

Unsupervised Feature Learning by Deep Sparse Coding

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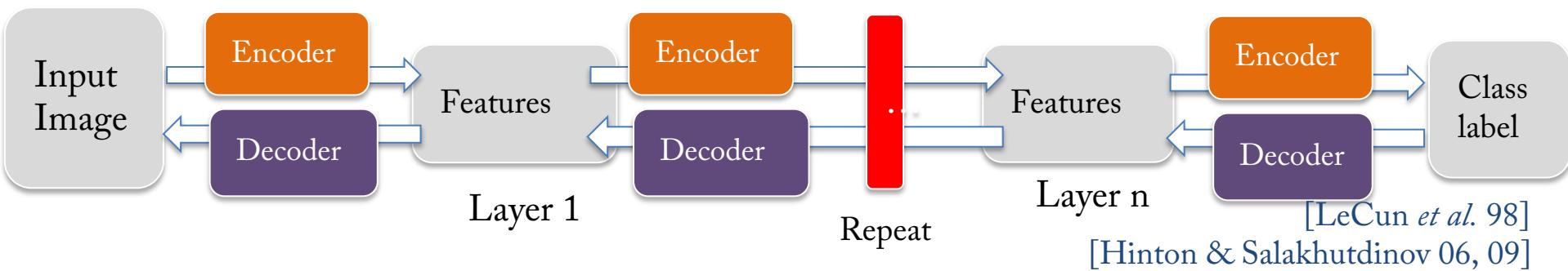
Yanjun Qi, University of Virginia

Two lines of Previous Research

- Bag of Visual words

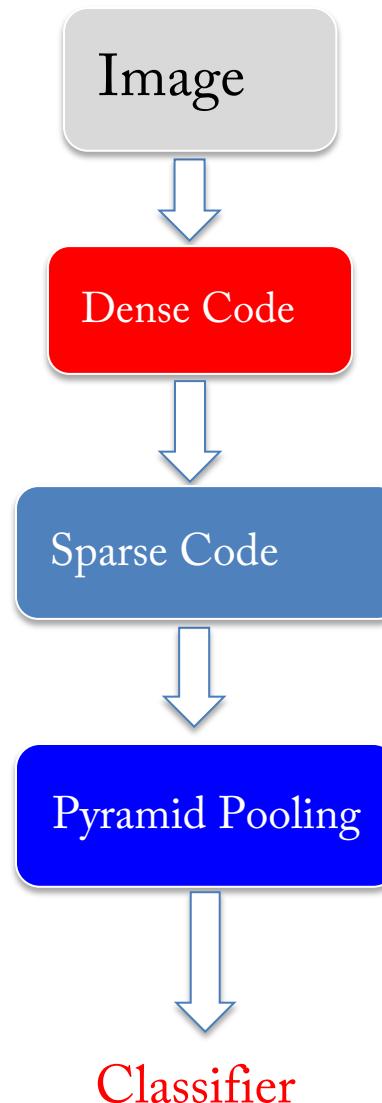


- Deep learning



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

Background: Bag of Visual words



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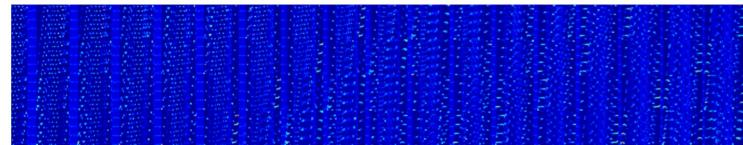
Image



Dense Code



Image



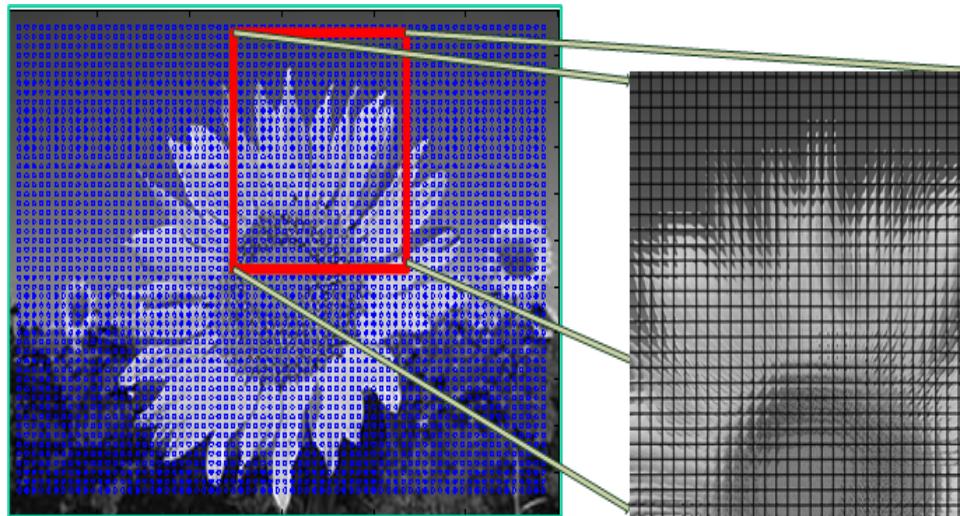
SIFT Descriptors

Background: Bag of Visual words

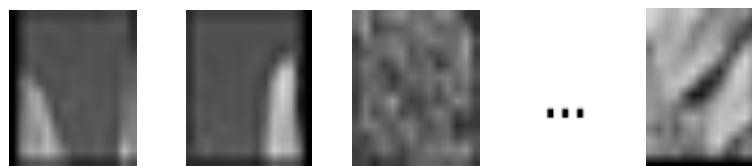
Image



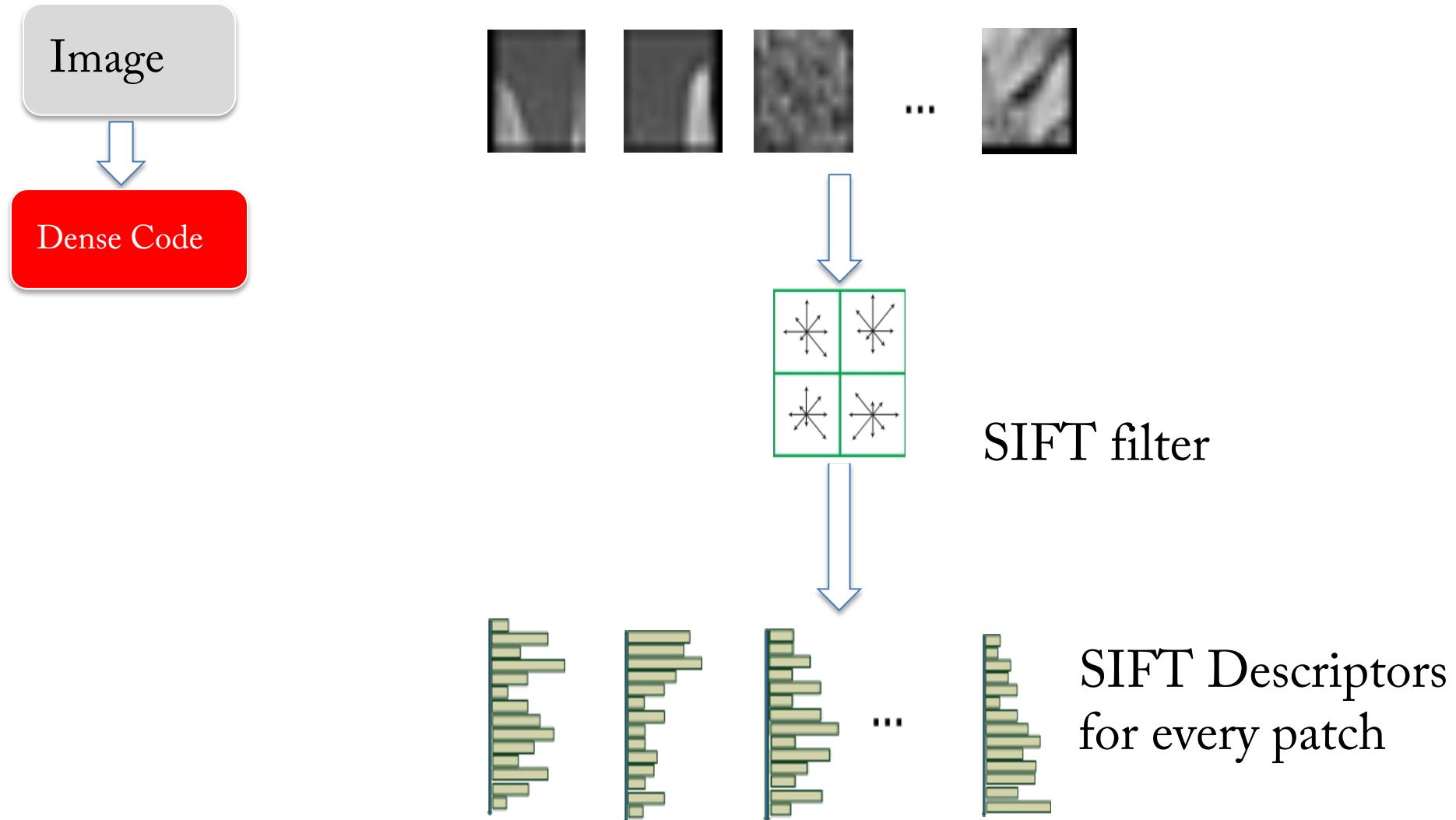
Dense Code



Example: 900 Patches

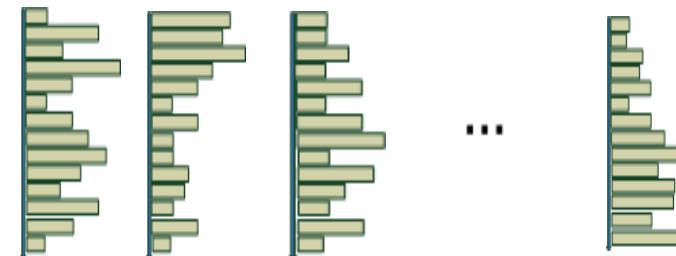


Background: Bag of Visual words



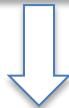
Background: Bag of Visual words

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



...

Dense Code

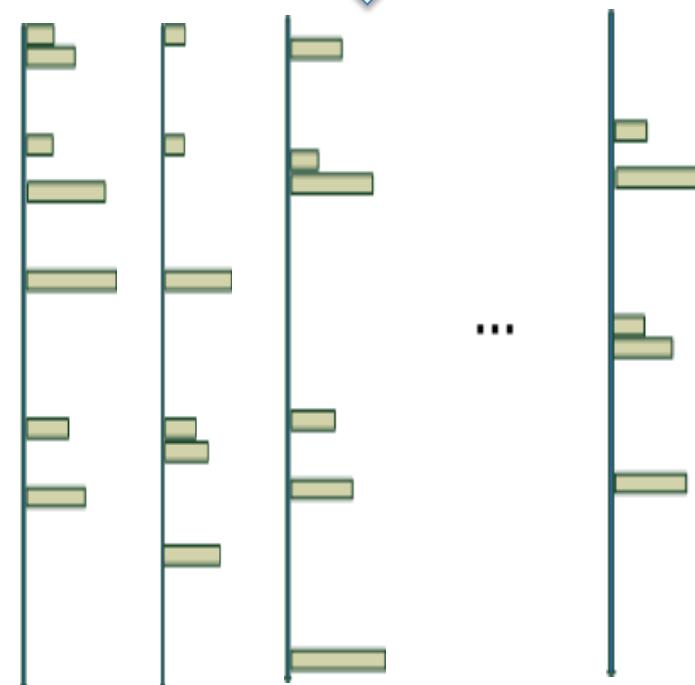


$$s = \operatorname{argmin}_s \|x - Bs\|^2 + \lambda \|s\|_1$$

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

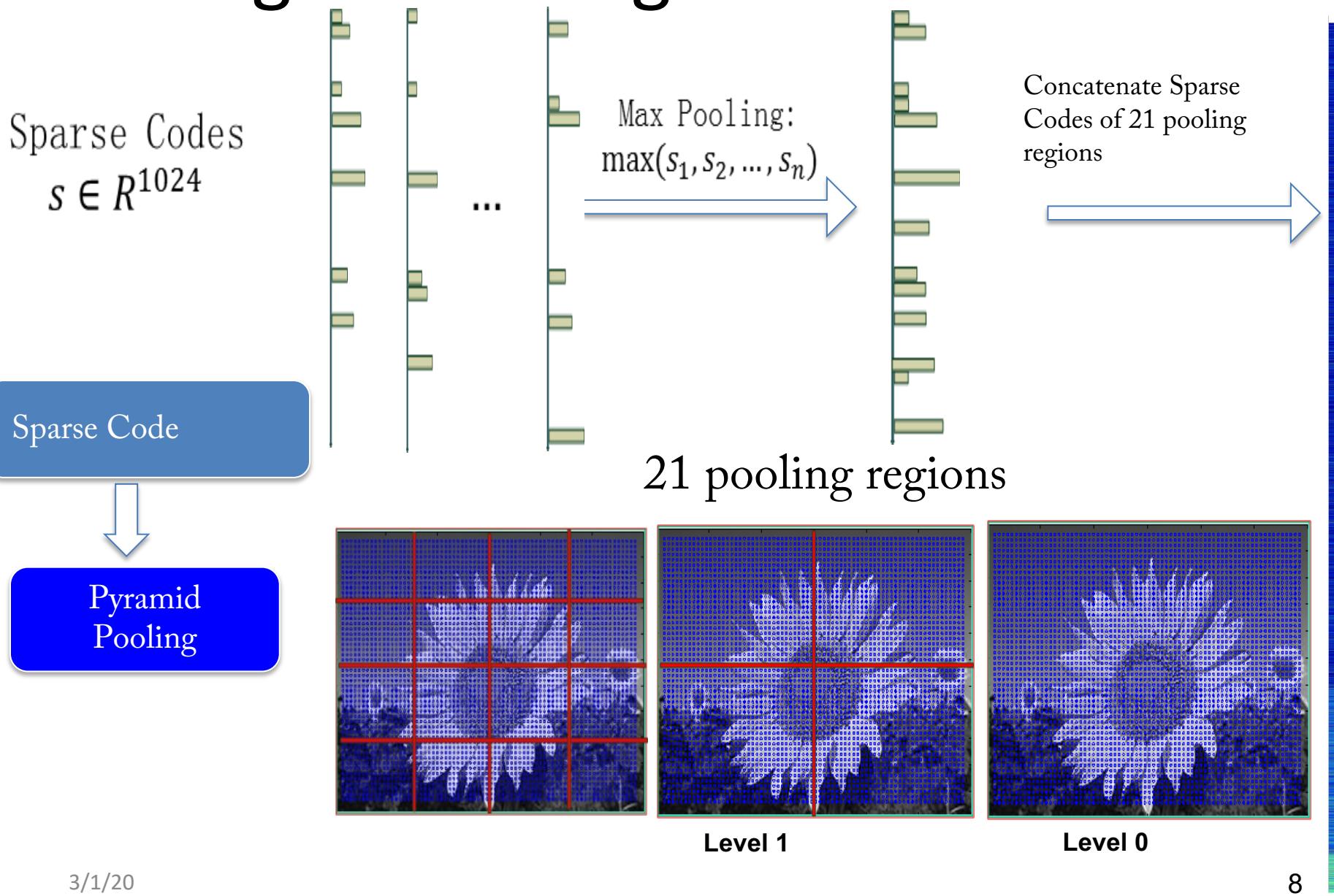
Sparse Code

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

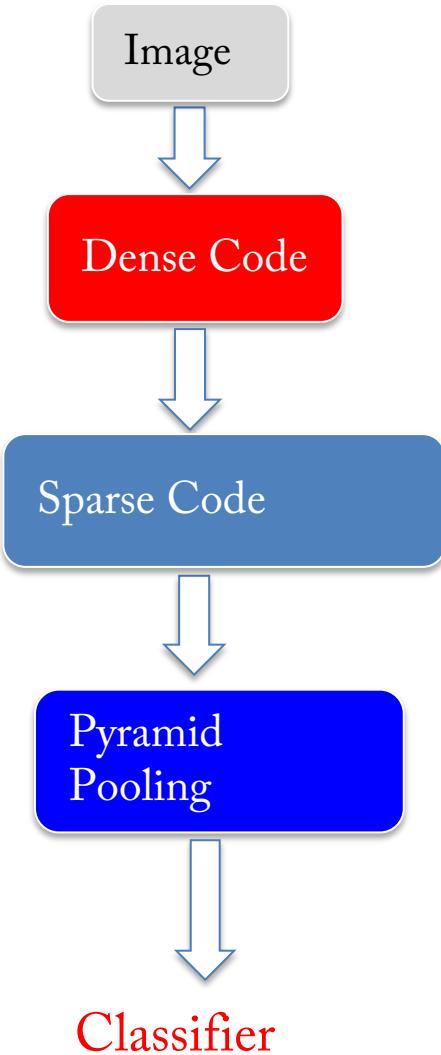


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Background: Bag of Visual words



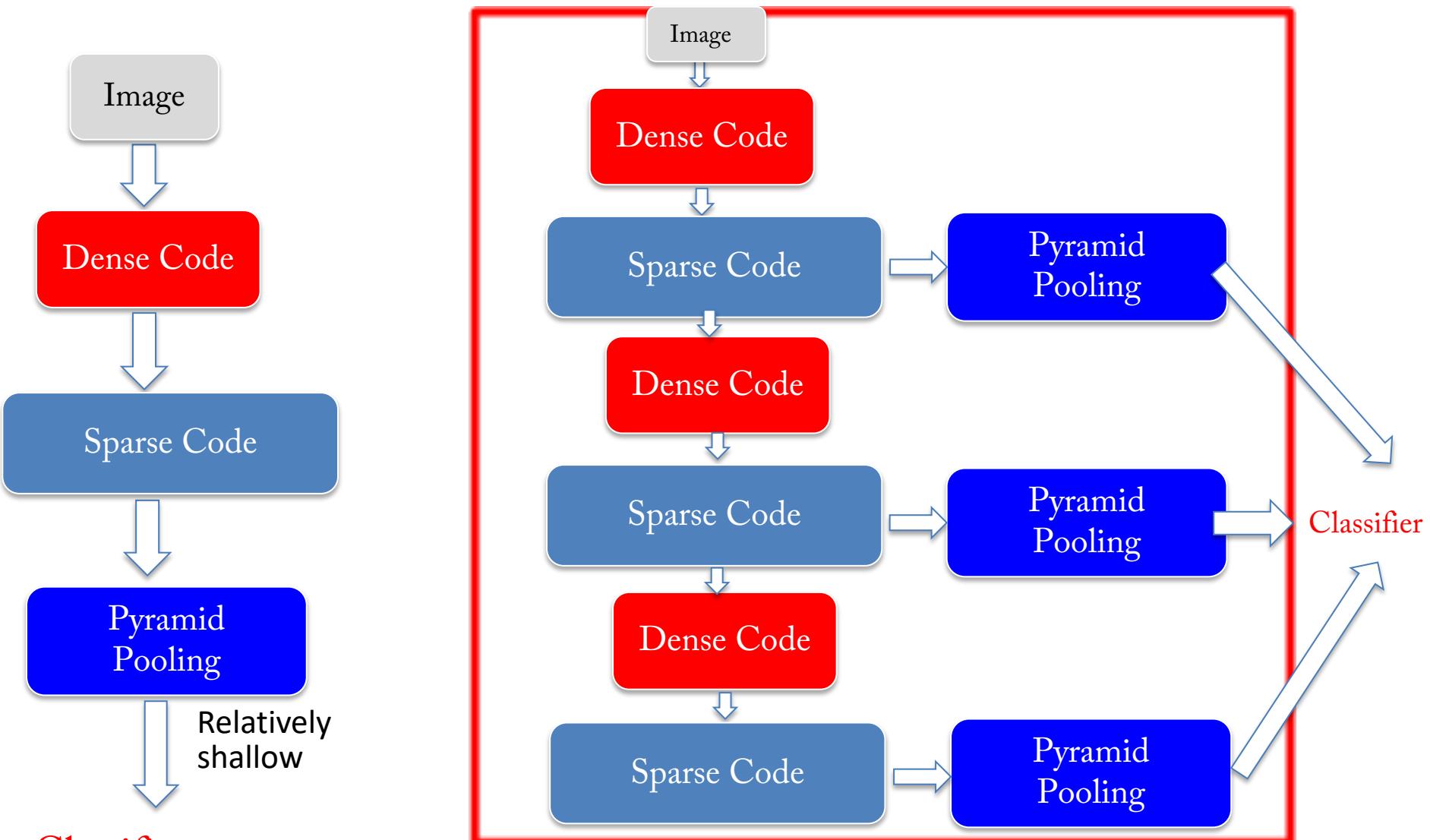
Background: Bag of Visual Words



Can we learn multiple layers of sparse representations?

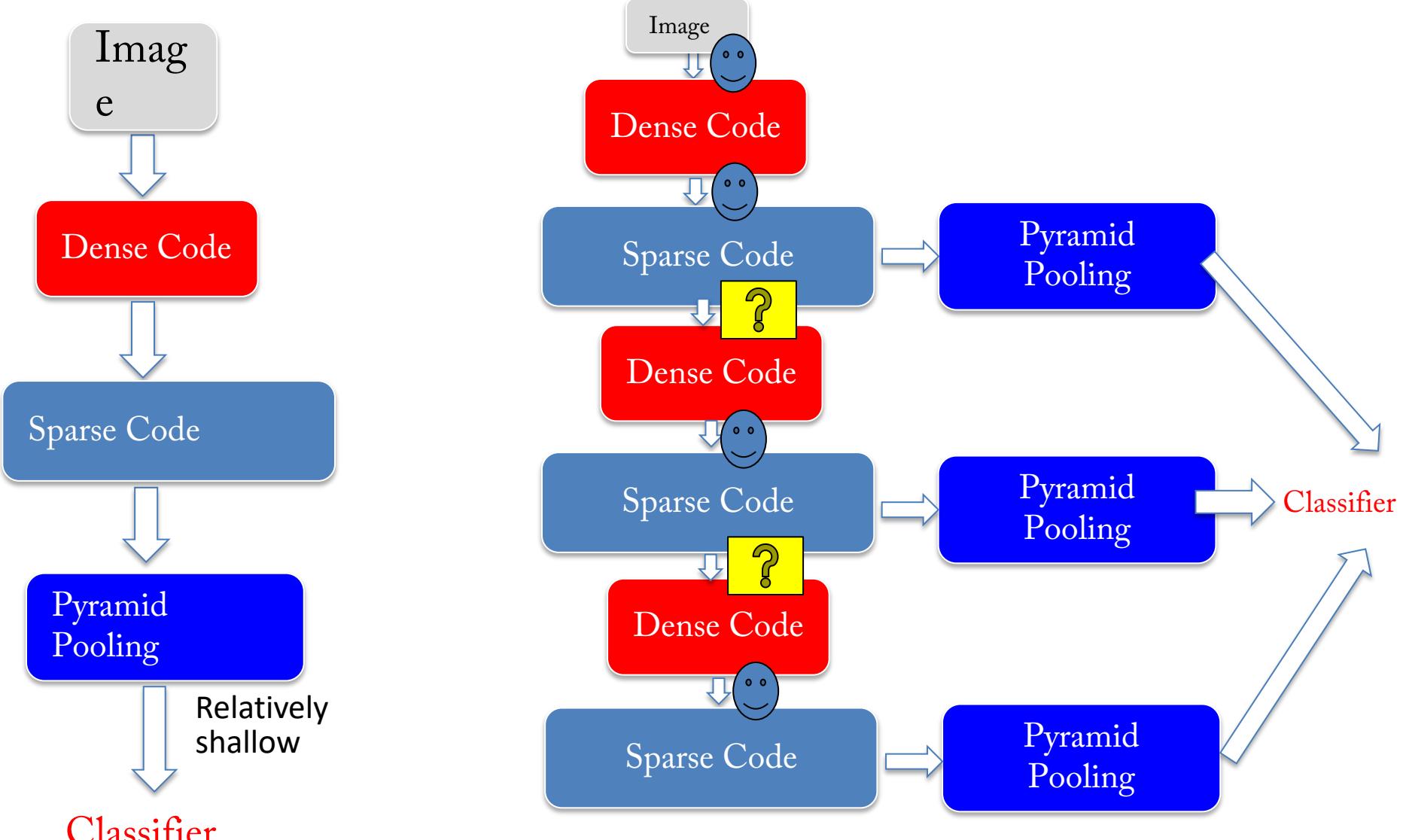
Relatively shallow

Deep Sparse Coding



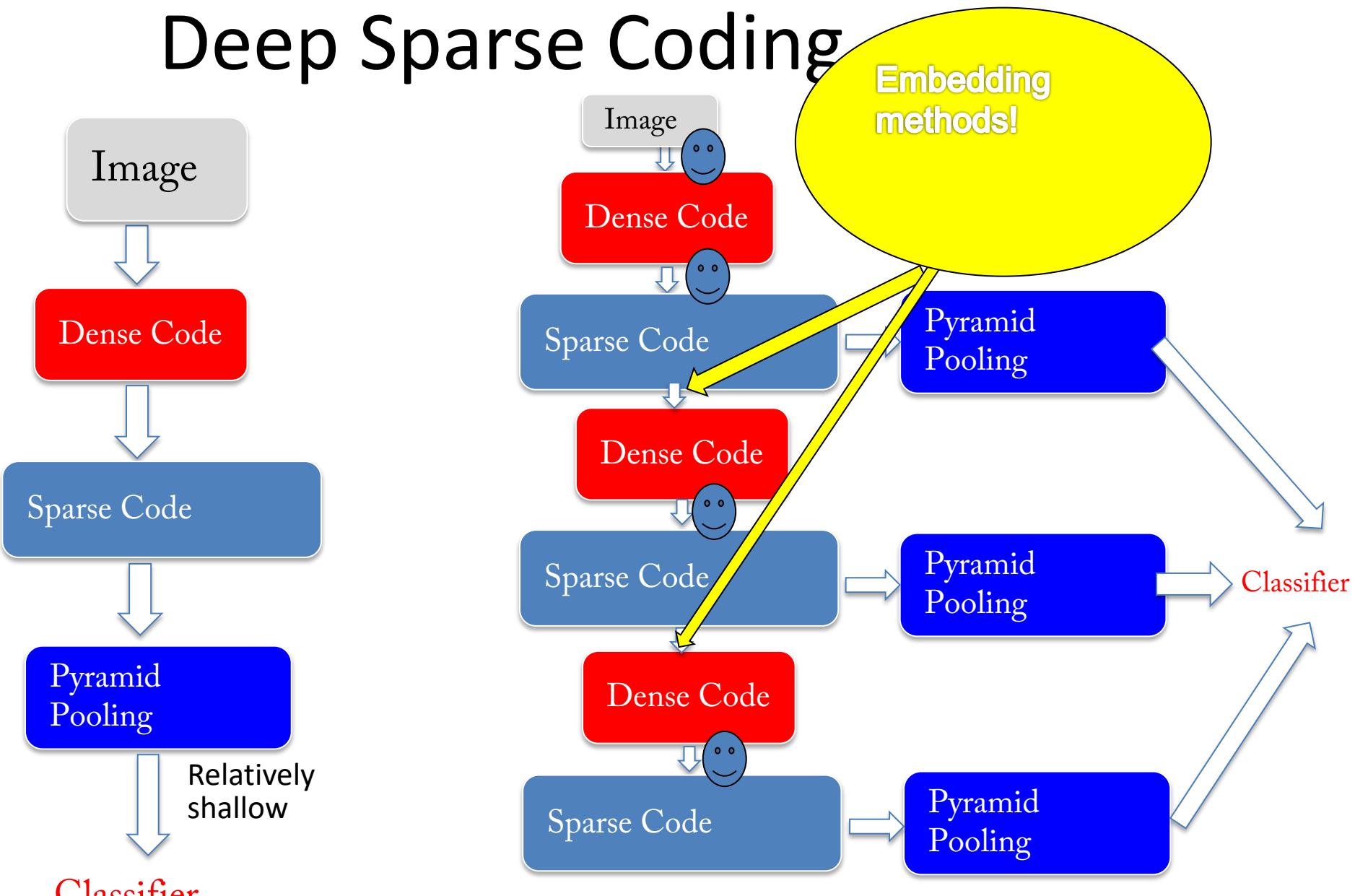
Classifier

Deep Sparse Coding



Unsupervised Deep Architecture!

Deep Sparse Coding

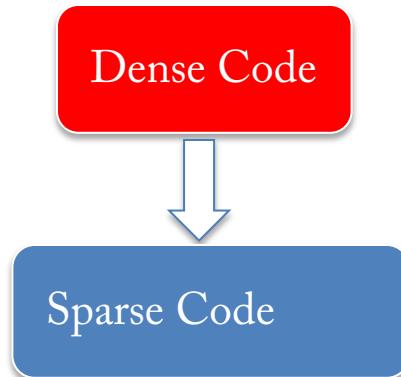


Unsupervised Deep Architecture!

Dimension expansion

Idea:

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



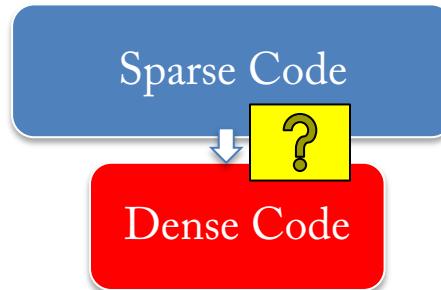
Method:

- Sparse coding

Dimension reduction

Idea:

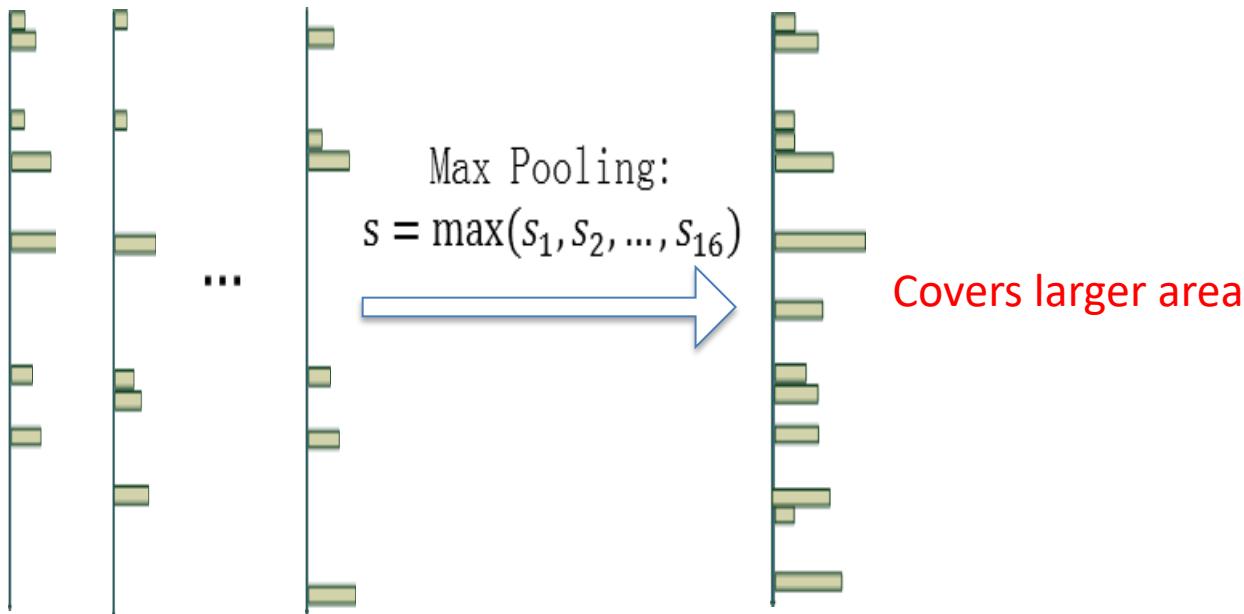
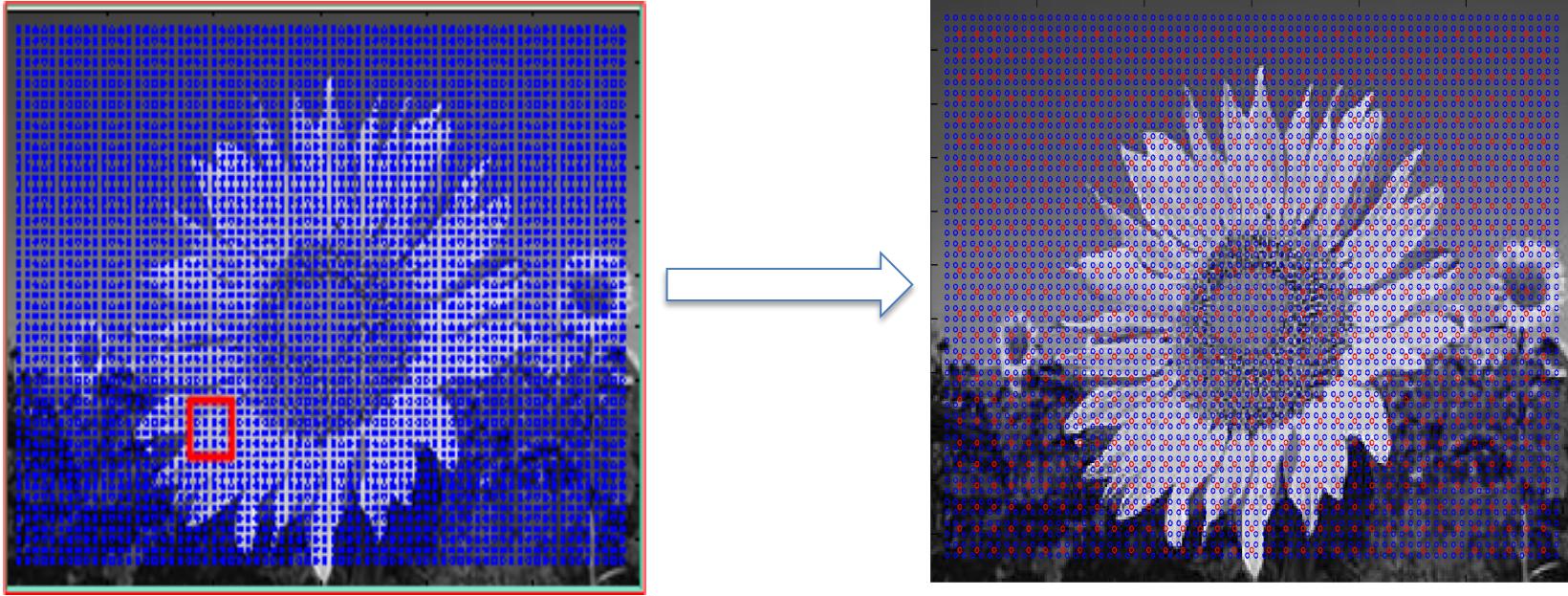
- Invariance
- Compositionality



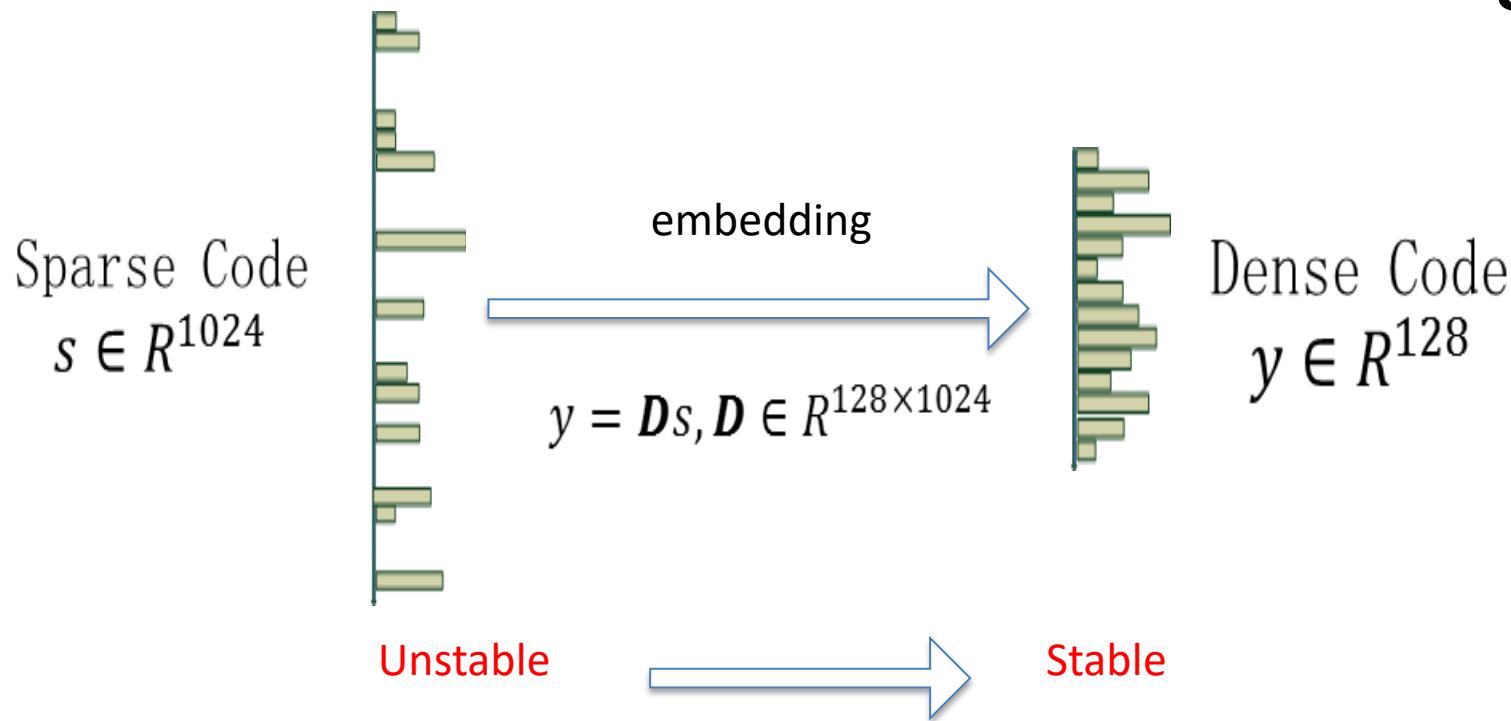
Method:

- Locally spatial pooling + Low dimensional embedding
- Both unsupervised

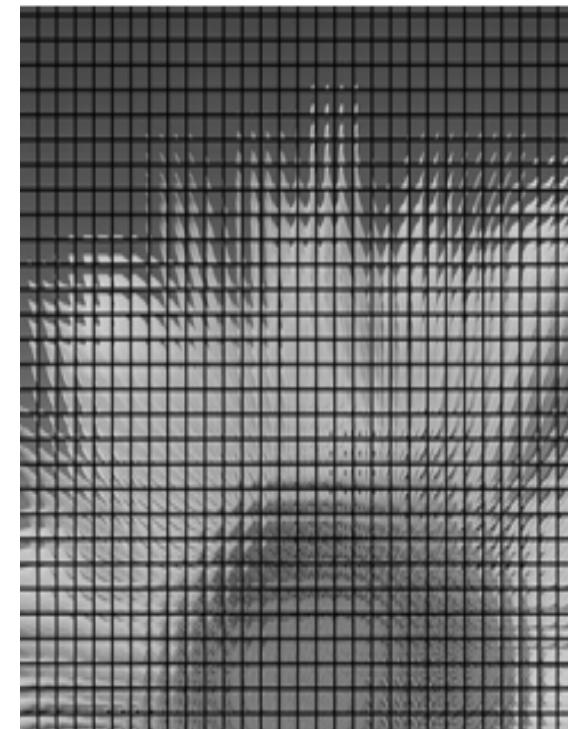
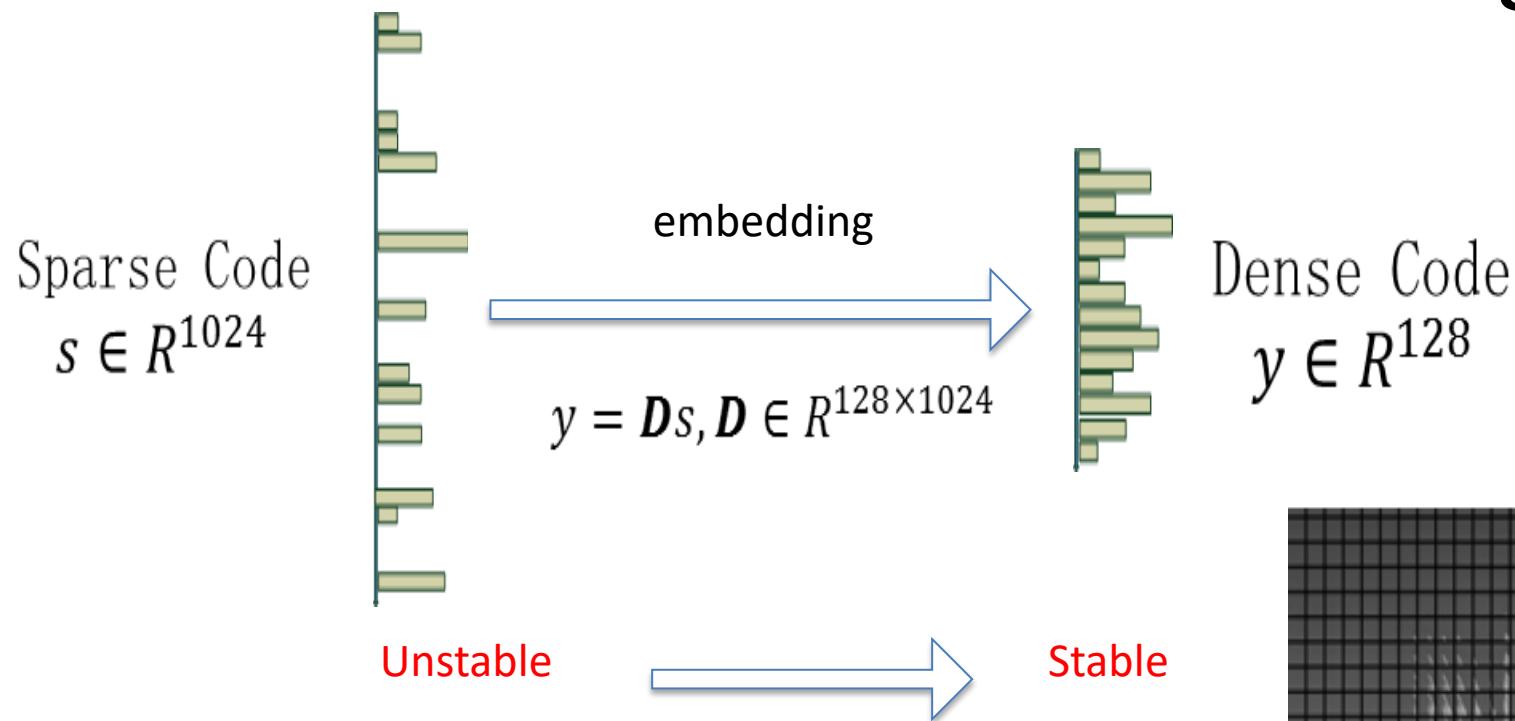
Locally spatial pooling



Low dimensional embedding

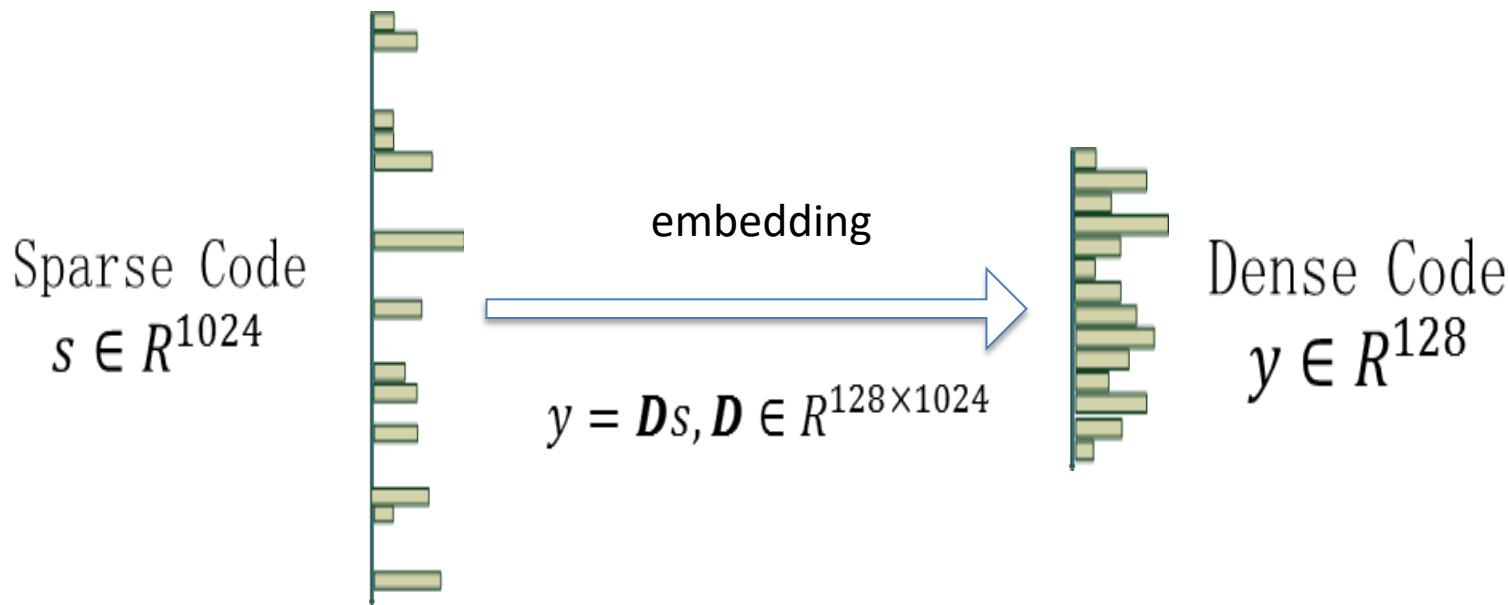


Low dimensional embedding



Idea: embedding with the help of spatial information

Low dimensional embedding



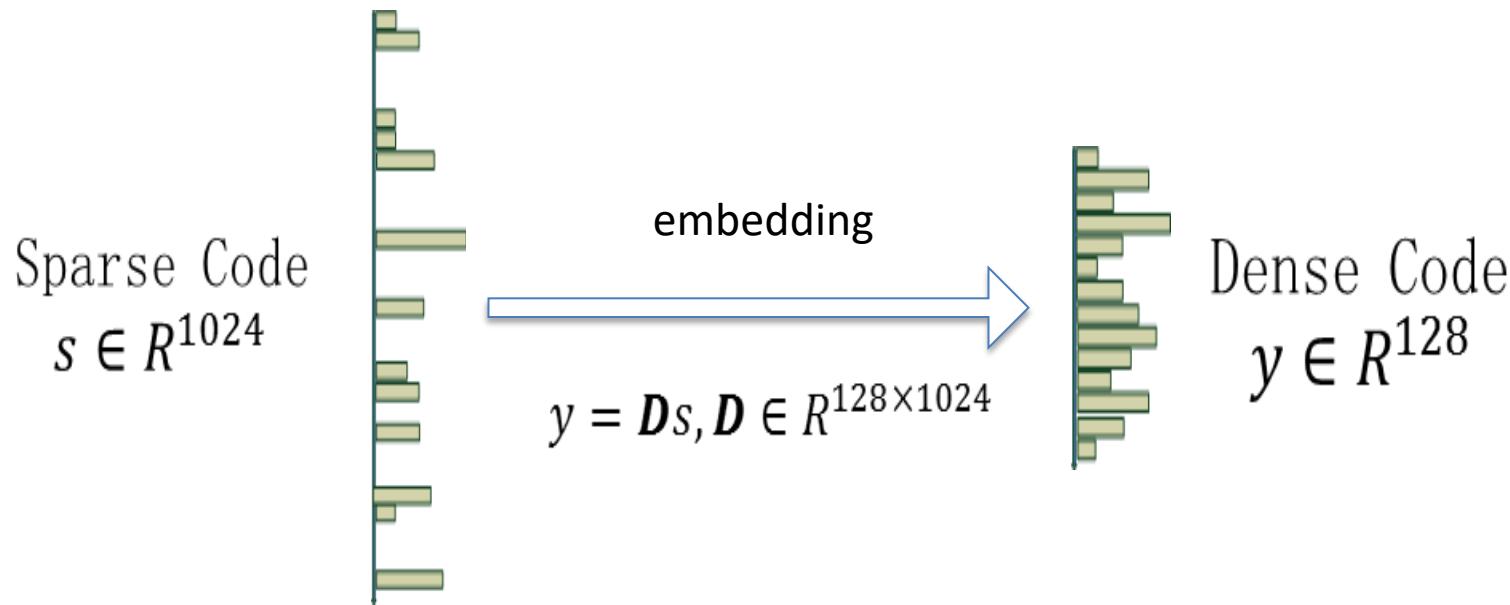
Dimensionality Reduction by Learning an Invariant Mapping (DR. LIM)

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

$w_{ij} = 1$, 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



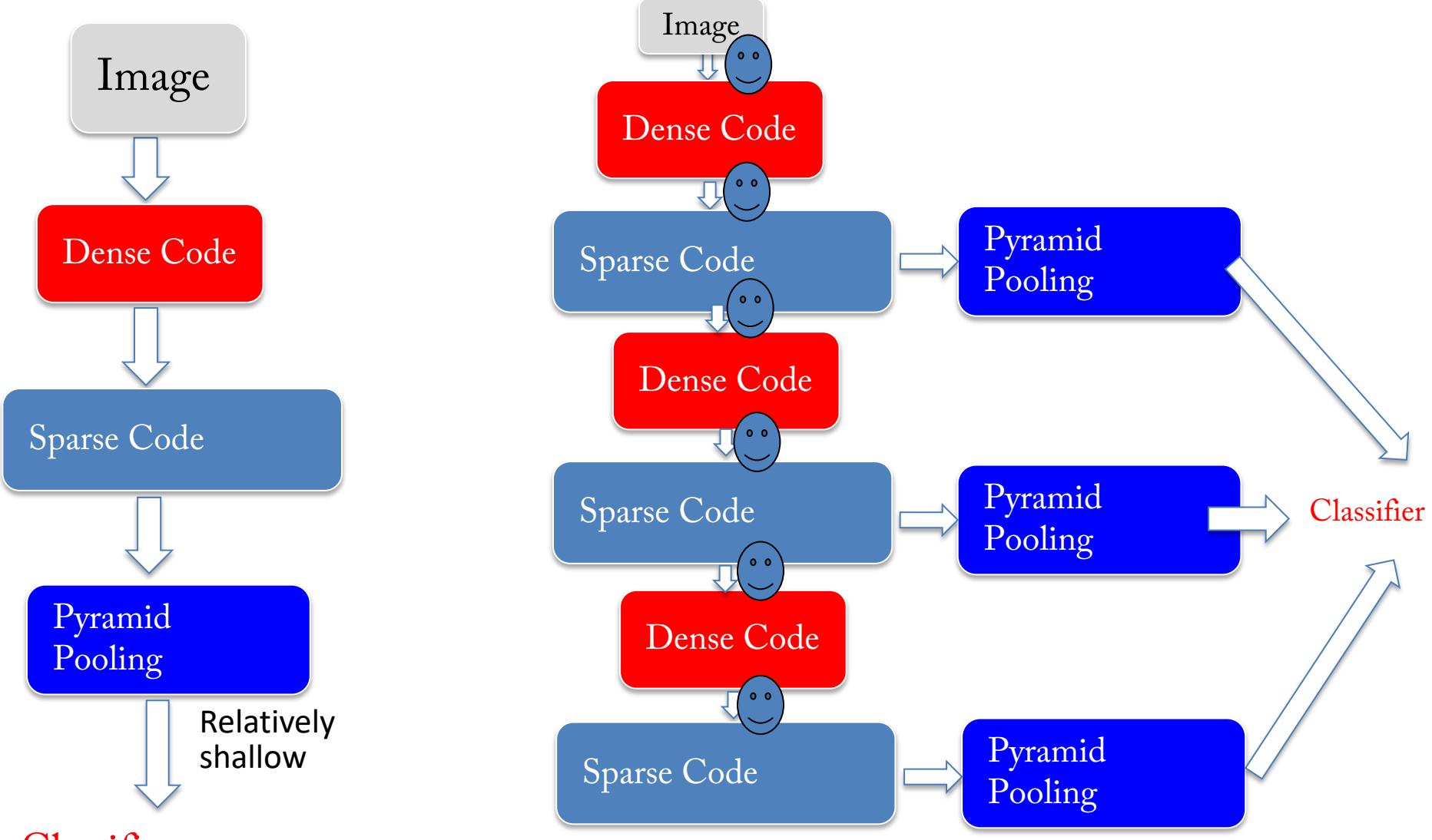
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Deep Sparse Coding



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