Robust multi-label learning for weakly labeled data

Author: Atanas Marinov Responsible professor: Lydia Chen Supervisors: Amirmasoud Ghiassi, Taraneh Younesian

Background information

- **Multi-label learning** (MLL) is a type of supervised learning in which each input example could be assigned to more than output label.
- Weak labels are a type of multi-label data corruption where not only all the relevant (true) labels are presented but also some of the irrelevant ones.

Bear

Weak

labels





Correct

labels

Example image

Overfitting problem - The deep neural networks used to solve MLL problems could be quite complex and often have a huge capacity. This enormous capacity, however, could also be a negative, as they tend to eventually **overfit** the undesirable corrupted labels.

Method: Co-ASL

Co-ASL solves the problem by combining the most prominent solutions in the two fields - **ASL**[1], the state-of-the-art MLL approach, and **Co-teaching**[2], an eminent robust training strategy.

- Initialize the two peer networks;
- The data is fed in a mini-batch manner;
- Each network forwards the entire minibatch and remembers the portion with the lowest loss (small-loss instance);
- The networks swap their small-loss instances and adjust their weights using only the instance of their peer.



Results

- **MS-COCO**[3] was used as a dataset for the experiments.
- It is a clean dataset and, therefore, an **artificial noise** was injected into it.
- The experiments were running for 80 epochs.

Conclusions

Co-ASL proves to be an MLL algorithm robust for weakly labeled data.

- Does not overfit even after 80 epochs.
- Achieves an improvement of **8-9 mAP** score over ASL.





CSE3000

fUDelft

80

70

60

20

10

mAP score 40 30

[1] Emanuel Ben-Baruch, Tal Ridnik. *Asymmetric loss for multi-label classification*[2] Bo Han, Quanming Yao. *Coteaching: Robust training of deep neural networks*[3] Tsung-Yi Lin, Michael Maire. *Microsoft COCO: common objects in context*

mAP score 25% noise

1 6 11 16 21 26 31 36 41 46 51 56 61 66 71 76

Epochs

Co-ASL1 Co-ASL2 ASL