

# Reg R-CNN: Lesion Detection and Grading under Noisy Labels

Using ordinal context in the training signal to improve lesion detection and grading by swapping a CE classifier for a regressor.

G. N. Ramien, P. F. Jaeger, Simon A. A. Kohl, K. H. Maier-Hein

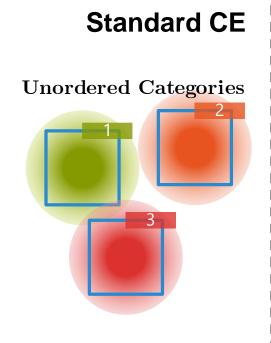
Presented by **David Zimmerer** 



#### **Motivation**

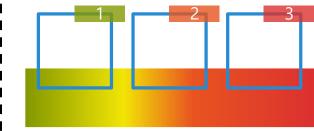
Discarding of ordinal context: cross-entropy loss penalizes predictions irrespective of distance to target.

Continuous Tissue Features Bins





**Ordinal Context** 



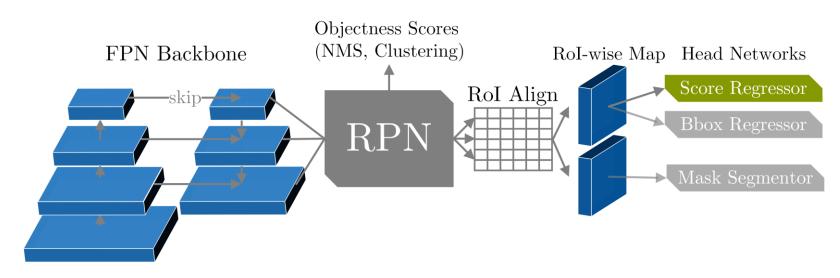


(left) NIH. Morphology & Grade: ICD-O-3 Morphology Codes, 2019



#### **Methods**

Reg R-CNN is Mask R-CNN with regressed instead of categorized object classed.



github.com/MIC-DKFZ/RegRCNN



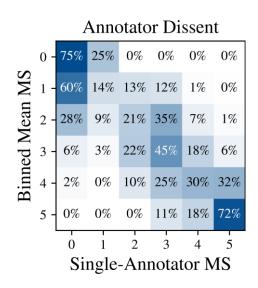


## **Experiments**

Reg vs. Mask R-CNN on public LIDC and specific toy data set.

#### **LIDC**

- 1026 patient CT scans
- 4 annotations per volume
- Raters often disagree



# Toy

- 3D cylinders
- 1.5k training, 1k test samples
- Task: detection and radius determination



<sup>1</sup> Jaeger et al., Retina U-Net, 2018

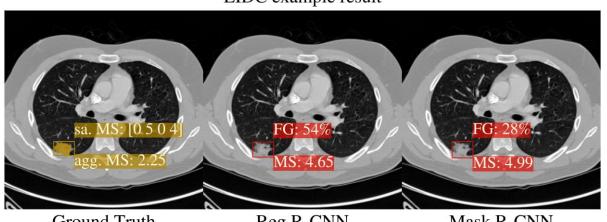


#### **Results**

#### Reg R-CNN outperforms Mask R-CNN in all cases except 3D toy AP.

Data & Dim	Network	AVP <sub>10</sub>	AP <sub>10</sub>	Bin Accuracy
LIDC 3D	Reg R-CNN	0.259±0.035	0.628±0.038	0.477±0.035
	Mask R-CNN	0.235±0.027	0.622±0.029	0.411±0.026
	$\Delta$ (Reg – Mask)	0.024±0.044	0.006±0.048	0.066±0.044
Toy 3D	Reg R-CNN	0.881±0.014	0.998±0.004	0.887±0.014
	Mask R-CNN	0.822±0.070	1.000±0.000	0.826±0.069
	∆(Reg – Mask)	0.059±0.071	-0.002±0.004	0.061±0.070

#### LIDC example result

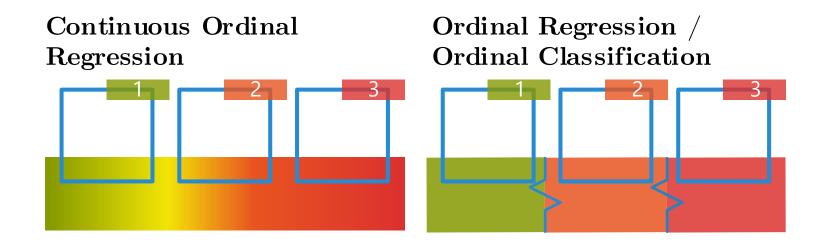


Ground Truth Reg R-CNN Mask R-CNN

#### **Discussion & Outlook**

Continuous regressor implies metric scale.

- We imply metric grading scale but is not guaranteed.<sup>1</sup>
- Future work: replace linear regressor by categorical ordinal regressor<sup>2</sup>.





<sup>1</sup>Liddell, Kruschke, Analyzing ordinal data, 2018; <sup>2</sup>Feindt, A neural Bayesian estimator, 2004





# Thank you for attending!



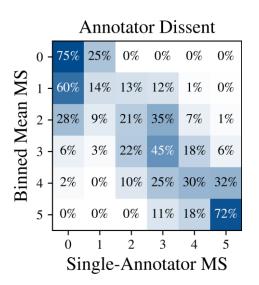


## **Experiments**

Reg vs. Mask R-CNN on public LIDC and specific toy data set.

#### LIDC

- 1026 patient
  CT scans
- 4 annotations per volume
- Raters often disagree



# Toy

- 3D cylinders
- 1.5k training, 1k test samples
- Task: detection and radius determination

<sup>1</sup> Jaeger et al., Retina U-Net, 2018

#### Setup:

- 2D (slice sampling) or 3D input
- 5-fold cross-validation
- Alternating single-rater (LIDC) or noisy (toy) labels during
- Averaged multi-rater (LIDC) or exact (toy) labels during testing
- Weighted ensemble and view aggregation (WBC¹)



#### **All Results**

Reg R-CNN outperforms Mask R-CNN in all cases except 3D toy AP.

