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Comparability of Raman Spectroscopic Configurations: A LargeScale Cross-Laboratory Study

Shuxia Guo

Institute of Physical Chemistry and Abbe Center of Photonics, University Jena, Germany.

Hugh Byrne

Technological University Dublin, hugh.byrne@tudublin.ie

Shiyamala Duraipandian

Technological University Dublin

See next page for additional authors

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Authors

Shuxia Guo, Hugh Byrne, Shiyamala Duraipandian, Fiona Lyng, Claudia Beleites, and Ute Neugebauer

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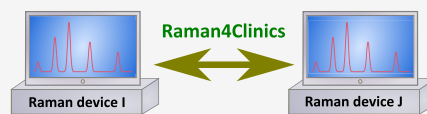
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ABSTRACT: The variable configuration of Raman spectroscopic platforms is one of the major obstacles in establishing Raman spectroscopy as a valuable physicochemical method within real-world scenarios such as clinical diagnostics. For such real world applications like diagnostic classification, the models should ideally be usable to predict data from different setups. Whether it is done by training a rugged model with data from many setups or by a primary-replica strategy where models are developed on a 'primary' setup and the test data are generated on 'replicate' setups, this is only possible if the Raman spectra from different setups are consistent, reproducible, and comparable. However, Raman spectra can be highly sensitive to the measurement conditions, and they change from setup to setup even if the same samples are measured. Although increasingly recognized as an issue, the dependence of the Raman spectra on the instrumental configuration is far from being fully understood and great effort is needed to address the resulting spectral variations and to correct for them. To make the severity of the situation clear, we present a round robin experiment investigating the comparability of 35 Raman spectroscopic devices with different configurations in 15 institutes within seven European countries from the COST (European Cooperation in Science and Technology) action Raman4clinics. The experiment was developed in a fashion that allows various instrumental configurations ranging from highly confocal setups to fibre-optic based systems with different excitation wavelengths. We illustrate the spectral variations caused by the instrumental configurations from the perspectives of peak shifts, intensity variations, peak widths, and noise levels. We conclude this contribution with recommendations that may help to improve the inter-laboratory studies.



INTRODUCTION

Raman spectroscopy is known as a noninvasive, label-free technology with high selectivity, and thus, it has been increasingly applied to biological studies.^{1–3} These studies include forensics,^{4,5} diagnostics,^{6–9} metabolism research,^{10–12} microbiology,^{13,14} clinical pharmacology,^{15–17} and food sci-

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Table 1. Number of Spectra for each Substance from each setup^a

setup ID	Wl./nm	#NeAr	#agar	#gelatine	#paracetamol	#polystyrene	#cyclohexane
ID01	514	10			10	10	10
ID02	514			10	10		
ID03	515	711			1053	709	361
ID04	532	200	10	300	30	300	100
ID05	532	151	135	116	121	11	177
ID06	532					20	
ID07	532	30	135	10	80	12	
ID08	532					120	
ID09	532			10			
ID10	532	11	12	12	13	11	
ID11	532	10	12	10			
ID12	532		12	10	12		12
ID13	532	200	70	190	10	100	200
ID14	532	1000	60	100	985	900	500
ID15	780	30	22		30		
ID16	785	99		60	150	20	
ID17	785	10		10	10	10	10
ID18	785						30
ID19	785			10			
ID20	785	40		100	40		40
ID21	785	10	10	10	10	10	10
ID22	785	10	10	10	10	10	
ID23	785	12	11		20	30	50
ID24	785			875	20		50
ID25							
ID26	785	200	58	93	4139	208	50
ID27	785	100	100	100	100	100	100
ID28	785		10				
ID29	785	10		10	10		10
ID30	785	11	10	11	12	11	
ID31	785	251	76	348	164	151	26
ID32	785	10			10	10	10
ID33	785					601	814
ID34	785		23	10	17		14
ID35	785					21	

^aThe setup ID was sorted according to the excitation wavelength.

ence.^{18,19} Most of these studies are yet at the proof-of-concept stage and are performed using a single or multiple similar Raman spectrometer(s). While the Raman spectra from any sample contain the vibrational fingerprint information of the molecules within the sample, they unfortunately also contain fingerprints of the analytical setup, e.g., Raman spectrometer, itself.²⁰ Therefore, the same sample can lead to different Raman spectra if measured on multiple setups, in different conditions, or at different times. To be clear, we will henceforth refer to ‘setup’ as an indication of all measurement related effects, be these temporal drifts, variations of measurement conditions, or instrumental configurations (viz. laser sources, spectral resolution).

Subtle differences in the Raman setup can degrade the reproducibility of the Raman spectroscopic signals, which makes it almost impossible to quantitatively compare measurements from different setups, and this is highly detrimental to cross-setup data analysis. The setup-dependence is very likely to have a larger influence in biological applications in which the spectral contributions of interest can be very subtle in the acquired Raman signal and thus are more easily masked. If Raman spectroscopy is expected to be utilized in, e.g., clinical scenarios, it is likely that one or more ‘primary’ setups hold a large database

of Raman spectral characteristics of, e.g., some disease states. Other Raman setups in different laboratories act as ‘replicate’ setups and send data to the ‘primary’ setup for testing against this ‘primary’ model/database. If the Raman setups are significantly different, these statistical models are likely to fail to correctly predict the newly measured samples on the ‘replicate’ setups. Building up individual statistical models for each setup could be a solution to this problem, but this is not likely to be acceptable in real-world scenarios, as the models would also be setup-dependent and this would further hamper the generalization of Raman spectroscopy.

A more feasible and attractive approach is to remove the setup-induced spectral variations from the spectral ‘database’. This is the task of pre-processing in general. The most straightforward way to remove the confounding influence of a setup is spectrometer calibration, including wavenumber and intensity corrections.^{20–22} Spectrometer calibration corrects for the influence of a setup on the Raman signals by retrieving the correct values from the measurements of known standard materials. In many cases, an interpolation is needed to deal with the different nominal spectral resolutions of different setups.²³ The extent to which an instrument software takes care of the spectrometer calibration varies greatly across manufacturers; so

does the possibility to access raw data, which allows the users to apply their own calibration functions from a known basis. Spectrometer calibration does decrease the setup-dependence but rarely removes it completely, due to the inaccuracy in the estimated instrumental response function. The remaining setup-derived spectral variations can still negatively affect the reliability of data analysis. In addition, the standard materials used for calibration may differ across laboratories (especially for biological materials), which can introduce additional spectral variations among different setups.

In addition to robust spectrometer calibration, standard-free approaches such as warping methods in which the unwanted spectral variations are removed by aligning all spectra against a given reference spectrum can also be used.²⁴ However, as a purely data-driven procedure, it does not necessarily give meaningful results, as these procedures may remove both sample- and setup-induced variations without distinction. In this context, methods based on bilinear modeling, such as replicate EMSC (extended multiplicative signal correction) and ASCA (ANOVA-simultaneous component analysis), make it possible to estimate the setup-induced spectral differences and remove them from the data while preserving the sample-induced variations.^{25–27} This seems to be a more reliable mechanism, although it requires the experiments to be well designed. Such a method is also limited if the sample- and setup-induced spectral variations are not independent of each other.

Considering all the issues stated above, setup-dependence is still a significant concern and a major challenge to Raman spectroscopy being translated into real-world applications. To investigate this analytical challenge, we designed a round robin experiment with researchers from approximately 50 European institutes, which was initiated within the COST action Raman4Clinics. Briefly, aliquots of the same samples were prepared in one partner laboratory and sent to other laboratories in which data were collected using Raman spectrometers (see Table S1) from various manufacturers, with various laser sources, different spectral resolutions (i.e., pixel size), and optical configurations ranging from highly confocal to fibre-optic probes (and hence different numerical apertures). We note that, of course, there can be spectral differences due to resonance excitation, and therefore, for the present study, we have chosen Raman setups and samples absent of resonance Raman contributions. In the end, data from 35 setups of 15 institutes were returned to a single laboratory for unified data processing and assessments. The comparability of the setups was evaluated from the perspectives of the peak positions, peak intensities, peak widths, and noise levels. We report these differences and provide recommendations for future cross-laboratory studies. Ultimately, the translation of Raman spectroscopy as a bioanalytical protocol to clinical applications will require regulatory certification, as per, for example, the European Medicine Agency “Guideline on bioanalytical method validation”, which details requirements of accuracy and precision.²⁸ We hope that this study will trigger more unified efforts from the Raman community to come to some harmonization on handling setup-dependence, for instance, rules of setup standardization, certification of devices to be used in clinics, etc. In the end, we hope to help establish Raman spectroscopy as a reliable tool for real-world applications.

EXPERIMENTAL AND METHODS

Raman Spectroscopy. Details of the experimental design are given in the Supporting Information, and the spectral data

are available via an open data repository (<https://doi.org/10.5281/zenodo.4152953>). Briefly, we received data from 35 setups for the analysis in this contribution. The information of Raman setups is summarized in Table S2, which includes the manufacturer, the source wavelength, and the nominal spectral resolution (i.e., pixel size). Raman spectra of the NeAr glow lamp and different substances were measured, including agar, gelatine, paracetamol, polystyrene, and cyclohexane. The number of replicate spectra obtained using each setup from the different substances is given in Table 1. The spectral data from setup ID25 were excluded from the analysis, due to recording errors in the wavenumber axis (see Figure S1).

Spectral Pre-Processing. The spectra from the NeAr lamp were normalized with respect to their maxima without any additional processing. All spectra from the five substances (i.e., agar, gelatine, paracetamol, polystyrene, and cyclohexane) were subjected to the same pre-processing steps. The details can be found in the Supporting Information. Briefly, any cosmic spikes were removed via a comparison between every pair of Raman spectra of the same substance. A wavenumber calibration was performed based on the spectra of the standard material paracetamol. Thereafter, all spectra were interpolated to an equidistant wavenumber grid of 1 cm^{-1} . This is followed by a sensitive nonlinear iterative peak (SNIP²⁹) clipping algorithm to remove any broad baseline artifacts. In the end, a vector normalization (l_2 norm) was performed. Note, a wavenumber calibration was conducted on 23 of the instruments where paracetamol measurements were reported.

Spectral Characterization. In order to assess any variations between the different setups, the pre-processed Raman spectra were analyzed to investigate the following four aspects: signal-to-noise ratio (SNR), peak shift, peak width, and specific peak ratios. These four metrics were chosen to benchmark analytical sensitivity, the reproducibility of peak positions, the spectral resolution, and the reproducibility of relative intensities of a measurement. We consider these four properties to be the most important for establishing whether there are variations between the different Raman platforms. The definition of these characteristics and their calculation are summarized in the following.

Signal-to-Noise Ratio (SNR). The SNR can be used to determine the lowest signal that can be reliably detected, by comparing an average signal to the noise level. Herein, the SNR was calculated in two ways using the samples of agar and gelatine, which featured the highest noise among all measured substances in this study. The mean spectra of these two substances from each measurement are given in Figures S2,S3, along with the standard deviation, as shadows. In the first case, we calculated the SNR for each spectrum individually as the ratio between the mean Raman intensity and the standard deviation of the estimated noise (see eq (S3.1)). The noise (I_n) was estimated as the difference between the spectrum and the output of a strong Savitzky–Golay (S-G) smoothing ($p = 2$, $n = 31$) of the spectrum.³⁰ Herein, the parameters p and n represent the degree of the polynomial function and the window size of the S-G filter, respectively. An example of spectrum before and after smoothing together with the estimated noise is given in Figure S8. As the noise could be upper-biased for the spectral region with sharp Raman bands, the SNR tends to be under-estimated in this case. In the second case, we calculated the SNR as the ratio between the mean and standard variation of the Raman intensities at a given wavenumber over 10 randomly selected Raman spectra (see eq (S3.2)) in which the peak intensities of

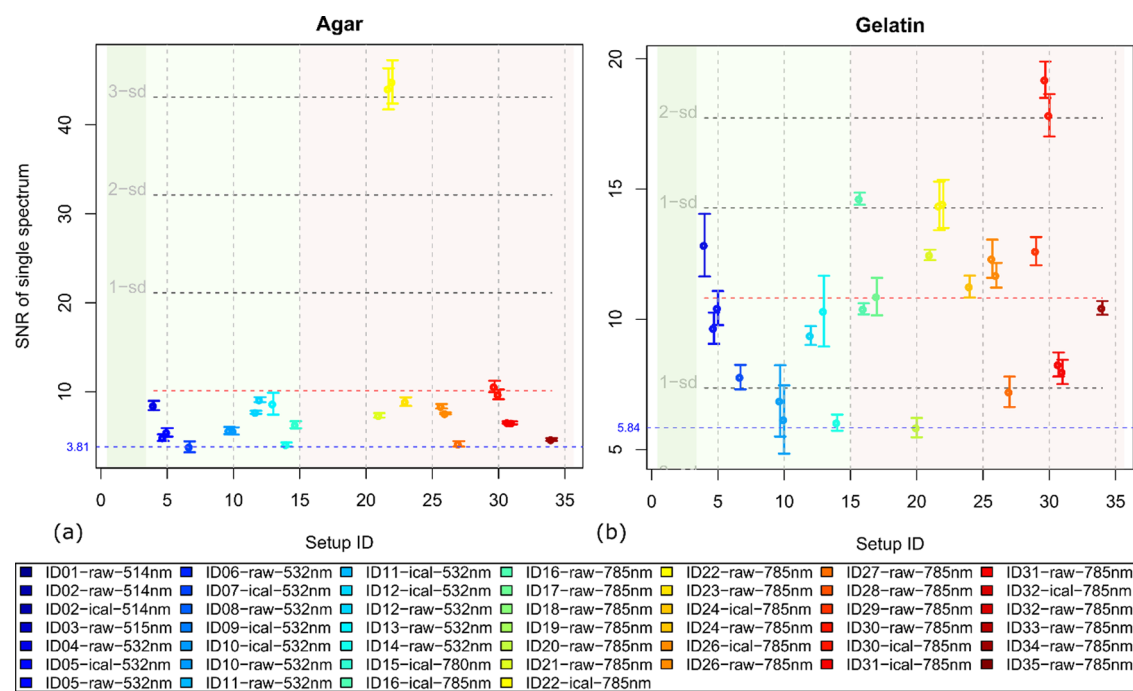


Figure 1. SNR of (a) agar and (b) gelatine: the color shades represent the setups with different source wavelengths. Each bar represents the measurement of a substance on a given setup. The bar contains the mean and standard deviation value of the SNR for each setup. The average and standard deviation of all setups are given using the red and gray dashed lines, respectively. The blue dashed line marks the minimal SNR of all setups. The identification of the setup, the type of data, and the source wavelength are given in the legend. In particular, the data types ‘ical’ and ‘raw’ denote the cases where the data were exported from the setup with (ical) and without (raw) intensity calibration, respectively.

the Raman bands around 846 and 1451 cm^{-1} were used for agar and gelatine, respectively. These two Raman bands were selected to be sufficiently intense and well separated from other bands.

Peak Shift. The peak shift characterizes the accuracy of the wavenumber axis (x axis, spectral axis) of a setup. It was defined as the deviation of a measured peak position from its theoretical position. To do so, we employed three substances having well-defined Raman bands: paracetamol, polystyrene, and cyclohexane. The mean spectra and standard deviations of the three substances for each setup are shown in Figures S4–S6 (a). The Raman bands used in our calculation are highlighted in Figures S4–S6 (b). Only the bands within the fingerprint region were considered, as the CH stretching region was not measured on all setups. To start, we fitted each of these Raman bands by a Gaussian peak according to eq 1 because Gaussian fits have been proven to work well for solid, powder, gel, or resin samples.³¹ The spectral region used for the fit was defined by $\tilde{\nu}_0 \pm 10\text{ cm}^{-1}$, where $\tilde{\nu}_0$ denotes the theoretical peak position. The parameter μ of the resulting Gaussian peak was considered to be the measured peak position of the Raman band being fitted, and the peak shift of this Raman band was determined by $\Delta\tilde{\nu} = \mu - \tilde{\nu}_0$. The final result of one single spectrum was determined as the mean absolute value of all peak shifts from the multiple Raman bands within this spectrum, which indicates the mean absolute deviation in the wavenumber axis of a measurement.

$$I(\tilde{\nu}) = A \cdot \exp\left(-\frac{(\tilde{\nu} - \mu)^2}{2\sigma^2}\right) \quad (1)$$

Peak Width. The peak width is considered as a straightforward metric to characterize the spectral resolution of a given setup and can be benchmarked by the full-width-at-half-maximum (FWHM). Herein, we employed the results of σ

from the peak fit described in the previous section and obtained the FWHM according to eq 2. The calculation was performed on both the NeAr emission and the Raman spectra of the three substances, i.e., paracetamol, polystyrene, and cyclohexane. The measured NeAr emission on different setups is visualized in Figure S7. To make the determination more precise, the FWHM was only calculated for the best-defined peak in each case; i.e., no significant shoulder peaks exist, and the peak does not overlap with its neighboring peaks. According to this criterion, we employed the emission at 626.56 and 878.2 nm for the NeAr measurement with laser sources of $514/532$ and 785 nm , respectively; the peaks at 1169 , 1602 , and 1028 cm^{-1} were used for paracetamol, polystyrene, and cyclohexane, respectively.

$$\text{FWHM} = 2\sigma \cdot \sqrt{2 \ln(2)} \quad (2)$$

Peak Ratio. The peak ratio is defined as the ratio between the areas of two given Raman bands. It quantifies the consistency of the relative Raman intensity across setups, and it is an important criterion for a reproducible Raman spectroscopic data analysis, either quantitative or qualitative. In particular, the area of a given Raman band was calculated via a simple integration of the peak region according to eq 3. We did not additionally subtract the baseline offset during this integration, as the spectra were already baseline corrected. The terms k_i and k_r denote the start and the end index of the band. These values were determined automatically as the positions around a band where the Raman intensity starts to increase (i.e., $I(\tilde{\nu}_{k_i-1}) > I(\tilde{\nu}_{k_i})$) or stops decreasing (i.e., $I(\tilde{\nu}_{k_i}) < I(\tilde{\nu}_{k_i+1})$), respectively. We based our calculation on the spectra of cyclohexane and obtained three peak ratios for each spectrum: the area of the bands at 1028 , 1266 , and 1444 cm^{-1} with respect to the area of the band at 801 cm^{-1} . The four bands are well-defined and do not show

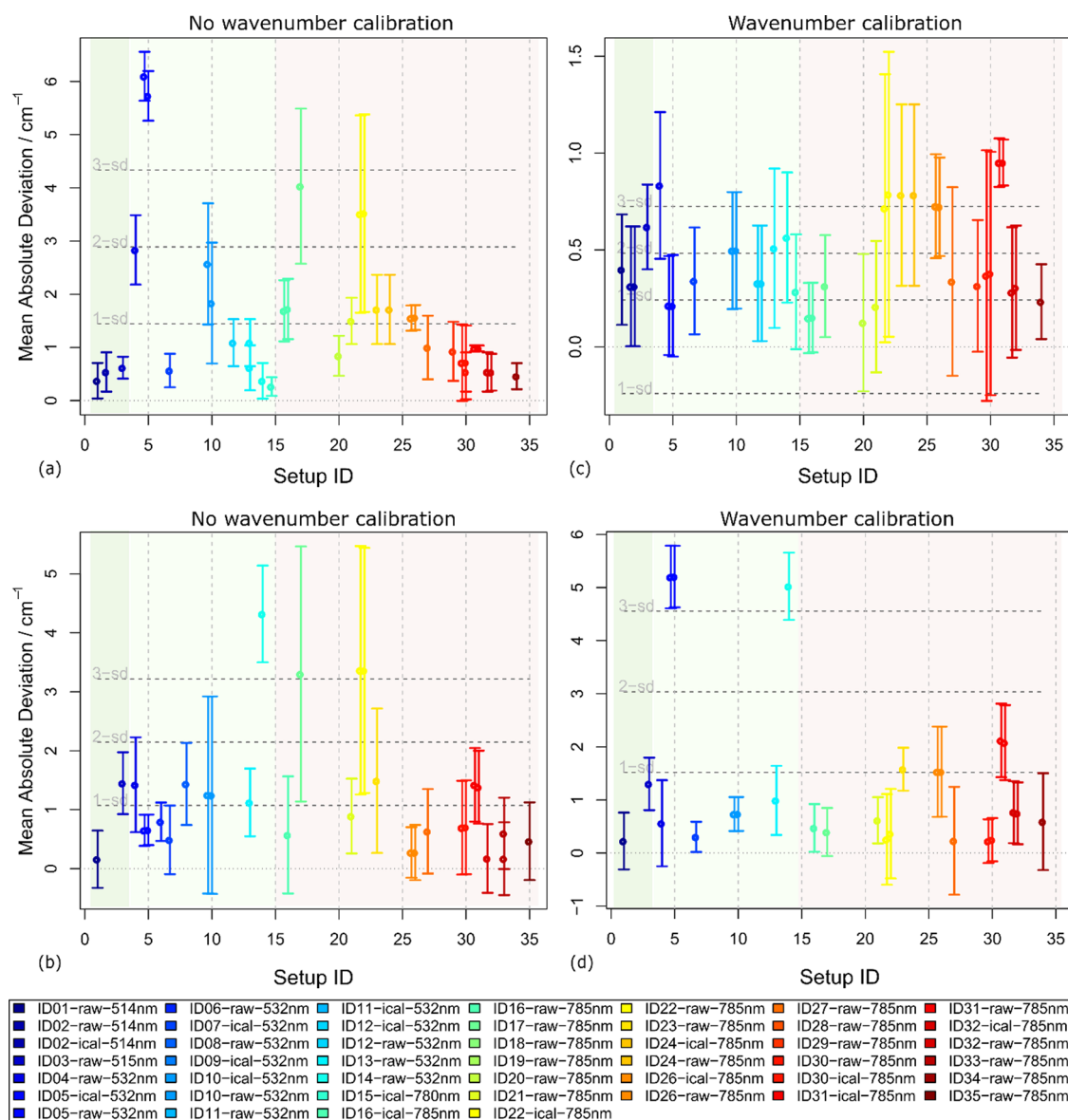


Figure 2. Results of peak shifts for (a, c) paracetamol and (b, d) polystyrene, without and with wavenumber calibration. The wavenumber calibration for polystyrene was only calculated if there is a paracetamol spectrum measured on the same setup. Therefore, a lower number of setups are shown for the calibrated results for polystyrene than without wavenumber calibration.

significant shoulders or overlap with neighboring bands, which helps to precisely estimate the peak area.

$$A = \sum_{i=k_1}^{k_2} I(\tilde{\nu}_i) \quad (3)$$

RESULTS AND DISCUSSION

We present here the results of assessment of the above-mentioned metrics calculated for different setups and substances. As our major focus in this study is the cross-setup comparability, we will mainly compare the results of different setups with each other but not specifically consider their reference values given in the literature. For such reference information, the interested readers are kindly referred to refs.^{32–34}

The SNR results using a single-spectrum and 10 replicate spectra are given in Figure 1 and Figure S9, respectively. The

color shades represent setups with different source wavelengths. The meta-information of the datasets is provided in the legend, including the identification number (ID) of the setup, the type of the data, and the excitation wavelength for each measurement. The two data types, represented as ‘raw’ and ‘ical’, differed in whether an intensity calibration was done by the setup itself or not, respectively. In particular, the SNR of each measurement was plotted as a single bar containing the information of the mean and standard deviation over the multiple spectra from the same measurement (the number of replicates for each experimental setup is provided in Table 1). The red dashed line marks the global mean of the SNR over all measurements, while the gray dashed lines show 1 to 3 times the global standard deviation of SNR over all measurements. The minimal mean SNRs for agar and gelatine, given using the blue dashed line, are 3.8 and 5.8, respectively. That is to say, the SNR is sufficient for all setups, which is reasonable considering a preliminary SNR filtering during the data upload (see experimental design in the

Supporting Information). This can be concluded from the SNR calculated from multiple spectra as well, as is shown in Figure S9.

The results of the peak shift calculations are plotted in Figure 2 and Figure S10, following the same structure as in Figure 1. It is implicit in the definition that the smaller the absolute value of the peak shift, the better the setup. An ideal measurement would give a zero peak shift; i.e., peaks from the measurement match perfectly with the theoretical values. As the ideal case is not easily achievable in reality, wavenumber calibration is important to minimize the peak shifts. To do so, we applied paracetamol as the standard material and calculated a calibration function based on the deviation between the measured and the theoretical positions for well-defined bands. This wavenumber calibration function was thereafter applied to the spectra of paracetamol, polystyrene, and cyclohexane. As shown in Figure 2 and Figure S10, the peak shifts were reduced closer to zero after the calibration, compared to before calibration. To make the conclusion clearer, we summarized the statistics of the peak shifts in Table 2. Therein, $\Delta\bar{\nu}$ denotes the mean peak shift of

Table 2. Statistics of Peak Shifts, Summarized from the Results of Figure 2

substance		mean($\Delta\bar{\nu}$)	<i>sd</i>	max(<i>sd</i> ₀)
paracetamol	no calibration	1.58 cm ⁻¹	1.44 cm ⁻¹	1.86 cm ⁻¹
	calibrated	0.45 cm ⁻¹	0.24 cm ⁻¹	0.73 cm ⁻¹
polystyrene	no calibration	1.20 cm ⁻¹	1.07 cm ⁻¹	2.16 cm ⁻¹
	calibrated	1.31 cm ⁻¹	1.52 cm ⁻¹	1.02 cm ⁻¹
cyclohexane	no calibration	1.33 cm ⁻¹	1.79 cm ⁻¹	1.37 cm ⁻¹
	calibrated	1.93 cm ⁻¹	2.16 cm ⁻¹	0.73 cm ⁻¹

each setup. The factors *sd* and *sd*₀ represent the standard deviation over all measurements and standard deviation for a single measurement, respectively. In particular, the decrease of $\Delta\bar{\nu}$ demonstrates an improved setup-independence after the wavenumber calibration, although we note that there are also

setups that are (extremely) negatively affected (ID05; ID14) or not improved (ID31). A generally better reproducibility within the same measurement was observed after wavenumber calibration, evidenced by a reduced maximal *sd*₀. The increase in the mean and standard deviation of $\Delta\bar{\nu}$ after calibration for polystyrene and cyclohexane will be explained in the next paragraph.

As it is apparent in the figures, the improvement due to the wavenumber calibration was not the same for the three substances: it was the most significant for paracetamol but much less so for polystyrene and cyclohexane. As is shown in Table 2, the mean of $\Delta\bar{\nu}$ was decreased for paracetamol but not for the other two substances. This is easy to explain, as the calibration function of a setup was 'learnt' from the spectra of paracetamol and 'transferred' to the other substances. This function is supposed to capture best the peak shift of the Raman spectra of paracetamol and less so for the other substances. The performance of the wavenumber calibration can also be degraded due to other issues. For example, the peak shifts can increase after wavenumber calibration if the spectra of paracetamol measured on the specific instrument do not truly reflect the peak shift of spectra to be calibrated. This was seen for the setup 'ID05', for which the spectra of paracetamol exhibited significantly larger peak shifts than those of polystyrene and cyclohexane. The wavenumber calibration in this scenario introduced additional errors in the peak positions and led to increased peak shifts (Figure 2 and Figure S10). Another possibility that makes the wavenumber calibration invalid was seen from setup 'ID14', for which the peak shifts in paracetamol were much smaller than the shifts in the spectra of polystyrene and cyclohexane. Therefore, the calibration function could not represent the peak shifts well, and thus, the wavenumber calibration hardly reduced the peak shifts (Figure 2 and Figure S10). These two scenarios may happen if the standard material (e.g., paracetamol) and the samples to be calibrated (e.g.,

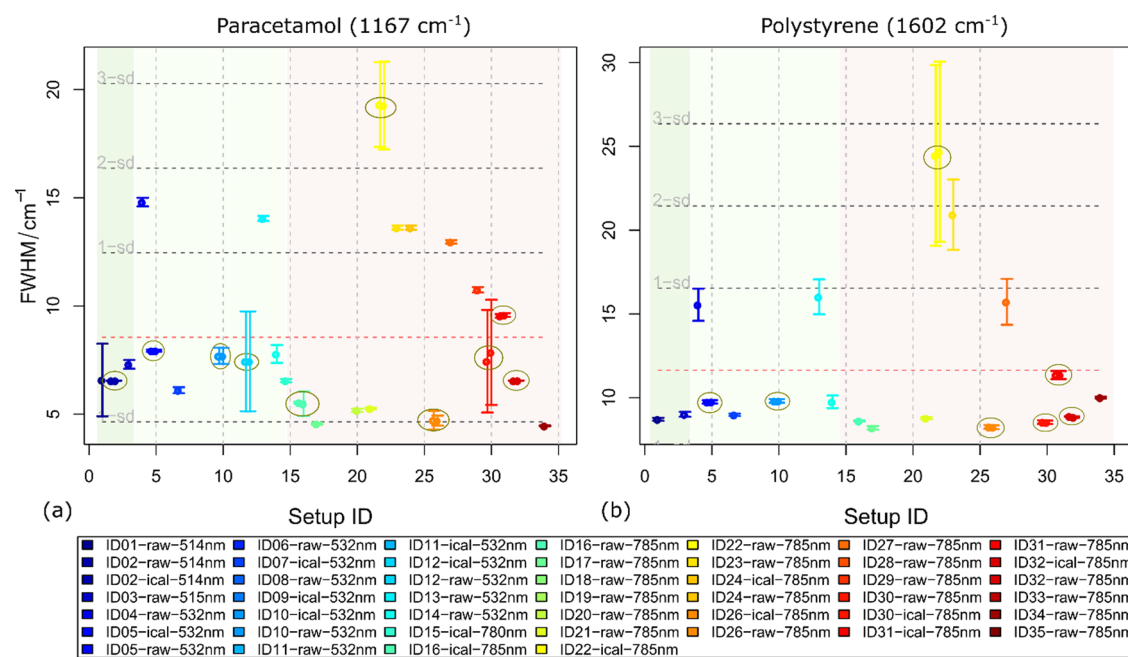


Figure 3. Results of FWHM calculated from (a) paracetamol and (b) polystyrene based on selected peaks. The setups from different source wavelengths were marked by the colored shades. Pairs of spectra with and without intensity calibration were highlighted by circles. The variations among different setups are clearly seen from the results.

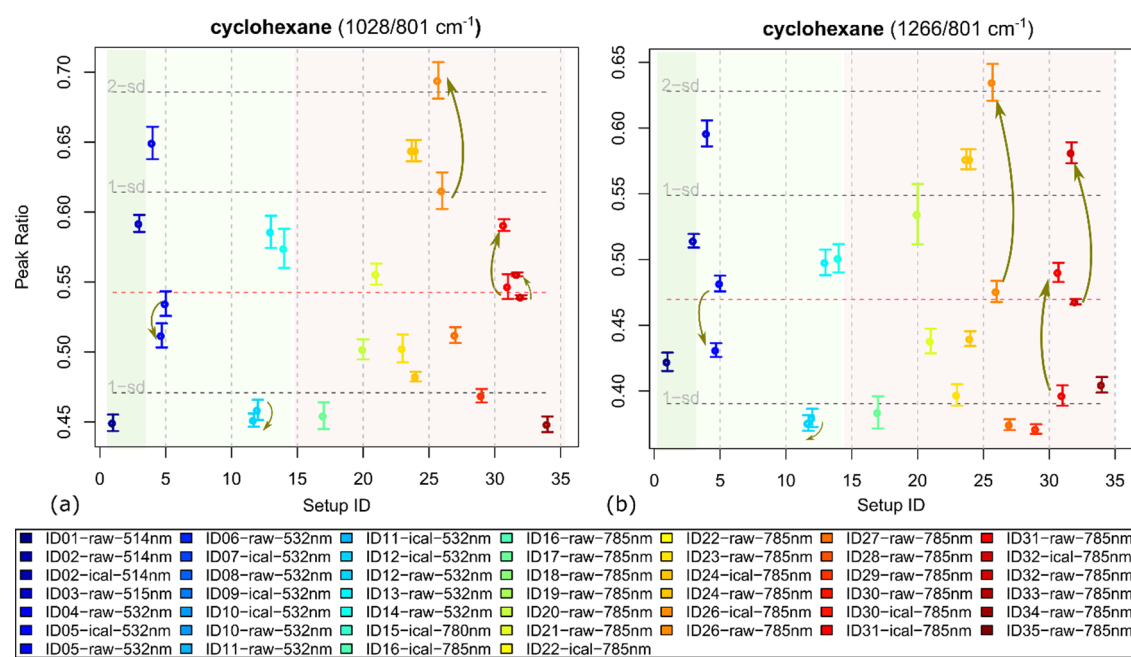


Figure 4. Ratios of peak intensities at (a) 1028 and (b) 1266 cm⁻¹ to the peak intensity at 801 cm⁻¹. The color shades represent the setups with different source wavelengths. The curved arrows in brown show the paired results without and with intensity calibration, pointing to the results with intensity calibration. It is clear that the peak ratios varied largely among measurements, and the setup intern intensity calibration does not improve the situation.

Table 3. Statistics of the Peak Ratios Summarized from Figure 4 and Figure S14 for the Setups with Different Source Wavelengths (514, 532, and 785 Nm)^a

raman bands (cm ⁻¹)	mean(\bar{r})			sd/mean(\bar{r})			max(sd ₀)/mean(\bar{r})		
	514 nm	532 nm	785 nm	514 nm	532 nm	785 nm	514 nm	532 nm	785 nm
1028/801	0.52	0.54	0.55	0.19	0.13	0.13	0.01	0.03	0.02
1266/801	0.47	0.47	0.47	0.14	0.17	0.18	0.02	0.02	0.05
1444/801	0.53	0.54	0.52	0.11	0.24	0.29	0.03	0.03	0.13

^aThe terms mean(\bar{r}), *sd*, and max(*sd*₀) represent the global average, global standard deviation, and the maximal standard deviation of the peak ratios, respectively. The variations in the relative intensities tend to increase between two peaks with longer distance along the wavenumber axis.

polystyrene and cyclohexane in this study) are measured under different conditions. It could also be the case that the setup bears short-term instability, leading to variations among spectra of standard material and samples. This could be a significant concern in applying Raman spectroscopy. However, we will not go into details of this issue as we did not have sufficient data from this respect.

An additional interesting observation is that *sd*, which benchmarks the comparability across setups, was increased by wavenumber calibration for polystyrene and cyclohexane. There are two likely reasons, to our assessment. First, the two cases with improper calibration as mentioned previously contributed to the increase in *sd*. Second, the variations in the spectra of paracetamol may be passed to the calibration procedure and add to the variation in the calibrated spectra. With all these observations, it is obvious and fair to say that the peak shifts remain a significant issue for the cross-setup reproducibility in Raman spectroscopy. Wavenumber calibration, as is already shown in a previous report,²¹ cannot completely remove the setup-induced shifts in the wavenumber axis of Raman spectroscopy.

The results of the FWHM are given in Figure 3(a,b) for paracetamol and polystyrene, respectively. Similar results are shown in Figure S11 (a,b) for cyclohexane and the NeAr lamp,

respectively. The cross-setup variations are clearly seen from all results. In addition, the setup 'ID22' was detected as an 'outlier', having significantly larger FWHM (i.e., lower spectral resolution), which turned out to differ from others as it was a fibre-optics based setup with low nominal spectral resolution (see Table S1). In addition, the intensity calibration, which was undertaken in some setups, did not significantly influence the peak widths. This can be seen from the similar results for all the encircled pairs in the plots, which correspond to the results from spectra of the same setup, with and without the intensity calibration.

To check the influence of nominal spectral resolution (i.e., pixel size in wavenumber) on the estimation of the peak shift and peak width, we visualized additionally the results of the peak shift and the FWHM from all setups with respect to their nominal spectral resolutions. As shown in Figures S12,13, no correlation can be concluded between the spectral resolution and the peak shift or peak width. This justifies our estimation of these two spectral characteristics, which were not systematically biased by the differences in the nominal spectral resolution (i.e., pixel size).

The results of the peak ratios are given in Figure 4 and Figure S14. The curved brown arrows indicate data pairs from the same setup, with (*ical*) and without (*raw*) intensity calibration. In

general, the peak ratios varied largely across setups. Additionally, there was no clear trend as to how the peak ratios change with the source wavelengths. Rather, however, the intensity calibration did not necessarily bring the peak ratios closer to the global mean; i.e., this correction did not improve the cross-setup comparability. In contrast, the influence of the intensity calibration seemed to be rather arbitrary. These facts very likely indicate the inaccuracy of the estimated intensity response function.

For clearer interpretation, we additionally summarized the statistics of the peak ratios in Table 3, corresponding to the results of the three source wavelengths: 514, 532, and 785 nm. Therein, \bar{r} and sd_0 represent the mean and standard deviation of each setup, while $\text{mean}(\bar{r})$ and sd represent the global mean and standard deviation. Taking $\text{mean}(\bar{r})$ as an approximation of the 'true' peak ratio, the cross- and within-setup variations can be benchmarked by $\max(sd_0)/\text{mean}(\bar{r})$ and $sd/\text{mean}(\bar{r})$, respectively. On this basis, a comparison among the three ratios showed that the cross- and within-setup variations increased when the two peaks involved are further apart in the spectra. That is to say, the variations in the relative peak intensities tend to increase between two points with longer distance along the wavenumber axis. This is understandable, as the spectrometer response function has a stronger influence if the two peaks are further apart.

With all the results shown above, we would like to stress the profound influence of the setup-dependence in Raman spectroscopy. The routinely applied spectrometer calibration clearly does not remove all setup-related effects on Raman spectra when the same chemical substance is measured on different Raman platforms. Inadequate calibration can be for multiple reasons: (i) it is difficult to duplicate the measurement conditions exactly between the standard material and real samples; (ii) Raman spectra can be angle and relative orientation dependent if the standard is crystalline, and the setup (laser/grating) is polarization sensitive; (iii) there may be influence from contaminations if the standard material is not refreshed regularly – for this particular experiment, the latter reason is clearly not the case, as one laboratory supplied all materials for analysis to all participating laboratories; (iv) the full technical details of built-in (automatic) calibrations are normally inaccessible, and thus, it is hard to ensure that exactly the same calibration methods are used by the different manufacturers or setups. The application of different calibration algorithms certainly constitutes an additional source of setup-dependence.

Based on all facts and reasons stated above, computational strategies to remove the setup-induced spectral variations are urgently needed. As a first attempt to solve this issue, we have developed recently model transfer methods to deal with cross-setup variations in Raman spectroscopy.^{26,35,36} However, a long-term effort is needed to completely solve the issue of setup-dependence, and this can only be achieved if cross-laboratory comparisons are undertaken, such as those reported in the present study and indeed elsewhere for surface-enhanced Raman scattering.³⁷

CONCLUSIONS

This contribution reports a large-scale and cross-laboratory round robin Raman experiment designed by researchers from approximately 50 institutes in Europe. The data were submitted by 15 different institutes and measured with 35 Raman setups, which were from five different manufacturers and configured with multiple laser sources. We have presented the results with

respect to four key metrics to assess Raman performance across setups. These include the SNR, the peak shift, the peak width, and the peak ratio for spectra of a set of five common substances measured on different setups. In this way, we could quantify and analyze the cross-setup comparability and variability in Raman spectroscopy, and we could verify the appropriateness to implement spectrometer calibration. More inspiring studies and efforts are certainly needed to standardize Raman spectroscopy and hence push Raman techniques closer to real-world applications. A detailed investigation on each of the different sources that harms the spectral reproducibility can be a good start. In addition, we recommend the following actions, which we believe are worthy for future investigations.

- First, we suggest that the manufacturers make spectrometer calibration a by-default included module and make the full technical details of the correction explicit and open access. This built-in process is, on the one hand, more likely to achieve 'duplicated' conditions for the real samples and the standard material and, on the other hand, makes calibration easier. In addition, by applying the same calibration procedure, the variations introduced by in-house written calibration programs in different labs are reduced. Moreover, further investigations of the temporal reliability of instruments and calibration are needed. Any short-term setup instabilities should be understood and well controlled for the development of clinical applications.
- Second, we recommend that the manufacturers provide access to the real raw data, before any processing steps are applied, so that it is more feasible to unravel data that are closer to the 'physical truth'. We would like to encourage both researchers and manufacturers to work together on standard operating procedures for verifying instrument calibration and performance.
- Third, we encourage researchers and scientists to make their data openly available and actively contribute to establishing larger databases. This has been done in other data-rich communities (e.g., ref³⁸) and would be a vast invaluable resource to build statistical models that are tolerant to unwanted spectral differences, like the spectral variations between seemingly identical analytical setups. In this way, reproducible predictions for different measurements from the same sample can be achieved. A larger database can also provide a better reference used for the spectral alignment approaches.
- Finally, we advocate a broader cooperation on the same scientific question in order to come up with 'global' solutions and reduce the variations in the setups used in different research groups answering the same question.

ASSOCIATED CONTENT

Supporting Information

The Supporting Information is available free of charge at <https://pubs.acs.org/doi/10.1021/acs.analchem.0c02696>.

(Table S1) Acceptable signal-to-noise ratio for specific band integral intensities across the at least 10 submitted spectra; (Table S2) information of Raman spectroscopic devices within the trial; (Figure S1) error in the measurement of the setup ID25; (Figure S2) mean spectra and standard deviation spectra of agar; (Figure S3) mean spectra and standard deviation spectra of gelatine; (Figure S4) mean and standard deviation spectra

and the representative Raman bands of paracetamol; (Figure S5) mean and standard deviation spectra and the representative Raman bands of polystyrene; (Figure S6) mean and standard deviation spectra and the representative Raman bands of cyclohexane; (Figure S7) mean spectra and standard deviation spectra of the NeAr lamp; (Figure S8) example of estimated SNR; (Figure S9) results of SNR for agar and gelatine; (Figure S10) results of peak shifts for cyclohexane without and with wave-number calibration; (Figure S11) results of FWHM calculated from cyclohexane and NeAr; (Figure S12) mean absolute deviation with respect to the nominal spectral resolution of different setups; (Figure S13) FWHM with respect to the nominal spectral resolution; and (Figure S14) results of the peak ratio between the cyclohexane peaks at 1444 and 801 cm^{-1} (PDF)

AUTHOR INFORMATION

Corresponding Author

Thomas Bocklitz – Institute of Physical Chemistry and Abbe Center of Photonics, University Jena, 07743 Jena, Germany; Member of Leibniz Health Technologies, Leibniz Institute of Photonic Technology Jena, 07745 Jena, Germany; orcid.org/0000-0003-2778-6624; Email: thomas.bocklitz@uni-jena.de

Authors

Shuxia Guo – Institute of Physical Chemistry and Abbe Center of Photonics, University Jena, 07743 Jena, Germany; Member of Leibniz Health Technologies, Leibniz Institute of Photonic Technology Jena, 07745 Jena, Germany; orcid.org/0000-0001-8237-8936

Claudia Beleites – Member of Leibniz Health Technologies, Leibniz Institute of Photonic Technology Jena, 07745 Jena, Germany; Chemometrix GmbH, 61200 Wölfersheim, Germany

Ute Neugebauer – Institute of Physical Chemistry and Abbe Center of Photonics, University Jena, 07743 Jena, Germany; Member of Leibniz Health Technologies, Leibniz Institute of Photonic Technology Jena, 07745 Jena, Germany; Center for Sepsis Control and Care, Jena University Hospital, D-07747 Jena, Germany

Sara Abalde-Cela – International Iberian Nanotechnology Laboratory (INL), 4715-310 Braga, Portugal; orcid.org/0000-0001-5615-594X

Nils Kristian Afseth – Nofima - Norwegian Institute of Food, Fisheries and Aquaculture Research, NO-9291 Tromsø, Norway

Fatima Alsamad – Université de Reims Champagne-Ardenne, Reims 51097, France

Suresh Anand – National Institute of Optics, National Research Council, 50019 Sesto Fiorentino, Italy

Cuauhtemoc Araujo-Andrade – ICFO-Institut de Ciències Fotoniques, The Barcelona Institute of Science and Technology, 08860 Barcelona, Spain

Sonja Aškračić – Institute of Physics Belgrade, University of Belgrade, Beograd, Serbia

Ertug Avci – Genetics and Bioengineering Department, Faculty of Engineering, Yeditepe University, Kayisdagi 34755, Ataşehir/Istanbul, Turkey

Monica Baia – Faculty of Physics, Babes-Bolyai University, Cluj-Napoca 400084, Romania

Malgorzata Baranska – Faculty of Chemistry, Jagiellonian University, 30-387 Krakow, Poland; Jagiellonian Centre for Experimental Therapeutics (JCET), 30-348 Kraków, Poland; orcid.org/0000-0001-8826-3144

Enrico Baria – Department of Physics, University of Florence, 50121 Firenze, Florence, Italy; European Laboratory for Non-linear Spectroscopy, 50019 Sesto Fiorentino, Florence, Italy

Luis A. E. Batista de Carvalho – Molecular Physical Chemistry R&D Unit, Department of Chemistry, University of Coimbra, 3004-535 Coimbra, Portugal; orcid.org/0000-0002-8059-8537

Philippe de Bettignies – HORIBA France SAS, 59650 Villeneuve-d'Ascq, France

Alois Bonifacio – Raman Lab, Dept. Engineering and Architecture, University of Trieste, 34127 Trieste, Italy; orcid.org/0000-0002-2251-7786

Franck Bonnier – Faculty of pharmacy, EA6295 NanoMédicaments et Nanosondes, University of Tours, 37000 Tours, France

Eva Maria Brauchle – NMI Natural and Medical Sciences Institute at the University of Tübingen, 72770 Reutlingen, Germany; Department of Women's Health, Research Institute of Women's Health and Cluster of Excellence iFIT (EXC 2180) "Image-Guided and Functionally Instructed Tumor Therapies", Eberhard Karls University Tübingen, 72074 Tübingen, Germany

Hugh J. Byrne – FOCAS Research Institute, Technological University Dublin, Dublin, Ireland

Igor Chourpa – Faculty of pharmacy, EA6295 NanoMédicaments et Nanosondes, University of Tours, 37000 Tours, France

Riccardo Cicchi – National Institute of Optics, National Research Council, 50019 Sesto Fiorentino, Italy; European Laboratory for Non-linear Spectroscopy, 50019 Sesto Fiorentino, Florence, Italy

Frederic Cuisinier – LBN, University Montpellier, 34000 Montpellier, France

Mustafa Culha – Genetics and Bioengineering Department, Faculty of Engineering, Yeditepe University, Kayisdagi 34755, Ataşehir/Istanbul, Turkey

Marcel Dahms – Institute of Physical Chemistry and Abbe Center of Photonics, University Jena, 07743 Jena, Germany; Member of Leibniz Health Technologies, Leibniz Institute of Photonic Technology Jena, 07745 Jena, Germany; Center for Sepsis Control and Care, Jena University Hospital, D-07747 Jena, Germany

Catalina David – HORIBA France SAS, 59650 Villeneuve-d'Ascq, France

Ludovic Duponchel – LASIRE – Laboratoire de Spectroscopie pour les Interactions, la Réactivité et l'Environnement, Univ. Lille, CNRS, F-59000 Lille, France; orcid.org/0000-0002-7206-4498

Shiyamala Duraipandian – FOCAS Research Institute and School of Physics & Clinical & Optometric Sciences, Technological University Dublin, Dublin, Ireland

Samir F. El-Mashtoly – Center for Protein Diagnostics (ProDi) and Department of Biophysics, Faculty of Biology and Biotechnology, Ruhr University Bochum, 44801 Bochum, Germany; orcid.org/0000-0001-6087-8817

David I. Ellis – Manchester Institute of Biotechnology, School of Chemistry, University of Manchester, M1 7DN Manchester, United Kingdom; orcid.org/0000-0002-7633-7019

- Gauthier Eppe** – Mass Spectrometry Laboratory, MolSys Research Unit, University of Liege, 4000 Liège, Belgium
- Guillaume Falgayrac** – MABLab, Marrow Adiposity and Bone Lab, Univ. Littoral Côte d'Opale, F-62300 Boulogne-sur-Mer, France; CHU Lille, F-59000 Lille, France; orcid.org/0000-0003-1806-1509
- Ozren Gamulin** – Department of Physics and Biophysics, School of Medicine, University of Zagreb, 10000 Zagreb, Croatia; Centre for Advanced Materials Science, 10000 Zagreb, Croatia
- Benjamin Gardner** – Physics and Astronomy, Mathematics and Physical Sciences, College of Engineering, Exeter EX4 4Q, United Kingdom; orcid.org/0000-0002-7223-9585
- Peter Gardner** – Manchester Institute of Biotechnology, School of Chemistry, University of Manchester, M1 7DN Manchester, United Kingdom; Department of Chemical Engineering and Analytical Science, School of Engineering, The University of Manchester, Manchester M1 3AL, United Kingdom; orcid.org/0000-0003-4034-3764
- Klaus Gerwert** – Center for Protein Diagnostics (ProDi) and Department of Biophysics, Faculty of Biology and Biotechnology, Ruhr University Bochum, 44801 Bochum, Germany
- Evangelos J. Giamarellos-Bourboulis** – 4th Department of Internal Medicine, National and Kapodistrian University of Athens, Medical School, Athens, Greece
- Sveinbjorn Gizurarson** – Faculty of Pharmaceutical Sciences, University of Iceland, Reykjavik, Iceland
- Marcin Gnyba** – Faculty of Electronics, Telecommunications and Informatics, Gdańsk University of Technology, 80-233 Gdańsk, Poland
- Royston Goodacre** – Department of Biochemistry and Systems Biology, Institute of Systems, Molecular and Integrative Biology, University of Liverpool, Liverpool L69 750 7ZB, United Kingdom
- Patrick Grysan** – Materials Research and Technology, Luxembourg Institute of Science and Technology, L-4422 Belvaux, Luxembourg, Belgium
- Orlando Guntinas-Lichius** – Department of Otorhinolaryngology, Jena University Hospital, 07743 Jena, Germany; orcid.org/0000-0001-9671-0784
- Helga Helgadóttir** – Faculty of Pharmaceutical Sciences, University of Iceland, Reykjavik, Iceland
- Vlasta Mohacek Grosev** – Centre for Advanced Materials Science, 10000 Zagreb, Croatia; Ruđer Bošković Institute, 10000 Zagreb, Croatia
- Catherine Kendall** – Biophotonics Research Unit, Gloucestershire Hospitals NHS Foundation Trust, Gloucester GL1 3NN, United Kingdom
- Roman Kiselev** – Member of Leibniz Health Technologies, Leibniz Institute of Photonic Technology Jena, 07745 Jena, Germany; St. Jude Children's Research Hospital, Memphis, Tennessee 38105, United States
- Micha Kölbach** – Renishaw GmbH, 72124 Pliezhausen, Germany
- Christoph Krafft** – Member of Leibniz Health Technologies, Leibniz Institute of Photonic Technology Jena, 07745 Jena, Germany
- Sivashankar Krishnamoorthy** – Materials Research and Technology, Luxembourg Institute of Science and Technology, L-4422 Belvaux, Luxembourg, Belgium; orcid.org/0000-0003-0410-1696
- Patrick Kubryck** – Renishaw GmbH, 72124 Pliezhausen, Germany
- Bernhard Lendl** – Institute of Chemical Technologies and Analytics, TU Wien, 1040 Wien, Austria; orcid.org/0000-0003-3838-5842
- Pablo Loza-Alvarez** – ICFO-Institut de Ciències Fotoniques, The Barcelona Institute of Science and Technology, 08860 Barcelona, Spain
- Fiona M. Lyng** – FOCAS Research Institute and School of Physics & Clinical & Optometric Sciences, Technological University Dublin, Dublin, Ireland
- Susanne Machill** – Chair of Bioanalytical Chemistry, TU Dresden, 01062 Dresden, Germany
- Cedric Malherbe** – Mass Spectrometry Laboratory, MolSys Research Unit, University of Liege, 4000 Liège, Belgium
- Monica Marro** – ICFO-Institut de Ciències Fotoniques, The Barcelona Institute of Science and Technology, 08860 Barcelona, Spain; orcid.org/0000-0003-4031-4605
- Maria Paula M. Marques** – Molecular Physical Chemistry R&D Unit, Department of Chemistry and Department of Life Sciences, University of Coimbra, 3004-535 Coimbra, Portugal; orcid.org/0000-0002-8391-0055
- Ewelina Matuszyk** – Jagiellonian Centre for Experimental Therapeutics (JCET), 30-348 Kraków, Poland
- Carlo Francesco Morasso** – Istituti Clinici Scientifici Maugeri IRCCS, 27100 Pavia, Italy
- Myriam Moreau** – LASIRE – Laboratoire de Spectroscopie pour les Interactions, la Réactivité et l'Environnement, Univ. Lille, CNRS, F-59000 Lille, France
- Howbeer Muhamadali** – Department of Biochemistry and Systems Biology, Institute of Systems, Molecular and Integrative Biology, University of Liverpool, Liverpool L69 750 7ZB, United Kingdom
- Valentina Mussi** – National Research Council, Institute for Microelectronics and Microsystems (IMM-CNR), 00133 Rome, Italy; orcid.org/0000-0003-1465-1739
- Ioan Notingher** – School of Physics and Astronomy, University of Nottingham, Nottingham NG7 2RD, United Kingdom; orcid.org/0000-0002-5360-230X
- Marta Z. Pacia** – Jagiellonian Centre for Experimental Therapeutics (JCET), 30-348 Kraków, Poland; orcid.org/0000-0002-0391-2502
- Francesco S. Pavone** – Department of Physics, University of Florence, 50121 Firenze, Florence, Italy; European Laboratory for Non-linear Spectroscopy, 50019 Sesto Fiorentino, Florence, Italy
- Guillaume Penel** – MABLab, Marrow Adiposity and Bone Lab, Univ. Littoral Côte d'Opale, F-62300 Boulogne-sur-Mer, France; CHU Lille, F-59000 Lille, France
- Dennis Petersen** – Department of Biophysics, Faculty of Biology and Biotechnology, Ruhr University Bochum, 44801 Bochum, Germany
- Olivier Piot** – Université de Reims Champagne-Ardenne, Reims 51097, France
- Julietta V. Rau** – Istituto di Struttura della Materia, Consiglio Nazionale delle Ricerche (ISM-CNR), 100-00133 Rome, Italy; Sechenov First Moscow State Medical University, 119991 Moscow, Russian Federation; orcid.org/0000-0002-7953-1853
- Marc Richter** – Renishaw GmbH, 72124 Pliezhausen, Germany
- Maria Krystyna Rybarczyk** – Chemical Faculty, Gdansk University of Technology, 80-233 Gdansk, Poland
- Hamideh Salehi** – LBN, University Montpellier, 34000 Montpellier, France

Katja Schenke-Layland – NMI Natural and Medical Sciences Institute at the University of Tübingen, 72770 Reutlingen, Germany; Department of Women's Health, Research Institute of Women's Health and Cluster of Excellence iFIT (EXC 2180) "Image-Guided and Functionally Instructed Tumor Therapies", Eberhard Karls University Tübingen, 72074 Tübingen, Germany; orcid.org/0000-0001-8066-5157

Sebastian Schlücker – Faculty of Chemistry, University of Duisburg-Essen, 45141 Essen, Germany; orcid.org/0000-0003-4790-4616

Markus Schosserer – Department of Biotechnology, Institute of Molecular Biotechnology, University of Natural Resources and Life Sciences, 1180, Austria; orcid.org/0000-0003-2025-0739

Karin Schütze – CellTool GmbH, 82327 Tutzing, Germany
Valter Sergio – Raman Lab, Dept. Engineering and Architecture, University of Trieste, 34127 Trieste, Italy; Faculty of Health Sciences, University of Macau, 999078 Macau, SAR China; orcid.org/0000-0003-3864-658X

Faris Sinjab – School of Physics and Astronomy, University of Nottingham, Nottingham NG7 2RD, United Kingdom

Janusz Smulko – Faculty of Electronics, Telecommunications and Informatics, Gdańsk University of Technology, 80-233 Gdańsk, Poland; orcid.org/0000-0003-1459-4199

Ganesh D. Sockalingum – Université de Reims Champagne-Ardenne, Reims S1097, France

Clara Stiebing – Member of Leibniz Health Technologies, Leibniz Institute of Photonic Technology Jena, 07745 Jena, Germany

Nick Stone – Physics and Astronomy, Mathematics and Physical Sciences, College of Engineering, Exeter EX4 4Q, United Kingdom; orcid.org/0000-0001-5603-3731

Valérie Untereiner – Université de Reims Champagne-Ardenne, S1097 Reims, France

Renzo Vanna – Istituti Clinici Scientifici Maugeri IRCCS, 27100 Pavia, Italy

Karin Wieland – Institute of Chemical Technologies and Analytics, TU Wien, 1040 Wien, Austria

Jürgen Popp – Institute of Physical Chemistry and Abbe Center of Photonics, University Jena, 07743 Jena, Germany; Member of Leibniz Health Technologies, Leibniz Institute of Photonic Technology Jena, 07745 Jena, Germany

Complete contact information is available at:

<https://pubs.acs.org/10.1021/acs.analchem.0c02696>

Notes

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