

# Image recognition

Jan van Gemert ([J.C.vanGemert@uva.nl](mailto:J.C.vanGemert@uva.nl))



## <sup>2</sup> Make a list of the types of recognition





### 3 Instance recognition: Dam Monument





# 4 Detection: are there people





5 Verification: is that John?





# 6 Object categories



**Building**

**Lamp**

**tree**

**People**



# 7 Scene and Context

Cityscape  
Outdoor

...





# 8 What distinguishes types? Challenges?





# Challenges: robustness



**Illumination**



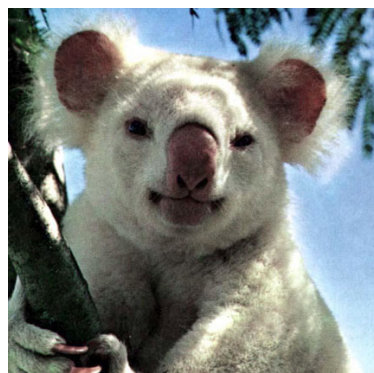
**Object pose**



**Clutter**



**Occlusions**



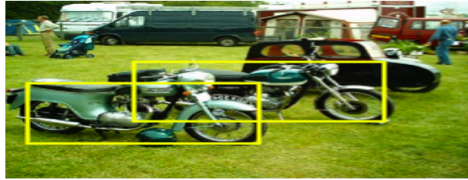
**Intra-class  
appearance**



**Viewpoint**



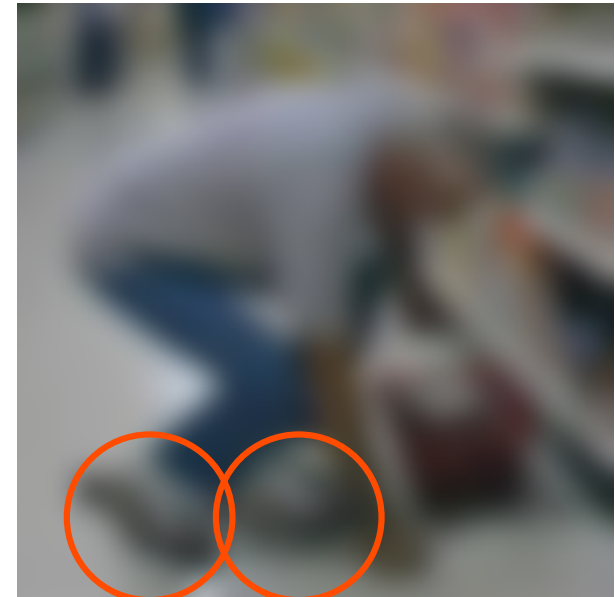
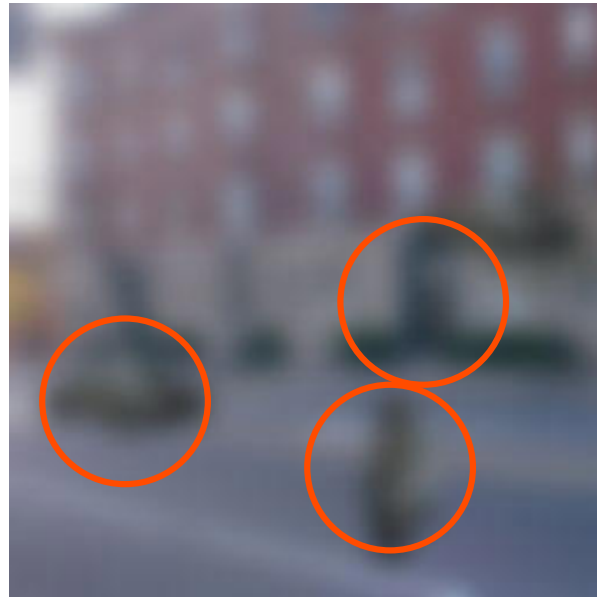
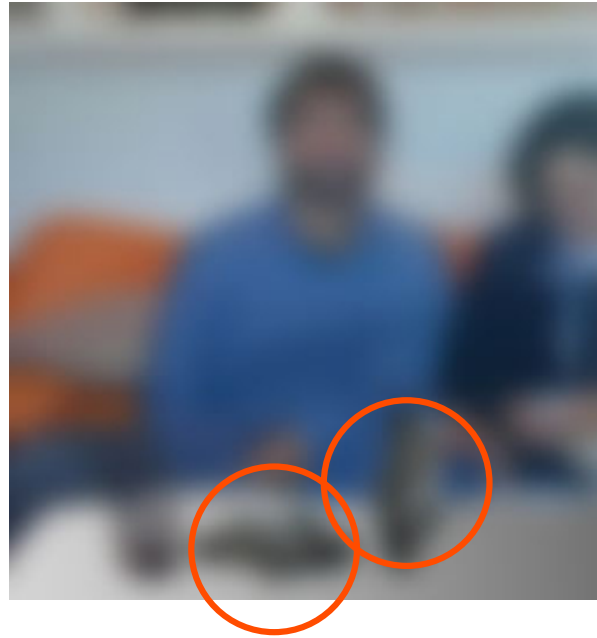
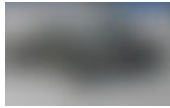
# Challenges: robustness



Realistic scenes are crowded, cluttered, have overlapping objects.



# Challenges: importance of context





# Bag-of-features models (Visual words)





# Bag-of-words model in text

- Orderless document representation: frequencies of words from a dictionary Salton & McGill (1983)



# Bag-of-words model in text

- Orderless document representation: frequencies of words from a dictionary Salton & McGill (1983)





# Bag-of-words model in text

- Orderless document representation: frequencies of words from a dictionary Salton & McGill (1983)





# Bag-of-words model in text

- Orderless document representation: frequencies of words from a dictionary Salton & McGill (1983)

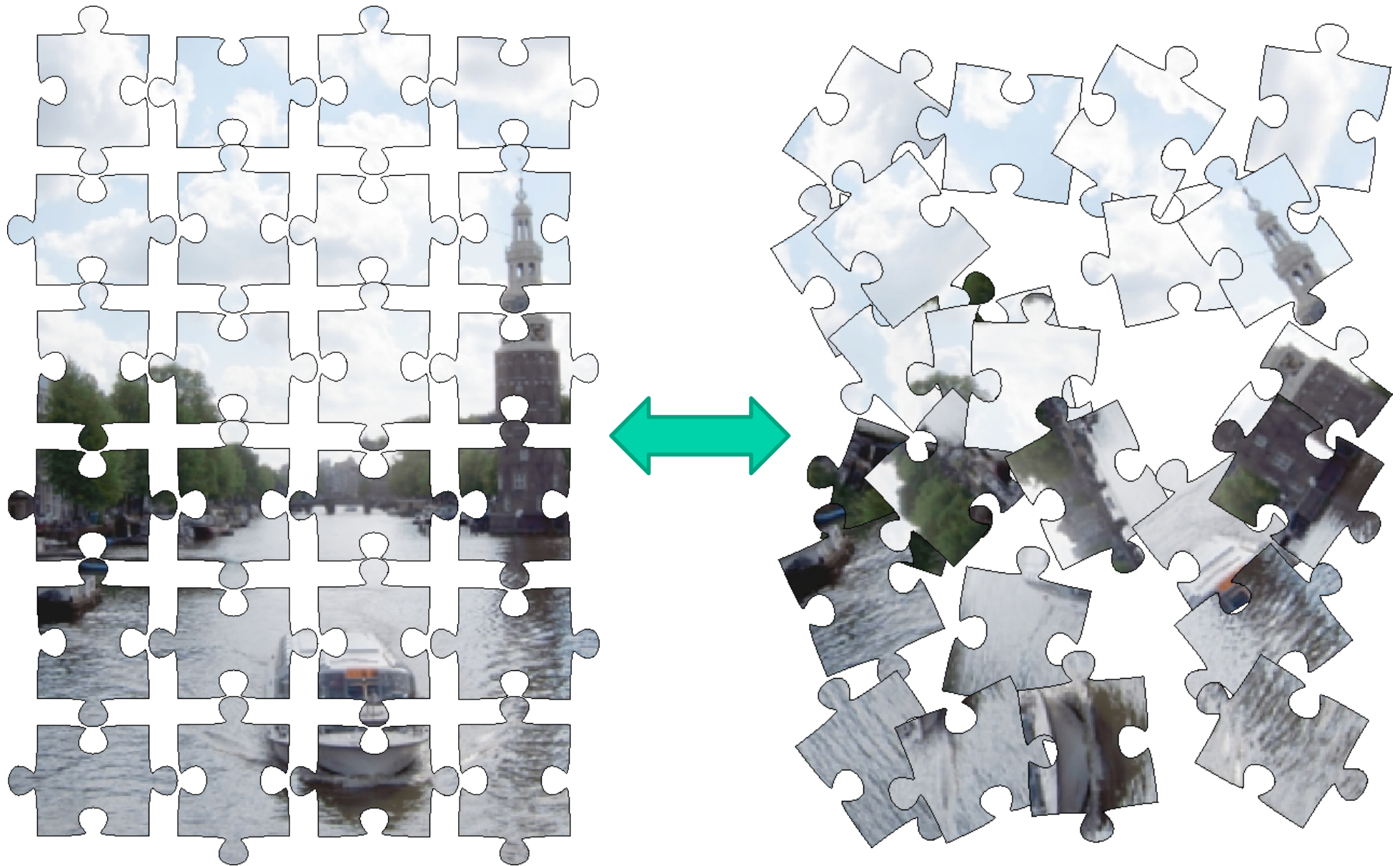


How would you apply this to images?

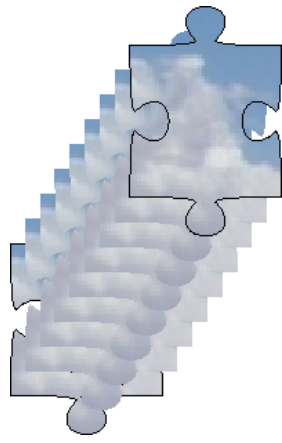
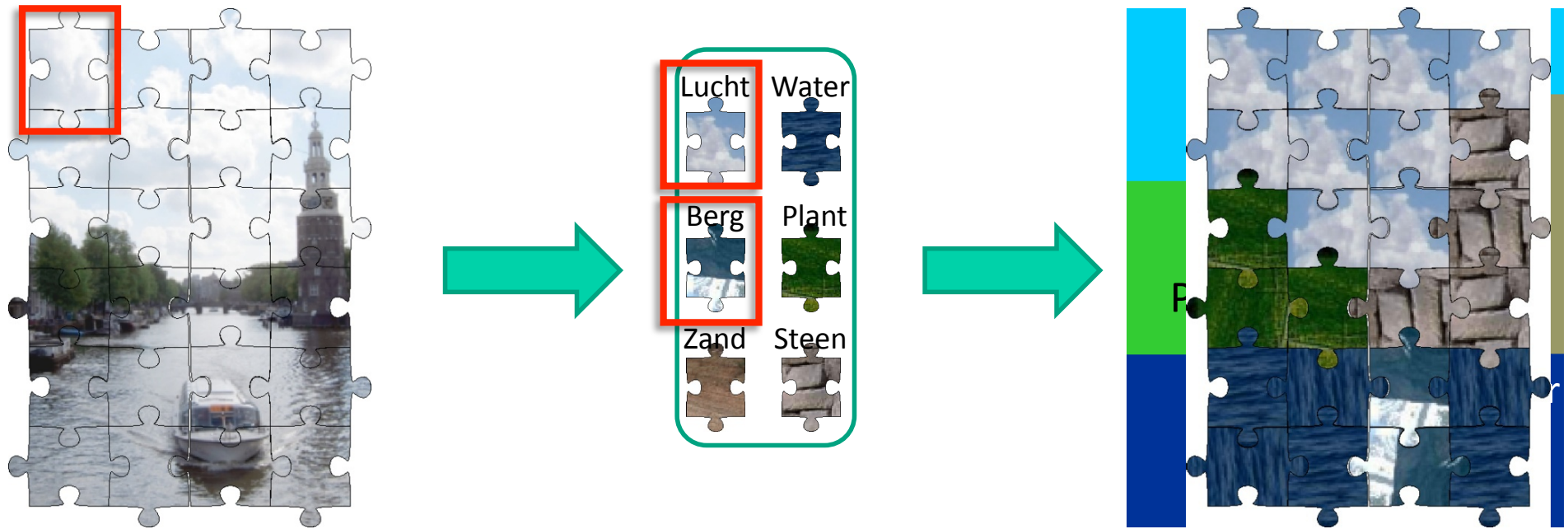




# Lokale Beeldkenmerken



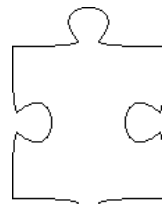
# Visueel woordenboek



Lucht  
9



Berg  
2



Zand  
0



Water  
6



Plant  
3

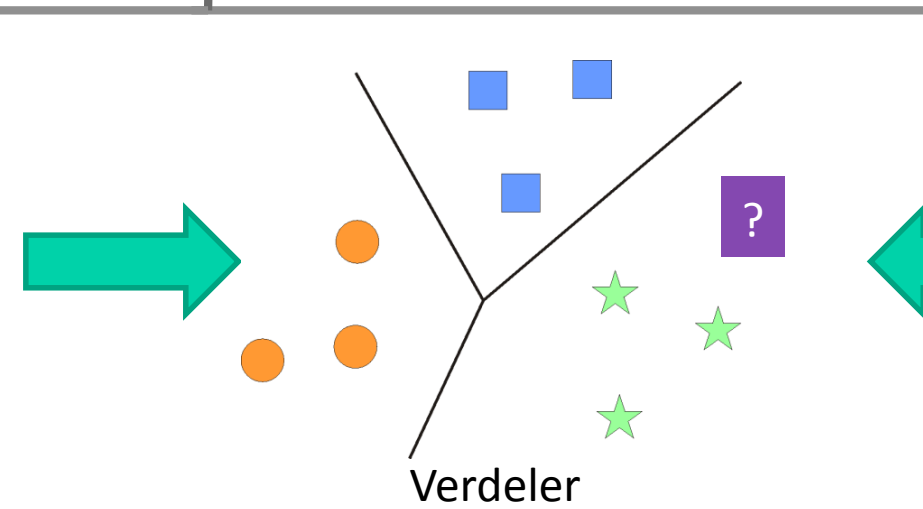
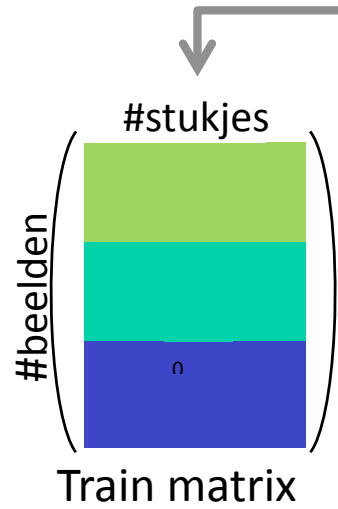
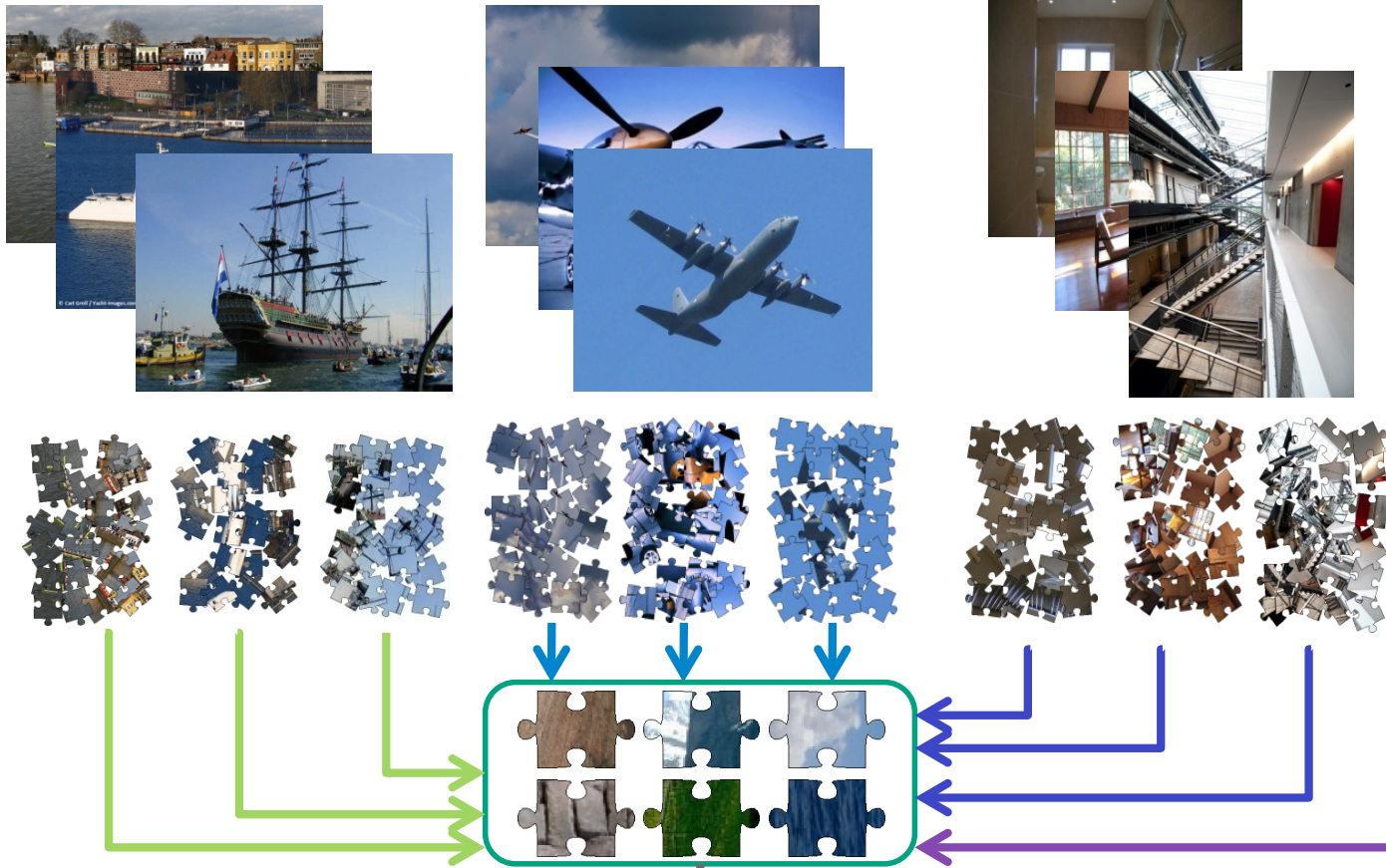


Steen  
4

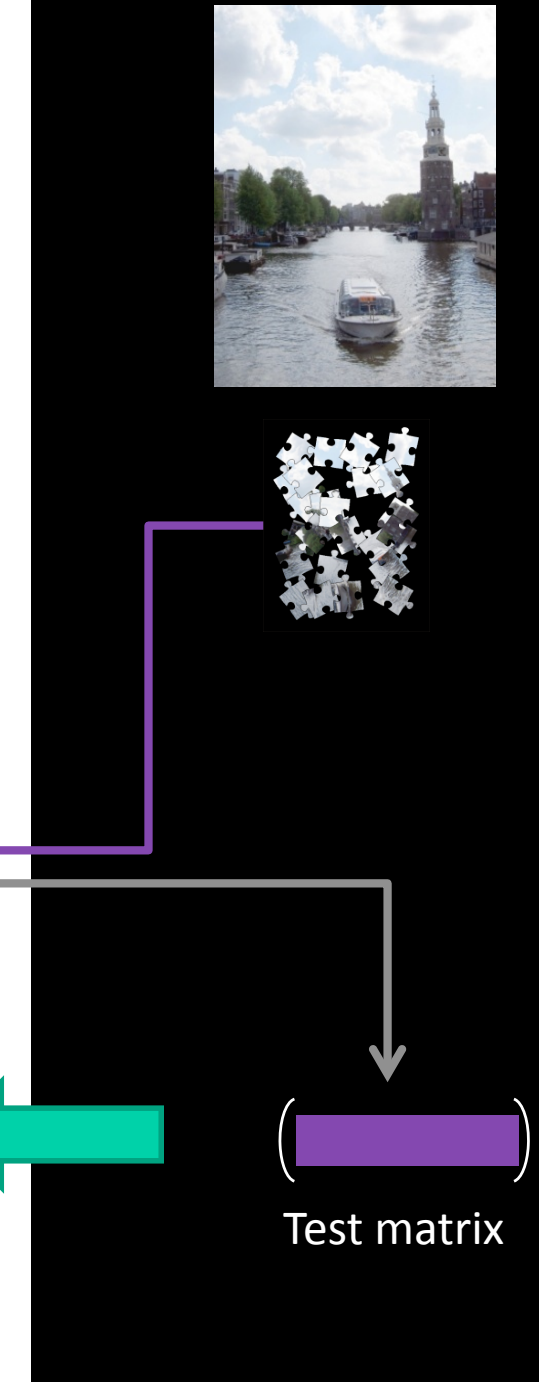
Hele beeld gerepresenteerd als vector van 6 getallen



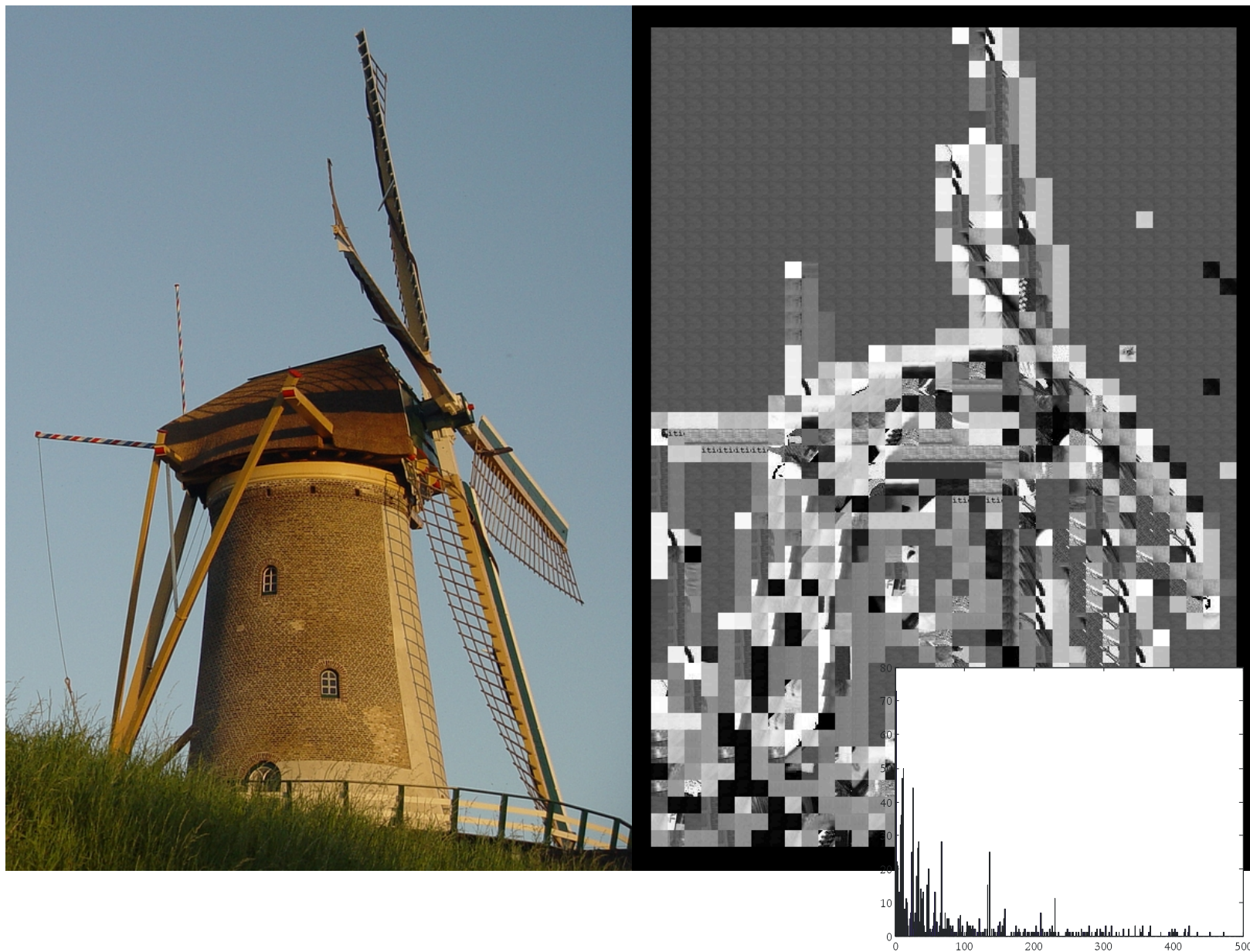
# 20 Leren



# Classificeren



# Example of Visual Words



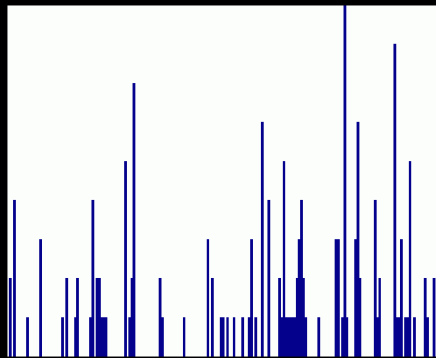


# Questions?

- Advantages?
  - Translation invariant
  - Robust matching of visual word instances
  - Fixed length feature vector
  - Good performance
- Disadvantages?
  - Loss of spatial position and object configuration
  - Quantization loses information

# Spatial pyramid representation

- Extension of a bag of features
- Locally orderless representation at several levels of resolution

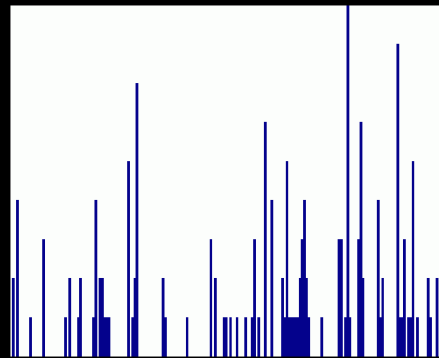


level 0

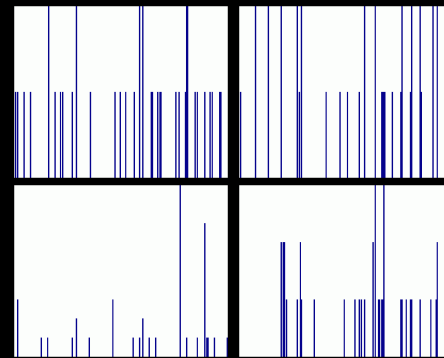


# Spatial pyramid representation

- Extension of a bag of features
- Locally orderless representation at several levels of resolution



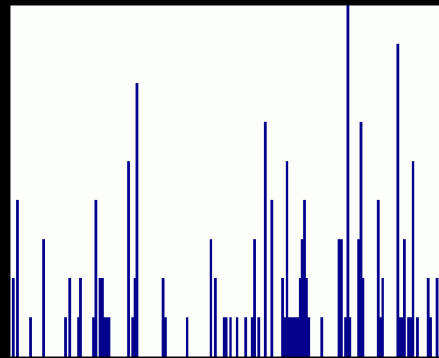
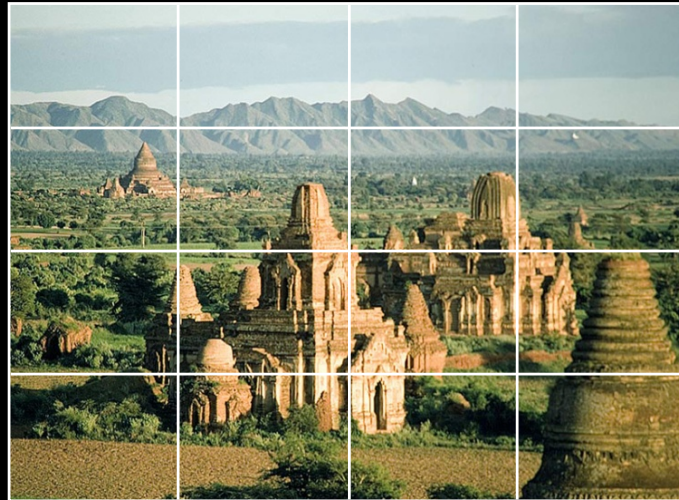
level 0



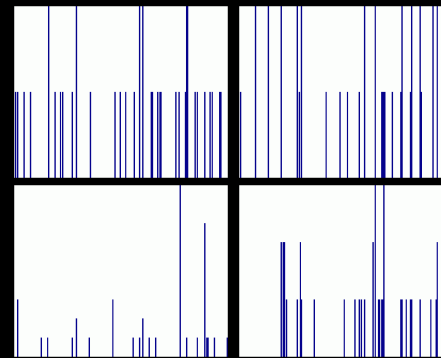
level 1

# Spatial pyramid representation

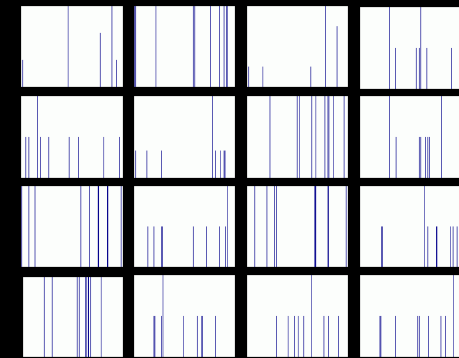
- Extension of a bag of features
- Locally orderless representation at several levels of resolution



level 0



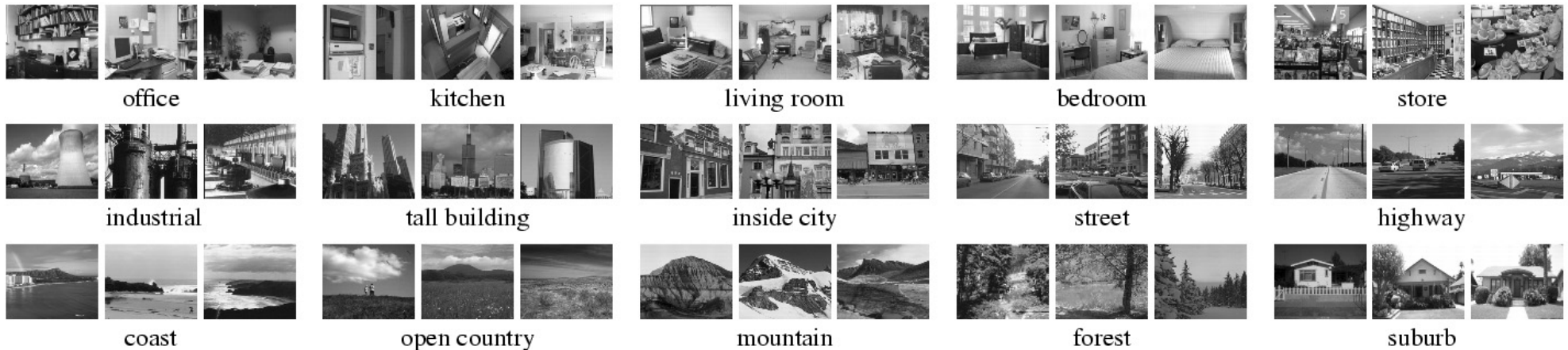
level 1



level 2



# Scene category dataset

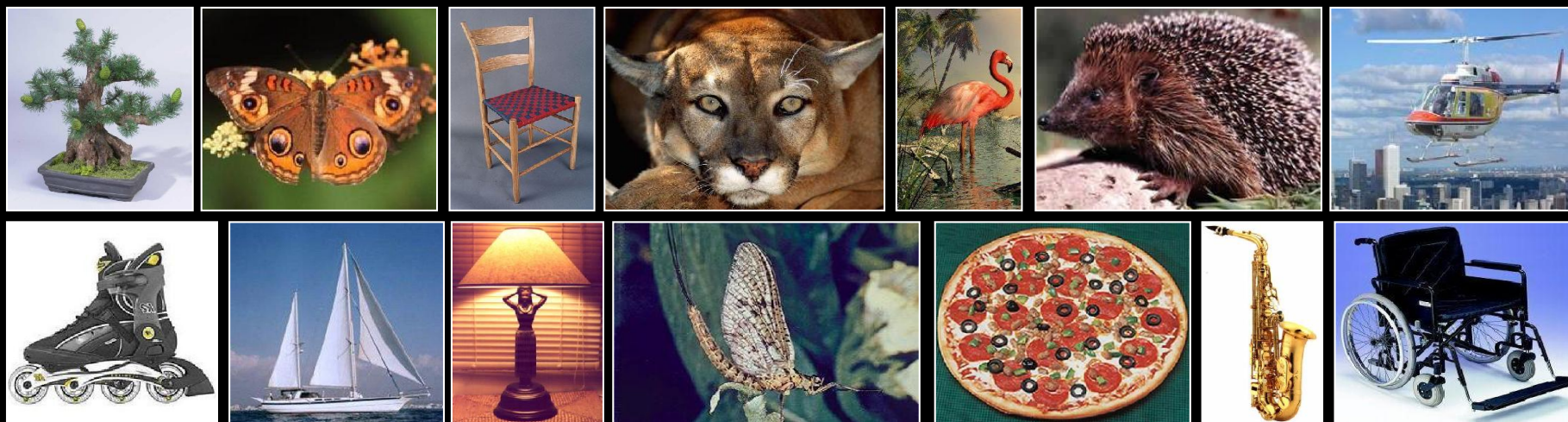


## Multi-class classification results (100 training images per class)

Level	Weak features (vocabulary size: 16)		Strong features (vocabulary size: 200)	
	Single-level	Pyramid	Single-level	Pyramid
0 (1 × 1)	45.3 ±0.5		72.2 ±0.6	
1 (2 × 2)	53.6 ±0.3	56.2 ±0.6	77.9 ±0.6	79.0 ±0.5
2 (4 × 4)	61.7 ±0.6	64.7 ±0.7	79.4 ±0.3	<b>81.1 ±0.3</b>
3 (8 × 8)	63.3 ±0.8	<b>66.8 ±0.6</b>	77.2 ±0.4	80.7 ±0.3

# Caltech101 dataset

[http://www.vision.caltech.edu/Image\\_Datasets/Caltech101/Caltech101.html](http://www.vision.caltech.edu/Image_Datasets/Caltech101/Caltech101.html)



## Multi-class classification results (30 training images per class)

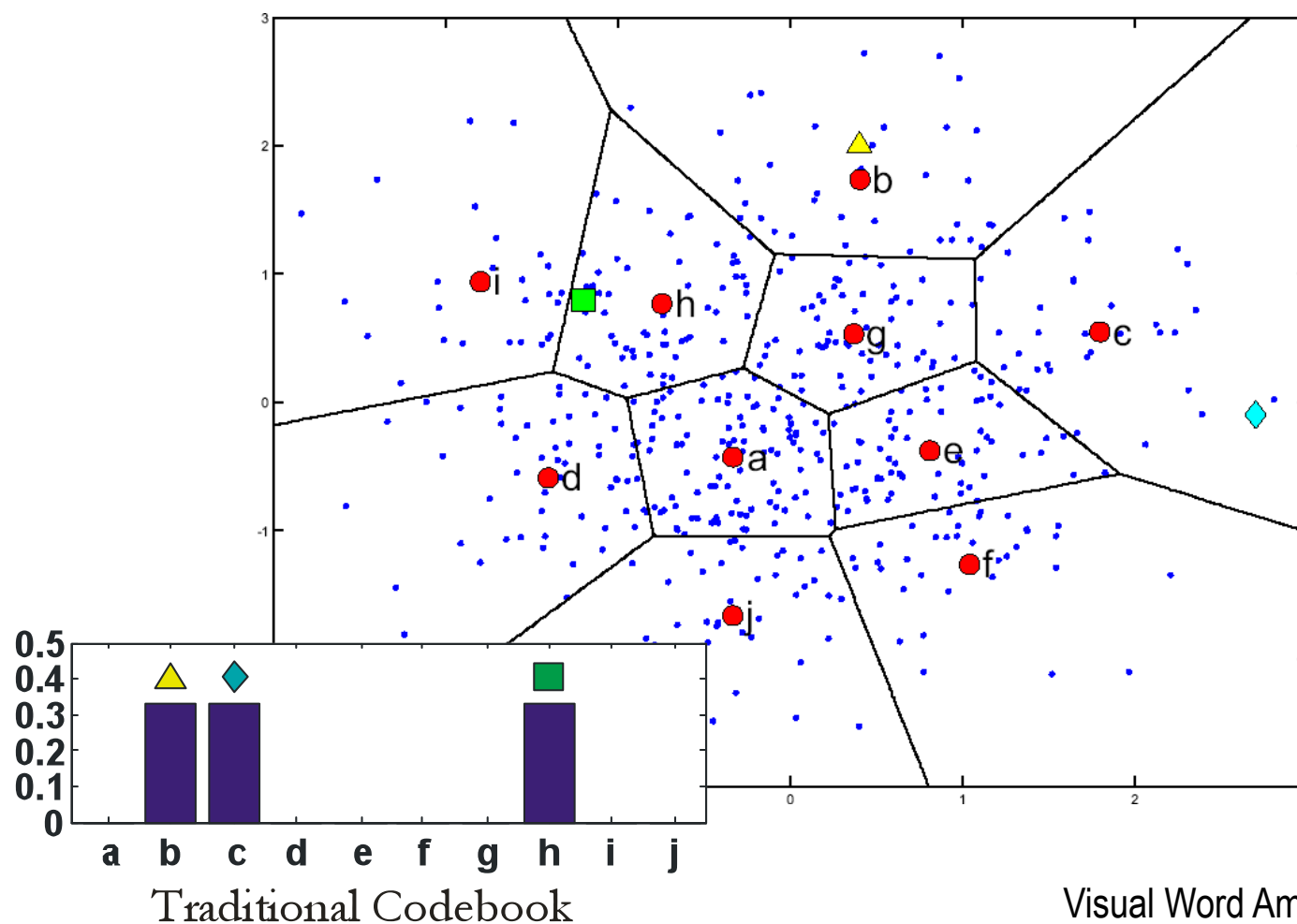
	Weak features (16)		Strong features (200)	
Level	Single-level	Pyramid	Single-level	Pyramid
0	15.5 $\pm$ 0.9		41.2 $\pm$ 1.2	
1	31.4 $\pm$ 1.2	32.8 $\pm$ 1.3	55.9 $\pm$ 0.9	57.0 $\pm$ 0.8
2	47.2 $\pm$ 1.1	49.3 $\pm$ 1.4	63.6 $\pm$ 0.9	<b>64.6</b> $\pm$ 0.8
3	52.2 $\pm$ 0.8	<b>54.0</b> $\pm$ 1.1	60.3 $\pm$ 0.9	64.6 $\pm$ 0.7



# Questions?

## Drawback to the Codebook Model

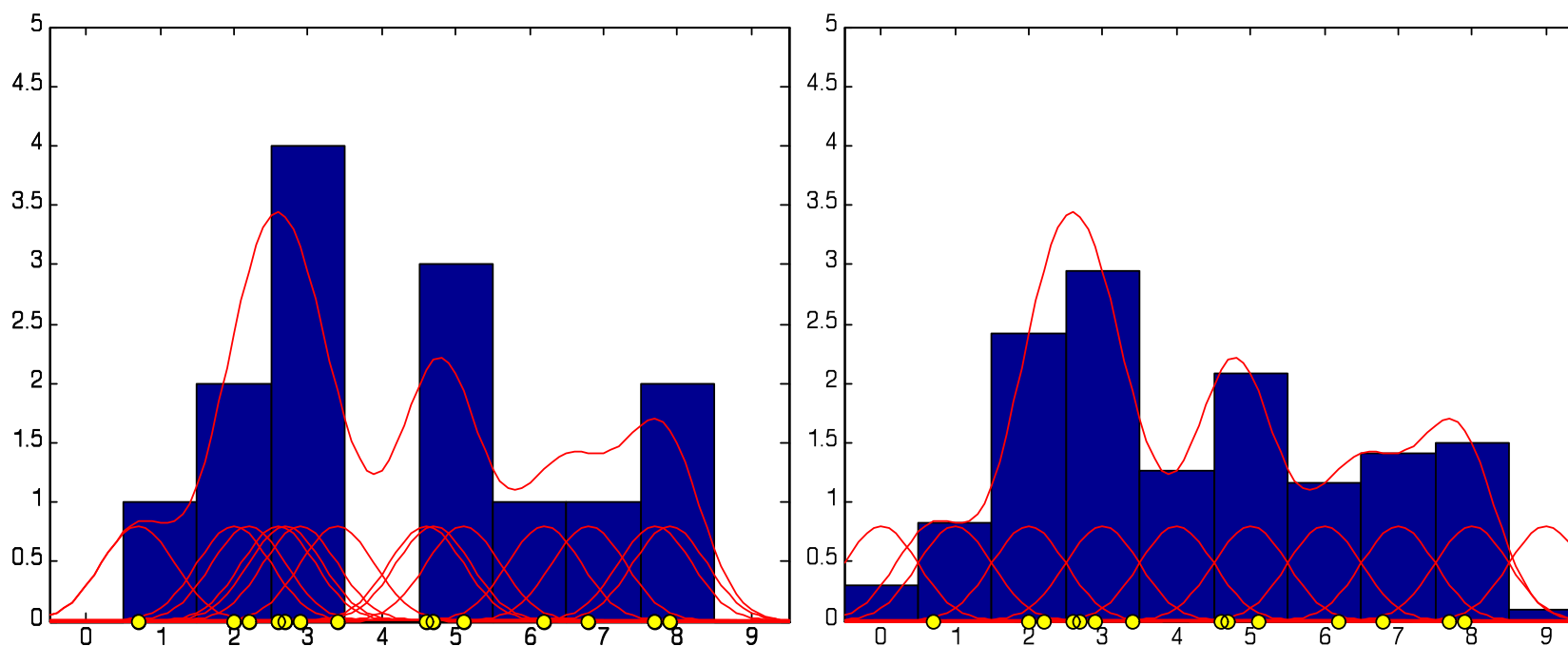
- Hard assignment of features to codewords
  - Codeword Uncertainty / Codeword Plausibility





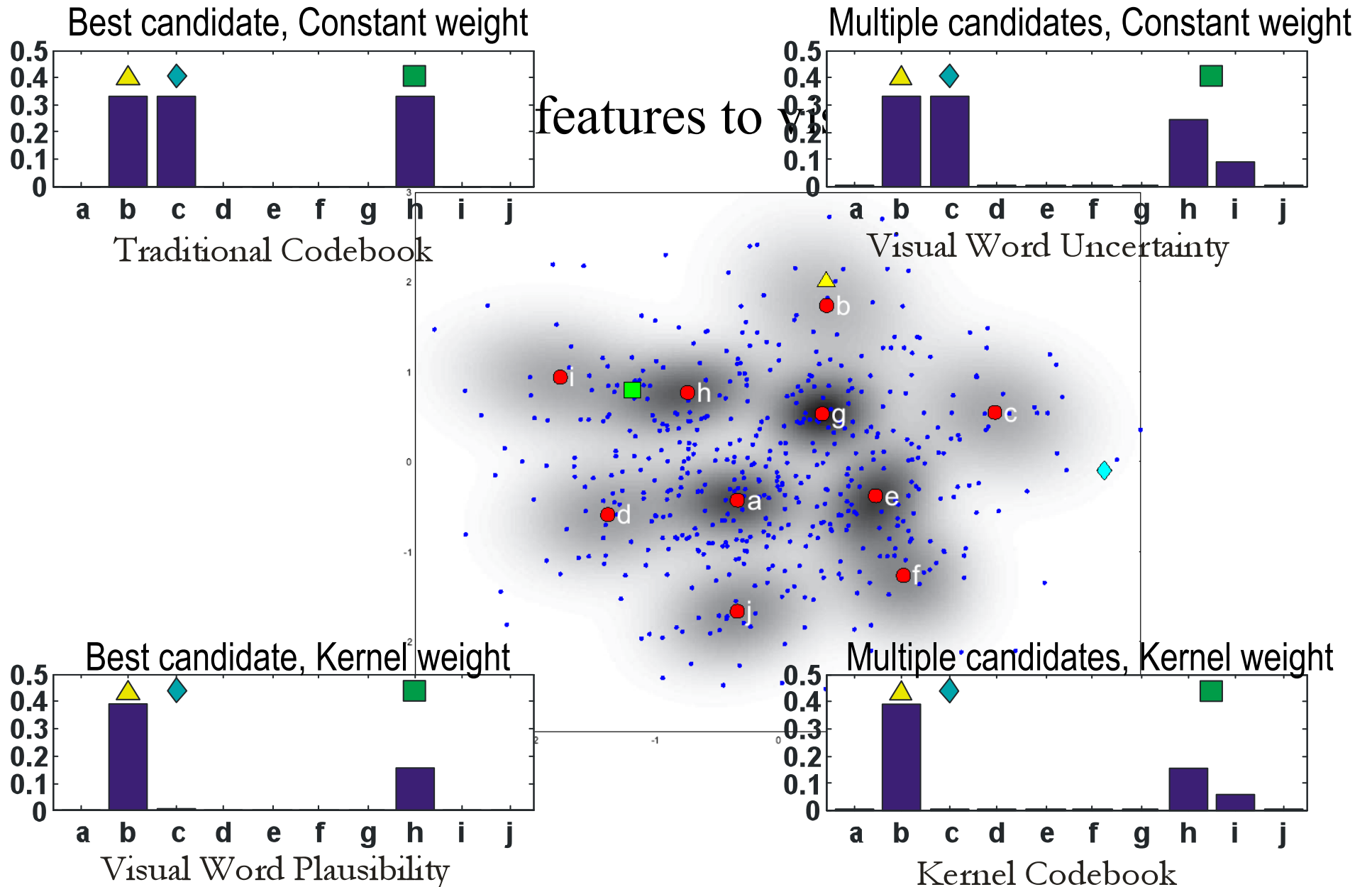
# Kernel Codebooks

- Codebook model is a histogram
  - Problem: Hard assignment to histogram bins
  - a solution: Kernel density estimation

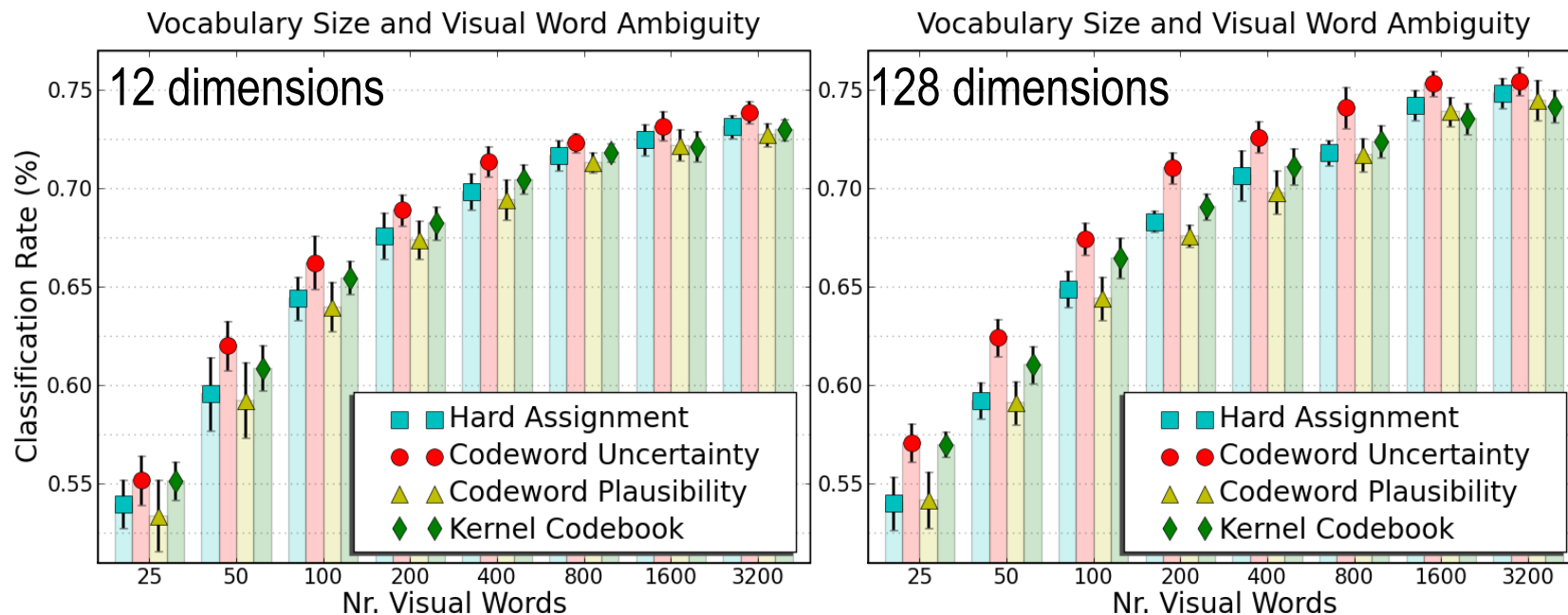


- Symmetric Kernel allows kernel codewords

# Ambiguity Modeling



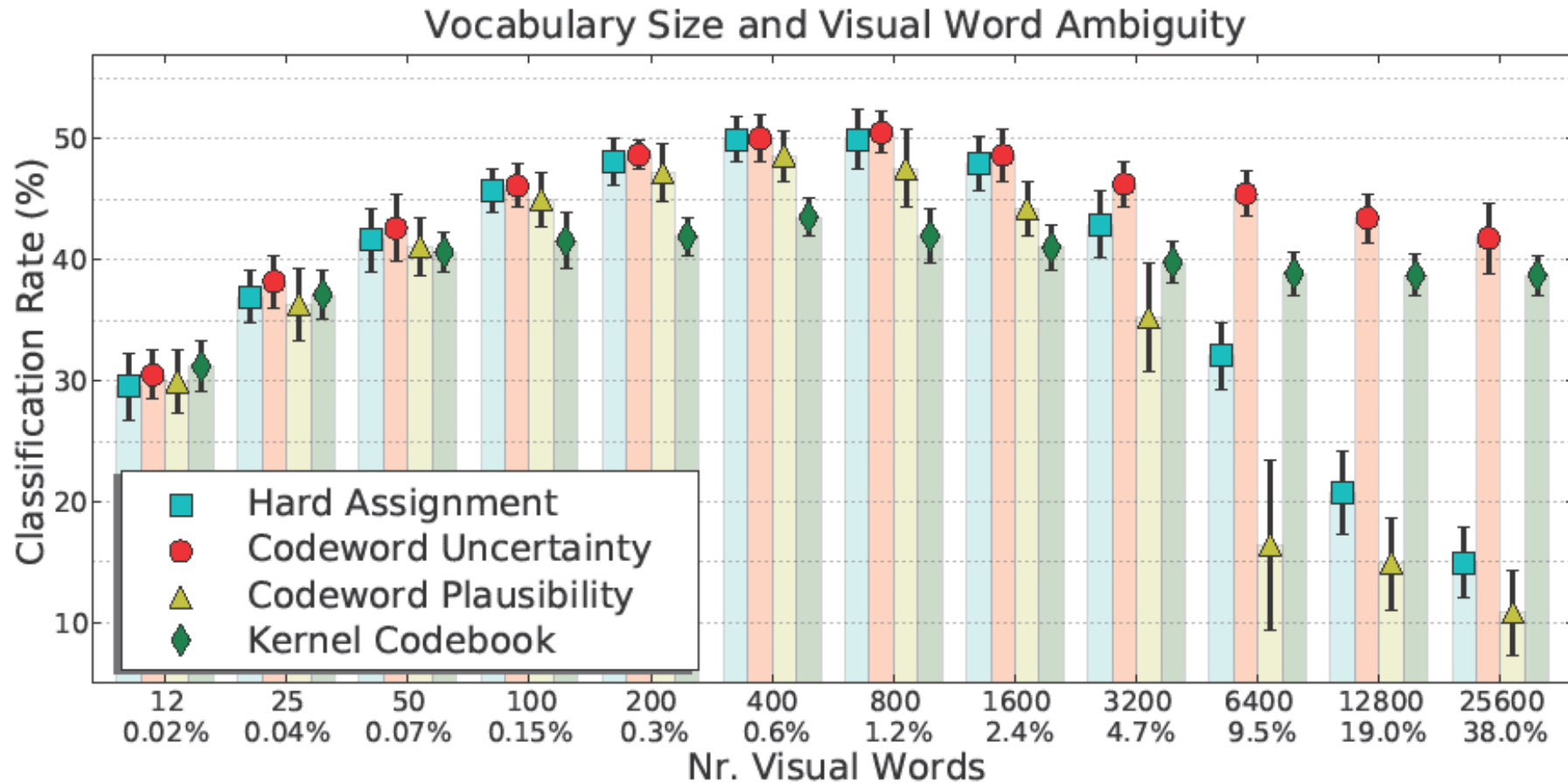
# Results Scene-15 Dataset



- Ambiguity has more effect in high-dimensions
- Codeword Plausibility always performs worst
- Codeword Uncertainty always performs best
- Ambiguity has more effect for small vocabularies
- Performance increases asymptotically with vocabulary size



# Results Scene-15 Dataset



- Hard assignment completely fails for too large vocabularies
- Soft assignment stays relatively constant

# Questions?

# Summary

- Types of Recognition, challenges
- Bag of Features, Visual-words
- Spatial-Pyramid
- Soft-assignment