

FREQUENCY DOMAIN ADVERSARIAL TRAINING FOR ROBUST VOLUMETRIC MEDICAL SEGMENTATION

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Introduction

- Semantic segmentation of organs, anatomical structures, or anomalies in medical images (e.g. CT or MRI scans) remains one of the fundamental tasks in medical image analysis.
- Deep learning and volumetric medical image segmentation
- Adversarial vulnerability of volumetric medical image segmentation
- Adversarial robustness of medical image models
- Understanding the vulnerability and robustness of medical image models

Contributions

- Volumetric Adversarial Frequency Attack (VAFA)
- Volumetric Adversarial Frequency Training (VAFT)

Voxel-domain Attacks vs. Frequency-domain Attacks

- Voxel-domain Attacks: Directly perturb input space
- Frequency-domain Attacks: Perturb the frequency-domain representation of an image e.g. DCT

Method

- Volumetric Adversarial Frequency domain Attack (**VAFA**)
- Volumetric Adversarial Frequency domain Training (**VAFT**)

Volumetric Adversarial Frequency Domain Attack (VAFA)

Volumetric Adversarial Frequency Domain Attack (VAFA)

$$\mathbf{x} \mapsto \mathcal{D}(\mathbf{x}) \mapsto \underbrace{\varphi(\mathcal{D}(\mathbf{x}), \mathbf{q})}_{\substack{\text{quantization,} \\ \text{rounding and} \\ \text{de-quantization}}} \mapsto \mathcal{D}_I(\varphi(\cdot)) \mapsto \mathbf{x}'$$

Volumetric Adversarial Frequency Domain Attack (VAFA)

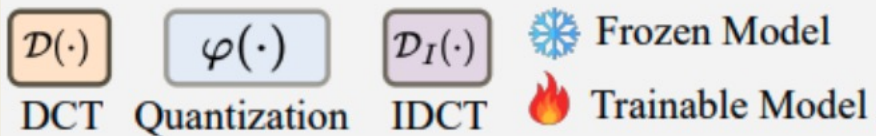
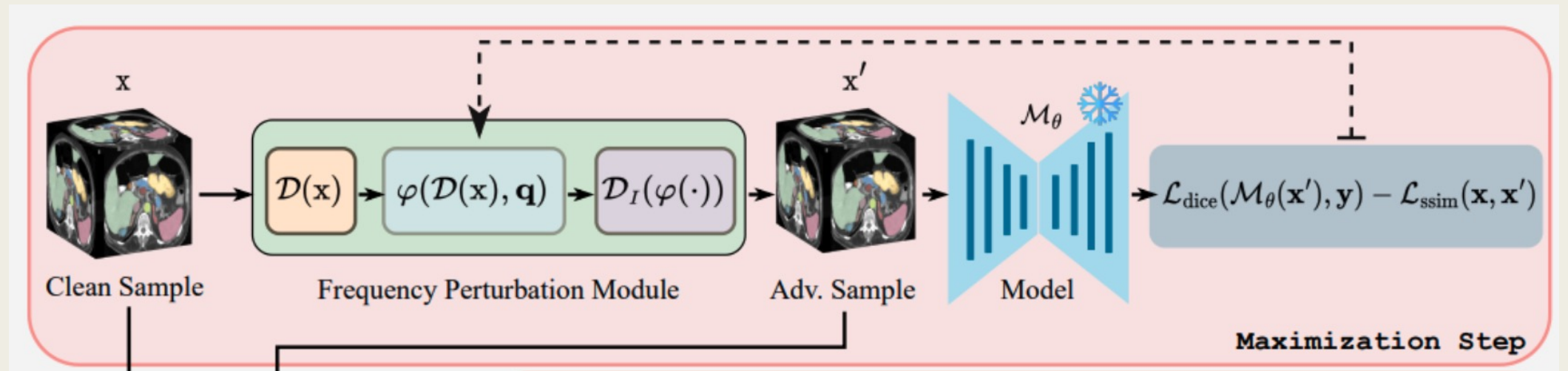
$$\varphi(\mathcal{D}(\mathbf{x}), \mathbf{q}) \quad := \quad \left\lfloor \frac{\mathcal{D}(\mathbf{x})}{\mathbf{q}} \right\rfloor \odot \mathbf{q}$$

$$\mathbf{q} \in \mathbb{Z}^{h \times w \times d}$$

Volumetric Adversarial Frequency Domain Attack (VAFA)

$$\begin{aligned} \underset{\mathbf{q}}{\text{maximize}} \quad & \mathcal{L}_{\text{dice}}(\mathcal{M}_{\theta}(X'), Y) - \mathcal{L}_{\text{ssim}}(X, X') \\ \text{s.t.} \quad & \|\mathbf{q}\|_{\infty} \leq q_{\text{max}}, \end{aligned}$$

Volumetric Adversarial Frequency Domain Attack (VAFA)



Volumetric Adversarial Frequency Domain Attack (VAFA)

Algorithm 1 Volumetric Adversarial Frequency Attack (**VAFA**)

```
1: Number of Steps:  $T$ , Quantization Threshold:  $q_{\max}$ 
2: Input:  $X \in \mathbb{R}^{H \times W \times D}$ ,  $Y \in \{0, 1\}^{\text{NumClass} \times H \times W \times D}$    Output:  $X' \in \mathbb{R}^{H \times W \times D}$ 
3: function VAFA( $X, Y$ )
4:    $\mathbf{q}_i \leftarrow 1 \quad \forall i \in \{1, 2, \dots, n\}$     $\triangleright$  Initialize all quantization tables with ones.
5:   for  $t \leftarrow 1$  to  $T$  do
6:      $\{\mathbf{x}_i\}_{i=1}^n \leftarrow \text{Split}(X)$     $\triangleright$  Split  $X$  into 3D patches of size  $(h \times w \times d)$ 
7:      $\mathbf{x}'_i \leftarrow \mathcal{D}_I(\varphi(\mathcal{D}(\mathbf{x}_i), \mathbf{q}_i)) \quad \forall i \in \{1, 2, \dots, n\}$     $\triangleright$  Frequency Perturbation
8:      $X' \leftarrow \text{Merge}(\{\mathbf{x}'_i\}_{i=1}^n)$     $\triangleright$  Merge all adversarial patches to form  $X'$ 
9:      $\mathcal{L}(X, X', Y) = \mathcal{L}_{\text{dice}}(\mathcal{M}_{\theta}(X'), Y) - \mathcal{L}_{\text{ssim}}(X, X')$ 
10:     $\mathbf{q}_i \leftarrow \mathbf{q}_i + \text{sign}(\nabla_{\mathbf{q}_i} \mathcal{L}) \quad \forall i \in \{1, 2, \dots, n\}$ 
11:     $\mathbf{q}_i \leftarrow \text{clip}(\mathbf{q}_i, \text{min}=1, \text{max}=q_{\max}) \quad \forall i \in \{1, 2, \dots, n\}$ 
12:  end for
13: end function
14: Return  $X'$ 
```

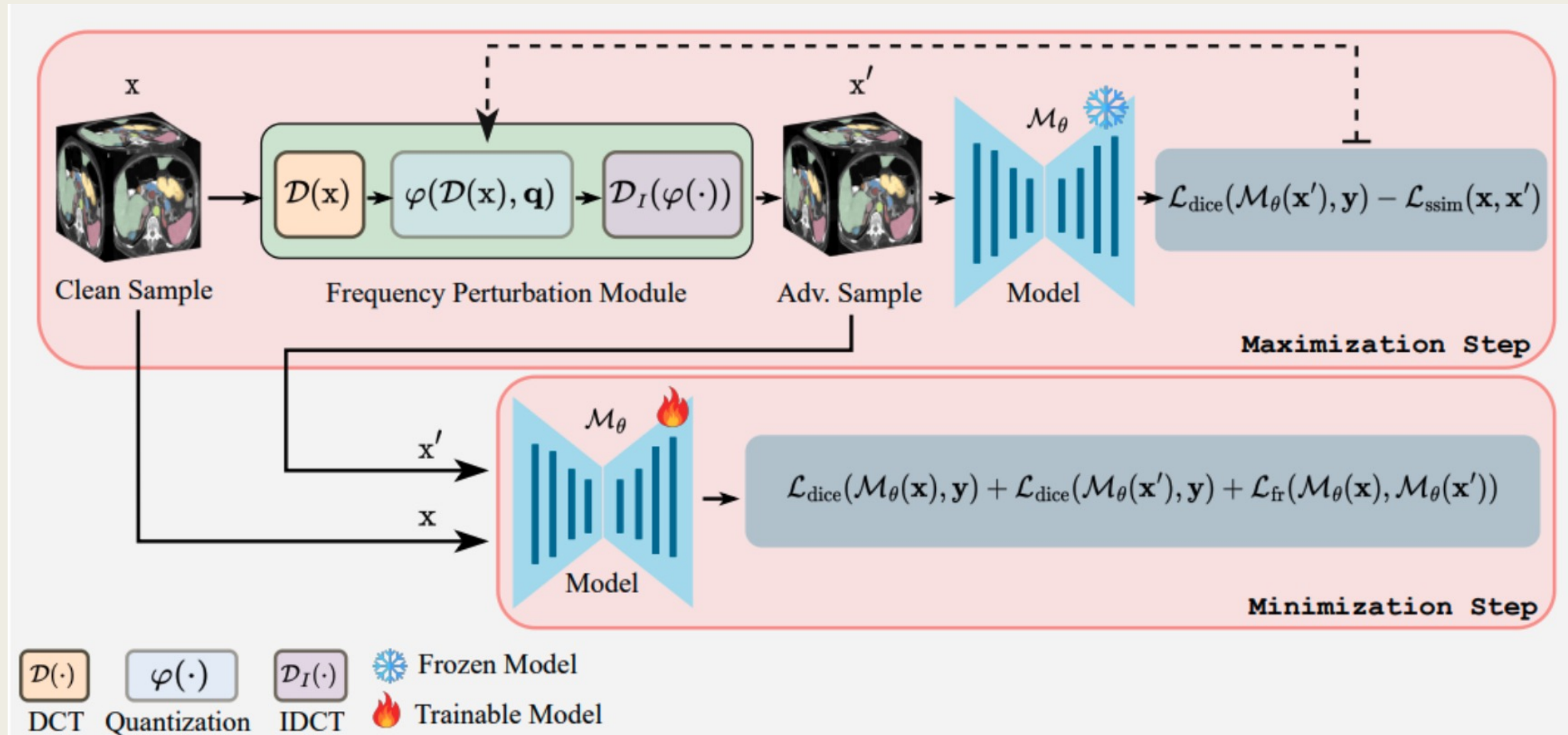
Volumetric Adversarial Frequency Domain Training (VAFT)

Volumetric Adversarial Frequency Domain Training (VAFT)

$$\underset{\theta}{\text{minimize}} \mathcal{L}_{\text{dice}}(\mathcal{M}_{\theta}(X), Y) + \mathcal{L}_{\text{dice}}(\mathcal{M}_{\theta}(X'), Y) + \mathcal{L}_{\text{fr}}(\mathcal{M}_{\theta}(X), \mathcal{M}_{\theta}(X')),$$

$$\mathcal{L}_{\text{fr}}(\mathcal{M}_{\theta}(X), \mathcal{M}_{\theta}(X')) = \|\mathcal{D}(\mathcal{M}_{\theta}(X)) - \mathcal{D}(\mathcal{M}_{\theta}(X'))\|_1$$

Volumetric Adversarial Frequency Domain Training (VAFT)



Volumetric Adversarial Frequency Domain Training (VAFT)

Algorithm 2 Volumetric Adversarial Frequency Training (**VAFT**)

```
1: Train Dataset:  $\mathcal{X} = \{(X_i, Y_i)\}_{i=1}^N$ ,  $X_i \in \mathbb{R}^{H \times W \times D}$ ,  $Y_i \in \{0, 1\}^{\text{NumClass} \times H \times W \times D}$ 
2: NumSamples= $N$ , BatchSize= $B$ , Target Model:  $\mathcal{M}_\theta$ , AT Robust Model:  $\mathcal{M}_{\bullet}$ 
3: for  $i \leftarrow 1$  to NumEpochs do
4:   for  $j \leftarrow 1$  to  $\lfloor N/B \rfloor$  do
5:     Sample a mini-batch  $\mathcal{B} \subseteq \mathcal{X}$  of size  $B$ 
6:      $X' \leftarrow \mathbf{VAFA}(X, Y) \quad \forall (X, Y) \in \mathcal{B} \quad \triangleright$  Adv. Freq. Attack on clean images.
7:      $\mathcal{L} = \mathcal{L}_{\text{dice}}(\mathcal{M}_\theta(X), Y) + \mathcal{L}_{\text{dice}}(\mathcal{M}_\theta(X'), Y) + \mathcal{L}_{\text{fr}}(\mathcal{M}_\theta(X), \mathcal{M}_\theta(X'))$ 
8:     Backward pass and update  $\mathcal{M}_\theta$ 
9:   end for
10: end for
11:  $\mathcal{M}_{\bullet} \leftarrow \mathcal{M}_\theta \quad \triangleright$  AT robust model after training completion.
12: Return  $\mathcal{M}_{\bullet}$ 
```

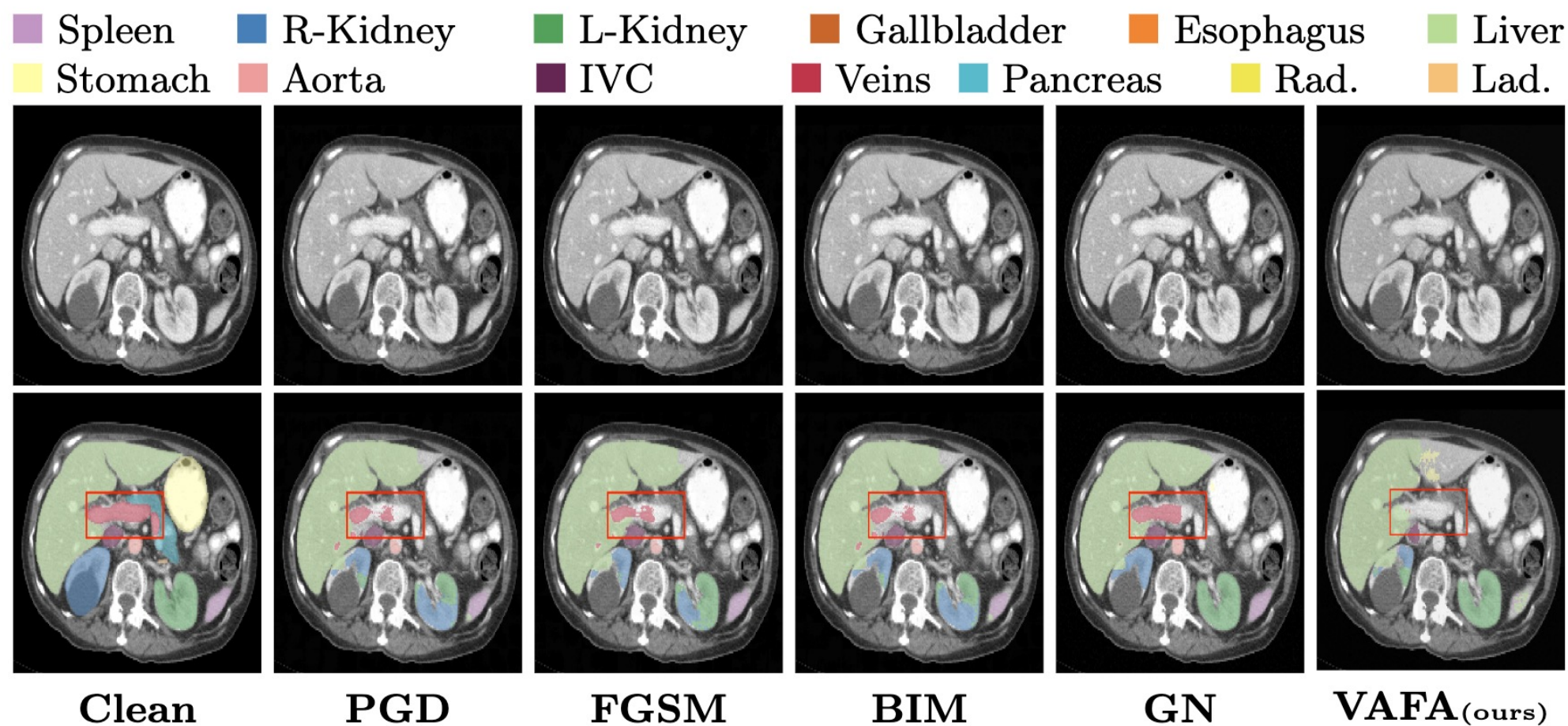
Results

- **Segmentation Models:** UNETR and UNETR++
- **Datasets:** Synapse and ACDC
- **Baseline Attacks:** PGD, FGSM, BIM, GN
- **Evaluation Metrics:** Dice Score, HD95 Distance, LPIPS
- **Programming Framework:** Pytorch

Volumetric Adversarial Frequency Domain Attack (VAFA)

Models → Attacks ↓		UNETR			UNETR++		
	DSC↓	HD95↑	LPIPS↑		DSC↓	HD95↑	LPIPS↑
Clean Images	74.3	14.0	-		84.7	12.7	-
PGD ($\epsilon = 4/8$)	62.7/50.8	40.4/64.5	98.9 /95.3		77.5/67.1	48.1/78.3	95.7/85.1
FGSM ($\epsilon = 4/8$)	62.8/53.9	34.8/48.7	98.8/94.7		73.1/67.1	37.3/43.2	94.7/82.2
BIM ($\epsilon = 4/8$)	62.8/50.7	39.9/ 65.8	98.8/95.3		77.3/66.8	46.6/78.1	95.8 /85.3
GN ($\sigma = 4/8$)	74.2/73.9	17.0/15.4	97.7/91.1		84.7/84.3	12.3/13.4	93.3/78.2
VAFA ($q_{\max} = 20/30$)	32.2/29.8	57.6 /59.9	97.5/ 96.9		45.3/39.3	73.9/85.2	94.2/ 94.7

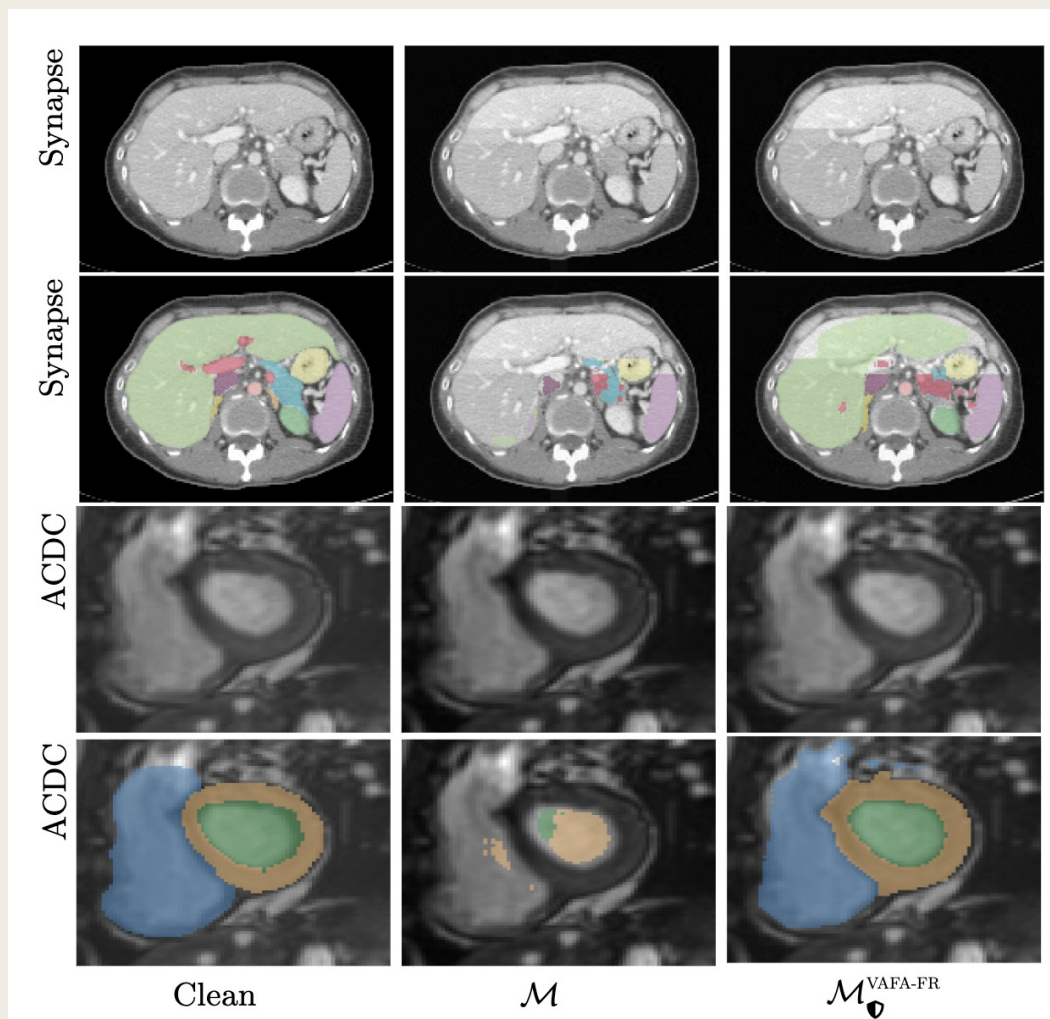
Volumetric Adversarial Frequency Domain Attack (VAFA)



Volumetric Adversarial Frequency Domain Training (VAFT)

	Attacks → Models ↓	UNETR					UNETR++				
		Clean	PGD	FGSM	BIM	VAFA	Clean	PGD	FGSM	BIM	VAFA
Synapse	\mathcal{M}^{PGD}	73.47	65.53	65.68	65.51	42.47	75.43	67.81	67.82	67.80	38.22
	$\mathcal{M}^{\text{FGSM}}$	72.44	64.80	66.31	64.76	39.02	81.06	73.84	74.76	73.77	37.48
	\mathcal{M}^{BIM}	75.12	67.78	68.32	67.78	45.97	74.80	67.58	67.46	67.57	35.72
	\mathcal{M}^{GN}	73.17	61.40	61.77	61.29	30.00	80.05	76.23	70.96	74.51	41.44
	$\mathcal{M}^{\text{VAFA}}$	74.67	64.83	65.49	64.73	66.31	81.88	69.09	65.40	68.90	76.47
	$\mathcal{M}^{\text{VAFA-FR}}$	75.66	65.90	66.79	65.83	66.33	82.65	70.61	67.00	70.41	78.19
ACDC	$\mathcal{M}^{\text{VAFA}}$	81.95	60.77	68.16	60.75	69.76	89.00	76.28	80.41	76.56	88.45
	$\mathcal{M}^{\text{VAFA-FR}}$	83.44	60.63	69.33	60.61	73.05	91.36	85.42	87.42	83.90	91.23

Volumetric Adversarial Frequency Domain Training (VAFT)



Github Repository

<https://github.com/asif-hanif/vafa>

Thank you !

VAFA – Ablation Study

- Impact of quantization threshold
- Impact of steps
- Impact of patch size

VAFA – Ablation Study

Impact of Quantization Threshold: q_{\max}

q_{\max}	DSC↓	HD95↑	LPIPS↑
-	74.31	14.03	-
10	65.95	26.25	99.10
20	56.24	35.92	98.70
30	50.96	44.09	98.33
40	49.58	43.66	97.90
60	48.83	44.55	96.60
80	48.76	45.30	94.50

VAFA – Ablation Study

Impact of Steps

Steps	DSC↓	HD95↑	LPIPS↑
-	74.31	14.03	-
10	61.33	33.20	98.85
20	56.24	35.92	98.70
30	54.37	38.00	98.64
40	53.31	37.76	98.59
50	52.97	39.23	98.54
60	52.25	39.19	98.52

VAFA – Ablation Study

Impact of Patch-Size

Patch Size	DSC↓	HD95↑	LPIPS↑
-	74.31	14.03	-
(4 × 4 × 4)	63.48	32.63	98.90
(8 × 8 × 8)	56.24	35.92	98.70
(16 × 16 × 16)	41.30	45.98	98.14
(32 × 32 × 32)	32.40	56.64	97.49
(48 × 48 × 48)	28.19	66.08	97.16
(96 × 96 × 96)	28.08	59.09	96.47