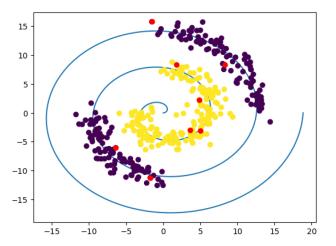
Graphical Semisupervised Learning We should probably write something here

Rohin Gilman Alex Johnson Kaitlynn Lilly May 22, 2023



$\begin{array}{c} \textbf{Introduction to Problem} \\ \textbf{Spiral} \end{array}$

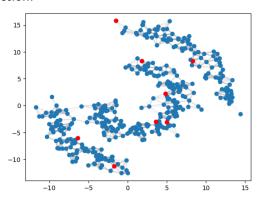
• Goal in SSL is, given $\{x_i, y_i\}_{i=1}^M$ and $\{x_i\}_{i=M+1}^N$, to predict $y_i(x_i)$ for $i = M+1, \ldots, N$ where $M \ll N$.



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- We leverage spectral geometric properties of the graph Laplacian matrix L.
- First, we construct our graphs by KNN or Proximity graph construction.



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Optimization Problem

Setup and loss functions

General optimization

$$ec{f^*} = rg \min_{ec{f} \in \mathbb{R}^m} \mathcal{L}(ec{f}; y) + \lambda ec{f}^T C^{-1} ec{f}$$

where \mathcal{L} is a loss function, either Probit or Regression loss in our case.

Probit loss

$$\mathcal{L}(\vec{f}, y) = -\sum_{j=1}^{M} \log \Psi(\vec{f_j} y_j)$$

Regression loss

$$\mathcal{L}(\vec{f}, y) = \sum_{i=1}^{M} \left(\vec{f_j} - y_j\right)^2$$

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Optimization Problem

Matérn kernel-regularization

General optimization

$$ec{f^*} = rg\min_{ec{f} \in \mathbb{R}^m} \mathcal{L}(ec{f}; y) + \lambda ec{f}^T C^{-1} ec{f}$$

where $C = (L + \tau^2 I)^{-\alpha}$, where L is the graph Laplacian.

 In class, we showed C is a kernel matrix. In fact, C belongs to a class of kernels called the Matérn family that have the form

$$K(x,y) = \kappa(\|x-y\|), \kappa(t) = \frac{2^{1-\nu}}{\Gamma(\nu)} \left(\sqrt{2\nu} \frac{t}{\gamma}\right) K_{\nu} \left(\sqrt{2\nu} \frac{t}{\gamma}\right)$$

where Γ is the Gamma function and K_{ν} is the modified Bessel function of the second kind.

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Three Cluster Case

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Conclusions and Further Questions

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