### Week 4

### Progress during 20-26th July

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#### Contents

- Unsupervised learning in star-galaxy classification/segmentation
  - Manifold learning for hidden layer clustering
  - Hypercolumns with kmeans for segmentation
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  - Applied to SDSS field datas (large scale)
  - TODOs



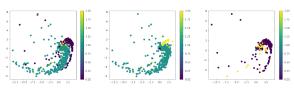
## Progress in this week

- Tried manifold learning for hidden layer clustering
- 2 Tried Hypercolumns with kmeans for segmentation
- Applied to SDSS field datas (large scale)
- Tried different VAE structures and hyperparameters for better performance
- 5 Tried different normalization method
- Opening Promising direction: hidden layer explaination



## Manifold learning for hidden layer clustering

### Manifold learning use geometric distance instead of Euclidian distance in clustering



Manifold Learning with 1000 points, 6 neighbors











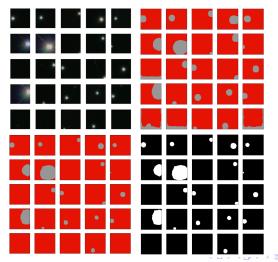






## Hypercolumns with kmeans for segmentation

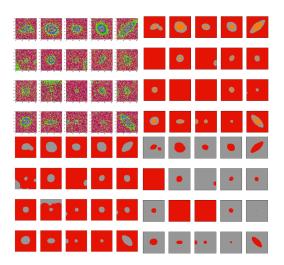
Then I use Hypercolumns to perform segmentation: (SDSS field data, cut to 64\*64 RGB)





### Hypercolumns with kmeans for segmentation

SDSS 64\*64 RGB image (color: different normalization method)





# Applied to SDSS field datas (large scale)

Apply the method above to the original field image (256\*256 RGB image), kmeans with 3 kernels:





### **Todos**

- Find knowledge from hidden layer(size, luminosity, gravity?) Figures below show the hidden variables when using the same VAE structure with different normalization methods.
- Try different normalization method
- 6 Hyperparameter adjustment
- Use other clustering methods instead of K-means
- Artificial convolution kernels

