

Computational tissue staining of non-linear multimodal imaging using supervised and unsupervised deep learning: supplement

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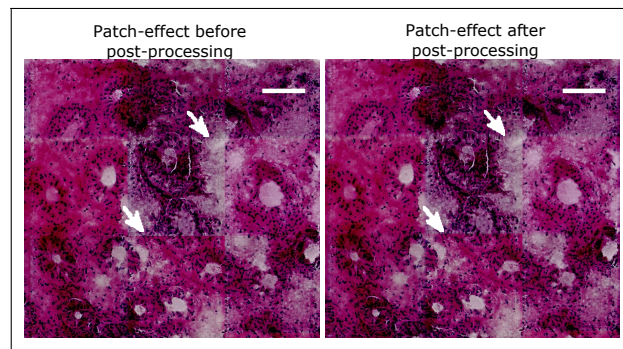


Fig. S1. This figure shows the effect of post-processing for removing 'patch-effect'. The 'patch-effect' was removed by interpolating pixel values of three neighbouring pixels at the end of every patch (256th pixel). This effect (shown in white arrows) was visible for few images and its removal did not significantly affect the performance metrics. The scale bar represents 100 μm

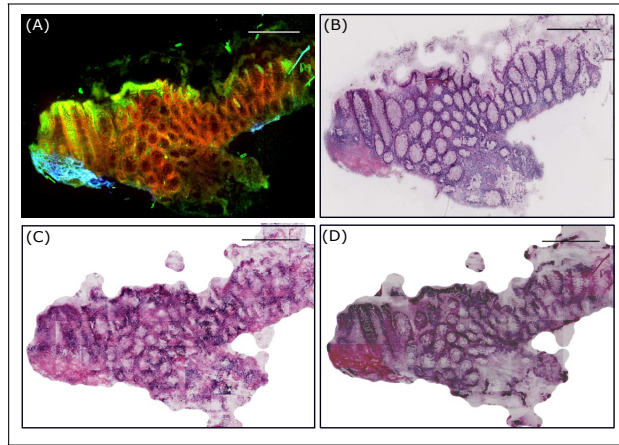


Fig. S2. (A) shows a non-linear multimodal image from test dataset, (B) visualizes corresponding histopathologically stained H&E image (unregistered), (C) shows the computational H&E image by the Pix2Pix (MSE = 4.4×10^3 , SSIM = 0.65, CSS = 0.94) and (D) depicts computational H&E image by the cycle CGAN model (MSE = 8.4×10^3 , SSIM = 0.63, CSS = 0.94). The contrast of the computational H&E image in (D) is reduced by a factor of 0.7. The images here are downsampled to 20% of original size for clarity. The scale bar represents 100 μm .

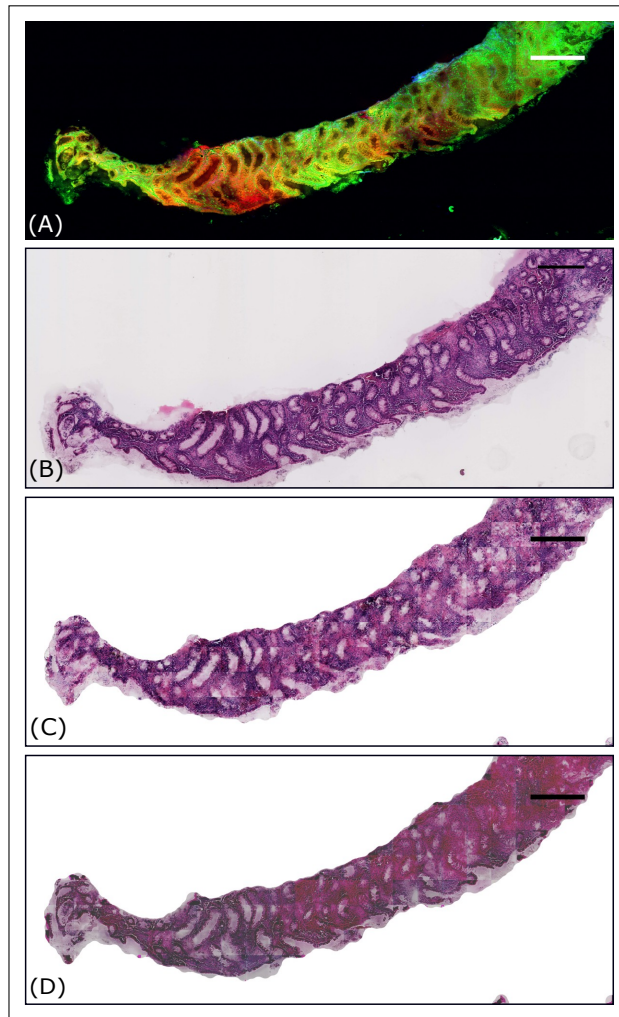


Fig. S3. (A) shows a non-linear multimodal image from training dataset, (B) visualizes corresponding histopathologically stained H&E image (unregistered), (C) shows the computational H&E image by the Pix2Pix (MSE = $2.8 \cdot 10^3$, SSIM = 0.74, CSS = 0.96) and (D) depicts computational H&E image by the cycle CGAN model (MSE = $5.9 \cdot 10^3$, SSIM = 0.72, CSS = 0.94). The contrast of the computational H&E image in (D) is reduced by a factor of 0.7. The images here are downsampled to 20% of original size for clarity. The scale bar represents $100 \mu\text{m}$.

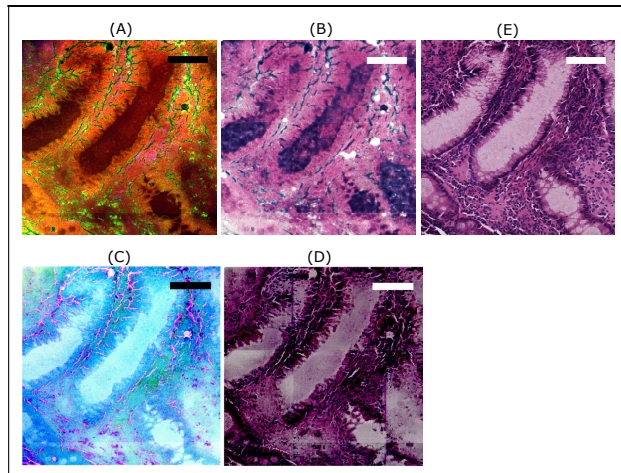


Fig. S4. This figure shows effect of ‘contrast inversion’ on the computational H&E images of the cycle CGAN model. (A) shows an original MM patch given to the cycle CGAN model and (B) visualizes the generated H&E patch. Here, we see that the colors in the generated H&E image are inverted when compared to the pathologically stained H&E patch shown in (E). Contrarily, (C) shows a contrast inverted MM patch and (D) depicts the generated H&E patch by the cycle CGAN model. Here, the color contrast is similar to the pathologically stained H&E patch shown in (E). The scale bar represents 100 μm .

Table S1. The quantitative metrics namely mean squared error (MSE), structure similarity index (SSIM) and color similarity index (CSS) evaluated for 19 images from the training and testing dataset is given for the Pix2Pix and the cycle CGAN models, respectively.

Image	Pix2Pix			Cycle CGAN		
	MSE	SSIM	CSS	MSE	SSIM	CSS
Train 1	3601.43	0.61	0.94	9373.09	0.57	0.92
Train 2	2800.83	0.74	0.96	5890.62	0.72	0.94
Train 3	4984.19	0.38	0.89	13955.34	0.33	0.89
Train 4	3337.39	0.56	0.93	8157.08	0.53	0.93
Train 5	4451.48	0.51	0.92	11509.74	0.47	0.90
Train 6	5100.78	0.49	0.92	9629.29	0.46	0.89
Train 7	4956.57	0.41	0.91	10389.17	0.38	0.90
Train 8	4167.94	0.52	0.92	8376.58	0.49	0.90
Train 9	7938.24	0.46	0.91	15823.56	0.41	0.91
Train 10	7657.41	0.39	0.88	9725.03	0.36	0.88
Train 11	3066.22	0.67	0.94	6749.37	0.65	0.93
Train 12	3057.30	0.65	0.94	8129.79	0.62	0.93
Train 13	5918.03	0.36	0.90	15783.79	0.30	0.90
Test 1	5617.82	0.60	0.92	6600.03	0.59	0.91
Test 2	3582.64	0.61	0.94	6633.59	0.60	0.94
Test 3	3827.09	0.65	0.94	4767.95	0.63	0.94
Test 4	3292.67	0.60	0.94	10118.23	0.58	0.93
Test 5	4416.22	0.64	0.94	8408.46	0.63	0.94
Test 6	4937.68	0.49	0.91	10265.39	0.48	0.93