

Data Shapely based Auto-labeling algorithm

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INTRODUCTION

- Training deep neural networks for health care tasks requires a large amount of correctly labeled data
- Manually annotation requires highly expertise domain knowledge, and hence is very expensive and time-consuming in medical domain
- Little research efforts have been devoted

AIM

 In this study, we developed an automated method to label data using the data Shapley algorithm.

METHODS

 Given a dataset D, the data value of data point i can be characterized by its Data Shapley [2] as following:

$$\phi = C \sum_{S \subset D - \{i\}} \frac{L(S \cup \{i\}) - L(S)}{\binom{n-1}{|S|}}$$

where L is the associated loss function

- Suppose we have a small amount of welllabeled dataset, and a large amount of unlabeled dataset.
- 1) Multiple copies of unlabeled dataset are generated with different pseudo labels
- 2) Combine with above labeled dataset

METHODS (CONTINUED)

3) One epoch of training from scratch, and record the relative loss change

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- 4) Repeated 3) by 1000 times, the averaged loss change can be considered as the negative Shapley value.
- 5) The copy with largest Shapley value indicates the correct label
- 6) Repeat this procedure for many times, a large amount of unlabeled dataset can be freely labeled by using only a small amount of well-labeled dataset.

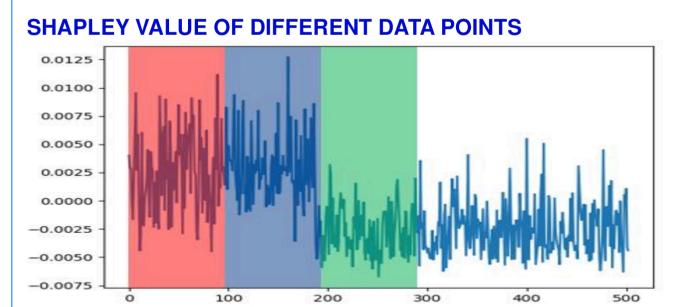
DATASETS

- The Osteosarcoma Tumor (OSTU) Identification Dataset with 344 images/3 labels was used.
- The OSTU datasets are split into 220(labeled)/30(test)/94(unlabeled) datasets.
- To explore the performance dependence on the size of the labeled dataset, different amount of labeled images are separately used as the labeled dataset.

CONCLUSIONS

- An auto-labeling algorithm was developed with more than 90% top-1 accuracy by using the data Shapley value.
- The auto-labeling performance can be further improved as we added more labeled images.

RESULTS



- The shading region represents unlabeled datasets.
- Green are correct label; the other two colors have wrong label copies.

EFFECT OF LABELED DATASET SIZE ON THE AUTO-LABELING ACCURACY

| Labeled dataset size | 0 | 31 | 63 | 94 | 126 | 157 | 189 | 220 |
|----------------------|-----|-----|-----|-----|-----|-----|-----|-----|
| Top-1 accuracy | 69% | 74% | 72% | 91% | 96% | 90% | 98% | 95% |

REFERENCES

- Shapley, L. S., Roth, A. E. et al. The Shapley value: essays in honor ofLloyd S. Shapley (Cambridge University Press, 1988).
- Ghorbani, A. and J. Y. Zou (2019). "What is your data worth? Equitable Valuation of Data." arXiv 1904.02868v2.

