GUARDIAN: Guarding Against Uncertainty and Adversarial Risks in Robot-Assisted Surgeries

Supplementary Material



Fig. A. Visualizing clean versus adversarial perturbed images with alpha and epsilon values set to 8 and a standard deviation of 0.1.



Fig. B. We conducted an ablation study to obtain optimal learning rates for the RDV model against the mAP for the triplet recognition across the initial 12 epochs, focusing on the volatility of learning rate trends to establish a robust starting point.

Table A. Impact of ep-

silon on image quality.

 $\frac{1}{|Attack|Epsilon|PSNR SSIM LPIPS|}$ On i

 Table B. Impact of alpha on image quality.
 Table C. Impact of steps on image quality.

	4	90 74	0 0 70	0.001										
PGD	4	38.14	0.878	0.901	Attack	Alpha	DSNR	SSIM	LDIDS					
	8	33.68	0.732	0.833	Attack	Alpha i Sidit SSIM LI II S				Attack	Num Step	s PSNR	SSIM	LPIPS
	16	28.70	0.517	0.769	DOD	2	33.68	0.732	0.833	PGD		94.14	0 740	0.947
	32	23.34	0.285	0.684		4	33.05	0.710	0.819		10	33.68	0.740	0.847
FGSM	4	36.59	0.843	0.876	PGD	8	32.04	0.675	0.803		15	33.49	0.726	0.827
	8	30.59	0.629	0.782		16	30.59	0.616	0.776		20	33.40	0.724	0.824
	16	24.64	0.364	0.669		2	34.31	0.762	0.840	BIM	5	33.52	0.732	0.829
	32	18.81	0.169	0.521		4	33.06	0.716	0.823		10	33.06	0.716	0.822
	4	1 90 00	0.900	0.000	BIM	8	39.11	0.687	0.805		15	33.16	0.720	0.823
BIM	4	38.08	0.869	0.895		0	02.11	0.007	0.805		20	33.10	0.718	0.822
	8	33.06	0.716	0.822		16	30.59	0.632	0.778					
	16	28.45	0.529	0.745										
	32	26.68	0.460	0.710										



Fig. C. Visualization of the impact of hyperparameters from PGD, FGSM, and BIM attacks on the RDV model's mAP: varying epsilon (row one), alpha (row two), and steps (row three), with other parameters held constant in each row.