## Supplementary Materials for Medical image segmentation via single-source domain generalization with random amplitude spectrum synthesis\*

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Fig. 1. Visualisation of different hyperparameter settings in RASS.

Algorithm 1 RASS for 3D and 2D Medical Image

1: Input:  $x_s$  $\triangleright$  2D or 3D medical image 2: **RASS parameters:**  $\alpha, \beta, \gamma$  $\triangleright$  Set values 3: if  $\dim(\boldsymbol{x_s}) == 3$  then  $\triangleright$  Process as 3D image  $\mathcal{F}(\boldsymbol{x_s}) \leftarrow \text{FFT3D}(\boldsymbol{x_s})$ ▷ Obtain 3D Fourier spectrum 4:  $\mathcal{A}(\boldsymbol{x_s}), \mathcal{P}(\boldsymbol{x_s}) \leftarrow \operatorname{Abs}(\mathcal{F}(\boldsymbol{x_s}), \operatorname{Ang}(\mathcal{F}(\boldsymbol{x_s})) \ \triangleright \operatorname{Amplitude} \text{ and phase spectrum}$ 5:6:  $\sigma_{H \times W \times D} \leftarrow \text{Meshgrid}([-H/2, H/2], [-W/2, W/2], [-D/2, D/2])$ for  $m \in [-H/2, H/2]$  do 7:for  $n \in [-W/2, W/2]$  do 8: for  $p \in [-D/2, D/2]$  do 9:  $\sigma[m,n,p] \leftarrow \left(2\alpha \sqrt{\tfrac{m^2+n^2+p^2}{H^2+W^2+D^2}}\right)^{\gamma} + \beta$  $\triangleright$  Calculate perturbation  $\sigma$ 10:  $\boldsymbol{\delta}_{H \times W \times D} \sim \mathcal{N}(1, \boldsymbol{\sigma}_{H \times W \times D}^2)$  $\triangleright$  Sample 11: $\mathcal{A}(\boldsymbol{x_s}) \leftarrow \mathrm{FFTShift}(\mathcal{A}(\boldsymbol{x_s}))$ 12: $\tilde{\mathcal{A}}(\boldsymbol{x_s}) \leftarrow \boldsymbol{\delta}_{H \times W \times D} \odot \mathcal{A}(\boldsymbol{x_s})$ 13:▷ Synthesize amplitude spectrum 14:  $\tilde{\mathbf{x}} \leftarrow \text{Inverse-FFT3D}(\tilde{\mathcal{A}}(\boldsymbol{x_s}), \mathcal{P}(\boldsymbol{x_s}))$  $\triangleright$  Recover the image 15: else if  $\dim(\boldsymbol{x_s}) == 2$  then ▷ Process as 2D image  $\mathcal{F}(\boldsymbol{x_s}) \leftarrow \mathrm{FFT2D}(\boldsymbol{x_s})$ 16: $\triangleright$  Obtain 2D Fourier spectrum  $\mathcal{A}(\boldsymbol{x_s}), \mathcal{P}(\boldsymbol{x_s}) \leftarrow \operatorname{Abs}(\mathcal{F}(\boldsymbol{x_s})), \operatorname{Ang}(\mathcal{F}(\boldsymbol{x_s})) \triangleright \operatorname{Amplitude} \text{ and phase spectrum}$ 17: $\sigma_{H \times W} \leftarrow \text{Meshgrid}(-H/2, H/2, -W/2, W/2)$ 18:for  $m \in [-H/2, H/2]$  do 19:for  $n \in [-W/2, W/2]$  do 20: $\sigma[m,n] \leftarrow \left(2\alpha \sqrt{\frac{m^2+n^2}{H^2+W^2}}\right)^{\gamma} + \beta$  $\triangleright$  Calculate perturbation  $\sigma$  for 2D 21: $\delta_{H \times W} \sim \mathcal{N}(1, \sigma_{H \times W}^2)$  $\mathcal{A}(\boldsymbol{x}_s) \leftarrow \text{FFTShift}(\mathcal{A}(\boldsymbol{x}_s))$ 22: $\triangleright$  Sample 23:24: $\tilde{\mathcal{A}}(\boldsymbol{x_s}) \leftarrow \boldsymbol{\delta}_{H imes W} \odot \mathcal{A}(\boldsymbol{x_s})$  $\triangleright$  Synthesize amplitude spectrum  $\tilde{\mathbf{x}} \leftarrow \text{Inverse-FFT2D}(\tilde{\mathcal{A}}(\boldsymbol{x_s}), \mathcal{P}(\boldsymbol{x_s}))$ 25: $\triangleright$  Recover the image 26: end if

Table 1. Ablation study on different backbone.

Backbone	FeTA2021	IOSTAR	LES-AV
U-Net	$76.03. \pm 0.36$	$65.33_{\pm 0.19}$	$72.33_{\pm 0.17}$
MedNeXt	$76.34. \pm 0.19$	$65.79 \pm 0.21$	$72.83 \pm 0.19$
SegResNet	$76.56_{\pm 0.23}$	$65.86_{\pm 0.12}$	$72.88_{\pm 0.07}$

**Table 2.** Ablation study of RASS on the FeTA2021 dataset.  $\gamma$  is fixed to 2.0.

parameter	$\beta = 0.15$	$\beta = 0.25$	$\beta = 0.45$
$\alpha = 2.0$	$76.51_{\pm 0.31}$	$76.32_{\pm 0.32}$	$76.16_{\pm 0.31}$
$\alpha = 3.0$	$76.52 \pm 0.28$	$76.56_{\pm 0.23}$	$76.08 \pm 0.26$
$\alpha = 9.0$	$75.52 \pm 0.29$	$75.67_{\pm 0.25}$	$75.31_{\pm 0.33}$



Fig. 2. Visualization of images after different mask sizes and numbers in RMS.