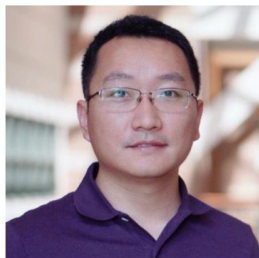


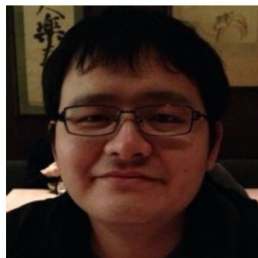
The 2nd Learning from Imperfect Data (LID) Workshop

Opening Remark

Welcome!



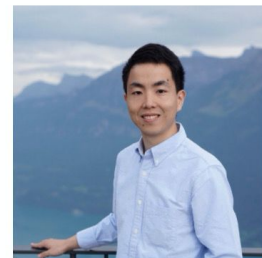
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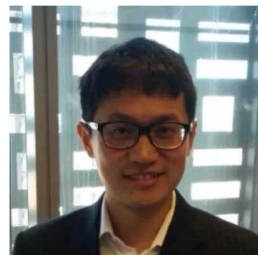
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Outline

Challenges

Motivation

Invited Talks

Oral Presentations

Best Paper Award

Sponsor

Motivation

We are often facing the problem that we are not able to acquire enough amount of perfect annotations (e.g., object bounding boxes and pixel-wise masks) for reliable training models.

A semantic segmentation system with 80+% mIOU performance on the benchmark dataset does not necessarily mean that it would not mislead the self-driving car crash into a white-flipped-over truck.



[source: cityscape dataset.](#)



[source:https://www.slashgear.com/](https://www.slashgear.com/)

Motivation

Ideal Machine Learning Problems	Real-world Applications
<ul style="list-style-type: none">● IID (Independent and identically distributed) assumption● Same distribution for training and test● Distributions fixed over time	<ul style="list-style-type: none">● Sample points are not drawn IID (e.g. lots of images are sampled from same video sequences)● Training sample is biased (e.g. perception dataset rarely contains the flipped-over truck on the road)● Training points with uncertain labels (e.g. imperfect polygon for segmentation, image-level label for object detection, mistakes in the annotations)● Multiple training sources (e.g. combine multiple datasets together for training)● Distribution may drift with time

Motivation

In the 2nd Learning from Imperfect Data challenge, we focus on Weakly supervised learning, which refers to address the challenging pattern recognition tasks by learning from weak or imperfect supervision, for example weakly supervised learning for semantic segmentation, scene parsing, or object localization.

Motivation

We organize this workshop to investigate current ways of building industry level AI system relying on learning from imperfect data.

Challenges: Weakly-supervised Semantic Segmentation (Track 1)

Dataset: ILSVRC-LID-200

Descriptions: This dataset is built upon the object detection track of ILSVRC, which totally includes 456,567 training images from 200 categories. We provide pixel-level annotations for 15K images, including 5,000 and 10,000 images for validation and testing, respectively.

Supervision in Training: Image-level labels

Target in Testing: Perform pixel-wise classification, i.e. semantic segmentation

The Key Eval Metric: mIoU

Challenges: Weakly-supervised Scene Parsing (Track 2)

Dataset: ADE20K-LID

Descriptions: This dataset is built upon the ADE20K dataset. There are 20,210 images in the training set, 2,000 images in the validation set, and 3,000 images in the testing set. We provide the additional point-based annotations on the training set.

Supervision in Training: one pixel annotation for each independent instance

Target in Testing: Perform pixel-wise classification, i.e. scene parsing

The Key Eval Metric: mIoU

Challenges: Weakly-supervised Object Localization (Track 3)

Dataset: ILSVRC

Descriptions: This dataset is built upon the ILSVRC dataset, including 1.2 million images for training. we annotate pixel-level masks of 44,271 images (validation/testing: 23,151/21,120)^[1] to facilitate the evaluation to be performed in a direct way

Supervision in Training: Image-level labels

Target in Testing: Highlight the target object in the class activation map

The Key Eval Metric: Peak IoU

[1] Rethinking Localization Map: Towards Accurate Object Perception with Self-Enhancement Maps. X Zhang, etc. arXiv preprint.

Challenge Results

10 Participants

Track1: Weakly-supervised Semantic Segmentation Challenge

1st	Guolei Sun, Wenguan Wang, Luc Van Gool. ETH Zurich
2nd	Wonho Bae*, Junhyug Noh*, Jinhwan Seo, and Gunhee Kim. Seoul National University, Vision & Learning Lab
3rd	Mariia Dobko, Ostap Viniavskyi, Oles Dobosevych. The Machine Learning Lab at Ukrainian Catholic University, SoftServe

1 Participant

Track2: Weakly-supervised Scene Parsing Challenge

1st	Hao Zhao, Ming Lu, Anbang Yao, Yiwen Guo, Yurong Chen, Li Zhang. Tsinghua University
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5 Participants

Track3: Weakly-supervised Object Localization Challenge

1st	Wonho Bae*, Junhyug Noh*, Jinhwan Seo, and Gunhee Kim. Seoul National University, Vision & Learning Lab
2nd	Chuangchuang Tan ^{1*} , Tao Ruan ^{1*} , Guanghua Gu ² , Shikui Wei ¹ , Yao Zhao ¹ . ¹ Beijing Jiaotong University, ² Yanshan University
3rd	Zhendong Wang, Zhenyuan Chen, Chen Gong. Nanjing University Of Science and Technology, LEAP Group@PCA Lab

Oral Presentations

Guolei Sun ETH Zurich

Oral 1: [The 1st Place of Track-1](#) Mining Cross-Image Semantics for Weakly Supervised Semantic Segmentation

Mariia Dobko UCU & SoftServe Team

Oral 2: [The 3rd Place of Track-1](#) NoPeopleAllowed: The 3 step approach to weakly supervised semantic segmentation

Hao Zhao Intel

Oral 3: [The 1st Place of Track-2](#) Pointly supervised Scene Parsing with Uncertainty Mixture

Wonho Bae Seoul National University

Oral 4: [The 1st Place of Track-3 & The 2nd Place of Track-1](#) Revisiting Class Activation Mapping for Learning from Imperfect Data

Chuangchuang Tan Beijing Jiaotong University & Mepro Team

Oral 5: [The 2nd Place of Track-3](#) Dual Gradients Localization framework for Weakly Supervised Object Localization

Zhendong Wang Nanjing University Of Science and Technology & LEAP Group@PCA Lab

Oral 6: [The 3rd Place of Track-3](#) Weakly Supervised Object Localization

Best Paper Award

The 2nd Learning from Imperfect Data (LD) Workshop

Best Paper Award

PRESENTED TO

Guolei Sun , Wenguan Wang , Luc Van Gool

FOR THEIR PAPER

MINING CROSS-IMAGE SEMANTICS FOR WEAKLY
SUPERVISED SEMANTIC SEGMENTATION

Sponsor

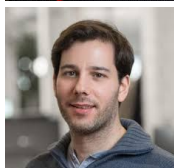


Invited Talks



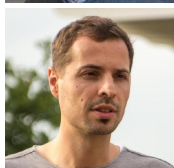
Alan Yuille Professor at Johns Hopkins University

Invited talk 1: You Only Annotate Once, or Never



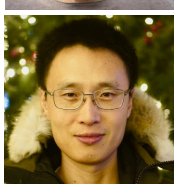
Iasonas Kokkinos CEO at Ariel AI Inc.

Invited talk 2: Learning 3D object models from 2D images.



Peter Kotschieder Director of Research at Mapillary

Invited talk 3: Computer Vision with Less Supervision



Boqing Gong Research Scientist at Google

Invited talk 4: Towards Domain Adaptation in the Wild: Long-Tailed Sources and Open Compound Targets

Invited Talks



Zhicheng Yan Staff Research Scientist at Facebook Research
Invited talk 5: Decoupling Representation and Classifier for Long-Tailed Recognition



Ming-Hsuan Yang Professor at University of California, Merced
Invited talk 6: Show, Match and Segment: Joint Weakly Supervised Learning of Semantic Matching and Object Co-segmentation



Chunhua Shen Professor at University of Adelaide
Invited talk 7: Single shot instance segmentation

Thank You!