

Supplementary Material: Multi Label Loss Correction against Missing and Corrupted Labels

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Combining MLLSC with ASL loss function

MLLSC can be easily integrated into different types of multi-label loss functions. Here we provide the resulting loss function for MLLSC + ASL (Ridnik et al., 2021). ASL is an asymmetric multi-label loss which weights positive and negative labels differently (see Eq.4 in the main paper). We define MLLSC combined with ASL as

$$\begin{cases} \mathcal{L}_k^+ = \mathbb{1}(p_k > \tau) (1 - \mathcal{P}'_{k,m})^{\gamma_+} \log(\mathcal{P}'_{k,m}) + (1 - \mathbb{1}(p_k > \tau)) \mathcal{P}'_{k,m}{}^{\gamma_-} \log(1 - \mathcal{P}'_{k,m}) \\ \mathcal{L}_k^- = \mathbb{1}(p_k < \tau') \mathcal{P}'_{k,m}{}^{\gamma_-} \log(1 - \mathcal{P}'_{k,m}) + (1 - \mathbb{1}(p_k < \tau')) (1 - \mathcal{P}'_{k,m})^{\gamma_+} \log(\mathcal{P}'_{k,m}) \end{cases} \quad (1)$$

where γ_+ and γ_- are the focus parameters for positive and negative labels, and $\mathcal{P}'_{k,m} = \max(p_k - m, 0)$ is the marginal probability with tunable hyper-parameter m . Besides, $\mathbb{1}(\cdot)$ is an indicator function. We set $\gamma_- = 4$, $\gamma_+ = 2$ and $m = 1$ according to the default values in (Ridnik et al., 2021).

References

Tal Ridnik, Emanuel Ben Baruch, Nadav Zamir, Asaf Noy, Itamar Friedman, Matan Protter, and Lihi Zelnik-Manor. Asymmetric loss for multi-label classification. In *ICCV*, pages 82–91. IEEE, 2021.