

Representation Focused Algorithm for Deep Networks (R., Jones & Harchaoui 2022)

Reduced objective

- Consider learning a feature representation $\phi(\cdot, \theta)$ and a linear predictor W on top of $\phi(\cdot, \theta)$,

$$\min_{\theta, W} \frac{1}{n} \sum_{i=1}^n \ell(W^\top \phi(x_i, \theta), y_i) + \Omega(\theta, W)$$

- For squared loss ℓ , penalty Ω , can define

$$f(\theta) := \min_W \frac{1}{n} \sum_{i=1}^n \ell(W^\top \phi(x_i, \theta), y_i) + \Omega(\theta, W)$$

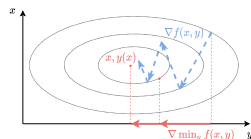
→ pseudo-likeli. [Besag \(1975\)](#) or Wiberg algo. [Wiberg \(1976\)](#)

Algorithm Idea

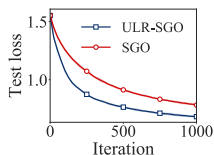
- Stochastic gradient descent on reduced objective $f(\theta)$
→ Biased oracles with bias controlled by mini-batch size
- Generalized to approx. minimizers for non-quad. losses
→ Gradient of $f(\theta)$ obtained by implicit diff.

[Blondel et al. \(2021\)](#)

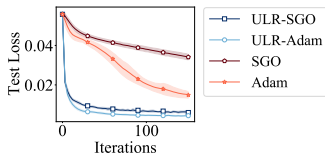
→ Can be plugged into e.g. Adam



Potentially circumvent oscillations



All-CNN on CIFAR10 multinomial loss



LeNet5 on MNIST squared loss

SGO: Stochastic Gradient Optimization

ULR-X: Proposed oracle with optim. algo. X