Multi-Label Image Classification via Knowledge Distillation from Weakly-Supervised Detection

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Project Page: https://yochengliu.github.io/MLIC-KD-WSD/



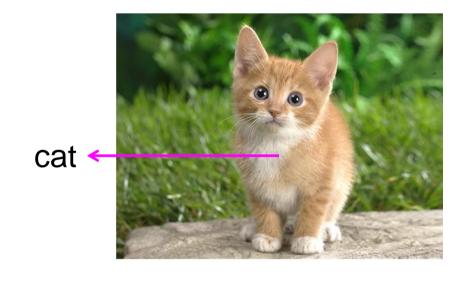
Outline

Introduction

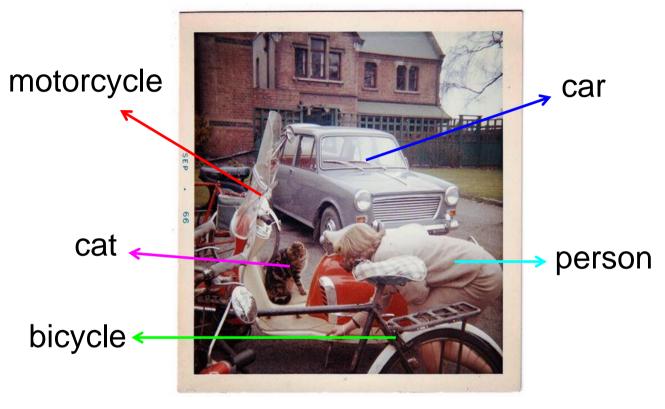
Related work

MLIC-KD-WSD

Introduction Problem



Single label, e.g., ImageNet



Multiple labels

real-world images general visual understanding

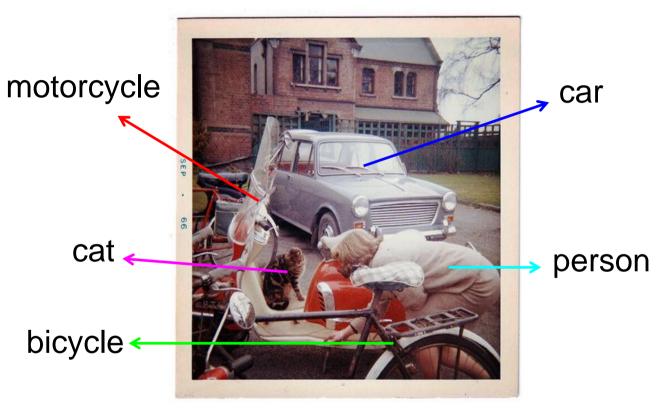
Introduction <u>Challenges</u>

Visual recognition

- inter-class similarity
- intra-class variation

Thorough understanding

- classes \(\to \) semantic regions
- class dependencies
 e.g., person & baby, cat & dog



Multiple labels

Introduction <u>Applications</u>

Scene recognition

HUAWEI NOV3 3

高颜值 爱自拍 2400万四摄* Al实力派



*前置双摄像头为2400万+200万像素,后置双摄像头为2400万+1600万像素 图片仅供参考,请以实物为准。

前置2400万[2] AI美拍[3]

前置2400万摄像头配合200万虚化镜头,支持放大 2m×1.5m海报[4]。Al识别多种典型环境,自动配 置拍照参数。











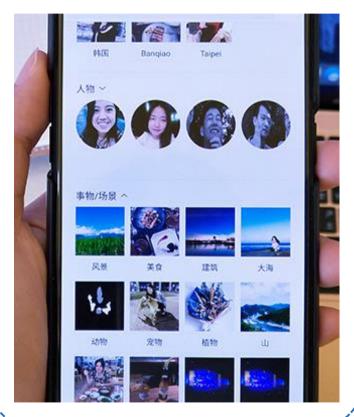




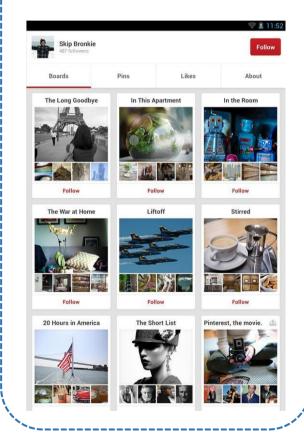




Smart albums

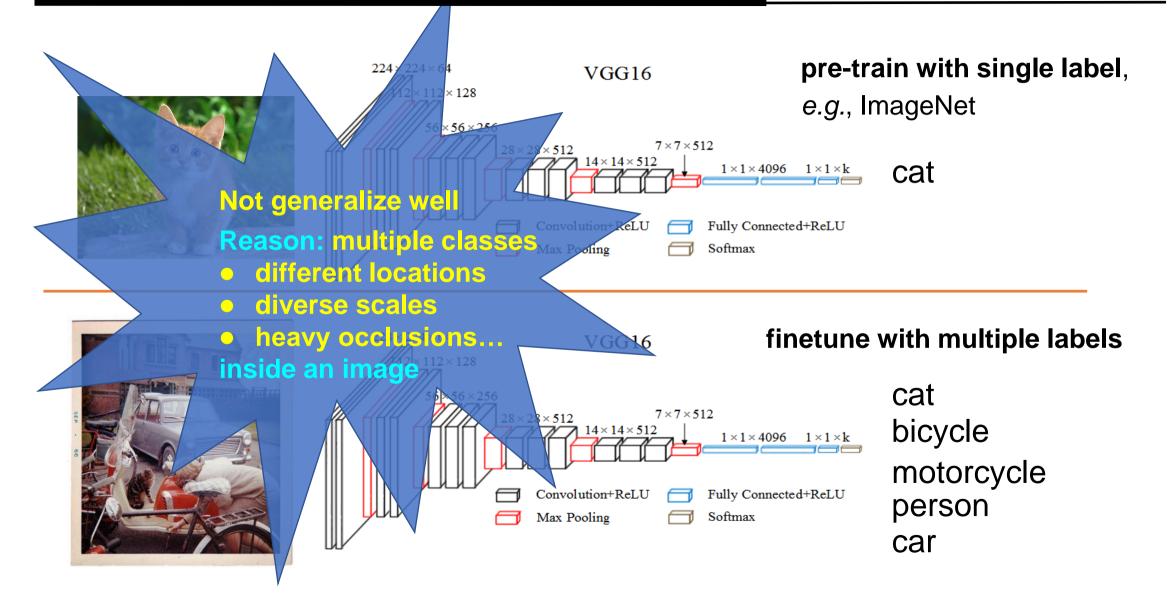


Social image analysis

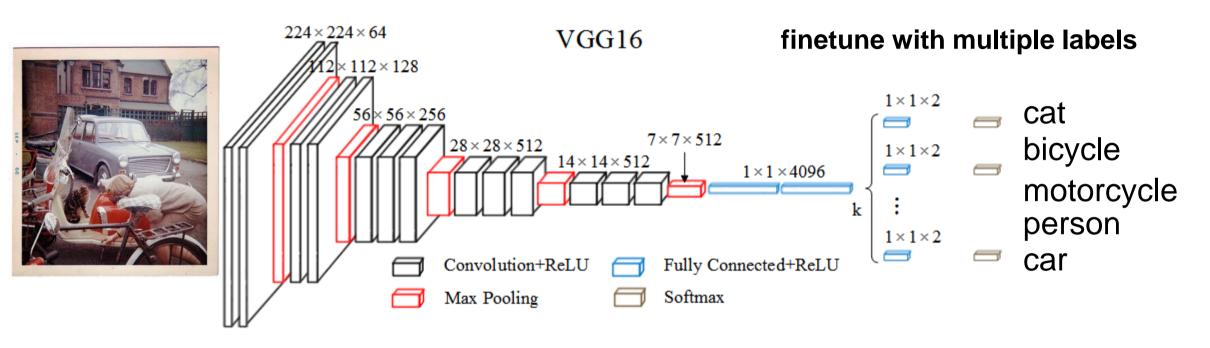


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Related Work Effortless method

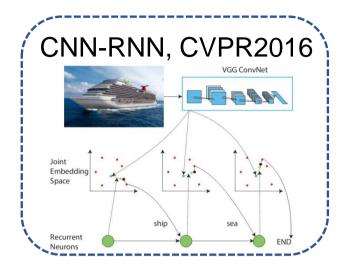


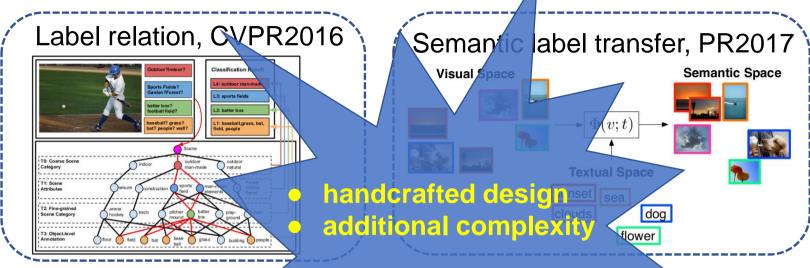
Related Work Binary decomposition

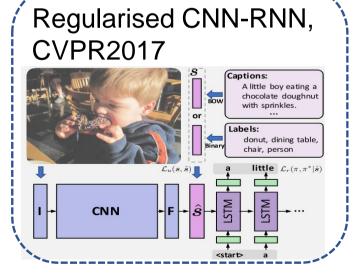


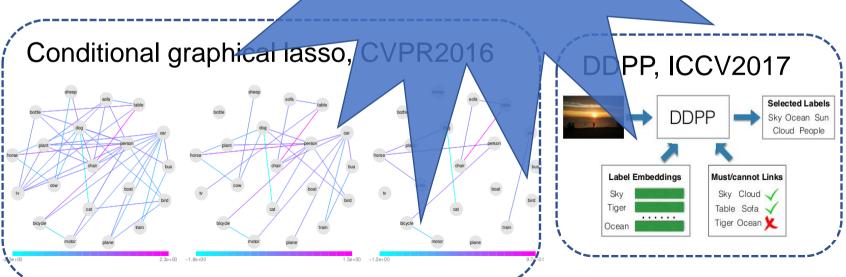
Ignoring class dependencies, e.g., person & baby, cat & dog

Related Work Class dependencies

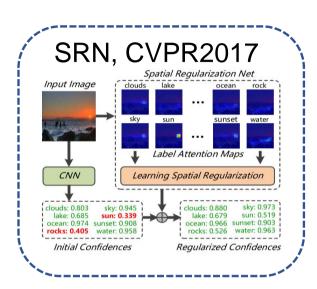


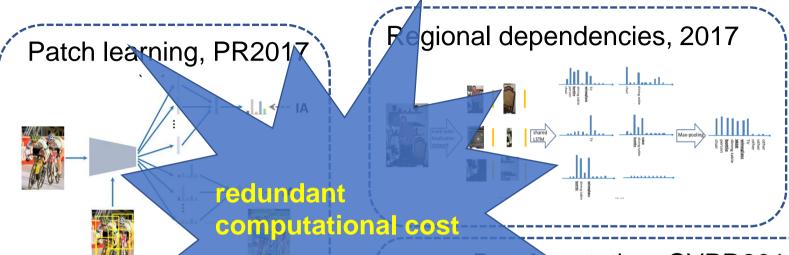


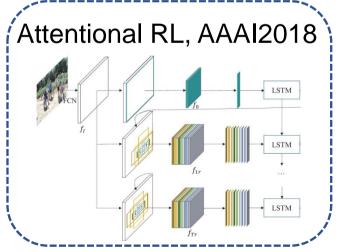




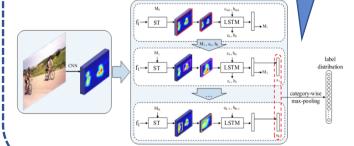
Related Work Localize semantic regions



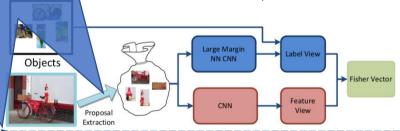




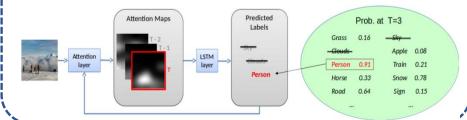
Attentional regions, ICC 12017



Box Annotation, CVPR2016



RNN with visual attention, AAAI2018

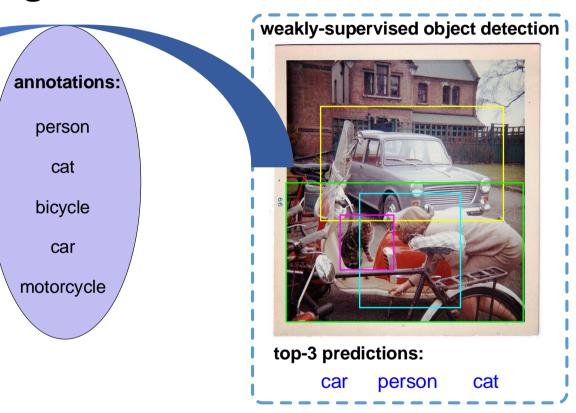


MLIC-KD-WSD Motivation

knowledge distillation



poor localization
for multiple instances



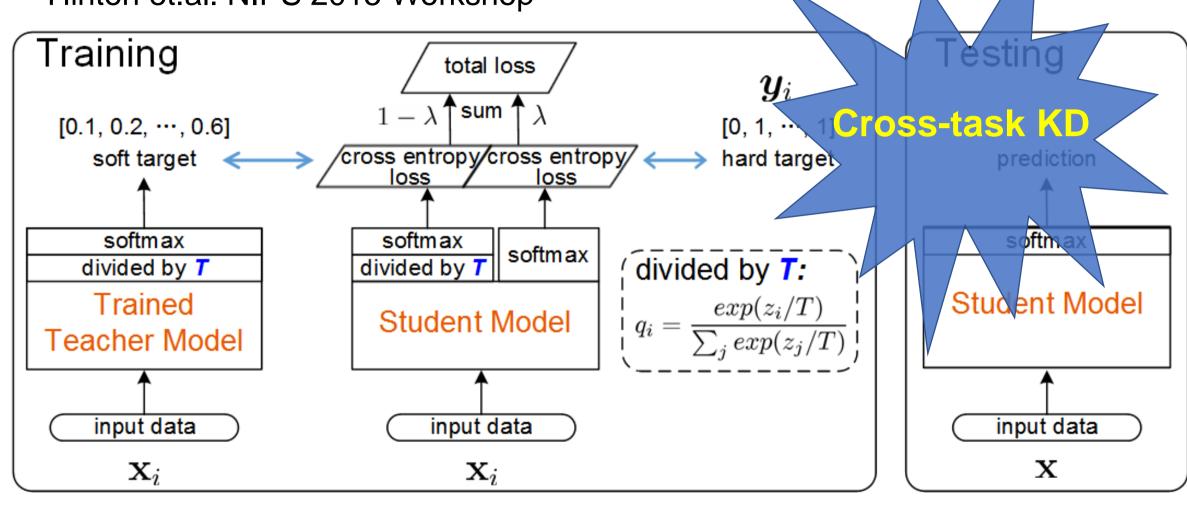
locate semantic regions

object-relevant informative
image-level annotation

MLIC-KD-WSD

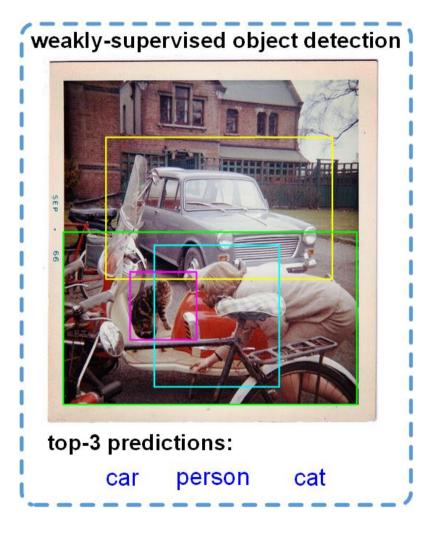
Knowledge distillation (KD)

Hinton et.al. NIPS 2015 Workshop



What is the unique knowledge in WSD?

- ✓ localization of semantic instances
- √ object-level features
- ✓ class dependencies



Step 1:

Weakly-Supervised Detection

Step 2:

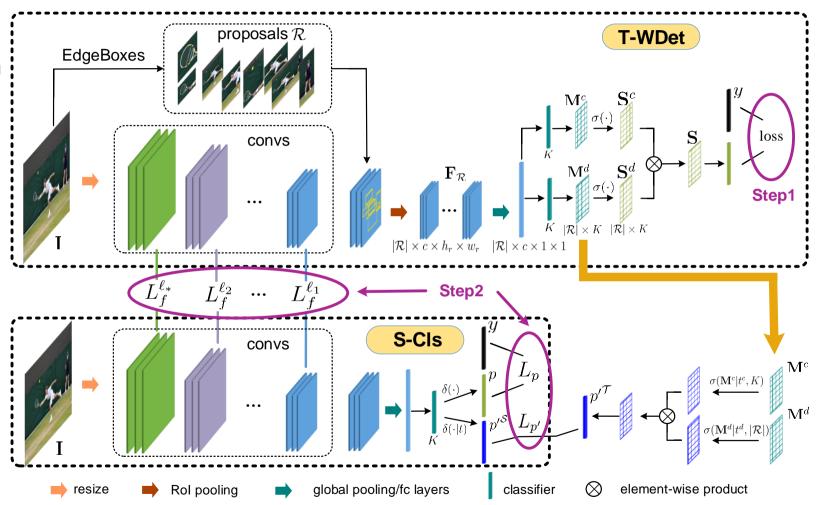
Knowledge Distillation

Stage 1:

Feature-level transfer
Only update convs' params

Stage 2:

Prediction-level transfer Update all params



Stage 1:

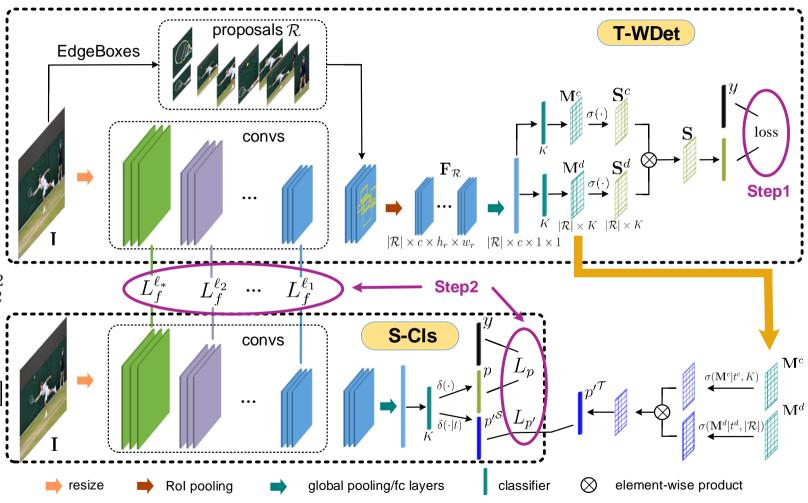
Feature-level transfer only update convs' params

- ✓ localization of instances
- √ object-level features

$$L_f(\mathbf{w}_{\text{conv}}^{\mathcal{S}}) = \frac{1}{2N} \sum_{n} \frac{1}{|\mathcal{R}'_n|} ||\mathbf{F}_{\mathcal{R}'_n}^{\mathcal{T}} \ominus \mathbf{F}_{\mathcal{R}'_n}^{\mathcal{S}}||_2^2$$

$$\mathbf{F}_{\mathcal{R}'_n}^{\mathcal{T}} = \mathbb{C}_{R \in \mathcal{R}'_n} \left[s'_R \odot \phi_{\text{RoI}}(\mathbf{F}_{\text{conv}}^{\mathcal{T}}; R) \right],$$

$$\mathbf{F}_{\mathcal{R}'_n}^{\mathcal{S}} = \mathbb{C}_{R \in \mathcal{R}'_n} \left[s'_R \odot \phi_{\text{RoI}}(\Psi(\mathbf{F}_{\text{conv}}^{\mathcal{S}}) | \mathbf{w}_{\text{conv}}^{\mathcal{S}}; R) \right]$$



Stage 2:

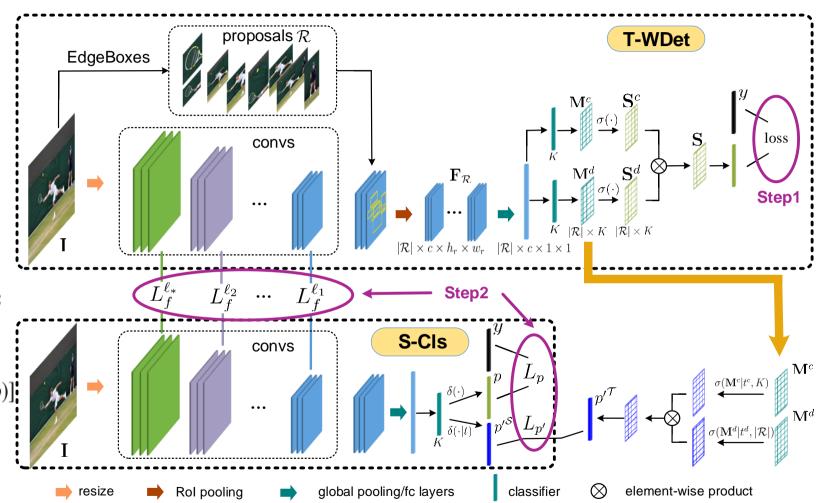
Prediction-level transfer

update all params

√ class dependencies

$$L_{p'}(\mathbf{w}^{\mathcal{S}}) = \frac{1}{2N} \sum_{n} ||p'^{\mathcal{T}} - p'^{\mathcal{S}}(\mathbf{w}^{\mathcal{S}})||_{2}^{2}$$

$$L_p(\mathbf{w}^{\mathcal{S}}) = -\frac{1}{N} \sum_n [y \log p + (1-y) \log(1-p)]$$



MLIC-KD-WSD

Quantitative results

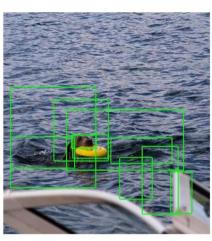
MS-COCO: object label

Method	All			Top-3	
	mAP	F1-C	F1-O	F1-C	F1-O
CNN-RNN [32]	-	-	-	60.4	67.8
CNN-LSEP [19]	_	62.9	68.3	-	-
CNN-SREL-RNN [21]	_	63.4	72.5	-	-
RMAM(512+10crop) [33]	72.2	-	-	<u>66.5</u>	71.3
RARLF(512+10crop) [5]	_	-	-	65.6	70.5
MIML-FCN-BB [39]	66.2	-	-	-	-
MCG-CNN-LSTM [43]	64.4	-	-	58.1	61.3
RLSD [43]	68.2	-	-	62.0	66.5
Ours-S-Cls (w/o)	70.9	63.6	67.0	60.7	66.7
Distillation [12]	71.3	64.7	69.3	61.5	67.6
FitNets [23]	72.5	<u>65.2</u>	70.9	62.3	68.3
Attention transfer [42]	71.4	64.6	69.8	61.6	67.8
Ours-S-Cls (w/)	74.6	69.2	74.0	66.8	72.7

NUS-WIDE: concept label

Method	All			Top-3	
	mAP	F1-C	F1-O	F1-C	F1-O
CNN-RNN [32]	-	-	-	34.7	55.2
Tag-Neighbors [15]	52.8	-	-	45.2	62.5
CNN-LSEP [19]	-	52.9	70.8	-	-
CNN-SREL-RNN [21]	-	52.8	<u>71.0</u>	-	-
MCG-CNN-LSTM [43]	52.4	-	-	46.1	59.9
RLSD [43]	54.1	-	-	46.9	60.3
KCCA [30]	52.2	-	-	_	-
Ours-S-Cls (w/o)	55.6	52.0	67.2	47.5	64.8
Distillation [12]	57.2	54.3	69.5	50.3	67.5
FitNets [23]	57.4	54.9	70.4	51.4	68.6
Attention transfer[42]	<u>57.6</u>	<u>55.2</u>	70.3	<u>51.7</u>	<u>68.8</u>
Ours-S-Cls (w/)	60.1	58. 7	73.7	53.8	71.1

MLIC-KD-WSD Results on MS-COCO



without:

bird surfboard person

with:

frisbee dog boat

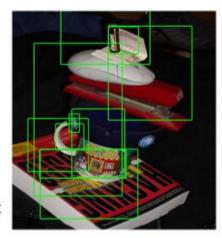


without:

bed toilet vase

with:

sink bottle potted plant

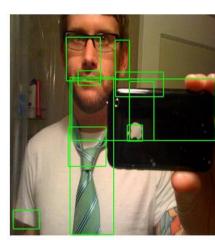


without:

chair person dining table

with:

book mouse cup

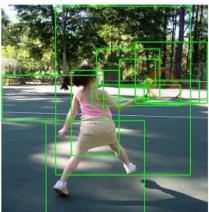


without:

person laptop bottle

with:

person cell phone tie



without:

person car frisbee

with:

person sports ba**ll** tennis racket

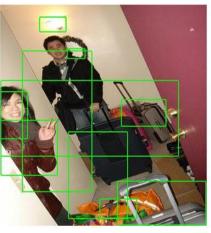


without:

umbrella person chair

with:

boat umbrella backpack

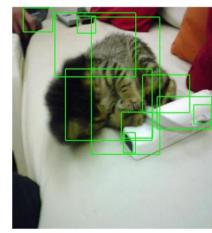


without:

person tie knife

with:

person suitcase handbag



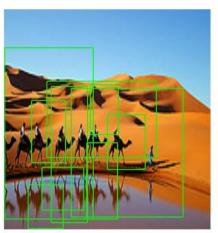
without:

bed laptop dog

with:

cat remote person

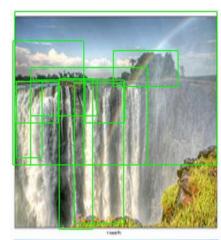
MLIC-KD-WSD Results on NUS-WIDE



without:

sky horses clouds

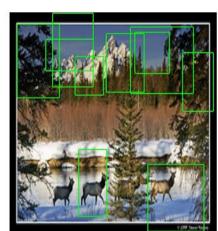
with: animal reflection sand



without:

sky grasss flowers

with: waterfall sky rainbow



without:

reflection cow ocean

with: mountain sky animal



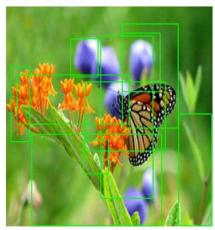
without:

sky food birds

with:

sky flowers

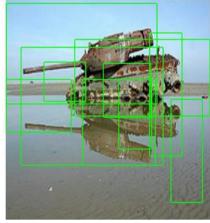
sunset



without:

plants rainbow food

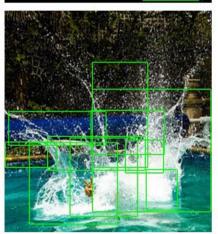
with: animal flowers leaf



without:

sky birds boats

with: military sky water



without:

lake waterfall plants

with: water sports swimmers

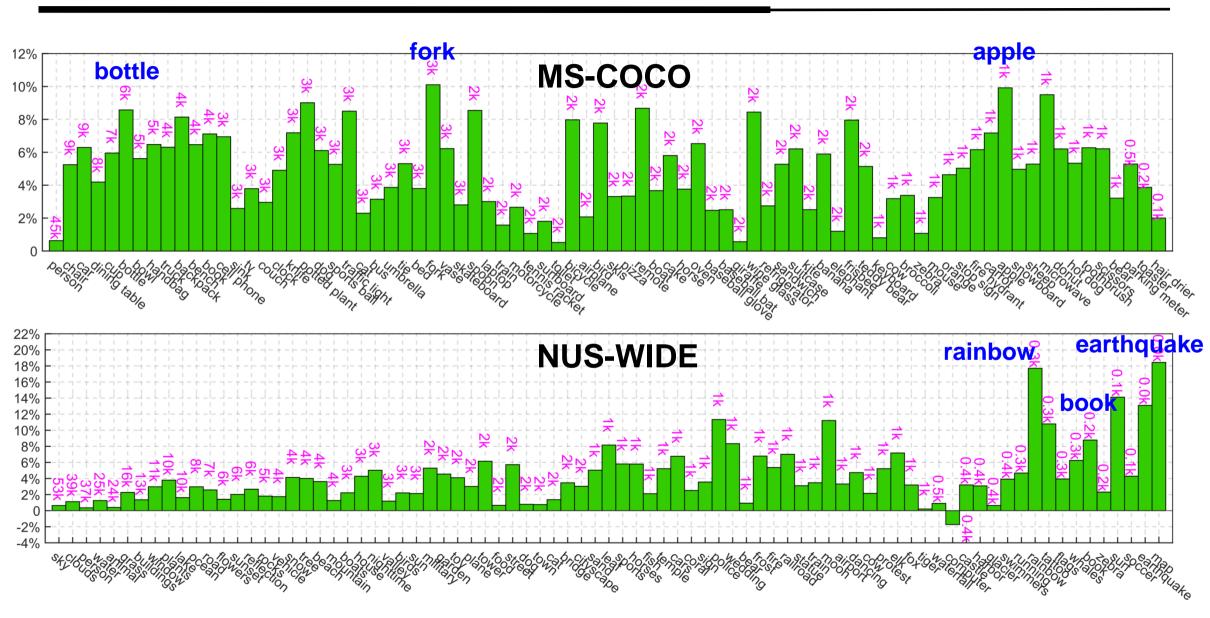


without: garden

garden earthquake waterfa**ll**

> with: rocks water valley

MLIC-KD-WSD Robustness



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Thanks for your attention!

Q & A