



Graph Learning Network

A Structure Learning Algorithm

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🔖 gitlab.com/mipl/graph-learning-network/

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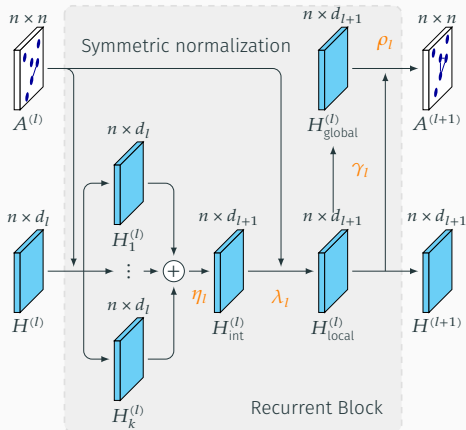
Existing Graph Neural Networks

-  Do not support extreme variations over the graph
-  Fixed graph's structure (over the network)




We overcome these issues by ...

- ✂ Iteratively predict features and adjacency
- 🔄 Repeat (until diminishing returns, five times according to our experiments)

Proposal: GLN



Loss Functions:




-  Intersection over Union (IoU) of adjacency
-  Class-balanced Cross-Entropy (HED)
-  Regularization

$$H_{\text{int}}^{(l)} = \sum_{i=1}^k \sigma_l(\hat{A}^{(l)} H_i^{(l)} W_i^{(l)})$$

$$H_{\text{local}}^{(l)} = \sigma_l(\hat{A}^{(l)} H_{\text{int}}^{(l)} U^{(l)})$$

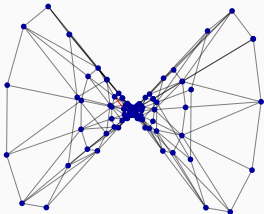
$$H_{\text{global}}^{(l)} = \sigma_l(H_{\text{local}}^{(l)} Z^{(l)})$$

$$A^{(l+1)} = \sigma_l(M^{(l)} H_{\text{local}}^{(l)} Q^{(l)} H_{\text{global}}^{(l)} M^{(l)\top})$$

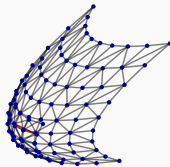
-  Learnable matrices
-  Non linearities
-  Embedding functions

We can predict surfaces

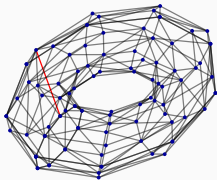
Elliptic hyperboloid



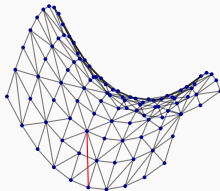
Elliptic paraboloid



Torus

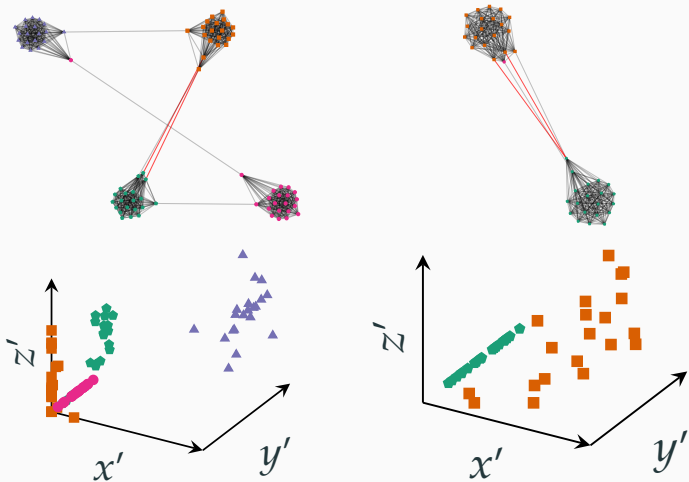


Saddle



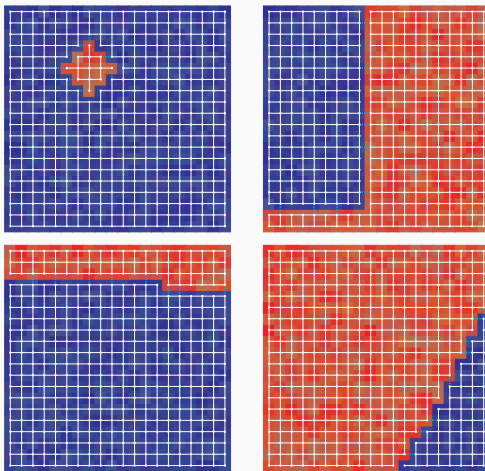
i Not predicted edges (FN), extra predicted edges (FP), and correctly predicted ones.

...and communities



i Not predicted edges (FN), extra predicted edges (FP), and correctly predicted ones.

...and segment (simple) images



i Not predicted edges (FN), extra predicted edges (FP), and correctly predicted ones.

Lets talk for more info

 Code available:

gitlab.com/mipl/graph-learning-network/

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