

Unsupervised Feature Learning by Deep Sparse Coding

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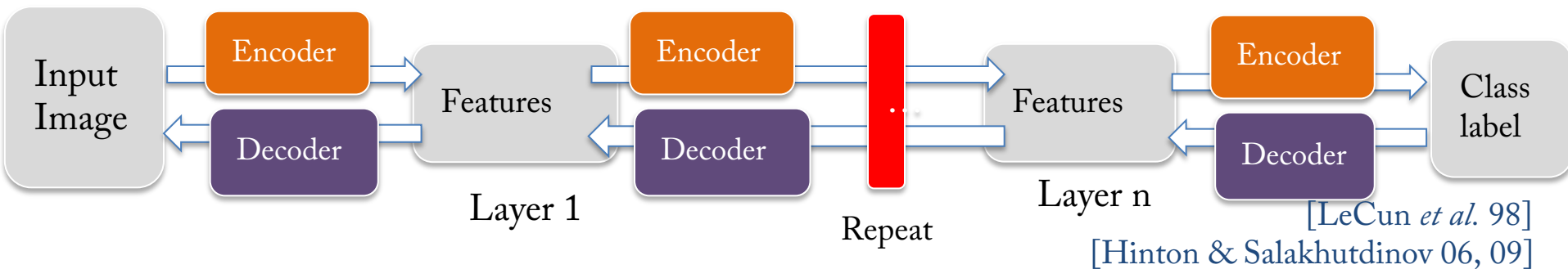
Yanjun Qi, University of Virginia

Two lines of Previous Research

- Bag of Visual words



- Deep learning



- Motivation: **Can we combine the power of two methods?**

Background: Bag of Visual words

Image



Dense Code



Sparse Code

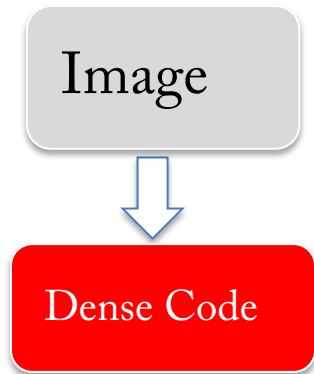


Pyramid Pooling

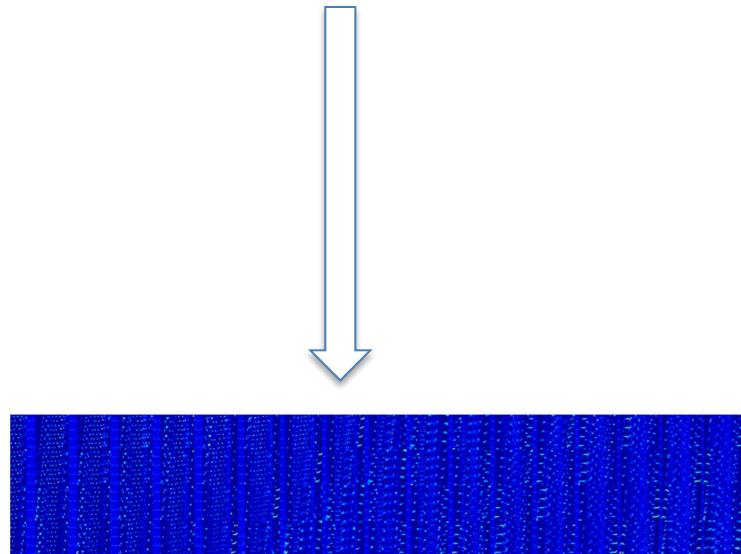


Classifier

Background: Bag of Visual words

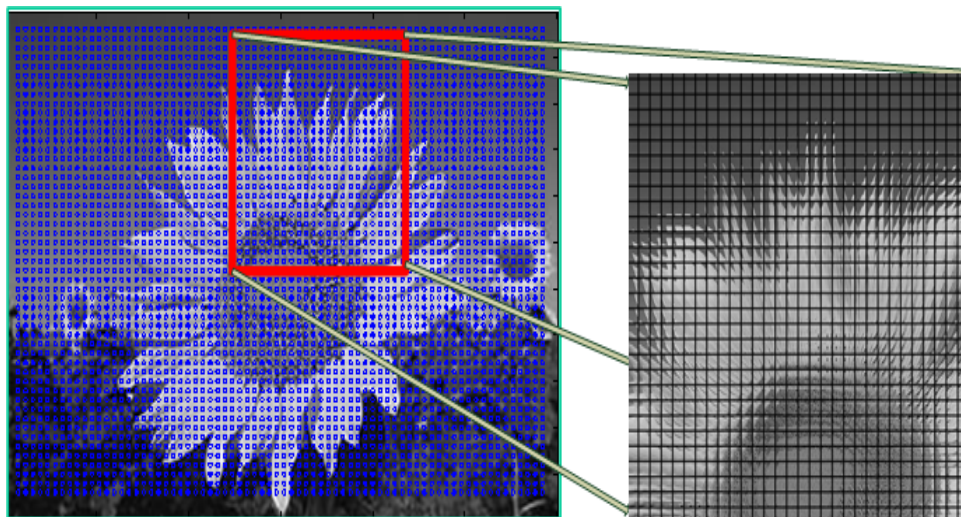
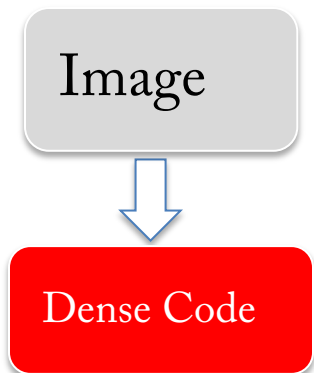


Image

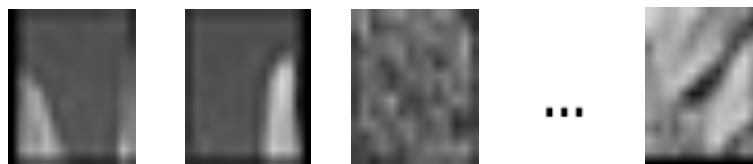


SIFT Descriptors

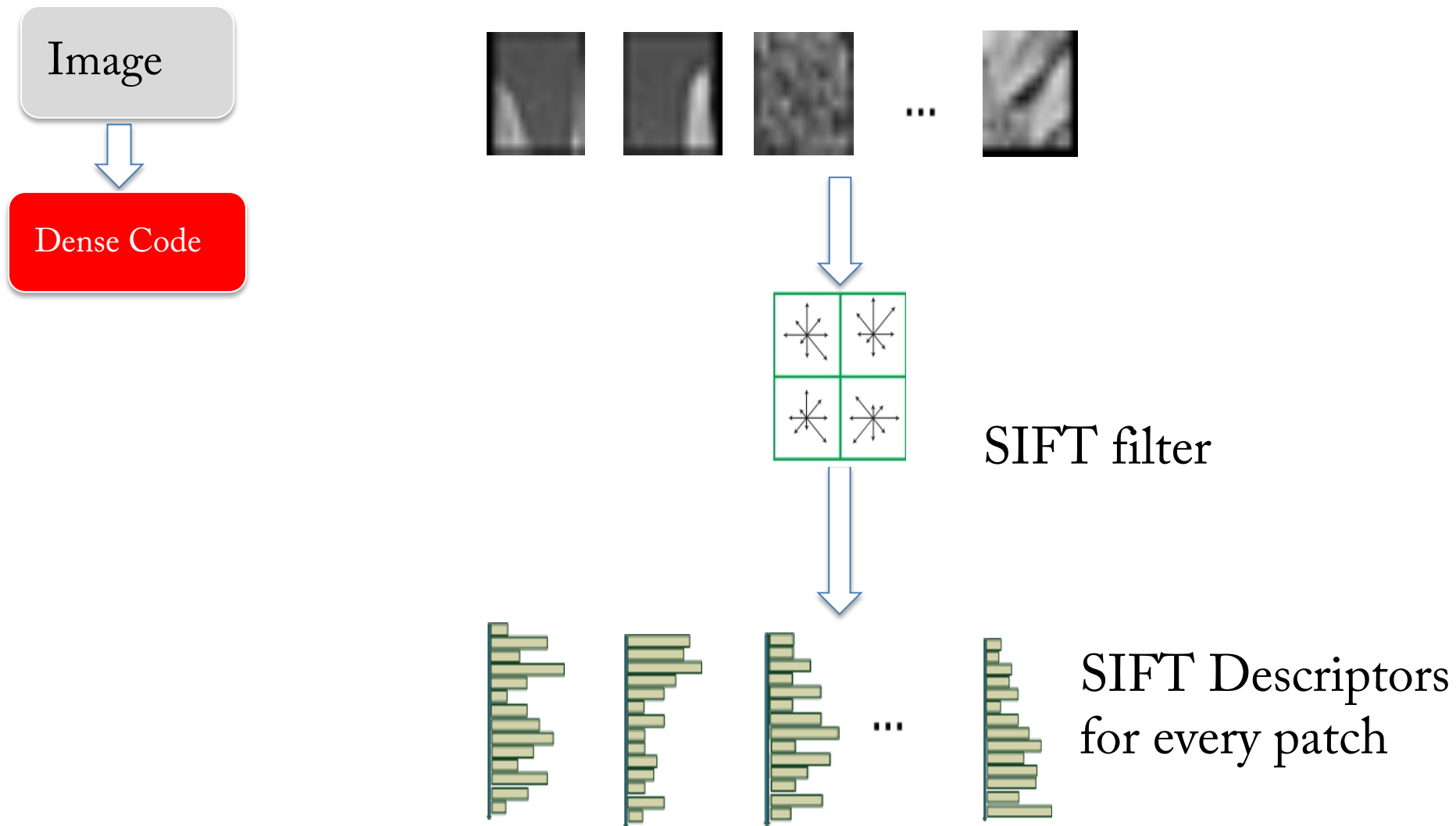
Background: Bag of Visual words



Example: 900 Patches

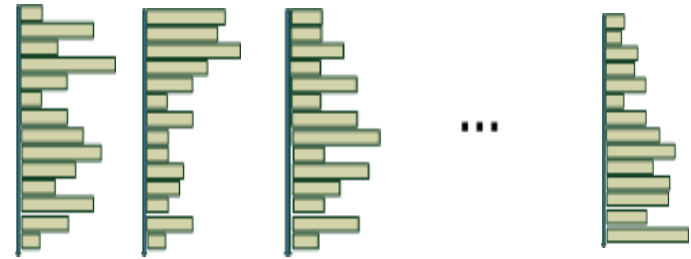


Background: Bag of Visual words



Background: Bag of Visual words

SIFT Descriptors
 $x \in R^{128}$, dense



$$s = \operatorname{argmin}_s \|x - Bs\|^2 + \lambda \|s\|_1$$

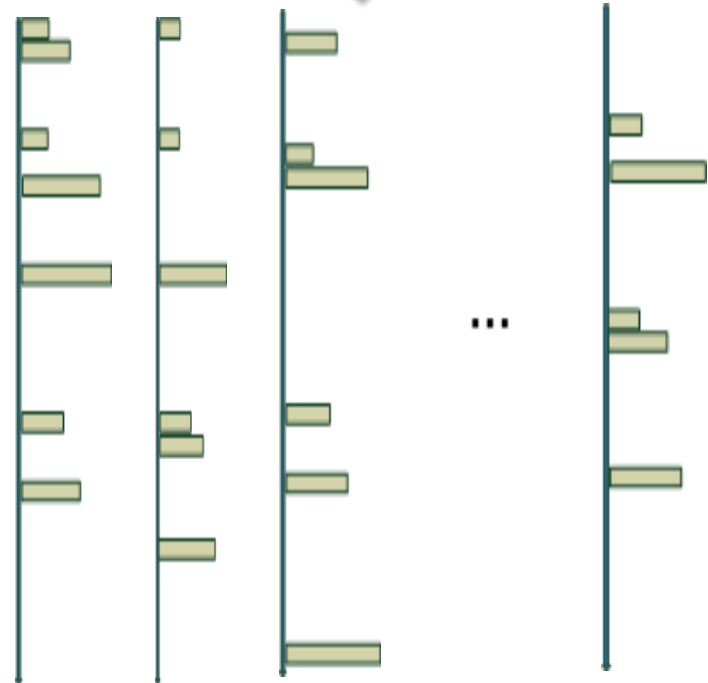
$B \in R^{128 \times 1024}$

Dense Code



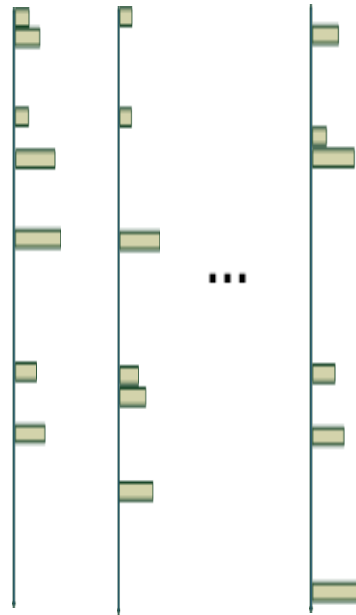
Sparse Code

Sparse Codes
 $s \in R^{1024}$



Background: Bag of Visual words

Sparse Codes
 $s \in R^{1024}$



Max Pooling:
 $\max(s_1, s_2, \dots, s_n)$



Concatenate Sparse
Codes of 21 pooling
regions

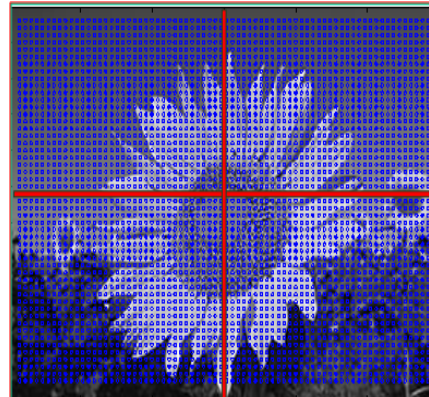
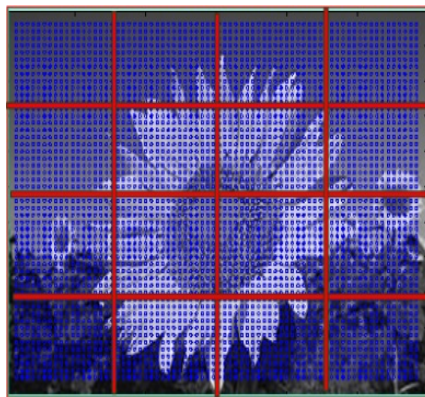


Sparse Code

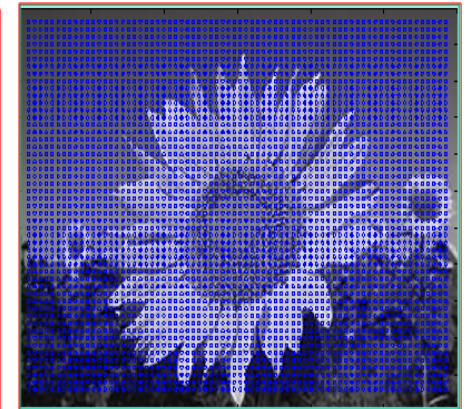


Pyramid
Pooling

21 pooling regions

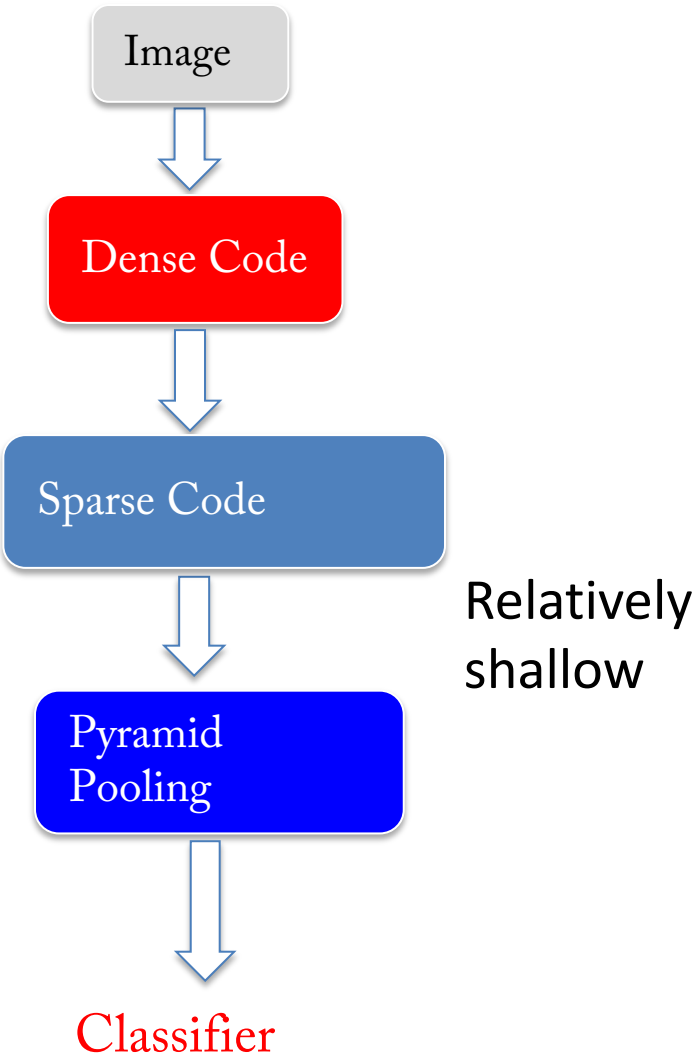


Level 1



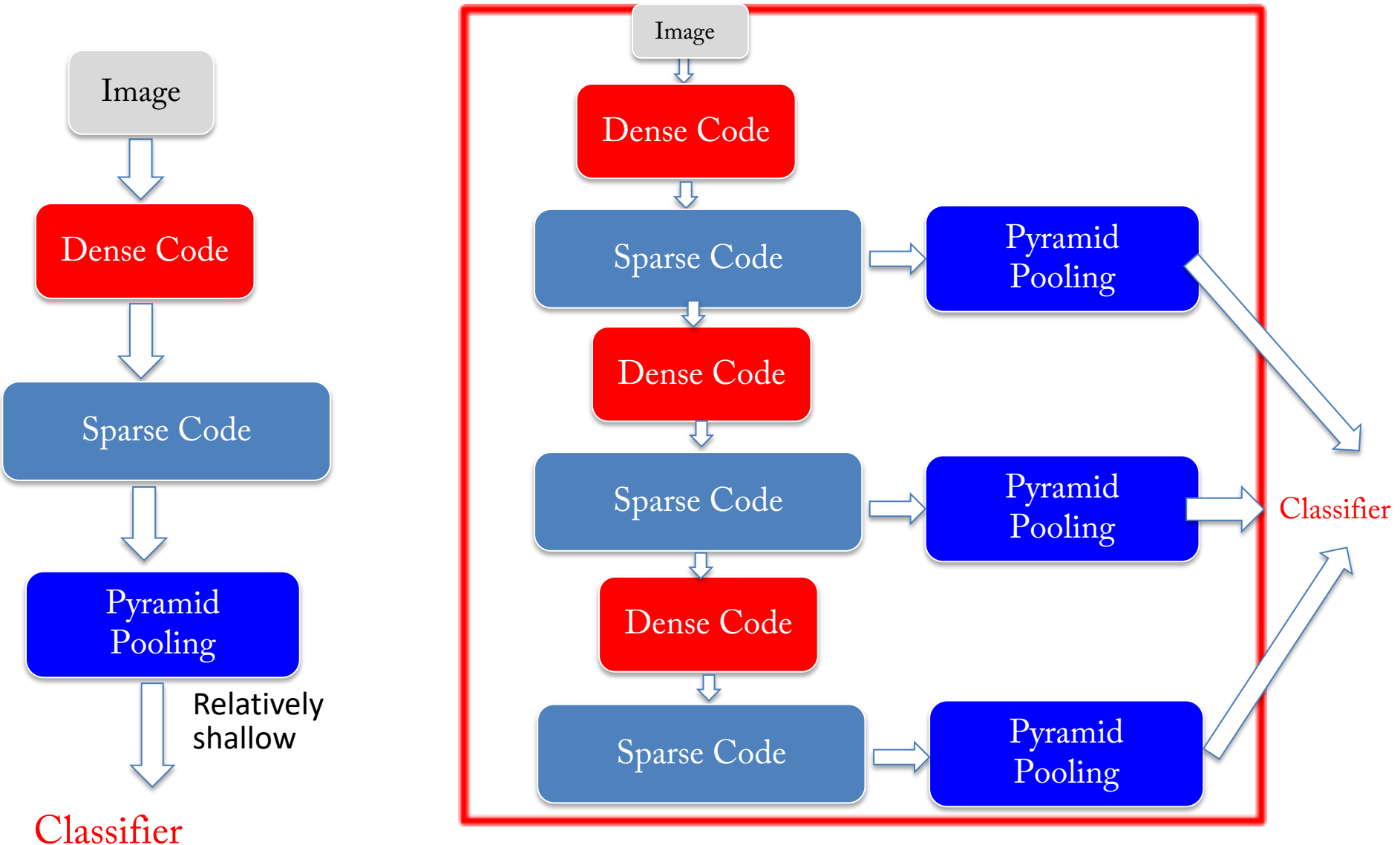
Level 0

Background: Bag of Visual Words

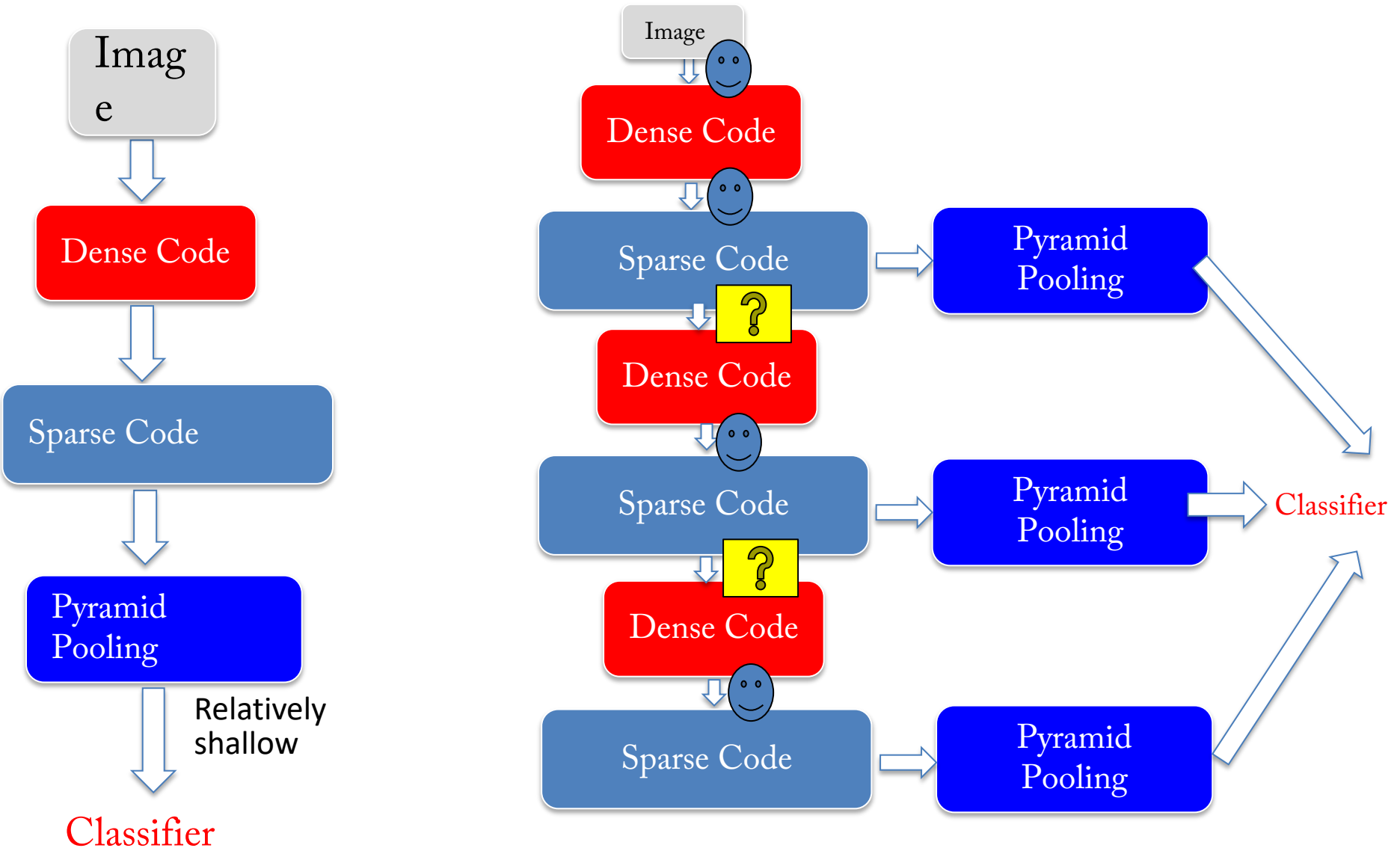


Can we learn multiple layers of sparse representations?

Deep Sparse Coding

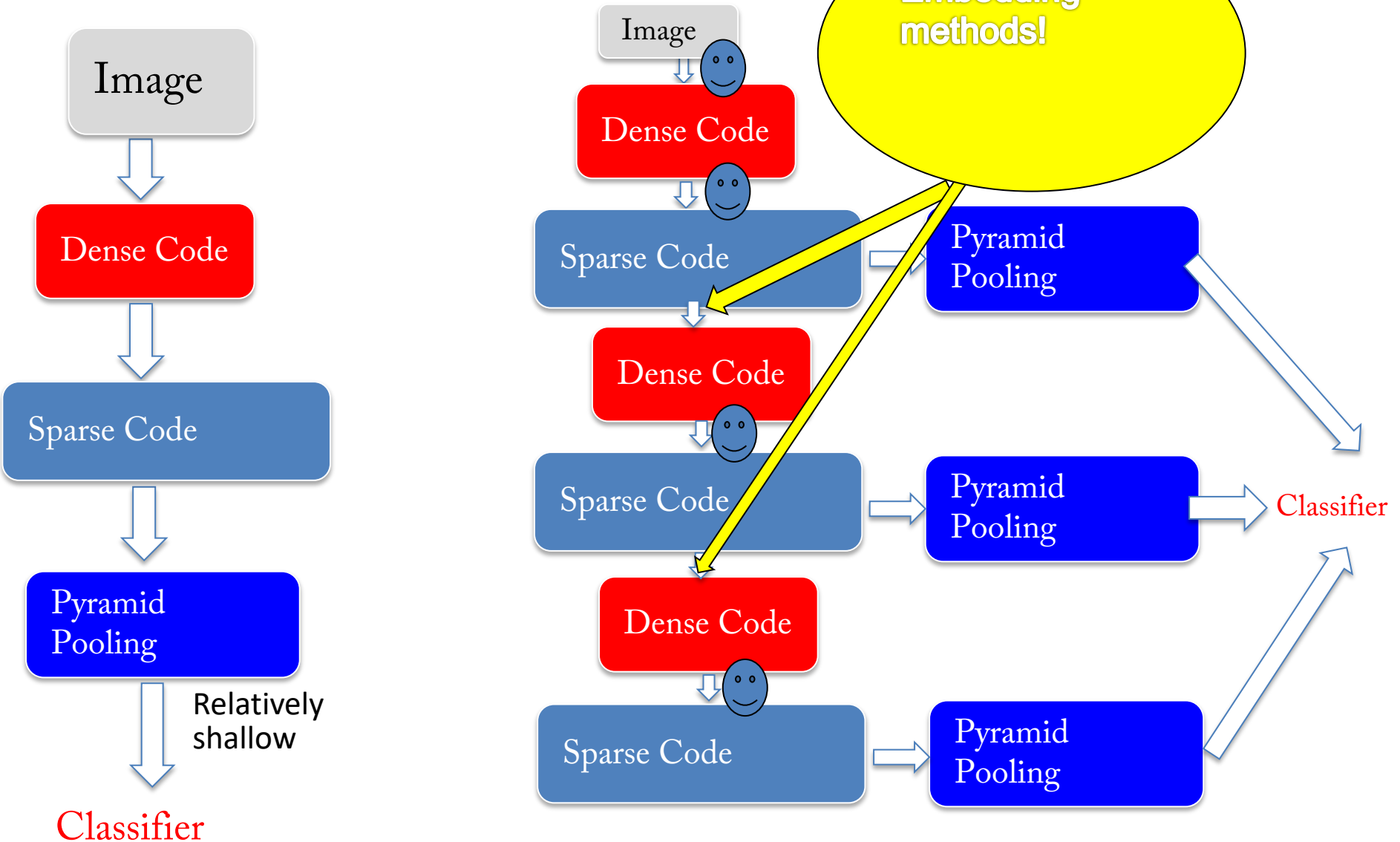


Deep Sparse Coding



Unsupervised Deep Architecture!

Deep Sparse Coding



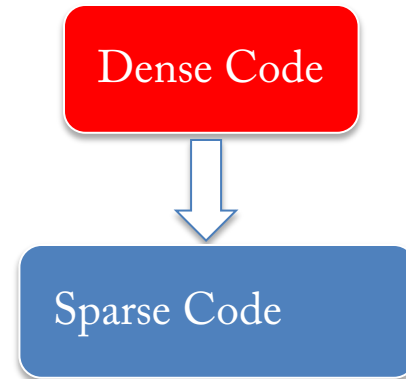
Dimension expansion

Idea:

- Separation (for classification)
- Representation (bag of visual words)

Method:

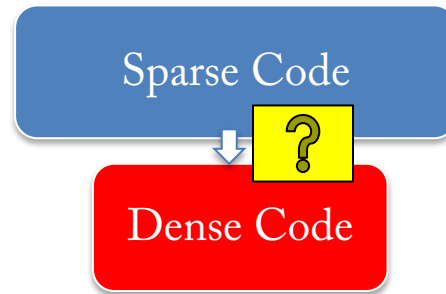
- Sparse coding



Dimension reduction

Idea:

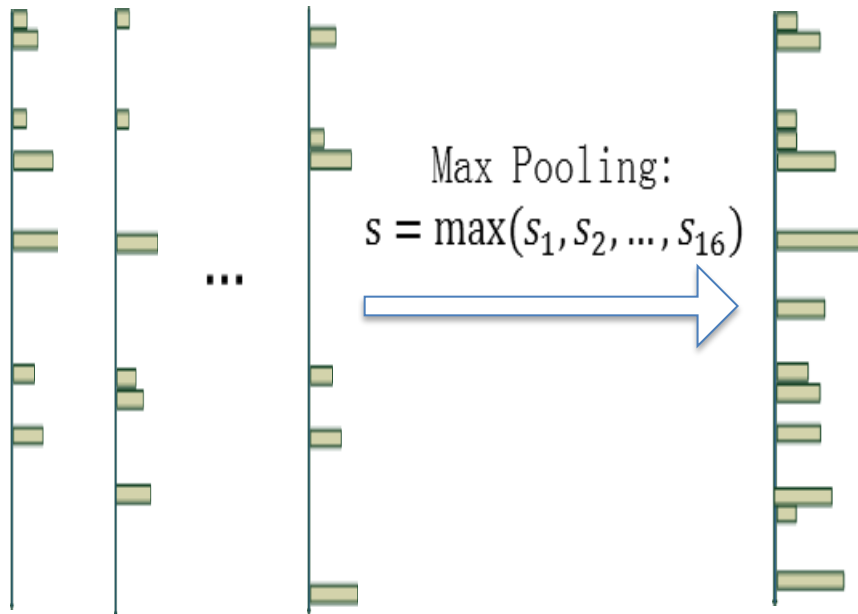
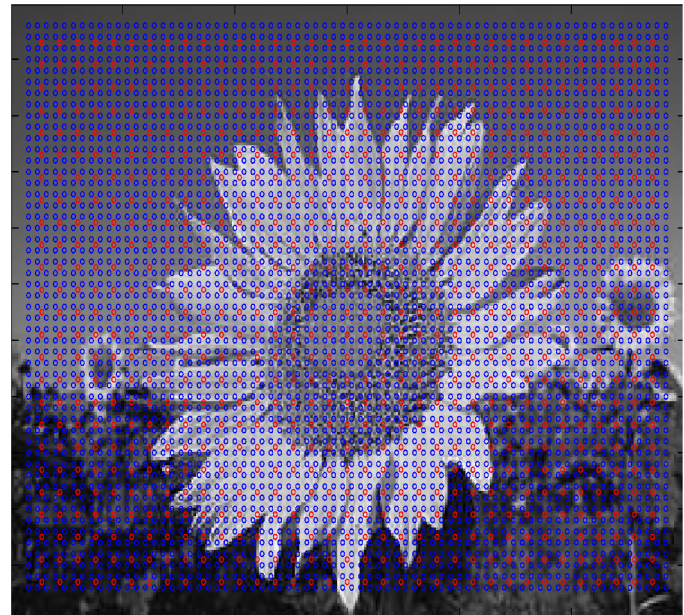
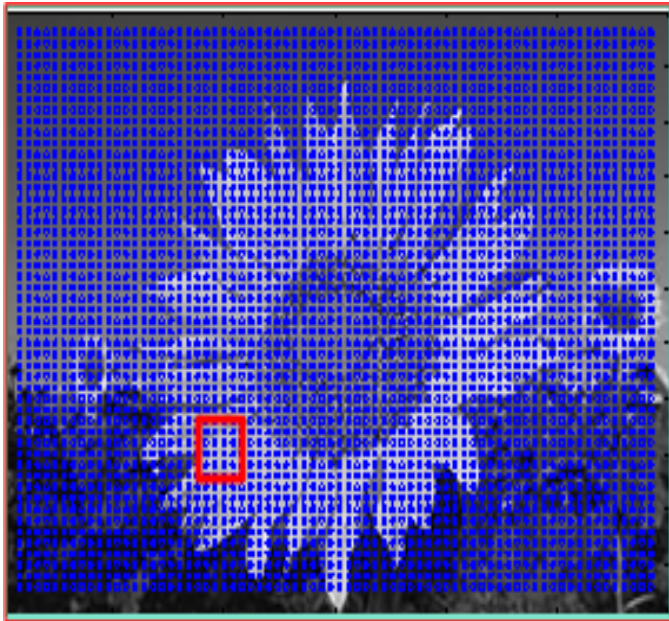
- Invariance
- Compositionality



Method:

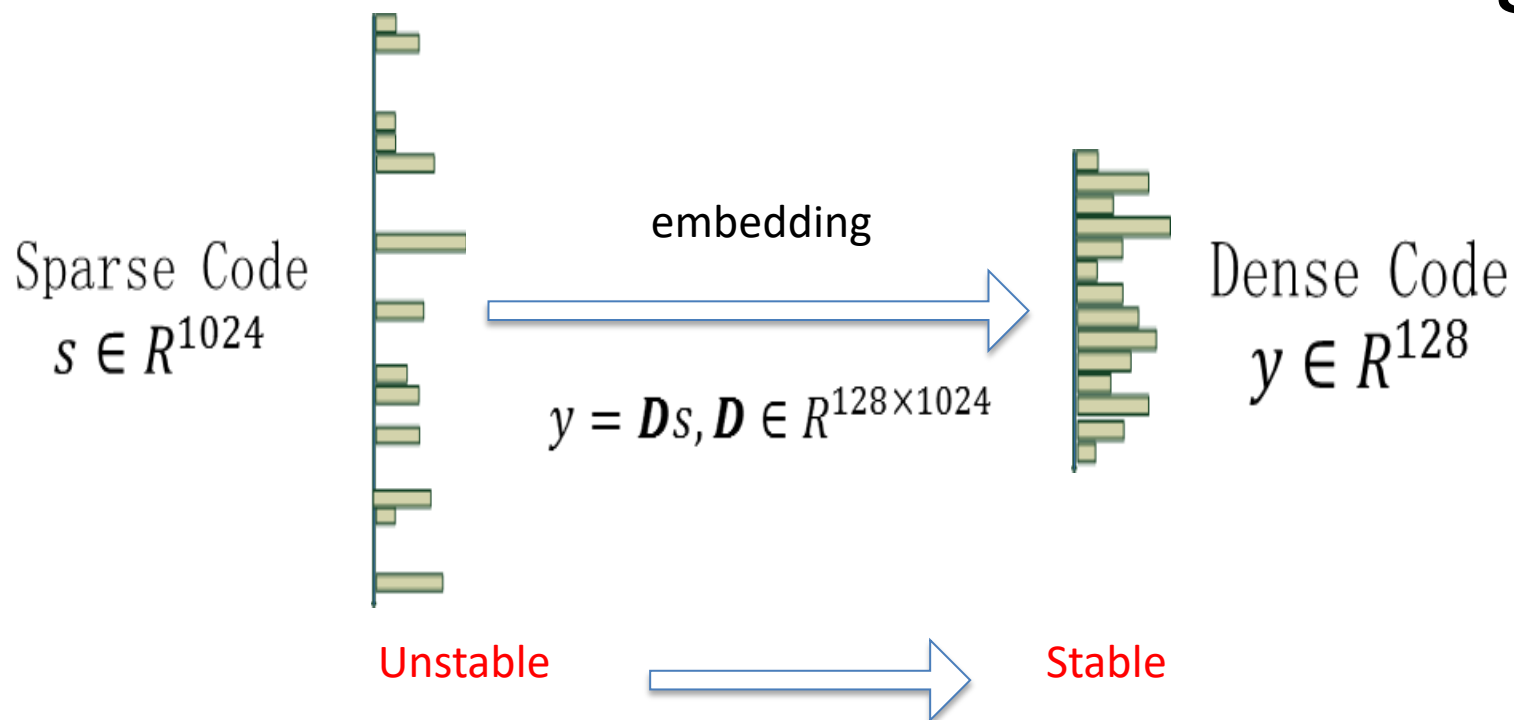
- Locally spatial pooling + Low dimensional embedding
- Both unsupervised

Locally spatial pooling

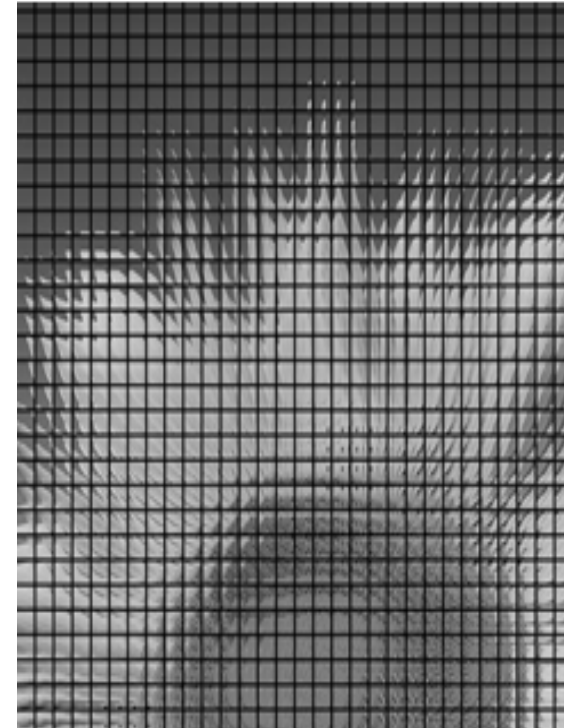
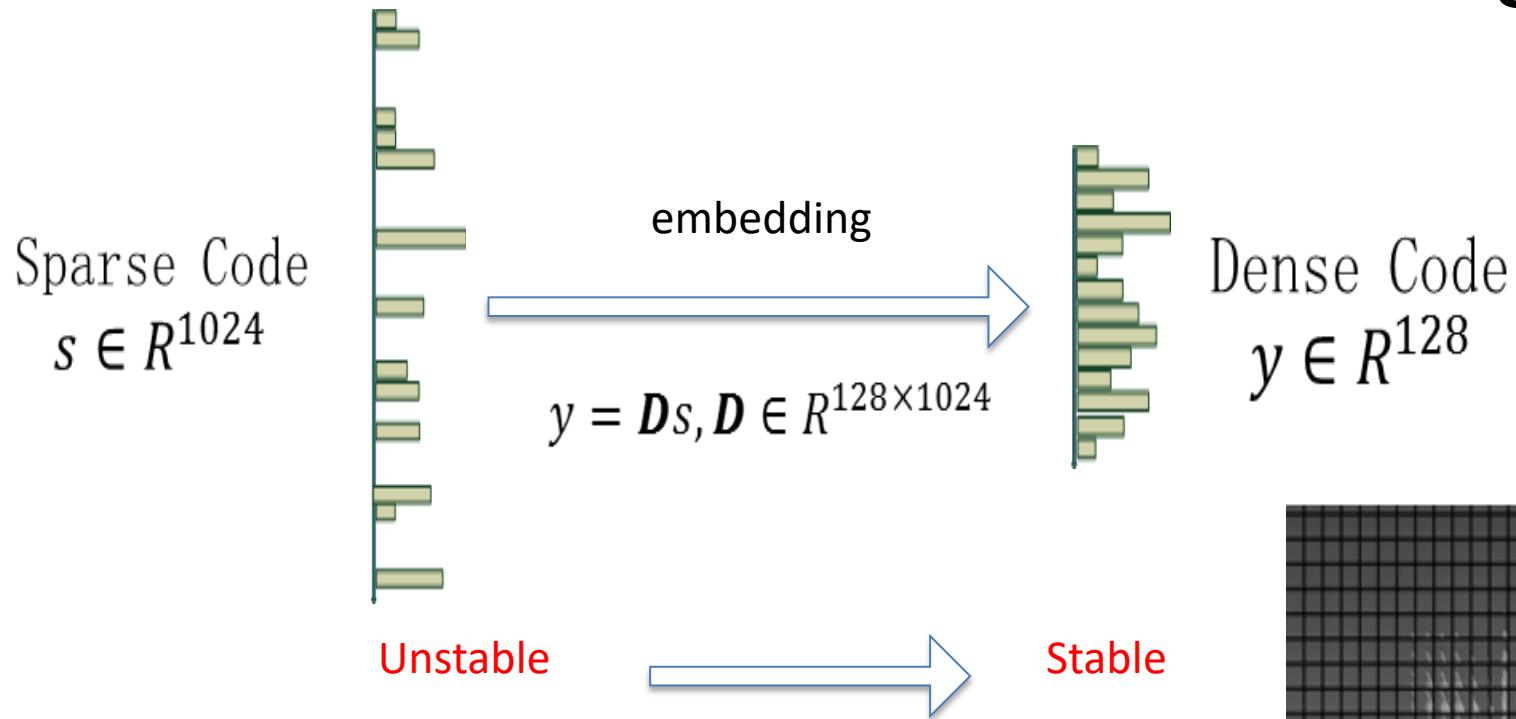


Covers larger area

Low dimensional embedding

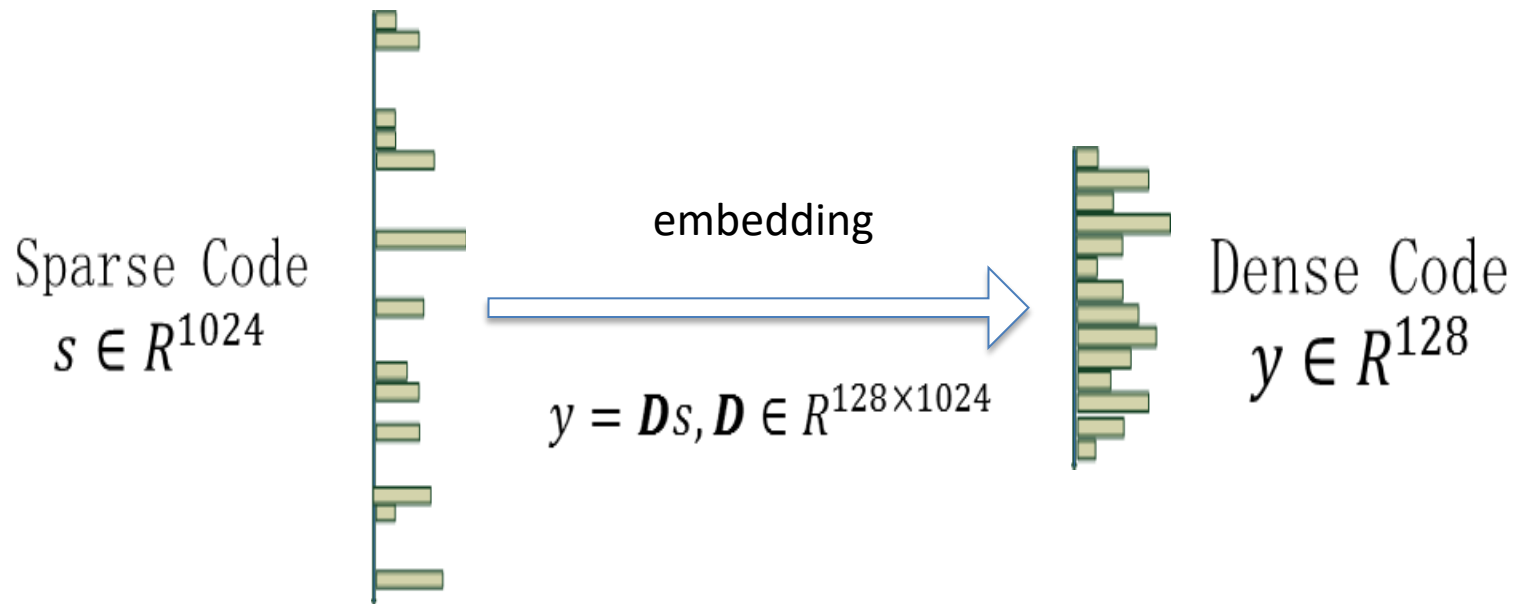


Low dimensional embedding



Idea: embedding with the help of spatial information

Low dimensional embedding



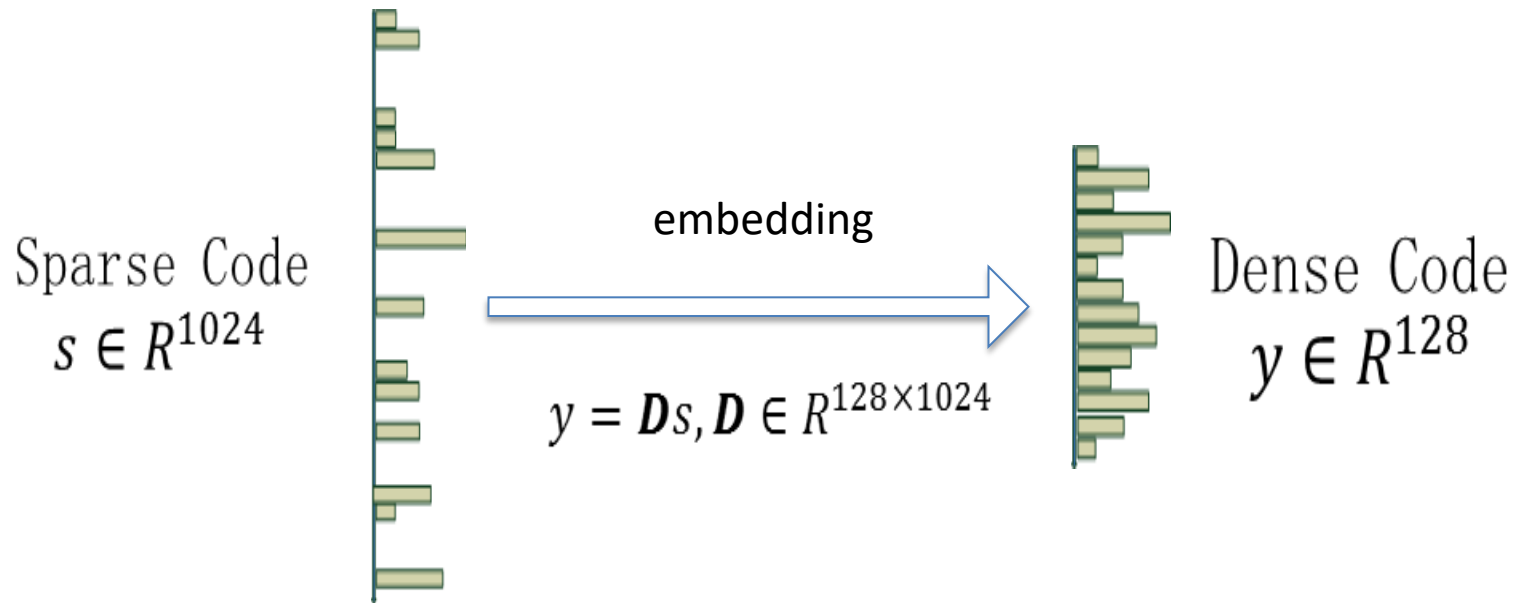
Dimensionality Reduction by Learning an Invariant Mapping (DR. LIM)

$$D = \operatorname{argmin}_D \sum_{i,j} w_{ij} \|Ds_i - Ds_j\|^2 + \sum_{i,j} (1 - w_{ij}) \max(0, \alpha - \|Ds_i - Ds_j\|)^2$$

$w_{ij} = 1$, if s_i and s_j are overlapping neighbors

$w_{ij} = 0$, if s_i and s_j are non-overlapping neighbors

Low dimensional embedding



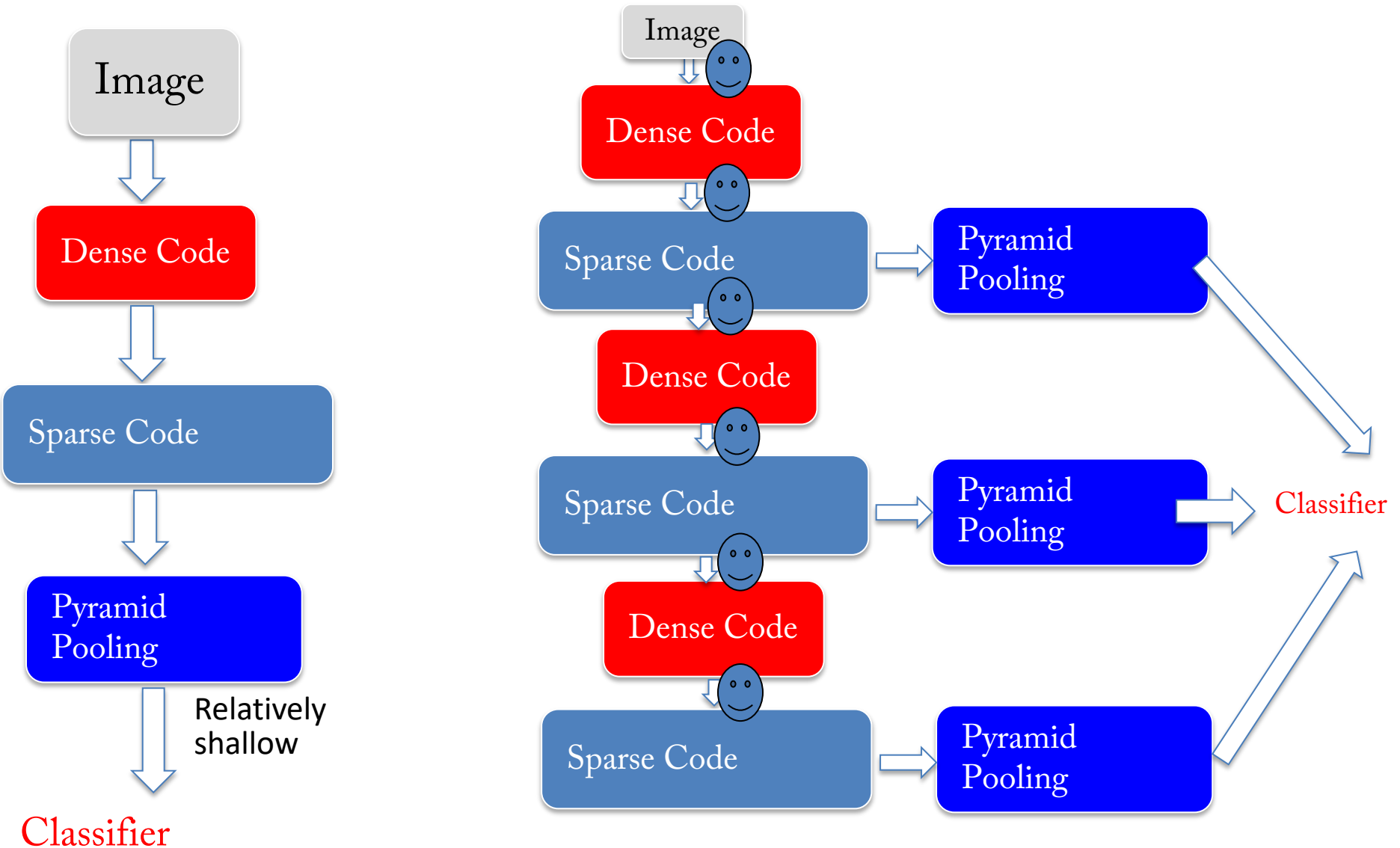
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Deep Sparse Coding



Unsupervised Deep Architecture!

Conclusion

- Combine sparse coding with deep learning
- Combine dimension expansion and dimension reduction
- A connecting function is learned by embedding method
- Combining multiple layers of sparse code achieves state-of-the-art performance on image classification tasks