MASTER'S THESIS PRESENTATION

QINGYAO SUN

Department of Statistics The University of Chicago

Beyond Invariance: Test-Time Label-Shift Adaptation for Addressing "Spurious" Correlations

THURSDAY, February 2, 2023, at 2:30 PM Jones 111, 5747 S. Ellis Avenue

ABSTRACT

Spurious correlations, or correlations that change across domains where a model can be deployed, present significant challenges to real-world applications of machine learning models. However, such correlations are not always "spurious"; often, they provide valuable prior information for a prediction. Here, we present a test-time adaptation method that exploits the spurious correlation phenomenon, in contrast to recent approaches that attempt to eliminate spurious correlations through invariance. We consider situations where the prior distribution p(y,z), which models the dependence between the class label y and the "nuisance" factors z, may change across domains, but the generative model for features $p(\mathbf{x}|\mathbf{y},z)$ is constant. We note that this corresponds to an expanded version of the label shift assumption, where the labels now also include the nuisance factors z. Based on this observation, we train a classifier to predict $p(\mathbf{y},z|\mathbf{x})$ on the source distribution, and propose a test-time label shift correction that adapts to changes in the marginal distribution $p(\mathbf{y},z)$ using unlabeled samples from the target domain. We evaluate our method, which we call "Test-Time Label-Shift Adaptation" (TTLSA), on two different image datasets -- the CheXpert chest X-ray dataset and the Colored MNIST dataset --- and show a significant improvement over baseline methods.