

UNSUPERVISED STACKED CAPSULE AUTOENCODER FOR HYPERSPPECTRAL IMAGE CLASSIFICATION

Erting Pan¹, Yong Ma^{1,2}, Xiaoguang Mei^{1,2}, Fan Fan^{1,2}, and Jiayi Ma^{1,2}

¹Electronic Information School,
Wuhan University,
Wuhan 430072, China

² Institute of Aerospace Science
and Technology, Wuhan university,
Wuhan, 430072, China



Outline

◆ Challenge & Motivation

- Challenge
 - Insufficient labeled samples
 - High computational cost
 - Unavoidable loss of valuable information
- Motivation
 - Capsule
 - Structural characteristics of hyperspectral data

◆ Methodology

- Overview
- Part Capsule Autoencoder
- Object Capsule Autoencoder

● Experiments

- Quantitative and qualitative results on Pavia Center dataset

◆ Conclusion



Challenge & Motivation

- Superiority performance of existing deep learning methods

- **Challenge**

- Insufficient labeled samples **caused by expensive cost in collecting and labeling hyperspectral data.**
- High computational cost **caused by complex network structure**
- Unavoidable loss of valuable information in CNNs **caused by the pooling layer**

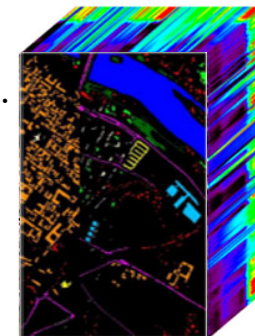
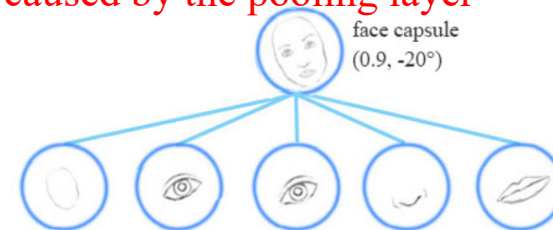
- **Motivation**

- Capsule

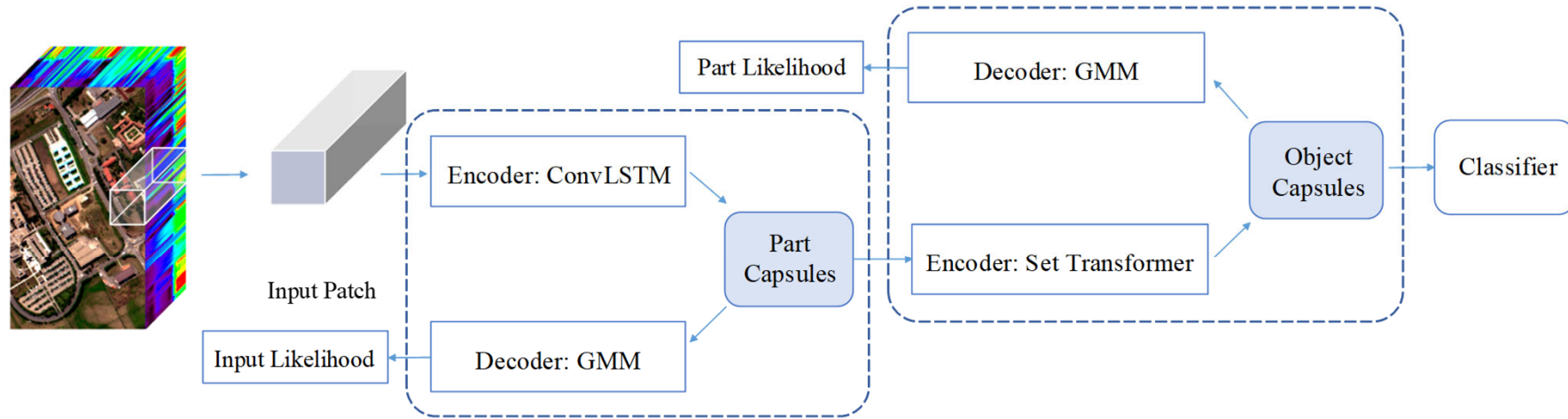
- It interprets an object by **the geometrical arrangements of parts**
- Representing attributes such as presence, position, scale orientation...

- Hyperspectral data

- Spectral dimension —— inherent properties of land-covers
- Spatial domain —— Spatial dependence and distributions



Methodology



Origin HSI data

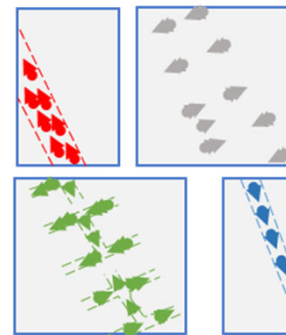


Part Capsule AutoEncoder



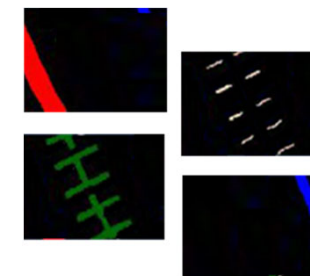
Parts with their pose

Object Capsule AutoEncoder



Objects predictions

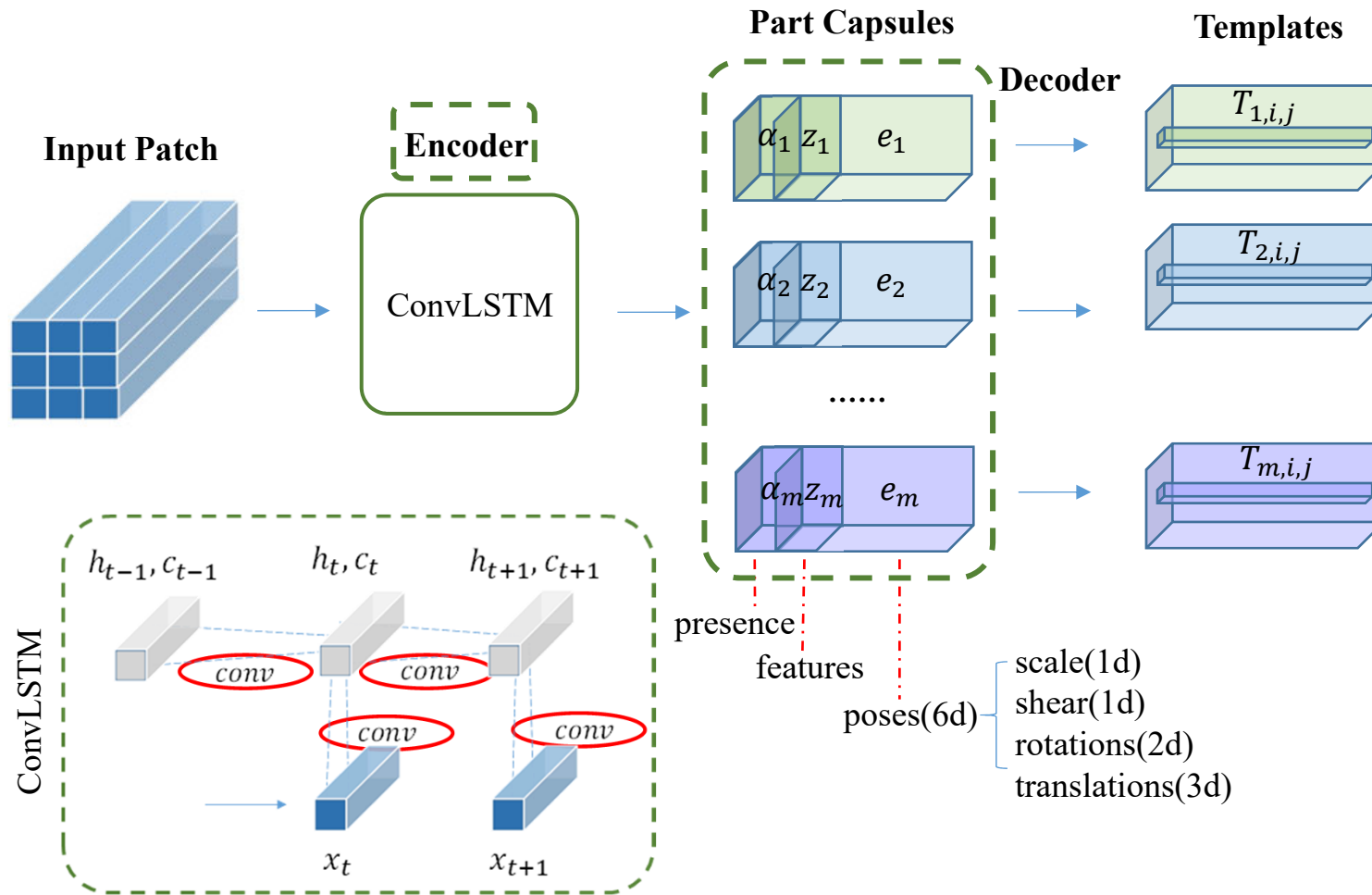
Result



Objects

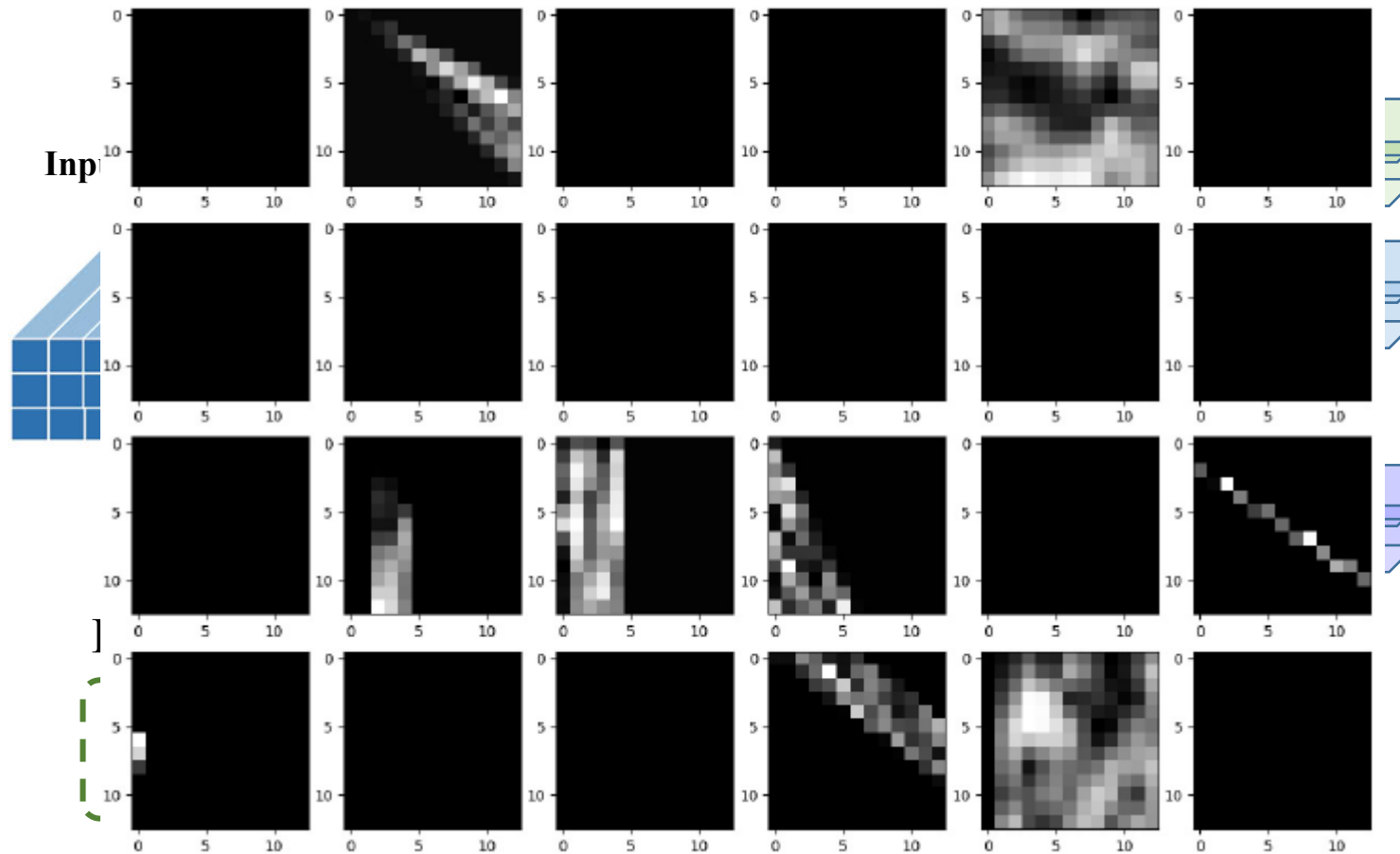


Part Capsule Autoencoder



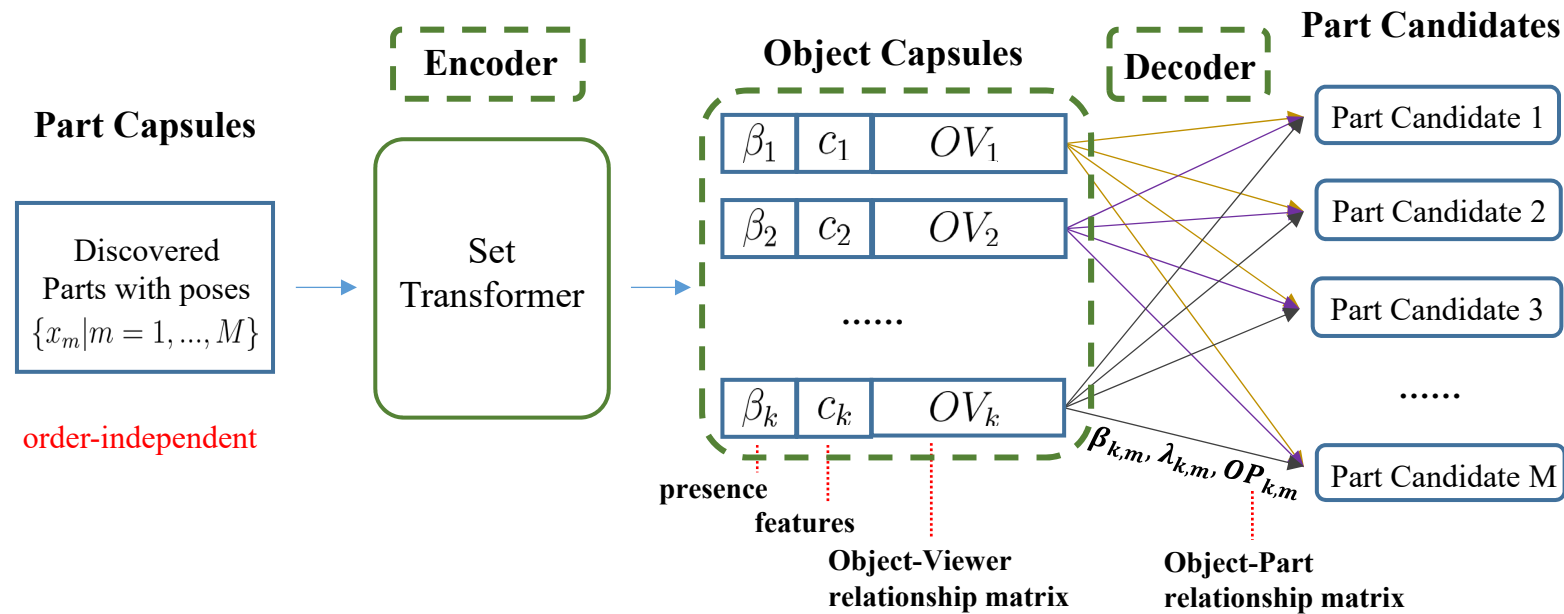


Part Capsule Autoencoder





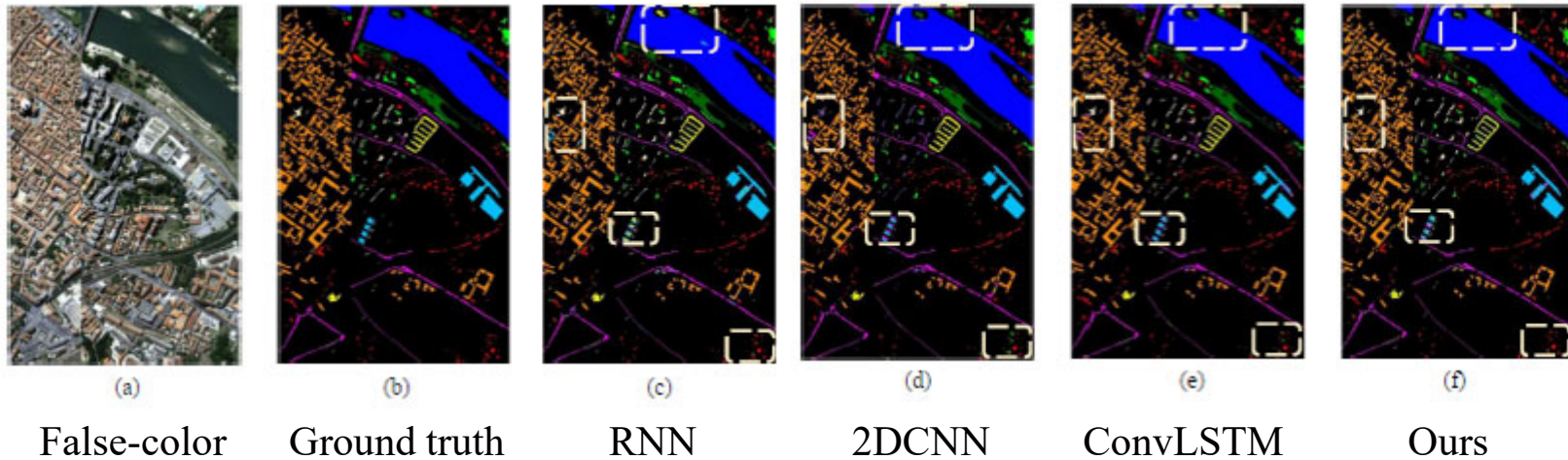
Object Capsule Autoencoder



$$\text{GMM: } p(z_m, \alpha_m) = \prod_{m=1}^M \left[\sum_{k=1}^K \frac{\beta_k \beta_{k,m}}{\sum_i \beta_i \sum_j \beta_{i,j}} \mathcal{N}(z_m | \mu_{k,m}, \lambda_{k,m}) \right]^{\alpha_m}$$

$$\mu_{k,m} = OV_k \cdot OP_{k,m}$$

Experiments



Label	KNN	RNN	2DCNN	ConvLSTM	our method
OA	88.29	90.75	92.75	93.47	96.39
AA	89.14	93.88	87.33	96.25	97.16
Kappa	82.59	87.66	89.94	90.16	95.88



Conclusion

Unsupervised stacked capsule autoencoder

- Modeling the spatial distribution of different land-covers
- Explore Object-Viewer and Object-Part correlations in HSI
- Employ ConvLSTM as the encoder of Part Capsule Autoencoder



THANKS FOR YOUR ATTENTION

Erting Pan
Electronic Information School
Wuhan University
panerting@whu.edu.cn