Supporting Information

Acceleration of high-quality Raman imaging via locality enhanced Transformer network

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Figure S1: Qualitative comparison of LETNet and bicubic, nearest neighbor, EDSR, RCAN, ELAN on 4× Raman image SR of example 1. The result images are shown through pseudo-color images. The acquisition time of HR image and SR image are 68.25 minutes and 4.27 minutes respectively.



Figure S2: Qualitative comparison of LETNet and bicubic, nearest neighbor, EDSR, RCAN, ELAN on 4× Raman image SR of example 2. The result images are shown through pseudo-color images. The acquisition time of HR image and SR image are 68.25 minutes and 4.27 minutes respectively.



Example3 HR



Bicubic





Nearest Neighbour





Figure S3: Qualitative comparison of LETNet and bicubic, nearest neighbor, EDSR, RCAN, ELAN on 4× Raman image SR of example 3. The result images are shown through pseudo-color images. The acquisition time of HR image and SR image are 68.25 minutes and 4.27 minutes respectively.



Example1 HR



ELAN



Figure S4: Grayscale image of HR image and SR images of ELAN and LETNet. Raman image at 1440 cm⁻¹

(represents lipids of breast cancer cells) was chosen to show difference.



Figure S5: The training loss and validation curves before 600 epochs. (a) The red curve is the loss-epoch curve of training set. (b) The blue curve is the loss-epoch curve of validation set.



Figure S6: Comparison result of another test example. The first column is corresponding whole HR image, and other columns contain the HR image and SR images within the red rectangle.



Figure S7: The whole differential images of HR image and SR images of different methods. The brighter pixels in the image represent more errors in SR images.

Supplementary Tables

The performance of LETNet with depth-wise convolution of different kernel size on $2\times$, $4\times$ and $8\times$ hyperspectral Raman image SR was presented in Table S1. The using of kernel size in this experiment comprehensively considers the SR performance of the model, including PSNR, SSIM and Params. The best results are shown in bold.

 Table S1: SR performance of depth-wise convolution with different kernel size on hyperspectral Raman image of breast cancer cells.

Methods	Scale	PSNR	SSIM	Params(K)
LETNet-3×3	2×	44.46	0.9624	17906
LETNet-7×7		44.34	0.9623	18146
LETNet-11×11		44.74	0.9666	18578
LETNet-13×13		44.54	0.9637	18866
LETNet-3×3	4×	42.01	0.9416	44912
LETNet-7×7		42.14	0.9436	45152
LETNet-11×11		42.05	0.9448	45584
LETNet-13×13		41.96	0.9436	45872
LETNet-3×3	8×	40.27	0.9292	152936
LETNet-7×7		40.36	0.9291	153176
LETNet-11×11		40.08	0.9267	153608
LETNet-13×13		39.93	0.9233	153896

The performance of LETNet with depth-wise convolution of different kernel size on 2- 8× threechannel Raman image SR. The highest PSNR and SSIM are achieved when the kernel size is 11, demonstrated that large DW convolution is efficient for SR of this dataset [1]. The best results are shown in bold.

Methods	Scale	PSNR	SSIM	Params(K)
LETNet-3×3	2×	47.04	0.9852	403
LETNet-7×7		47.24	0.9854	461
LETNet-11×11		47.26	0.9855	565
LETNet-13×13		47.19	0.9854	634
LETNet-3×3	4×	40.12	0.9449	404
LETNet-7×7		40.23	0.9462	462
LETNet-11×11		40.27	0.9464	566
LETNet-13×13		40.25	0.9463	635
LETNet-3×3	8×	36.06	0.896	408
LETNet-7×7		36.10	0.8965	466
LETNet-11×11		36.13	0.8971	570
LETNet-13×13		36.09	0.8964	639

 Table S2: SR performance of depth-wise convolution with different kernel size on three-channel Raman image of brain tumor tissues.

Supplementary Reference

[1] Ding, X., Zhang, X., Zhou, Y., Han, J., Ding, G., & Sun, J. (2022). Scaling Up Your Kernels to 31×31: Revisiting Large Kernel Design in CNNs. 2022 IEEE/CVF Conference on Computer Vision and Pattern Recognition (CVPR), 11953-11965.