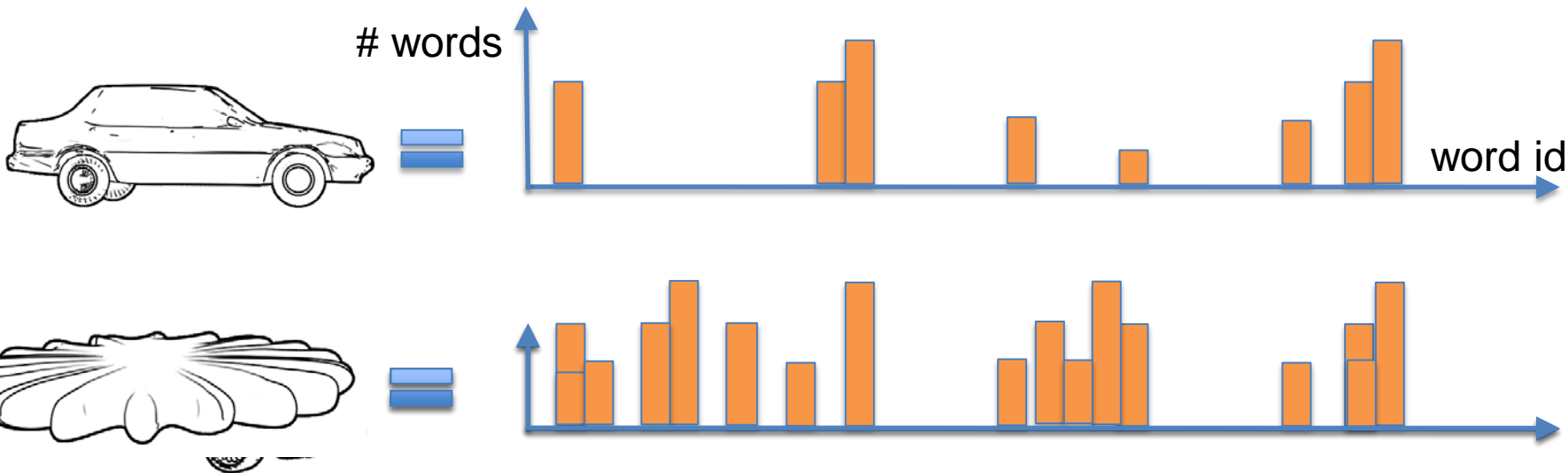
quantize

2
0
0
7
3
0
0
3

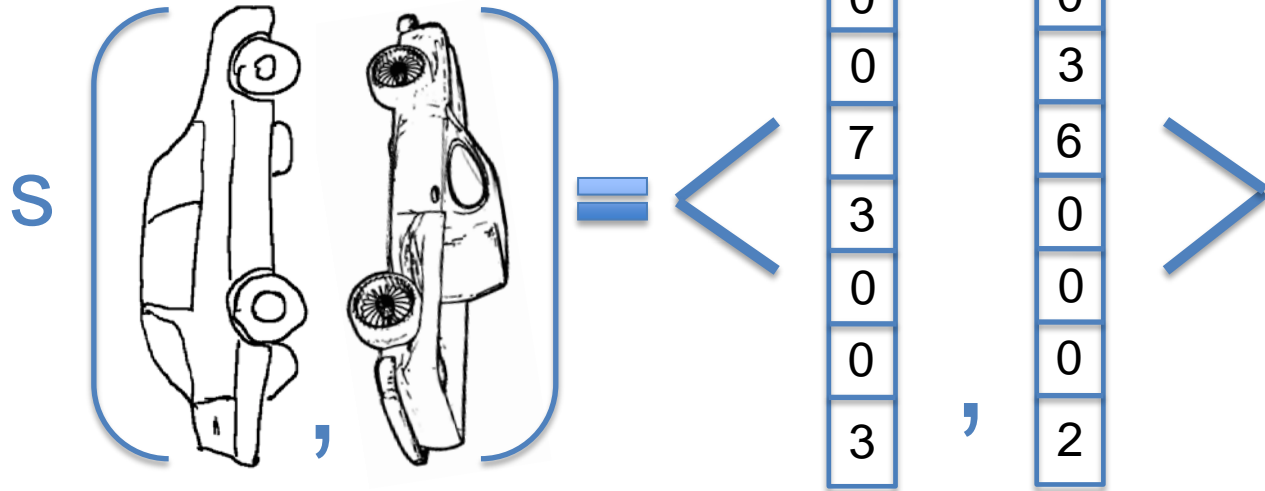
Online search



- Images as (sparse) histograms of visual words



Online search



- Similarity as angle in high-dimensional space
- Vectors sparse: use inverted index

Outline



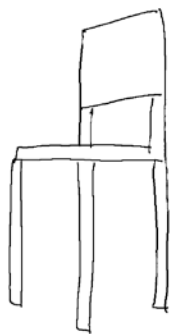
- Background
- Overview
- Framework
- **Results**
 - **Images**
 - **Discussion**

Results

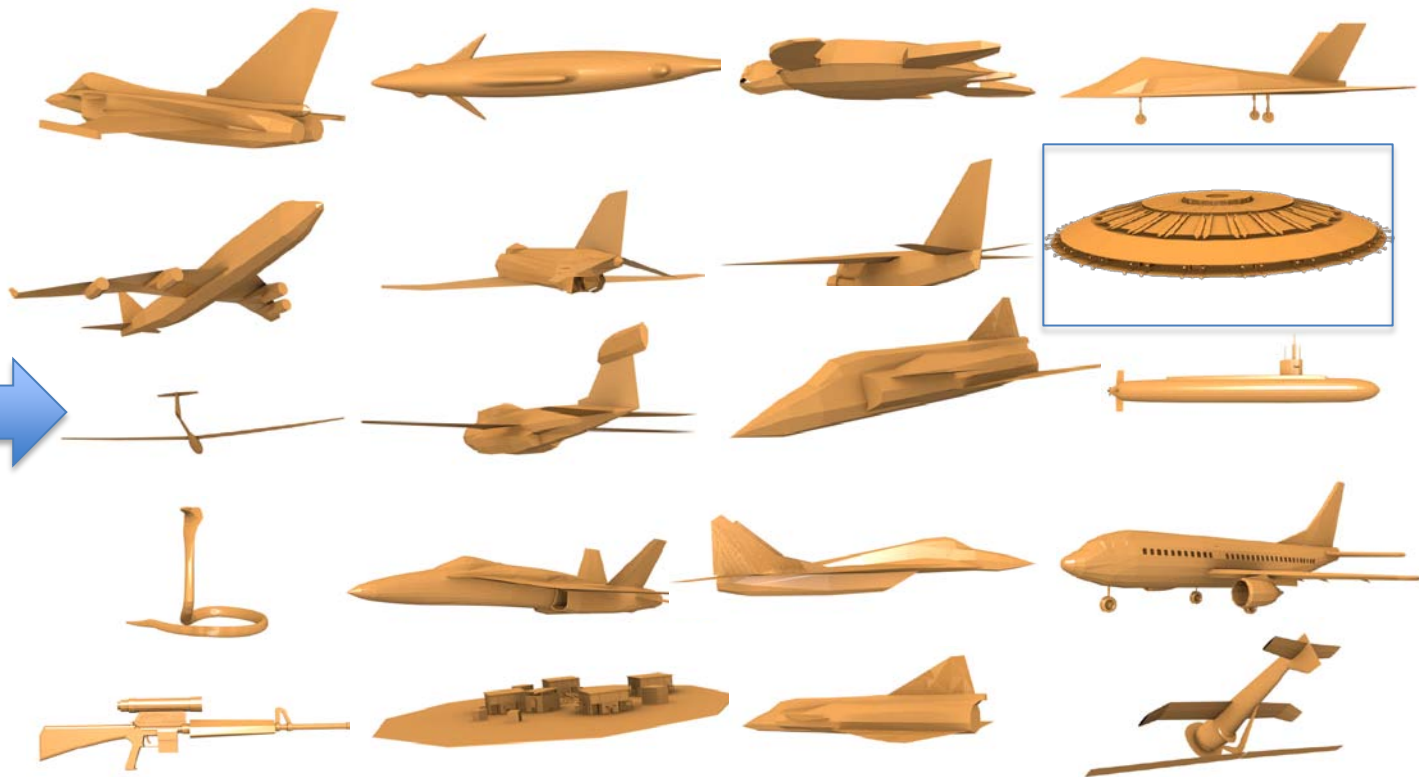


- Based on Princeton shape db (~2k models)
 - ~10ms for a search

Results



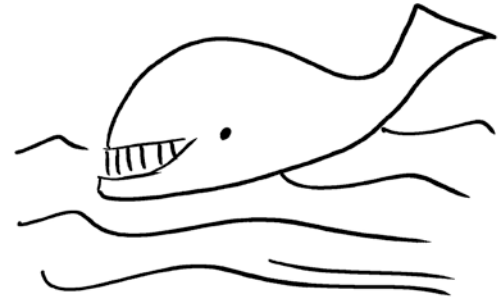
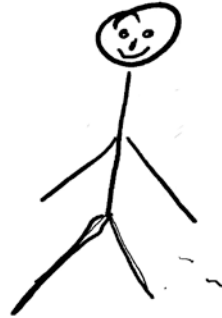
Results



Failure cases



- NPR methods require high resolution meshes
- Sketches from “real users” can be quite abstract



Future work



- View generation
 - canonical, “salient” views – which provides best retrieval?
- Feature representation
 - multi-scale, rotation-invariance?
- Larger datasets than the PSB models
 - Method fast enough to handle millions of models
 - Will it remain effective?

Thanks



More information : <http://www.telecom-paristech.fr/~boubek/papers/SBSR/>

- Acknowledgements

- Princeton shape benchmark [Shilane'04]
- RTSC tool by Doug DeCarlo, Szymon Rusinkiewicz
- Cited authors for images from their papers



