

Leveraging Uncertainty for Deep Interpretable Classification and Weakly-Supervised Segmentation of Histology Images

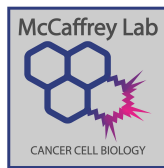
#4

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MIDL 2022 - Short



LIVIA
LABORATOIRE
D'IMAGERIE, DE VISION
ET D'INTELLIGENCE
ARTIFICIELLE

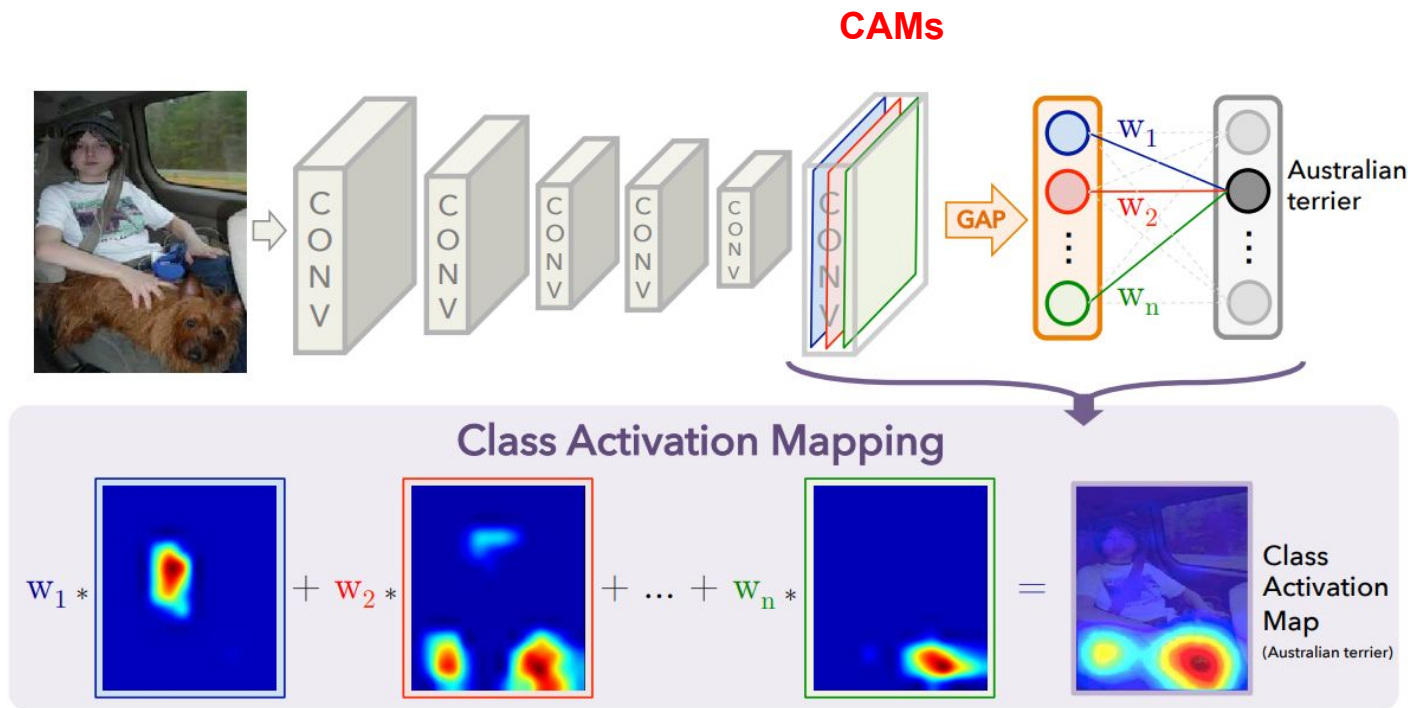


The Goodman Cancer Research Centre



McGill

Classifier, ROI localization, and interpretability via global labels

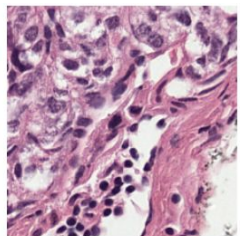


CAMs' challenges in histology images

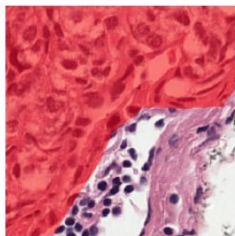
- Non-salient object → visual similarity between foreground/background



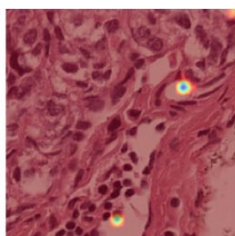
High false positives/negatives



Input



Ground truth

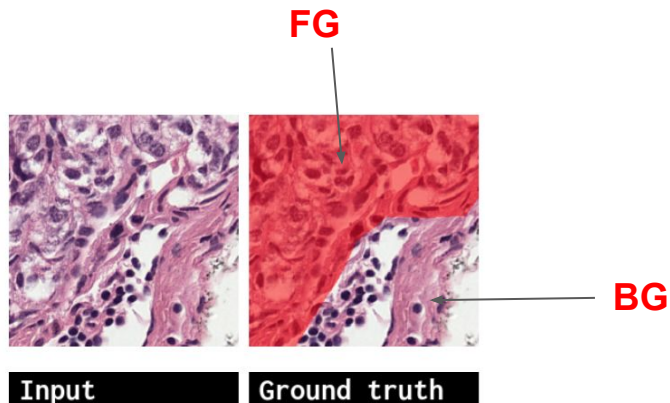


CAM

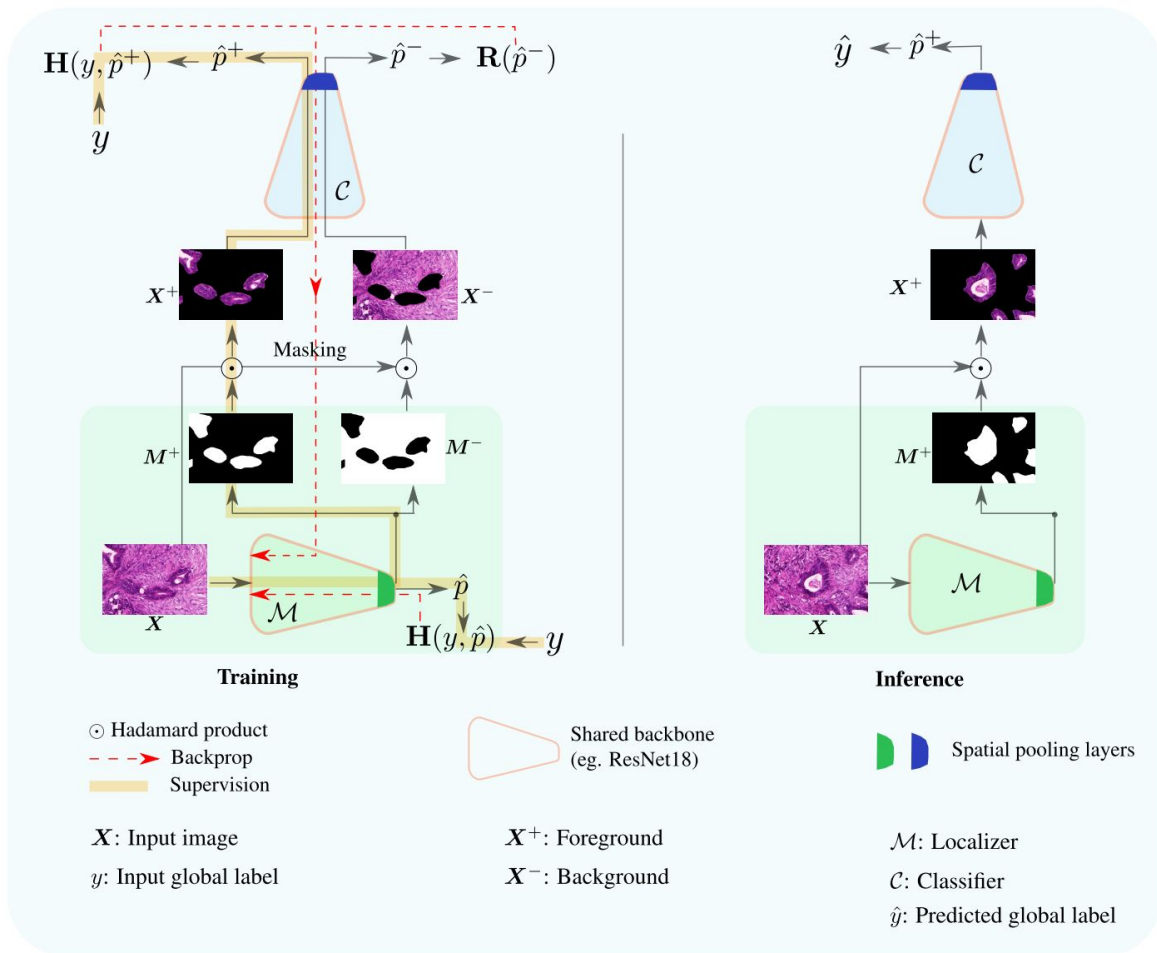
*Deep Weakly-Supervised Learning Methods for
Classification and Localization in Histology Images: A
Comparative Study. 2022. arxiv.org/abs/1909.03354*

Our work: Constrain CAMs

- Explicit modeling foreground/background map
- Constrain the presence of both FG / BG using size constraints
- Ensure that each map is consistent using classifier response.



Our architecture



Training loss

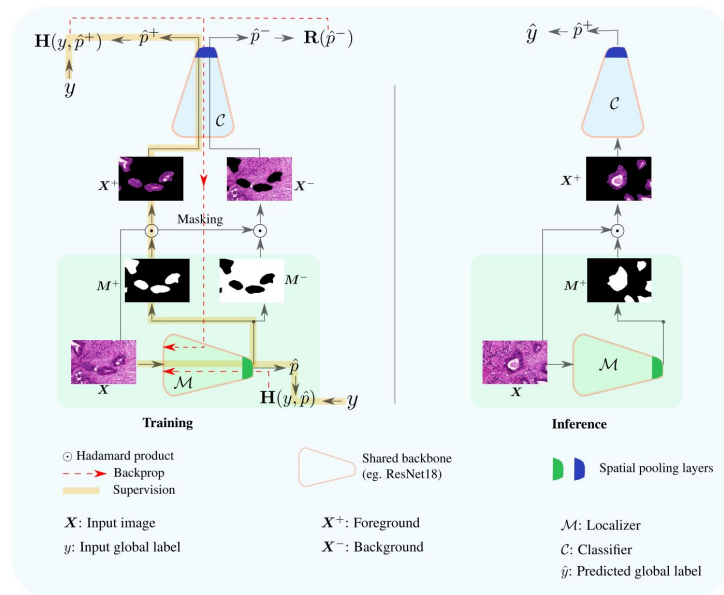
$$\min_{\theta_c} \underbrace{\mathbf{H}(p, \hat{p}^+)}_{\text{Maximize classifier Response over FG}} + \lambda \underbrace{\mathbf{R}(\hat{p}^-)}_{\text{The BG has no discriminative regions left}} - \underbrace{\frac{1}{t} [\log s^+ + \log s^-]}_{\text{Ensure both FG/BG are present: max size.}}$$

ASC: Absolute
Size Constraint

Maximize
classifier
Response
over FG

The BG has no
discriminative
regions left

Ensure both
FG/BG are
present:
max size.



Training loss

$$\min_{\theta_c} \underbrace{\mathbf{H}(p, \hat{p}^+)}_{\text{Maximize classifier Response over FG}} + \lambda \underbrace{\mathbf{R}(\hat{p}^-)}_{\text{The BG has no discriminative regions left}} - \underbrace{\frac{1}{t} [\log s^+ + \log s^-]}_{\text{Ensure both FG/BG are present: max size.}}$$

Maximize
classifier
Response
over FG

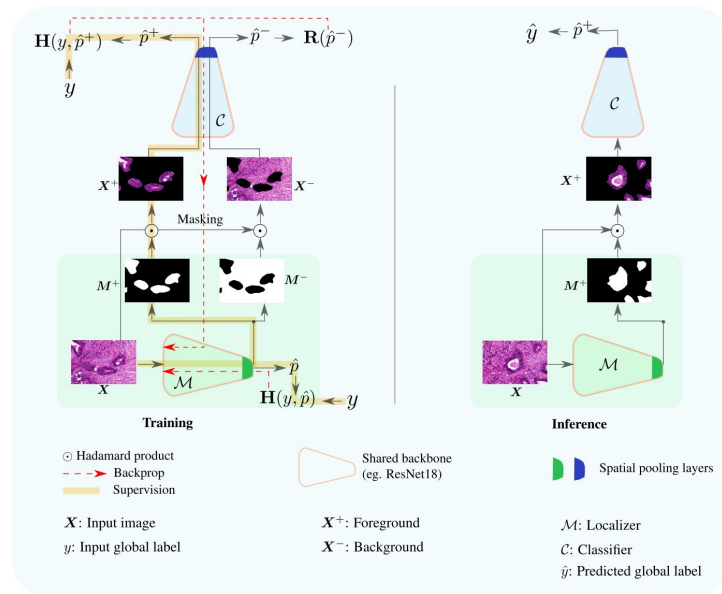
The BG has no
discriminative
regions left

Ensure both
FG/BG are
present:
max size.

$$\mathbf{R}(\hat{p}^-) = -\mathbf{H}(\hat{p}^-); \quad \text{or} \quad \mathbf{R}(\hat{p}^-) = \mathbf{H}(q, \hat{p}^-),$$

Explicit Entropy
Maximization (EEM)

Surrogate for explicit Entropy
Maximization (SEM). q : uniform dist.



Training loss

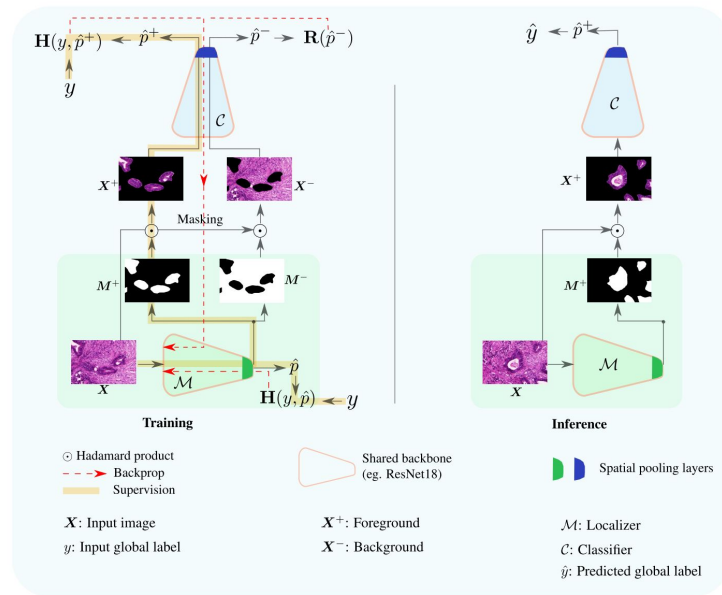
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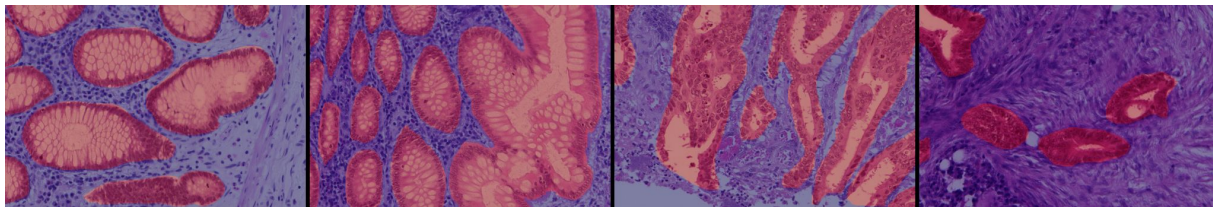


$$s^+ = \sum_{z \in \Omega} M^+(z), \quad s^- = \sum_{z \in \Omega} M^-(z)$$

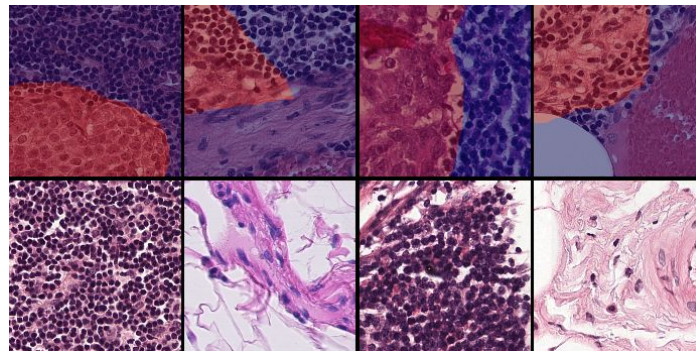
Sizes

Experiments

- Task: classify and localize ROI
- 2 public datasets: GlaS, Camelyon16 patches.



GlaS: colon cancer diagnosis



Camelyon16 patches: breast cancer

Results

Method	Image level	Pixel level	
	Cl. error (%)	F1 ⁺ (%)	F1 ⁻ (%)
All-ones (Lower-bound)	--	66.01	00.00
PN [39]	--	65.52	24.08
ERASE [80]	7.50	65.60	25.01
CAM-Max [52]	1.25	66.00	26.32
CAM-LSE [57, 74]	1.25	66.05	27.93
Grad-CAM [64]	0.00	66.30	21.30
CAM-Avg [88]	0.00	66.90	17.88
Wildcat [20]	1.25	67.21	22.96
Deep MIL [33]	2.50	68.52	41.34
Ours (EEM)	0.00	72.11	69.07
Ours (SEM)	0.00	71.94	69.23
U-Net [60] (Upper-bound)	--	90.19	88.52

GlaS

Method	Image level	Pixel level	
	Cl. error (%)	F1 ⁺ (%)	F1 ⁻ (%)
All-ones (Lower-bound)	--	59.44	00.00
PN [39]	--	31.15	37.36
ERASE [80]	8.61	31.30	42.48
CAM-Max [52]	10.06	48.28	81.92
CAM-LSE [57, 74]	1.51	64.31	63.78
Grad-CAM [64]	2.40	62.78	79.05
CAM-Avg [88]	2.40	62.75	79.05
Wildcat [20]	1.48	62.73	72.59
Deep MIL [33]	1.93	59.01	36.94
Ours (EEM)	6.26	67.98	88.80
Ours (SEM)	6.95	68.26	88.55
U-Net [60] (Upper-bound)	--	71.11	89.68

Camelyon16

Visual results

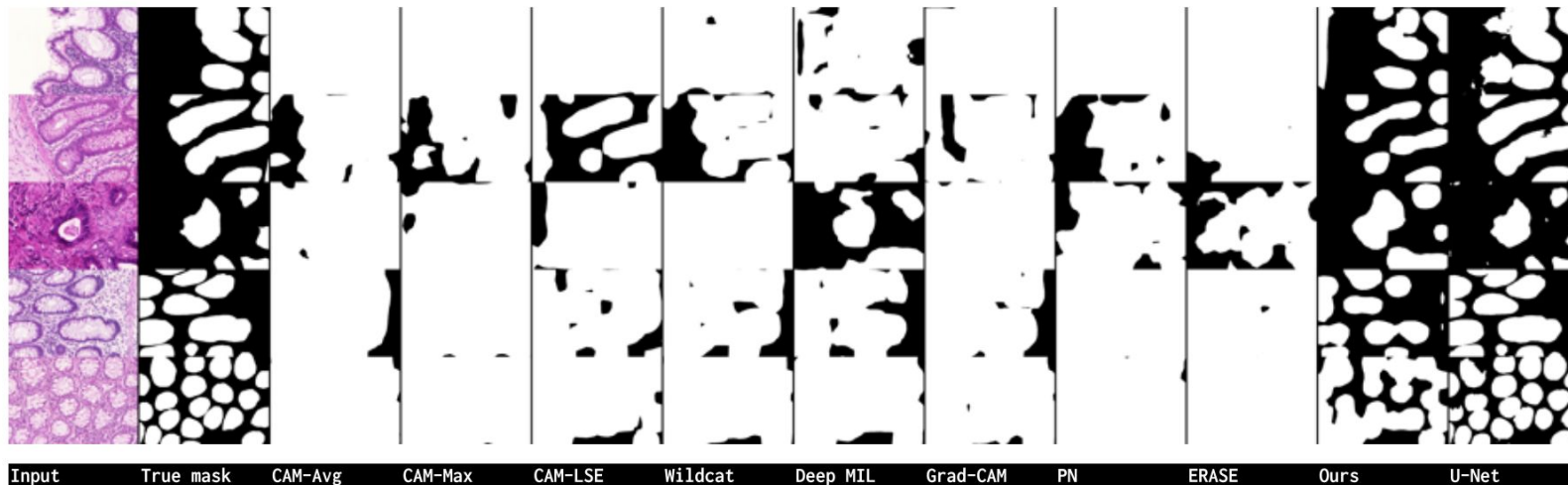


Figure 2: **GlaS dataset:** Qualitative results of the predicted binary mask for each method on several GlaS test images. Our method, referred to as *Ours*, is the SEM version with the ASC regularization term. (Best visualized in color.)

Visual results

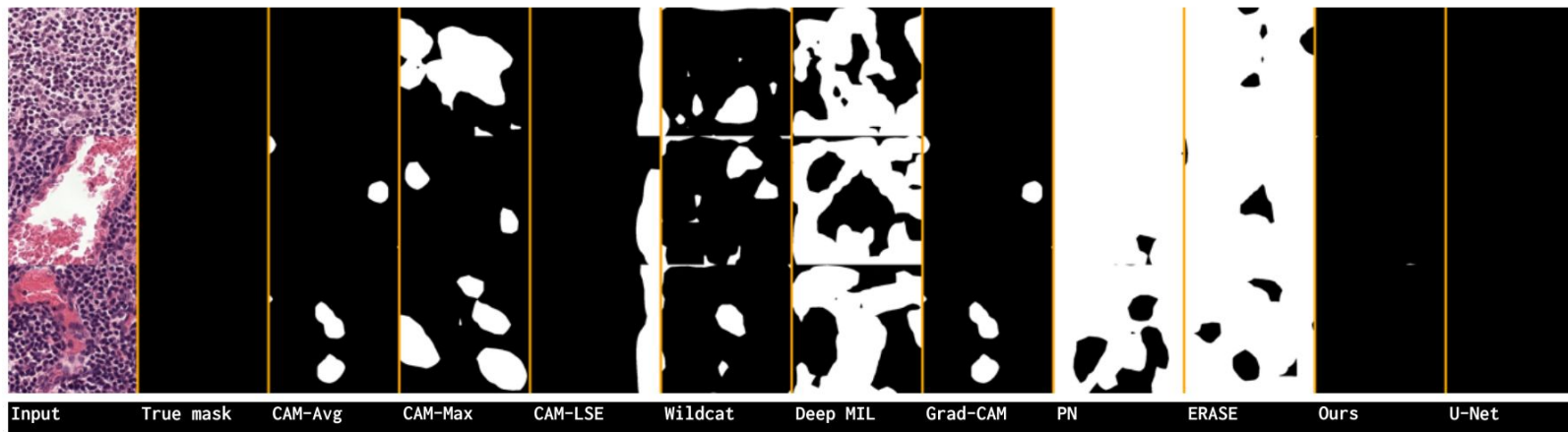


Figure 3: **Camelyon16-P512 benchmark**: Examples of mask predictions over **normal** samples from the testing set. White pixels indicate metastatic regions, while black pixels indicate normal tissue. This illustrates false positives. Note that normal samples do not contain any metastatic regions. Ours is SEM version with the ASC regularization. (Best visualized in color.)

Results

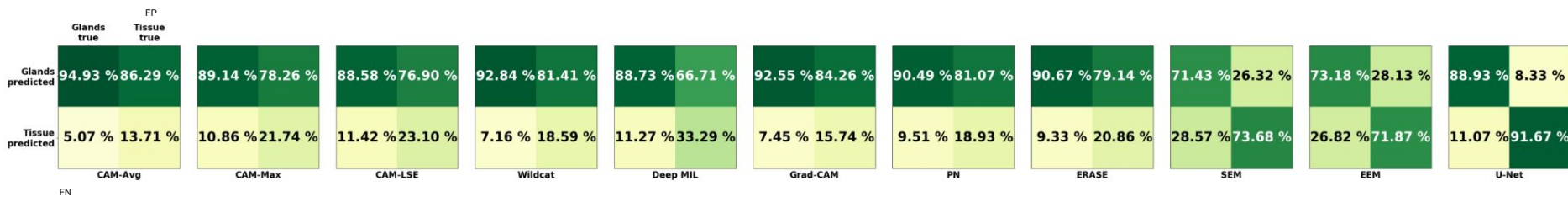


Figure 5: **GlaS dataset**: Confusion matrix over entire pixels of test set. (Best visualized in color.)

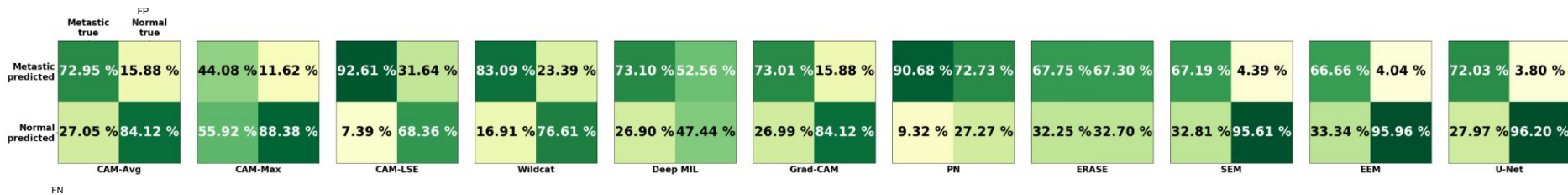
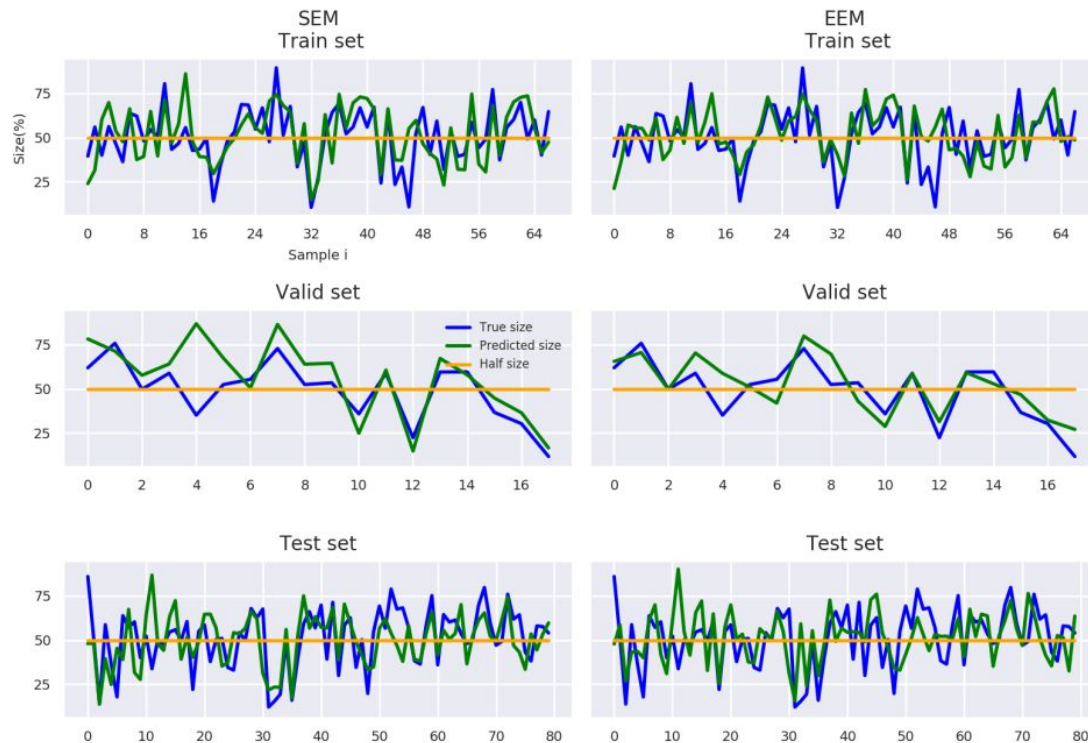


Figure 6: **Camelyon16-P512 dataset**: Confusion matrix over entire pixels of test set. (Best visualized in color.)

Size constraint



GlaS

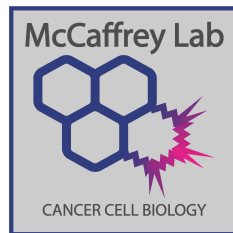
Ablation

Method	Image level	Pixel level	
	Cl. error (%)	F1 ⁺ (%)	F1 ⁻ (%)
Wildcat [20]	1.25	67.21	22.96
FG only	1.25	71.54	49.23
FG + BG (EEM)	1.25	72.54	61.82
FG + BG (EEM) + ASC	0.00	72.11	69.07
FG + BG (SEM)	1.25	72.96	61.95
FG + BG (SEM) + ASC	0.00	71.94	69.23

Thanks! Questions?

Please visit us at #4 – Short

Code: <https://github.com/sbelharbi/deep-wsl-histo-min-max-uncertainty>



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