

Weakly Supervised Segmentation in Medical Imaging: A Counterfactual Approach



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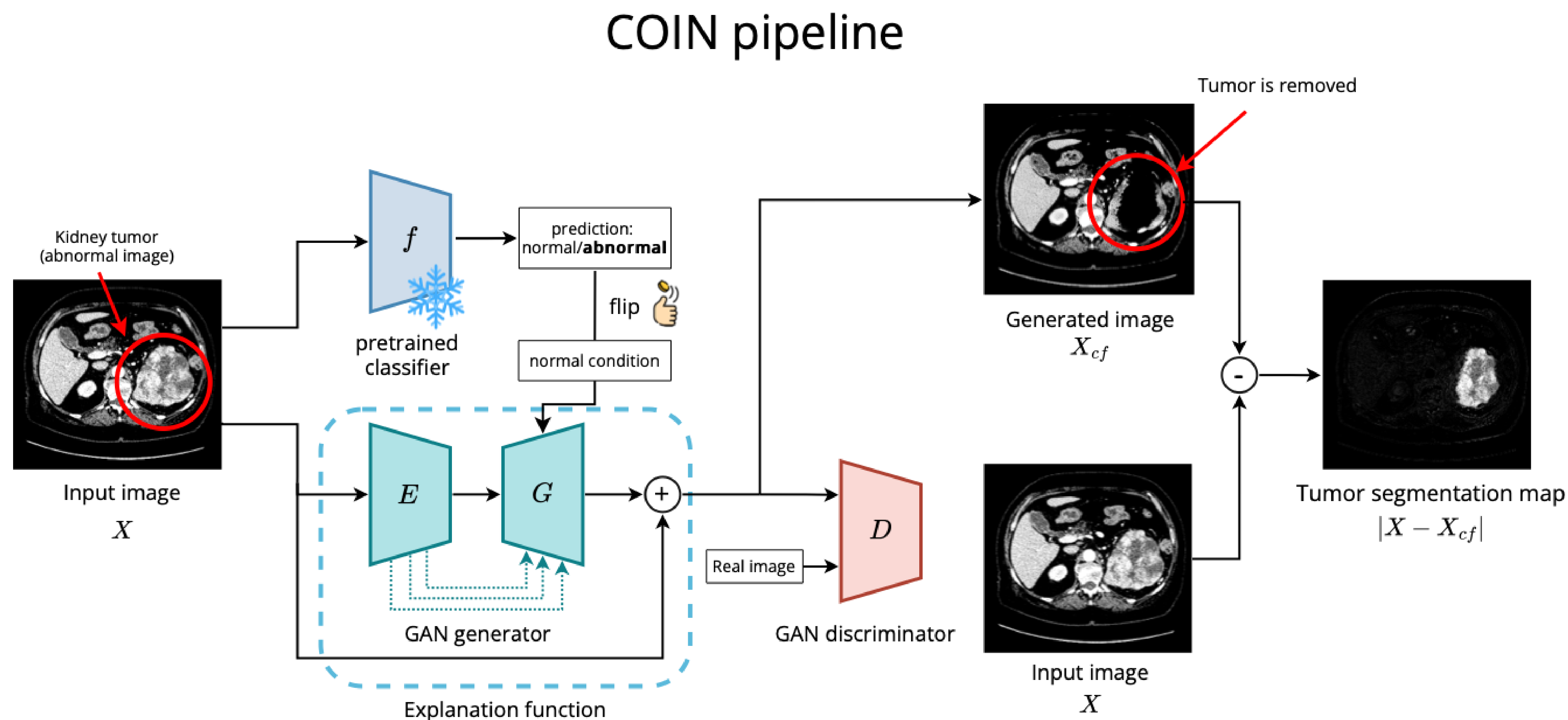


Figure 1: Overview of the proposed Counterfactual Inpainting (COIN) pipeline [1]. Given the input image X and image classifier f , the GAN model inpaints the pathology. The absolute difference of the original image X and counterfactual image X_{cf} results in a weak tumor segmentation map.

Comparison of COIN with other methods

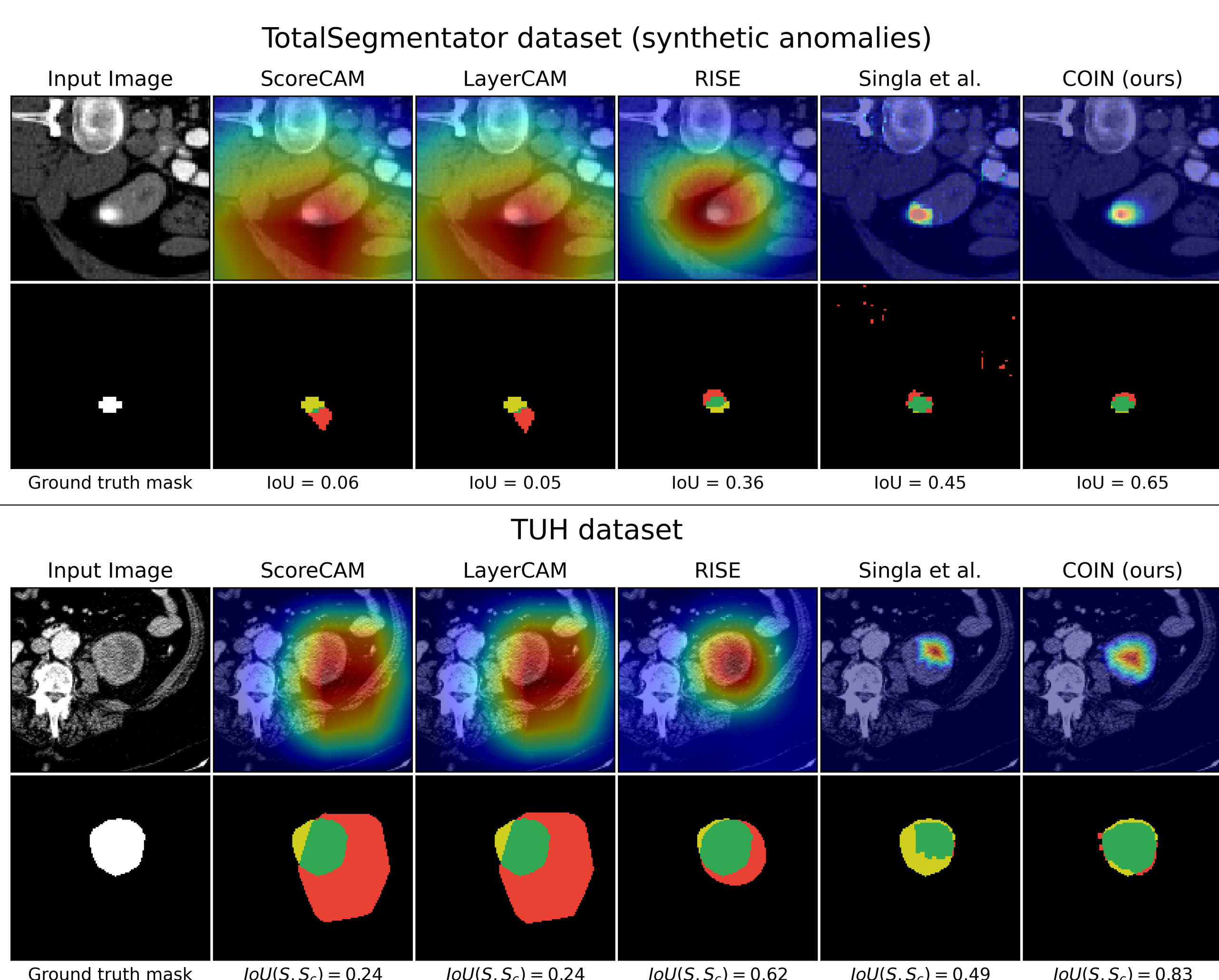


Figure 2: Visualization of the attribution and COIN pipeline methods' predictions on TotalSegmentator and TUH datasets [1]. The colors represent **ground truth** (white), **true positives** (green), **false positives** (red), and **false negatives** (yellow).

Datasets	Methods	FID ↓	CV ↑	IoU ↑
TotalSegmentator	ScoreCAM	-	-	0.030
	LayerCAM	-	-	0.026
	RISE	-	-	0.397
	Singla et al.*	0.047	0.998	0.445
	COIN	0.003	0.997	0.646
Tartu University Hospital	ScoreCAM	-	-	0.293
	LayerCAM	-	-	0.296
	RISE	-	-	0.294
	Singla et al.*	0.203	0.992	0.352
	COIN	0.036	0.980	0.432

Table 1: Metric results for the attribution methods and the COIN pipeline on TotalSegmentator and TUH datasets [1].

Conclusion

Our innovative Counterfactual Inpainting (COIN) approach inspired by the work of Singla et al. [2] accurately segments pathology regions in CT scans without reliance on the existing segmentation masks. COIN's main architectural improvements are:

- 1) **Perturbation-based** image generation yields great fidelity counterfactual images.
- 2) **Simplified conditioning** to focus only on the inpainting counterfactual generation.
- 3) **Skip-connections** for improving the image generation quality of the counterfactuals.
- 4) **New loss function** for enforcing smoothness among counterfactual images.

COIN examples on Tartu University Hospital dataset

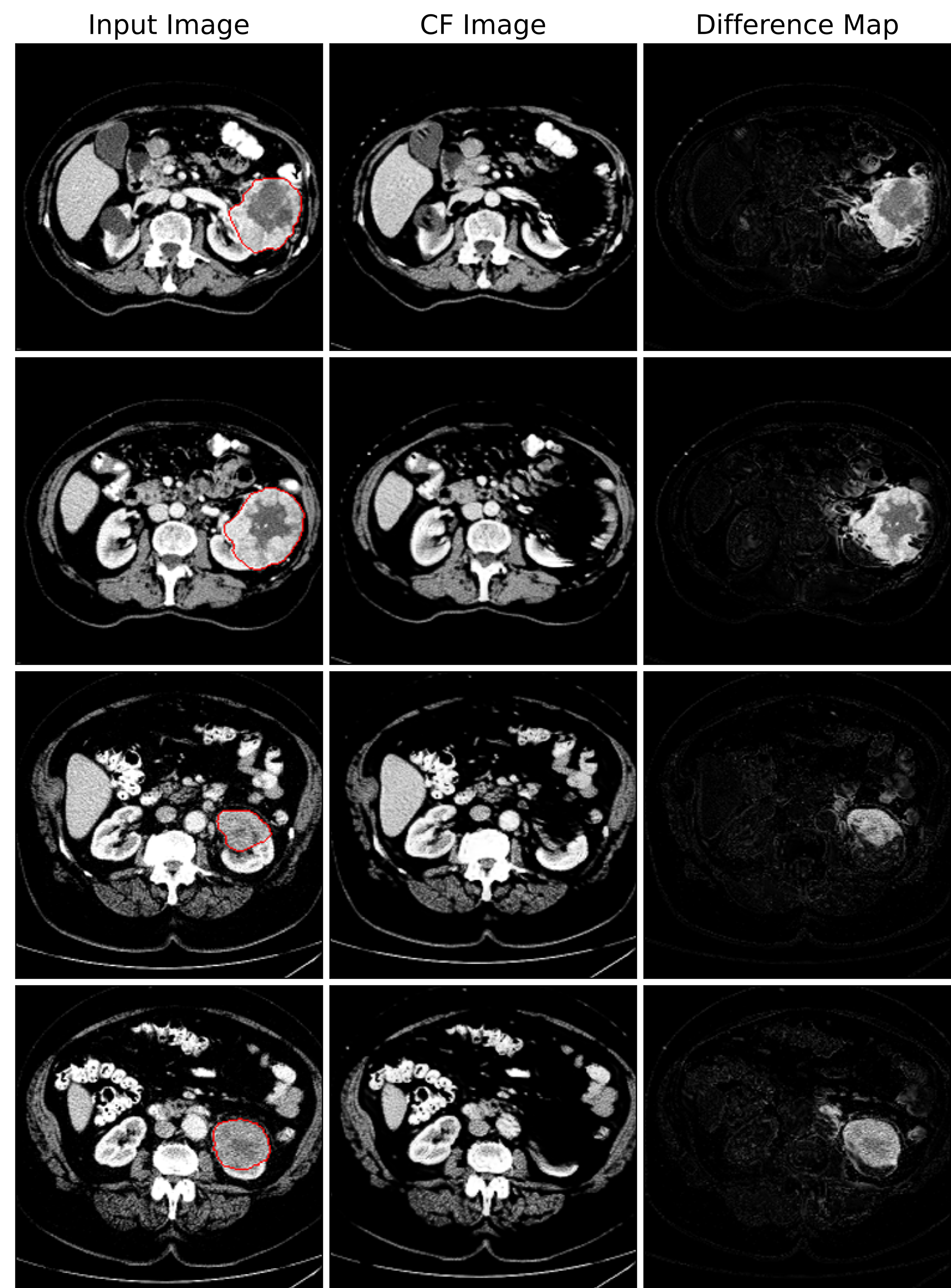


Figure 3: Visualization of counterfactual examples generated with COIN for TUH validation set [1]. The model correctly inpaints majority of the tumors (red contours).

Acknowledgements

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References

- [1] D. Shvetsov et al. *COIN: Counterfactual inpainting for weakly supervised semantic segmentation for medical images*. Apr. 19, 2024. arXiv: 2404.12832 [cs].
- [2] S. Singla et al. *Explaining the Black-box Smoothly- A Counterfactual Approach*. Nov. 18, 2022. arXiv: 2101.04230 [cs, eess].