

ADVERSARIAL MACHINE LEARNING

DEFENSE: VIRTUAL ADVERSARIAL TRAINING

Uses both labeled and unlabelled datapoints

Loss function:

Regularizer:

Objective Function:

LDS
$$(x_*, \theta) := D\left[p(y|x_*, \hat{\theta}), p(y|x_* + r_{\text{vadv}}, \theta)\right]$$

$$r_{\text{vadv}} := \underset{r; ||r||_2 \le \epsilon}{\text{arg max}} D\left[p(y|x_*, \hat{\theta}), p(y|x_* + r, \theta)\right],$$

where $x_* \in \{D_l, D_{ul}\}$

 $\mathbf{v}_{\mathrm{vadv}}(\mathcal{D}_{l}, \mathcal{D}_{ul}, \theta) := \frac{1}{N_{l} + N_{ul}} \sum_{x_{*} \in \mathcal{D}_{l}, \mathcal{D}_{ul}} \mathbf{v}_{ul}$

 $LDS(x_*, \theta).$

$$\ell(\mathcal{D}_l, \theta) + \alpha \mathcal{R}_{\text{vadv}}(\mathcal{D}_l, \mathcal{D}_{ul}, \theta),$$

Miyato, T., Maeda, S. I., Koyama, M., & Ishii, S. (2017). Virtual Adversarial Training: a Regularization Method for Supervised and Semi-supervised Learning.

NO DEFINITIVE ANSWER

DEFENSE: VIRTUAL ADVERSARIAL TRAINING

- Uses both labeled and unlabelled datapoints
- Loss function:

$$LDS(x_*, \theta) := D\left[p(y|x_*, \hat{\theta}), p(y|x_* + r_{vadv}, \theta)\right]$$

$$r_{vadv} := \underset{r; ||r||_2 \le \epsilon}{arg \max} D\left[p(y|x_*, \hat{\theta}), p(y|x_* + r, \theta)\right],$$
where $x_* \in \{D_l, D_{ul}\}$

Regularizer:

$$\mathcal{R}_{\text{vadv}}(\mathcal{D}_l, \mathcal{D}_{ul}, \theta) := \frac{1}{N_l + N_{ul}} \sum_{x_* \in \mathcal{D}_l, \mathcal{D}_{ul}} \text{LDS}(x_*, \theta).$$

Objective Function:

$$\ell(\mathcal{D}_l, \theta) + \alpha \mathcal{R}_{\text{vadv}}(\mathcal{D}_l, \mathcal{D}_{ul}, \theta),$$