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**PAPER**

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Single particle inductively coupled plasma mass  
spectrometry with nanosecond time resolution



reaction gas.<sup>17</sup> Additionally, Kálomista *et al.* were able to distinguish gold nanorods from spherical gold NP by comparing the shape of the oversampled ion cloud profiles of both particle types,<sup>18</sup> as well as to determine the structure and composition of Au–Ag bimetallic nanoparticles.<sup>19</sup> Because it was suggested that the cytotoxicity of gold nanoparticles and the tendency of titanium dioxide particles to permeate the blood–brain-barrier is shape-dependent,<sup>20–22</sup> this application could be of rising interest for future investigations. Another very recently demonstrated advantage of microsecond dwell times is the significant reduction of the  $^{12}\text{C}^+$  and  $^{13}\text{C}^+$  background signals, which was exploited by Bolea-Fernandez *et al.*, Liu *et al.*, and Gonzalez de Vega *et al.* to detect microplastic particles as small as 1.0  $\mu\text{m}$ , 0.8  $\mu\text{m}$ , and 0.6  $\mu\text{m}$ , respectively, with dwell times of 100  $\mu\text{s}$ .<sup>23–25</sup> Mozhaveva and Engelhard showed that fast data acquisition and a tailored data processing method for spICP-MS enable to detect silver nanoparticles (AgNP) in the presence of up to 7.5  $\mu\text{g L}^{-1}$  ionic silver.<sup>26</sup> Naturally, these promising results led to an increasing usage of  $\mu\text{s}$ -spICP-MS (microsecond spICP-MS) for a wide range of applications, which to date makes up for ca. 45% of all reported applications of spICP-MS since 2014.<sup>3</sup>

Only recently, Duffin *et al.* recorded time stamps of individual detector output pulses of a multicollector ICP-MS with an accuracy of 0.5 ns by means of a time-to-digital-converter and utilized the time between two successive ion arrivals at the detector to distinguish between ions produced from particles and ions stemming from a continuous background. They observed a significant shortening of the time gap between ions during detection of an ion burst from a particle in comparison to longer gaps between background ions. The authors used an algorithm based on boxcar averages in combination with a reasonably set threshold value to determine the beginning and the end of a detected ion cloud in transient plots that present the cumulative count of detected ions vs. the measurement time stamp. This new approach allowed for the detection of 5 nm AuNPs in a citrate-buffered solution. The correlation between the number of detected counts per ion cloud vs. the theoretical number of atoms per particle was shown to be linear for gold particles from 5–20 nm. Above that size, however, the response curve flattens, which the authors attribute to non-detection of ions due to pulse pileup.<sup>27</sup>

In this feasibility study, we report spICP-MS data with a time resolution in the nanosecond range with single quadrupole ICP-MS instrumentation, investigate the possibilities of using dwell times that are significantly shorter than individual transients from AuNP entering the ICP using a standard sample introduction system, and describe a concept on how to process a continuous stream of nanosecond time-resolved spICP-MS (ns-spICP-MS) data for size calibration of nanomaterials.

## 2. Theory

When the SEM is operated in pulse counting mode, each pulse is registered as a detector event (DE) by the pulse detection electronics (above a certain discriminator level). Accordingly, each DE is denoted as one “count”. Only two logical states are possible; either a DE occurs (1 count) or it does not (0 counts).

Conventional pulse counting data acquisition systems for ICP-MS count the number of DE during certain, adjustable dwell times (nowadays as low as a few microseconds up to several seconds)<sup>8,15</sup> and eventually store and display an output signal as a data point (dp) with the accumulated number of counts as y-value and the measurement time of the dwell as x-value. The typical upper limit of the linear dynamic range of the pulse counting stage of an SEM for ICP-MS is ca.  $10^6$  counts per second. A “simultaneous mode” SEM offers the possibility to switch to the analog detection mode above that count rate, which extends the linear dynamic range to ca.  $10^{10}$  counts per second. In this work, the maximum count rate was expected to be below the linear range of the analog detection mode and therefore, only the pulse counting output current was sampled.

During the measurement of solutions with high analyte concentration, many analyte ions are detected during each dwell of the detector and the output signal is accordingly high. When long dwell times are set and the analyte in the solution is highly concentrated, the output signal is assumed to be normally distributed with an integer median value and a standard deviation. For example, a measurement of the  $^{115}\text{In}^+$  ion signal during aspiration of a 1  $\mu\text{g L}^{-1}$  dissolved indium solution with a dwell time of 1 s might produce dp with a mean y-value of 350 000 counts and a typical standard deviation of  $\pm 1\%$  (Fig. 1A). According to normal distribution statistics, between 360 500 and 339 500 DE are counted during each dwell of 1 s with a probability of 99.7% ( $\pm 3\sigma$ ).

The lowest possible absolute number of counts is 0. Most of the dp will be 0, when the analyte ion concentration is extremely low and DE occur rarely during the measurement, or if the dwell time is sufficiently short. At this point, the probability distribution for a certain number of counts to occur during one dwell interval is best described by Poisson statistics. If the same indium solution as described above is analyzed with a dwell time of 5  $\mu\text{s}$ , which can be achieved with our previously presented inhouse-built microsecond data acquisition unit ( $\mu\text{sDAQ}$ ),<sup>8</sup> the probability that a dp has the y-value 0 (no DE



Fig. 1 Theoretical probability histograms for the ICP-MS signal distribution of a solution with homogeneous dissolved analyte ion distribution (e.g., the  $^{115}\text{In}^+$  signal of a 1 ppb In solution). The assumed average signal intensity is 350 000 counts per second. (A) Data acquisition with 1 s dwell time (normal distribution, RSD = 1%). (B) Data acquisition with 5  $\mu\text{s}$  dwell time (Poisson distribution). (C) Data acquisition with 4 ns dwell time (Poisson distribution). The probability that more than one ion hits the detector during one dwell is approx.  $5 \times 10^{-8}\%$ . In practice, the only probable signal intensity values per dwell would be 1 (0.15%) and 0 (99.85%). Please note the logarithmic scale of the y-axis in plot (C).





during dwell interval) is *ca.* 17% (Fig. 1B). The probability that more than 5 analyte ions are detected during one dwell is below 1%.

The nanosecond data acquisition unit (nanoDAQ), which is presented here, allows the acquisition and storage of the pulse counting signal with a recording time interval as low as 4 ns. For comparison, the same calculation as described above for 5  $\mu\text{s}$  dwell time is now performed with 4 ns instead. According to Poisson statistics, the probability that no DE event occurs during one recording interval of 4 ns is *ca.* 99.85%. The probability that exactly one ion hits the detector during one dwell is less than 0.15%, and the coinciding detection of more than 1 ion has a probability of roughly  $5 \times 10^{-8}\%$ . Therefore, the risk that a dp has a *y*-value other than 0 or 1 is virtually eliminated and the transient output signal that is obtained with this nanoDAQ should in theory represent the actual pulse counting signal of the SEM.

In spICP-MS measurements with dwell times shorter than the duration of an analyte ion cloud from a nanoparticle, the ion cloud signal is “oversampled”, *i.e.*, more than one dp is acquired per ion cloud. Different approaches have been described to identify and extract data from individual ion clouds (or particle events) from the raw data of such a time-resolved measurement.<sup>15,26,28</sup> In an exemplary measurement in this study we analyzed ion cloud signal traces of AuNP that were acquired with 5  $\mu\text{s}$  time resolution using the  $\mu\text{s}$ DAQ (data not shown). They were found to have an average duration of 435  $\mu\text{s}$  and comprised 87 dp (average values of 639 ion clouds from 80 nm AuNP). The average maximum number of DE per 5  $\mu\text{s}$  (*i.e.*, per dp) was 40 counts. If the same measurement were to be carried out with 4 ns dwell time, the probability that more than one DE would occur during one dwell would be *ca.*  $6 \times 10^{-4}\%$ . So even during the measurement of nanoparticles up to a size of at least 80 nm, which produce ion clouds with very high analyte ion densities, the acquired data values of 0 or 1 should be reliable.

### 3. Experimental section

#### 3.1 Sample preparation

Monodisperse spherical AuNP in aqueous dispersions with tannic acid coating, an initial total mass concentration of *ca.* 0.05 mg mL<sup>-1</sup>, and nominal sizes of 5, 7, 10, 20, 40, and 100 nm were obtained from nanoComposix (San Diego, CA, USA), as well as citrate-coated AuNP with an initial mass concentration of 0.05 mg mL<sup>-1</sup> and nominal sizes of 5, 10, 30, 50, 80, and 100 nm. Citrate-coated AuNP with sizes of 70 nm (0.05 mg mL<sup>-1</sup>) and 125 nm (0.06 mg mL<sup>-1</sup>) were obtained from Nanopartz Inc. (Loveland, CO, USA). For particle number concentration (PNC) calibration and transport efficiency (TE) determination, quality control material LGCQC5050 (spherical AuNPs with a diameter of  $32.7 \pm 2.0$  nm, a total mass concentration of  $45.1 \pm 1.5$  mg kg<sup>-1</sup>, and a PNC of  $1.47 \times 10^{11} \pm 2.8 \times 10^{10}$  NP per g in 1 mM citrate buffer) were obtained from LGC Limited (Middlesex, UK). For NP size validation, the NIST reference material 8011 (spherical AuNP with a diameter of  $8.9 \pm 0.1$  nm (TEM) and a total mass concentration of  $51.56 \pm 0.23$

$\mu\text{g g}^{-1}$ ) were obtained from the NIST National Institute of Standards and Technology (Gaithersburg, Maryland, USA). Tri-sodium citrate dihydrate (p.a.,  $\geq 99\%$ ) was obtained from Carl Roth GmbH + Co. KG (Karlsruhe, Germany) and diluted in bi-distilled water to obtain a 1 mM citrate buffer solution.

NP dispersions were stored at 4 °C in darkness until the sample dilutions were prepared, except for the NIST RM 8011, which was stored at r.t., as recommended by the manufacturer. Before analysis, NP standard dispersions were let come to r.t., shaken by hand for 30 s, and directly diluted to the desired PNC in 1 mM citrate buffer. The approximate PNC [number of particles per mL (# mL<sup>-1</sup>)] for size calibration solutions was calculated from the mass concentration and particle diameter provided by the manufacturer under the assumption that the AuNP are monodisperse, spherical, and have a density of 19.3 g cm<sup>-3</sup>. Exact NP diameters for size calibration of selected NP standards were determined *via* in-house S-TEM analysis with a minimum number of 100 analyzed particles per sample and are listed in Table 1, as well as the size values as declared by the manufacturers (determined *via* TEM). Nitric acid ( $\geq 99\%$ , analytical reagent grade, Fisher Scientific, Loughborough, UK) was diluted to obtain a 2% HNO<sub>3</sub> solution for rinsing the ICP-MS sample introduction system before each sample.

#### 3.2 Instrumentation

All experiments were carried out on a model iCap Qc ICP-Q-MS (Thermo Fisher Scientific, Bremen, Germany). The Qtegra ISDS software (2.10.3324.131, Thermo Fisher Scientific) was used to control the instrument. Sample introduction was achieved with a model ESI SC-2 DX autosampler (ESI Elemental Service & Instruments GmbH, Mainz, Germany), a MicroFlow PFA-ST nebulizer (Thermo Fisher Scientific) with a sample flow rate of  $165 \pm 1$   $\mu\text{L min}^{-1}$  (determined gravimetrically), and a Peltier-cooled cyclonic quartz spray chamber (cooled to 3 °C). The plasma torch injector inner diameter was 1 mm and the sampling position was set to 1 mm, as well. The high-sensitivity skimmer cone insert “2.8” (Glass Expansion, Melbourne, Australia) for the nickel skimmer cone was used to increase the sensitivity of the instrument. The radio frequency (RF) generator power was set to 1450 W and extraction lens 1 and 2 voltages were 0 V and -154 V, respectively. A second roughing pump was attached to achieve an interface pressure of 1.4 mbar for highest NP detection efficiency. While a second roughing pump was used here, it is not a requirement for the nanosecond time resolution approach to work. The NP transport efficiency (TE) of the system was determined with the particle frequency method using three different concentrations of the reference material LGCQC5050 (3 replicates per concentration) as particle number calibration standard and was found to be  $6.8 \pm 0.2\%$  (RSD = 3%,  $n = 3$ ).

Each NP sample dispersion was analyzed for 30 or 60 s. To avoid carry-over, the <sup>197</sup>Au<sup>+</sup> signal was observed *via* a live monitor view with 5  $\mu\text{s}$  dwell time and the introduction system was rinsed with 2% HNO<sub>3</sub> between samples until almost no more particle signals were detected (at least 30 s) and subsequently rinsed with bi-distilled water to avoid dissolution of NP





the analysis of ion clouds with temporarily high analyte ion densities, detector events are well spaced out on the nanosecond time scale and the majority of dp have the value 0 (*cf.* Theory section above).

Raw data of the nanoDAQ is visualized in this work with a binary viewer program in a simplified way (Fig. 2). The acquired binary data, which is a succession of 0 and 1 (bits), is shown as bytes (8 bits) by the binary viewer program for visualization purposes. Depending on the value of each byte, its respective pixel in the graph is assigned with a customized colour. Here, each byte with a value of 0 (no DE) is represented by a white pixel and each byte with a value other than 0 (*i.e.*, one or up to 8 DE) is represented in blue. The first dp of the data set is in the upper left corner, subsequent dp are shown as pixels from left to right to complete a row and the data rows fill down from top to bottom to the last dp. When the analyte ion density in the plasma is low, ions that emerge from the mass analyzer region are very likely to hit the SEM spaced out in time. Such temporal gaps will appear as a white space in the display. Contrarily, analyte ions arrive at the detector in quick succession over a limited time span, if the analyte ion density is high. In spICP-MS, regions in the data display with a temporarily high density of blue pixels (*i.e.* representing DE) can be safely assumed to originate from the detection of an ion cloud with the SEM. Display settings can be changed to view a succession of ion clouds (Fig. 2A) or to zoom into a singular ion cloud (Fig. 2B) by choosing how many bytes should be binned for one logical pixel.

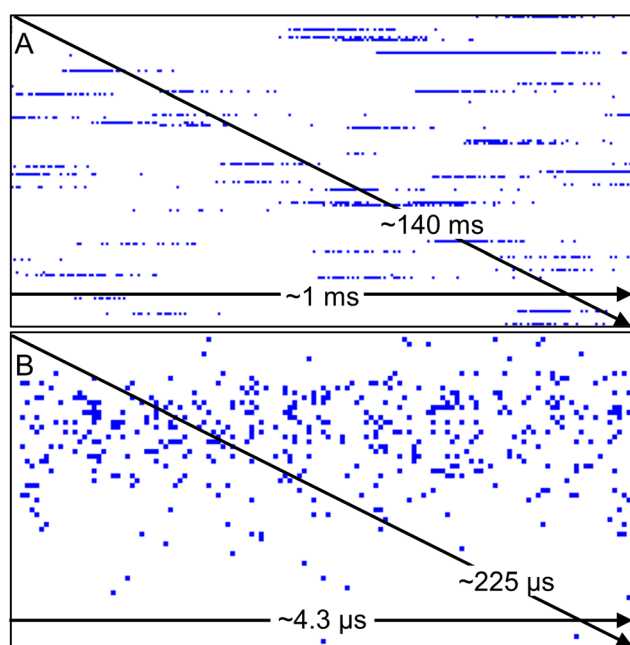


Fig. 2 Simplified visualization of spICP-MS data of a dispersion of 41 nm AuNP with nanosecond time resolution with a binary viewer program. (A) Several ion cloud signals during ~140 ms of acquisition. (B) Detector events in one individual ion cloud during ~225  $\mu$ s data acquisition. Note: each blue logical pixel represents a byte with a value unequal to 0 (1 or more DE), each white logical pixel represents a byte with the value 0 (no DE).

In both representations, similarities can be observed between this data and transient signal profiles of ion clouds that were produced from colloids in an ICP with microsecond time-resolved analysis. In  $\mu$ s-spICP-MS, if an appropriate amount of data points across a peak is available, one typically observes a mass-to-charge ( $m/z$ ) signal trace with a fast “rising edge”, *i.e.*, the ion density in space increases rapidly until a maximum is reached, which then gradually tails off to the background level. With nanosecond time-resolved analysis, it was similarly observed that the SEM pulse density in time, *i.e.*, the density of DE in Fig. 2, seems to correlate with the ion density in space upstream the mass analyzer. Here, earlier studies using time-resolved measurements of particles and single droplets are helpful to understand these transient signals. Transient ion cloud signal profiles have been described previously by Stewart and Olesik *via* experiments and simulations.<sup>30</sup> Both simulated and measured ion cloud signal traces in Stewart and Olesik’s study support our assumption that the data visualized with the binary viewer shows signals that originated from nanoparticles. Further evidence to support this was obtained by additional data processing, which is explained in the next sections.

The number of successive dp with 0 counts between two DE is proportional to the time gap between the two respective detector events. Further in the manuscript, this will be referred to as “event gap (EG)”. If the dwell time ( $\sim 4$  ns) and other SEM settings are kept constant throughout an experiment, multiplication of the observed EG with the dwell time provides transient information on the gap between two DE (*e.g.*: an EG value of 1000 would correspond to a lag of  $\sim 4$   $\mu$ s between two ion impacts at the detector).

Acquired ns-spICP-MS data files are usually too large to be adequately processed with common data processing software as, for example, Origin and Excel. Therefore, a custom data processing software for ns-spICP-MS was developed that compresses the file size by removing all dp with the value 0 but preserving the timing information. Specifically, each individual dp with a value of 1 is now labeled with the number of preceding 0 s as ordinate (*i.e.*, the numerical value of the EG) and with its consecutive number of dp (# dp) in the data set as abscissa (*i.e.*, the timing information). In the following example the pulse counting raw data consisting of 8 dp would be converted to 3 dp in the new following format: raw: (1,0); (2,0); (3,0); (4,1); (5,0); (6,0); (7,1); (8,1)  $\rightarrow$  converted: (4,3); (7,2); (8,0). The factor of data compression depends on the number of DE in the data file, because only the dp with 0-values are removed, *i.e.*, the higher the analyte ion concentration, the larger the converted file size.

The converted data can be imported into Excel or Origin for further processing, for which there are several options: one possibility for data visualization is to plot the dp value (1 or 0) *versus* the absolute measurement time, which is obtained by multiplication of the sequential number of the dp in the data set with the dwell time, as described above. The resulting graph resembles the structure of a bar code. Processed transient data of 1 s analysis of a AuNP dispersion with the nanoDAQ are shown in Fig. 3A, while Fig. 3B shows a zoom into a time window, during which an individual ion cloud was detected.







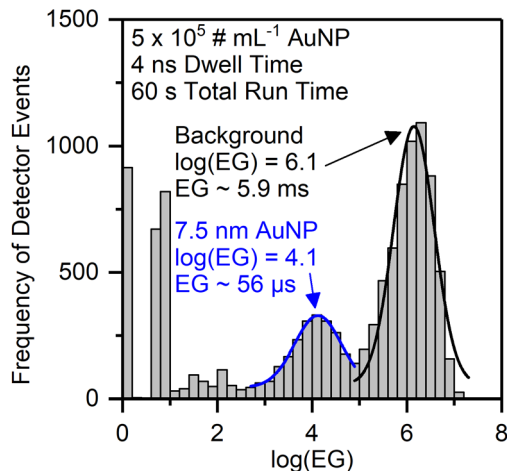


Fig. 5 Histogram of the frequency of detector events vs. the logarithm of the event gap ( $\log(\text{EG})$ ). In the bimodal distribution, the right peak with longer event gaps between detector events (avg.  $\sim 5.9$  ms) corresponds to the background (black) while the left shoulder with a shorter mean event gap ( $\sim 56$   $\mu\text{s}$ ) corresponds to ions from 7.5 nm AuNP (blue).

constructed from all detector events, so they only represent the merged information on ions from all nanoparticles, which is currently due to limitations in the data processing software and we hope to be able to improve this in a follow-up study to provide both nanosecond time resolution and size information on all individual particles. Hence, the mean  $\log(\text{EG})$  value of the NP signal population represents the mean temporal gap between two ions arriving at the detector, averaged over all ions in ion clouds from the respective NP population during the entire measurement. The calculation of the size resolution capability of the method is therefore not applicable.

We successively analyzed dispersions of AuNP with different sizes and processed the nanosecond time-resolved data as described above. With our nanoDAQ system in combination with decreased mass resolution we were able to distinguish signals from AuNP as small as 7.5 nm from the background in combination with a commercial single quadrupole ICP-MS (Fig. 5). Signals from particles with a size of 5.3 nm were also detected but the separation from the background signal was considered to be not sufficient.

Because of these promising results above, the mean  $\log(\text{EG})$  value of the particle population (Gauss fit) was calculated for each particle size and it was found that the mean  $\log(\text{EG})$  is inversely proportional to the particle size (Fig. 6A). Ion clouds from small particles show a higher mean  $\log(\text{EG})$  than large particles, *i.e.*, the mean time gap between two DE in an ion cloud from a small particle is longer. Smaller particles require less energy for complete vaporization, atomization, and ionization than larger particles and consequently, their ion clouds start to diffuse earlier in the plasma, which decreases their spatial ion density (in addition to broadening effects downstream). When an ion cloud with a low spatial ion density emerges from the mass analyzer, this results in the detection of ions with long temporal gaps between them. Based on this

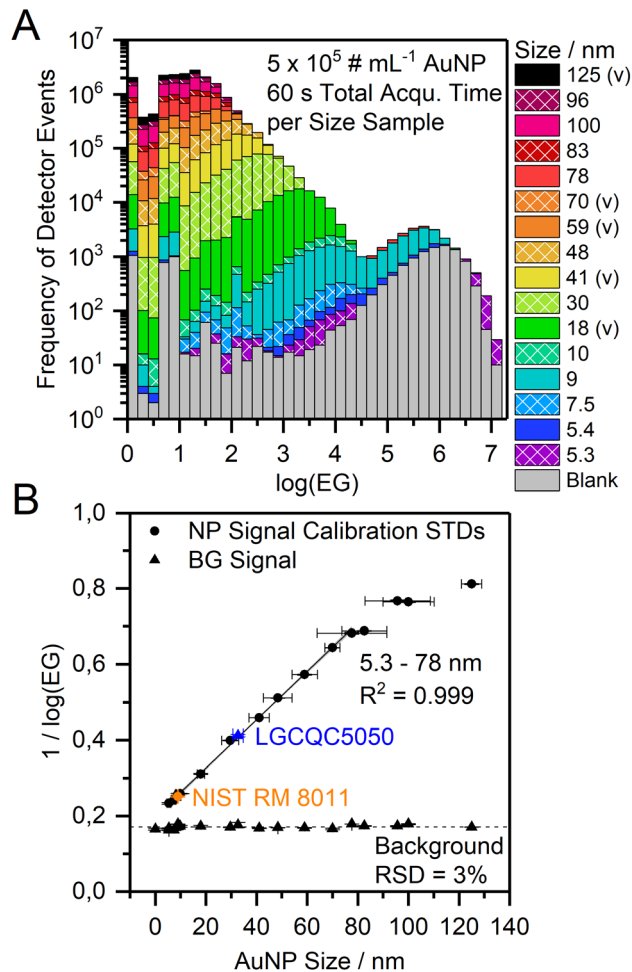


Fig. 6 (A) Frequency histograms of the logarithm of the event gap ( $\log(\text{EG})$ ) of the individual measurements of AuNP with 16 different sizes and an approx. PNC of  $10^5$  #  $\text{mL}^{-1}$  and a blank solution. NP sizes as declared by the manufacturer are indicated with (v). Please note the logarithmic scale on the y-axis. (B) Plot of AuNP size versus the mean inverse logarithm of the event gap of 16 AuNP size standard solutions (black circles) and two certified reference materials (NIST RM 8011: 8.9 nm; LGCQC5050: 32.7 nm). Mean values of three repetitive measurements (60 s) per size are shown. The x-error bars correspond to the SD of the particle size and the y-error bars indicate the SD of  $1/\log(\text{EG})$  from three repetitive measurements. The mean inverse logarithm of the event gaps of the background signal distribution is independent of the particle size and varies by an RSD of  $\pm 3\%$  among the 17 samples (black triangles).

and our observation of the EG values one likely explanation is that the mean spatial ion density in ion clouds from small particles is lower at the detector than in ion clouds from large particles (assuming no other contributing factors such as space-charge effects, ion focusing *etc.*). Moreover, we empirically found that the correlation between inverse  $\log(\text{EG})$  ( $1/\log(\text{EG})$ ) and original particle size is linear over a range of 5.3–78 nm ( $R^2 = 0.999$ ) and may, therefore, be a novel approach for NP size calibration (Fig. 6B). Apart from the qualitative considerations on the relationship between particle size and mean event gap of the corresponding detected ion









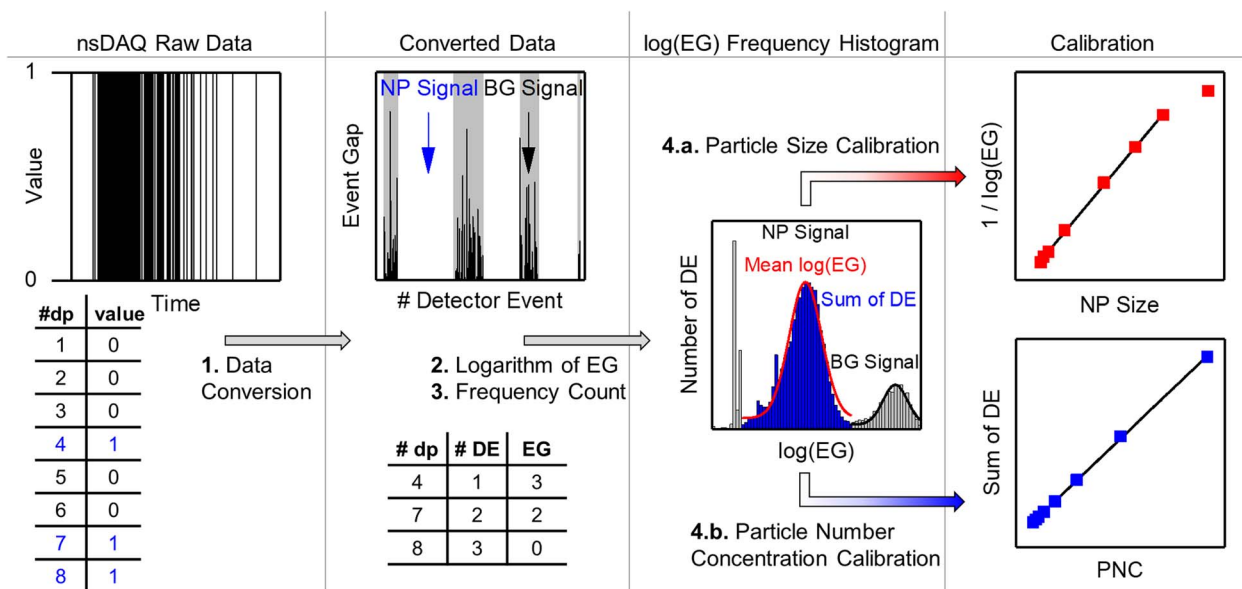
**Fig. 8** Comparison of spICP-MS data acquired simultaneously with (A) 4 ns (nanoDAQ), (B) 5  $\mu$ s ( $\mu$ sDAQ), and (C) 0.1 ms (vendor software) dwell time. Dispersions of AuNP with nominal sizes of 18 and 7.5 nm and a blank solution were measured for 30 s, respectively. Please note that the background signal of nanoDAQ data is on the right side (high  $\log(\text{EG})$ ), while the background in plots from conventional spICP-MS is on the left.

corresponds to an average of all detected particles. Information on possible outliers or on the degree of monodispersity cannot be drawn from this with the current software prototype. Contrarily, the  $\mu$ sDAQ and vendor software can be used to show a representative particle size distribution and can distinguish different particle sizes in a polydisperse NP sample.

The clear advantage of the nanoDAQ over the other methods lies in the rapid analysis and low size detection capabilities. Because the  $\log(\text{EG})$  histograms are created from data of each detected ion from all ion clouds, it naturally contains more data points than the particle-based (ion cloud-based) histograms of the alternative methods. Therefore, a statistically relevant number of data points can be gathered in a shorter time span. In comparison: during 30 s of measurement,  $\sim 400$  and  $\sim 560$  particle signals were detected with the vendor software and the  $\mu$ sDAQ, respectively, while during the same time *ca.* 14 000 detector events were recorded by the nanoDAQ.

### 4.3 Analysis of more complex nanoparticle dispersions

Mixtures of AuNP with two different sizes with a PNC ratio of 1 : 1 were analyzed with the nanoDAQ to assess the methods capability to characterize more complex nanoparticle samples. As already outlined above, the  $\log(\text{EG})$  histograms obtained from the nanoDAQ data show the merged  $\log(\text{EG})$  values of all ions that were detected during the entire acquisition time. Results from the measurement of test mixtures confirm that the created  $\log(\text{EG})$  frequency histograms show a convolution of the event gaps between ions in the ion clouds from both types of particles, as well as the BG signal distribution (see ESI, Fig. S1†). However, the  $\log(\text{EG})$  distributions are relatively broad in general, because the event gap is not even constant during the detection of one ion cloud. As shown above, the ion density in an individual ion cloud is high at the rising edge (low EG) and decreases towards the end (high EG). Therefore, each ion cloud creates a respective distribution of  $\log(\text{EG})$  values, and the measurement of many ion clouds creates a cumulated distribution of  $\log(\text{EG})$  values. The range of  $\log(\text{EG})$  values from measurements of monodisperse particle solutions ranges from *ca.* 1–5.5 (*ca.* 40 ns to 1.3 ms), which means that the distributions of all particle sizes overlap. Because smaller particles produce orders of magnitudes less ions than larger particles,



**Fig. 9** Workflow diagram for the processing of nanosecond time-resolved spICP-MS data.



the amplitudes of the log(EG) distribution of small particles are much lower than those of larger particles. Those two factors lead to difficulties in distinguishing two different particle sizes in a mixture, because the distribution of the larger particle size will overlap the distribution of the smaller particle type. The issue might be redeemed to some extent by increasing the particle number concentration of the smaller particles and to ensure a large size discrepancy between the two different types, which would severely impede the applicability of this approach to unknown samples.

#### 4.4 Summary of the data processing workflow for spICP-MS with the nanoDAQ

Based on the findings in this work, the optimized data processing workflow for nanosecond time-resolved spICP-MS data is visualized in Fig. 9 and can be summarized into the following steps:

(1) The nanoDAQ raw data is converted to remove data points (dp) with the value “0” and labels each detector event (DE) with the respective event gap (EG).

(2) The logarithm of the EG is calculated to obtain log(EG).

(3) The log(EG) values are binned and a frequency count is performed to create a histogram.

(4a) The inverse mean log(EG) of a Gaussian fit (red curve) to the particle signal distribution in the histogram is used for size calibration.

(4b) The sum of DE of the particle signal distribution (blue) is used for particle number concentration calibration.

## 5. Conclusions

These proof-of-principle experiments with nanosecond time resolution spICP-MS show that it is possible to acquire, store, and process data that represents the SEM pulse counting signal and to utilize the temporal gaps between individual detector events from ion clouds originating from AuNP in the plasma to quantify nanoparticle characteristics. Our method enables rapid nanoparticle size and PNC determination of monodisperse, non-complex particle dispersions with low size detection limits with a common single quadrupole ICP-MS instrument and a standard sample introduction system. In combination with reduced mass resolution for  $^{197}\text{Au}^+$ , we could distinguish signals from 7.5 nm AuNPs from the background signal in 60 s of net measurement time (excluding sample uptake and rinsing times). Because the data acquisition system and data processing method are independent of the sample introduction system, the analysis speed could be increased significantly by exchanging the currently used autosampler with a syringe-driven autosampler with shorter sample tube lengths. The currently applied pathway for particle sizing provides only average size information on the complete dataset of one measurement. While this is a drawback for conventional spICP-MS, which demands for particle-by-particle analysis, it does immensely increase the number of data points in the size histogram (here, the number of detected ions in the measurement is counted instead of the number of particles). Therefore,

the total measurement time could be reduced even further (*e.g.*, to 30 or 10 s) and still a statistically significant number of data points would be obtained for the size histogram. Furthermore, the sum of detected events can be used for particle number concentration calibration of monodisperse particles with known sizes, provided that measurement parameters are constant.

A custom-built data acquisition unit for spICP-MS with nanosecond time resolution enabled the recording of individual ion clouds at the SEM detector of the mass spectrometer. Time-resolved pulse counting signals were obtained and give some insights on ion arrival times at the detector. Time spans with no or very little counts were assumed to be background noise and time spans with a high event density were assumed to originate from dense ion clouds.

With our tailored data processing method, we could visualize and quantify the temporal gap or event gap, with which the ions arrive at the detector during the analysis of gold nanoparticle dispersions. We found empirically that the average inverse logarithm of the EG of the ions that stem from ion clouds is linearly proportional to the mean of the original particles' size. This proportionality can be utilized for a particle size calibration for unknown particle dispersions. However, the average value of the event gap is considered a measure for the ion density in the ion cloud. Large EG values are obtained from ion clouds, in which there is both a large temporal and, correspondingly, a large spatial gap between ions. Because we found a linear relationship between  $1/\log(\text{EG})$  and particle size, we can conclude that the average ion density in an ion cloud from a particle appears to be inversely proportional to the original particle's diameter.

Our nanosecond data acquisition system allows us to sample the SEM pulse counting signal above a certain discriminator level within time intervals of circa 4 ns (4.167 ns). The detector dead time, however, is assumed to be in the order of 50 ns.<sup>33</sup> Consequently, pulses (*i.e.* ion events) that occur within the dead time of the SEM cannot be acquired with the system. If the ion density in an ion cloud is sufficiently low, relatively few ions will be lost due to the dead time and the distances between detector events will remain mainly unaffected. Contrarily, the non-detection of an ion during the detector dead time will falsely increase the event gap between its proceeding and succeeding ion, which in consequence leads to an overestimation of the mean event gap of the particle signal distribution. This is a probable explanation for the underestimation of the  $1/\log(\text{EG})$  value of larger NP, *i.e.*, the AuNP with 100 nm diameter, which we observed in our data.

Similarly, the nanoDAQ data processing method does not yet correct for pulse oversampling effects of the SEM pulse counting output. Specifically, if the duration of a pulse from an incident ion exceeds 4 ns above the discriminator level, it will be sampled twice, and two (or more) successive data points will be stored with a value of 1. During data conversion, the second (or third) data point will be extracted as a detector event with an event gap of 0. Because it is not a valid mathematical operation to take the logarithm of 0, these DE will not be incorporated into the frequency count histograms.



Furthermore, with the current prototype version of the nanoDAQ we observed both periodically and irregularly occurring pauses during the data acquisition, which lead to data losses. Because the transient information presented above (Fig. 2–4) was calculated by multiplying the indices of DE in the data file with the dwell interval of 4 ns, the resulting ion cloud durations or event gaps in  $\mu\text{s}$  or ms can only be understood as rough estimates.

The subject of further studies will be to improve the data acquisition and processing (a) to remedy the data losses during acquisition, (b) to extend the linear size calibration range by applying a suitable dead time correction, (c) to increase the signal-to-noise ratio for analyses of samples with higher ionic background, (d) to distinguish between particles of different sizes in polydisperse mixtures, and (e) to utilize the event gap information to gain insights into the initial particle's geometry.

## Author contributions

The manuscript was written through contributions of all authors. All authors have given approval to the final version of the manuscript.

## Conflicts of interest

There are no conflicts to declare.

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