## 1 Supplementary Information for:

First experience in clinical application of hyperspectral endoscopy for evaluation of
 colonic polyps

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#### 20 Supplementary Methods

#### 21 Photobiological safety evaluations

22 In order to address photobiological safety of HySE, we compared the spectral irradiance and illumination angle of HySE to that of an FDA-approved clinical colonoscope 23 (Olympus Colonoscope CF-H290). For the spectral irradiance calculation, spectral profiles 24 25 and optical power were measured. Spectral profiles of a supercontinuum source coupled to HySE and illumination light of the colonoscope were measured using a spectrometer 26 (AvaSpec-ULS2048L, Avantes). For the optical power measurement in the worst-case 27 scenario, we used a calibrated thermal power meter (A-01-D12-USB, Laserpoint) and put the 28 distal end of the HySE or clinical colonoscope directly against the surface of the power meter. 29 Then, spectral irradiance was calculated by the following equation: 30

31 Spectral irradiance( $\lambda$ ) = ( $S(\lambda) \times P$ )/( $A \times D$ )

32 , where  $S(\lambda)$  is the signal (a.u.) at a wavelength of  $\lambda$  measured by the spectrometer, *P* is an 33 optical power (mW) of the illumination measured by the calibrated thermal power meter, *A* is 34 the total spectral area under the curve (a.u.×nm), and *D* is the area (cm<sup>2</sup>) of the calibrated 35 thermal power meter. The spectral irradiance of the compact supercontinuum source was 36 within that of the clinically approved colonoscope (Fig. 1d).

For the measurement of the illumination angle, we measured diameter of the illumination light at a known working distance (Fig. 1e). The illumination angle was determined by the following equation:

41 , where  $\alpha$  is a half illumination angle, *WD* is working distance, and *r* is a radius of the 42 illuminating area. The measured illumination angles of HySE and the colonoscope were 53.2° 43 - 65.0° and > 150°, respectively, which means that HySE illumination was again within that of 44 the clinical colonoscope.

Photothermal safety was also tested to check whether HySE could damage on epithelial tissue. The surface temperature of the polyscope was measured for 1 hour every 10 minutes using a non-contact thermometer (RS-8662, RS Components). The measured surface temperature of the polyscope coupled to the supercontinuum source and room temperature were  $22.09 \pm 0.26$  and  $22.1 \pm 0.17$ °C respectively, which demonstrates that HySE will not introduce any thermal risks related to high surface temperature during the endoscopy procedure (Fig. 1f). As the supercontinuum laser is a nanosecond laser source (pulse duration: ~ 1 ns, pulse duration: ~10kHz), the single-pulse energy was calculated to confirm laser safety under clinical conditions. The single-pulse energy was estimated to be  $8.56 \times 10^{-3}$  mJ/cm<sup>2</sup>, which is much lower than maximum permissible exposures (0.02 J/cm<sup>2</sup>) recommended by the American National Standards Institute and International Electotechnical Commission.

#### 57 Data analysis

58 Matlab R2020a and imageJ were used for hyperspectral image processing. GraphPad 59 Prism was used for statistical analysis. In total, 92 measurements obtained from eight patients 60 were analysed, and each measurement acquired 100 spectral images consisting of 200 61 spectral signals. Therefore, in total 200 × 100 × 92 spectra were analysed.

#### 62 Wide-area hyperspectral image reconstruction

The overall hyperspectral image reconstruction process was described previously<sup>1</sup> 63 and was tested for this study to evaluate whether it was sufficient to operate under clinical 64 conditions, which provides low contrast images and random image distortions during 65 measurement. Briefly, geometric transformation matrices (GMs) were estimated by extracting 66 and comparing scale-invariant features from the wide-field colour images. Then, the wide-area 67 hyperspectral image was reconstructed using the spectral images and estimated GMs. The 68 69 identification of regions of interest containing normal mucosa (8 patients), polyp (5 patients), 70 and post-resection tissue (2 patients), were confirmed by clinicians using a combination of the colonoscopy video and the wide-field colour images; locations were then mapped onto the 71 72 reconstructed hyperspectral images.

#### 73 Reflection and absorption calculation

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In order to calculate reflectance, measured spectra were normalised according to:

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$$\operatorname{Reflectance}(x, y, \lambda) = \frac{I_{\operatorname{sample}}(x, y, \lambda) - I_{dark}(x, y, \lambda)}{I_{white}(x, y, \lambda) - I_{dark}(x, y, \lambda)}$$

76 , where  $I_{sample}(x,y,\lambda)$  is the measured intensity at the wavelength of  $\lambda$  at the point x and y in the 77 image, and  $I_{dark}$  is the dark counts measured from the EMCCD in the absence of illumination, 78 and  $I_{white}$  is the intensity measured from a standard white reflectance target (Spectralon Diffuse 79 Reflectance Standards, LabSphere) using the white light source from the standard-of-care 80 colonoscope.  $I_{white}$  and  $I_{dark}$  were measured in every experiment to avoid variations introduced 81 depending on experimental conditions.

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Absorbance was calculated by taking the logarithm of reflectance via the following

83 equation:

Absorbance(x, y, 
$$\lambda$$
) =  $-\log_{10}\left(\frac{I_{sample}(x, y, \lambda) - I_{dark}(x, y, \lambda)}{I_{white}(x, y, \lambda) - I_{dark}(x, y, \lambda)}\right)$ 

In hyperspectral image analysis, reflectance and absorbance were selectively exploited for better data interpretation. As the measured intensity changes depending on the distance between the illumination fibre and the target, it is difficult to accurately compute the reflectance and absorbance under *in vivo* conditions. In our previous report<sup>2</sup>, we found that inaccurate intensity measurement alters the scale and offset of reflectance and absorbance, respectively. Therefore, reflectance was exploited for quantitative analysis, and absorbance was shown when the graph shape is important.

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### 93 Coefficient of variation

The coefficient of variation (CV) was calculated to evaluate intra- / inter-patient variability in hyperspectral imaging quality. Hyperspectral images from normal mucosa (5 patients) and polyps (2 patients), which were clearly diagnosed by the clinicians, were selected for the accurate calculation of the variability. In a patient, 10 - 30 spectral images which has 10 - 20 spectral data points were tested for calculating CV. Intra- / inter-patient CVs were calculated via following equations<sup>3</sup>:

100 Intra - patient 
$$CV(\lambda) = \sum_{m=1}^{N} \frac{\text{standard deviation}\left(\text{Reflectance}_{m}(\lambda)\right)}{\text{mean}(\text{Reflectance}_{m}(\lambda))} \times 100$$

101 Inter - patient 
$$CV(\lambda) = \sum_{p=1}^{N} \frac{\text{standard deviation } (\text{mean}\left(\text{Reflectance}_{p}(\lambda)\right)}{\text{mean}(\text{mean}(\text{Reflectance}_{p}(\lambda)))} \times 100$$

102 , where m and p indicate a spectral image number in a patient and patient number, 103 respectively.

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#### 105 Spectral angle mapping

106 For quantitative comparison of spectra acquired from different tissue types, spectral 107 angle mapping (SAM) was exploited<sup>4</sup>. The spectral angle,  $\alpha$ , was calculated by the following 108 equation:

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$$\alpha = \cos^{-1}\left(\frac{\sum_{\lambda=1}^{n} t_{\lambda} r_{\lambda}}{\left(\sum_{\lambda=1}^{n} t_{\lambda}^{2}\right)^{0.5} \left(\sum_{\lambda=1}^{n} r_{\lambda}^{2}\right)^{0.5}}\right)$$

, where  $t_{\lambda}$  and  $r_{\lambda}$  are values of target and reference spectral signals at the wavelength of  $\lambda$ , 110 respectively. n indicates the total number of spectral samples. The average spectra of normal 111 tissue, polyps and post-resection tissue were used as the reference spectral signals in order 112 to compare differences among these tissue types. Therefore, three spectral angles were 113 calculated from one spectral signal via exploiting three reference spectral signals 114 (Supplementary Fig 1a,b). To enhance the visualisation of SAM results, a synthetic RGB SAM 115 image was created by assigning green, blue, and red colours to the three SAM values obtained 116 via the reference spectral signals of normal mucosa, polyp, and post-resection tissue, 117 respectively. However, SAM analysis shows small values if the target and reference signals 118 are similar; thus, 1-SAM values were assigned as RGB values of a synthetic RGB SAM image 119 to make normal mucosa, polyps, and post-resection tissue as green, blue, and red colours. 120 (Supplementary Fig 1c-e). 121

#### 122 Machine learning

123 To investigate the potential for discrimination of the different tissue types from which spectra were collected, we employed a k-nearest neighbour (k-NN, k = 4) classification model. 124 The k-NN algorithm predicts the class of newly observed data by choosing the most frequent 125 class labels of k-nearest neighbour data points in the feature space<sup>5,6</sup>. We exploited SAM 126 values obtained by calculating three reference signals of normal mucosa, polyps, and post-127 resection tissue, which clearly shows a distinct distribution of spectral signals of each tissue 128 type. The established classifier was validated via k-fold cross-validation using data that was 129 not utilized for training. The training, cross-validation, and prediction were performed using 130 Matlab function: fitcknn, crossval, and predict, respectively. 131

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## 151 Supplementary Figures



Supplementary Fig. 1. Creation of a synthetic RGB SAM image. a A min-max normalized
reflectance signal obtained from polyp. b Three reference spectral signals for SAM calculation.
c Spectral angles calculated based on three references. d Bar graph showing 1-SAM results.
e Representative images of green (normal mucosa), blue (polyp) and red (post-resection
tissue) channels and synthetic RGB image to aid visualisation of the spatial data.

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Supplementary Fig. 2. Wide-field white-light images of polyps. Blue dashed lines indicatepolyps diagnosed by the clinicians.





## 166 Supplementary Movies (single frame shown for reference)



**Supplementary Movie 1.** Wide-field image registration process by extracting features and 169 calculating geometric transformation matrices.



**Supplementary Movie 2.** Slice images of the reconstructed wide-area hyperspectral image.



- **Supplementary Movie 3.** A representative movie that shows wide-field image scanning and
- 177 quantitative analysis of corresponding spectral images based on SAM and kNN.

## 180 Supplementary Table

# 181 Supplementary Table 1. Summary of recruited patients

Patient	Date	Optic Fibre	Notes
number			
P1	26.Aug.2019	Optic Fibre A	System optimization
P2	27.Aug.2019	Optic Fibre A	System optimization
P3	15.Jan.2020	Optic Fibre A	Polyp
P4	15.Jan.2020	Optic Fibre B	Polyp
P5	17.Jan.2020	Optic Fibre A	Fail to data acquisition
P6	17.Jan.2020	Optic Fibre B	Normal
P7	21.Jan.2020		Didn't perform endoscopy
P8	22.Jan.2020	Optic Fibre A	Polyp
P9	22.Jan.2020	Optic Fibre B	Normal
P10	22.Jan.2020		Didn't perform endoscopy
P11	24.Jan.2020		Didn't perform endoscopy
P12	27.Jan.2020	Optic Fibre A	Polyp
P13	27.Jan.2020	Optic Fibre B	Polyp

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