Modulated 3D Cross-Correlation Dynamic Light Scattering Applications for Optical Biosensing and Time-Dependent Monitoring of Nanoparticle-Biofluid Interactions

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Featured Application: Advanced optical sensing for biomedical, food and drug development, monitoring of nanoparticle interactions.

Abstract: This paper reviews dynamic light scattering (DLS) of gold nanoparticle-protein interactions for the model protein bovine serum albumin (BSA), as well as in complex biofluids, at the example of mouse serum. DLS data of nanorods of various aspect ratio, of proteins and of mouse serum are discussed in terms of the analysis of their hydrodynamic radii, leading to the distinction of rotational and translational motion as well as to the detection of agglomerates. We address in particular advances obtained by modulated 3D cross correlation dynamic light scattering and recent progress using the CORENN algorithm for analysis of DLS data.

Keywords: dynamic light scattering; plasmonic nanoparticles; gold nanorods; nanoparticle-protein interaction; monitoring; BSA; mouse serum; biofluid; hydrodynamic radius

1. Introduction

Dynamic light scattering (DLS) is widely used for the study of nanoparticles, colloidal particles, for the study of proteins, as well as for the field of nanoparticle suspensions, such as nanoparticle inks [1,2]. The technique is widely used in research, and still generally associated to the need of dilution to low concentrations, and lengthy tests for obtaining results that can be reliably compared. Considerable advances were made by the introduction and development of modulated 3D-cross correlation dynamic light scattering (3D-ccDLS), which allows to perform DLS experiments even in turbid samples [3], as well as in the development of algorithms for the analysis of hydrodynamic radii from the optical data that are recorded in DLS measurements also in the case of polydisperse particle distributions, e.g., [4–6]. These technical advances on the experimental and on the analysis side provide thus timely tools, to address the investigation of nanoparticle interactions with complex fluids, such as biofluids even at high concentration.

At the examples of gold nanoparticles, nanorods, model proteins and mouse serum, this review article gives an overview of recent advances to study their motion and interaction by the use of 3D-ccDLS and reviews major considerations for DLS data analysis which illustrate that modulated 3D cross-correlation dynamic light scattering can be successfully used for the analysis of complex biofluids and for monitoring nanoparticle-protein interactions.
2. General Methods and Materials

2.1. Dynamic Light Scattering and Modulated 3D Cross-Correlation Dynamic Light Scattering

The principle of a dynamic light scattering experiments for nanoparticle sizing is based on Brownian motion and Rayleigh scattering from particles which are significantly smaller than the laser wavelength. The scattered intensity is recorded as a function of time and the autocorrelation function is analysed in order to determine diffusion coefficient(s) and to relate their values to hydrodynamic radii of the particles in motion [7].

While in a typical dynamic light scattering experiment multiscattering events can strongly affect the results, a modulated 3D cross-correlation experiment eliminates multiscattering events during the measurement and takes into account only single scattering events [3]. This can be successfully implemented by using two laser beams who alternately impinge on the analyzed area and two high speed detectors. As a result, samples can be measured even at high concentration and the need of dilution tests in order to obtain reliable results is significantly reduced [3]. Using this method, we have recently studied nanoparticle suspensions and inks [8]. In particular time-dependent behavior, such as particle-particle interactions, protein-corona formation, or formation of agglomerates can thus directly be measured and monitored at high concentration [9,10].

For our own studies, presented in this article, we have used a 3D Nanolab set-up (LS Instruments, Fribourg, FR, Switzerland), equipped with two laser beams (685 nm) and two detectors, temporally isolated from each other; the high-speed intensity modulators work at a frequency of 5 MHz, the scattering angle is 90°, data were recorded at 20 °C. Data analysis was performed using the LsLab RC software version from 2019 (LS Instruments, Fribourg, FR, Switzerland). As the focus of the article is to give an overview on several studies, experimental details for sample preparations and experimental procedures, are either given in the respective sections of this review or they follow the procedures which are specified in the cited references.

2.2. Gold Nanoparticles, Nanorods, Model Protein and Mouse Serum

Gold nanoparticles and gold nanorods are of broad interest in the field of sensor development and biomedical applications [11], are thus widely investigated for the study of interactions with proteins, also using DLS, e.g., [12].

For our studies discussed in this article, mouse serum was purchased from BioWest, Nuaillé, France, and model proteins bovine serum albumin (BSA) from Sigma Aldrich Chemie GmbH, Buchs, SG, Switzerland. Gold nanoparticles, nanospheres and nanorods, stabilized by citrate in aqueous suspension, were purchased from NanoComposix, San Diego, CA, USA); aspect ratios, nanoparticle dimensions and concentrations of the suspensions are summarized in Table 1:

| Gold Nanoparticles 1 | Aspect Ratio | Dimensions L, D (nm) 2 | Concentration (µg/mL) |
|---------------------|-------------|------------------------|-----------------------|
| Nanospheres         | 1.0         | D: 40                  | 50                    |
| Nanorods-600        | 2.6         | L: 48, D: 18           | 31                    |
| Nanorods-800        | 3.6         | L: 55, D: 15           | 16                    |
| Nanorods-980        | 5.6         | L: 85, D: 15           | 18                    |

1 Numbers refer to the wavelengths of absorbance (in nm); 2 Nominal dimensions—L: length, D: diameter.

3. Monodispersed Nanorods: Bimodal Distributions due to Translation and Rotation

This section reviews the analysis methods of DLS data in case of monodispersed nanorods. First the basic analysis of the autocorrelation function of particle suspensions is summarized, leading to
an average diffusion coefficient, hydrodynamic radius and a polydispersity index according to the Cumulants method.

Second, analysis methods for the decay function in case of polydispersed samples thus involving several diffusion coefficients are summarized.

Third, the analysis of rotational and translational contributions to the diffusion of particles of non-spherical shape, such as nanorods, is summarized and discussed at the examples of DLS measurements of gold nanorods. These methods are also applicable to other non-spherical particles such as proteins.

3.1. Basic Analysis of Particle Suspensions: Hydrodynamic Radius and Polydispersity Index

For monodispersed spherical particles, the autocorrelation function decays exponentially. The decay parameter is proportional to the translational diffusion coefficient $D_t$. The Stokes-Einstein equation describes the (translational) diffusion coefficient $D_t$ in relation to the hydrodynamic particle radius $R_{Ht}$, expressed by:

$$D_t = \frac{kT}{6\pi\eta R_{Ht}}, \quad (1)$$

where $T$ is the absolute temperature, $k$ the Boltzmann constant, and $\eta$ the dynamic viscosity of the liquid. If $k$, $T$, and $\eta$ are known, the hydrodynamic particle radius can thus be determined from Equation (1) [7].

Inversely, if $R_{Ht}$ is known, it becomes possible to determine the viscosity of the liquid from DLS experiments using well defined monodispersed spherical tracer particles [13,14]. Such viscosity measurements of BSA protein suspensions have recently been demonstrated using modulated 3D cross-correlation DLS [14]. Their analysis based on microviscosity measurements using DLS agrees well with results of macroviscosity measurements and shows that viscosity measurements can be performed with small amounts of liquid.

The diffusion coefficient, $D_v$, is extracted from the autocorrelation function of the detected scattered light intensity as measured by DLS. In case of spherical particles, analysis is based on an exponential fit of the autocorrelation function, and generally referred to as a Cumulants analysis, and defined in ISO 13321 and ISO 22412. This analysis also gives information on the polydispersity index PDI, which reflects the width of the size distribution.

Although widely used as standard, Cumulants analysis has significant disadvantages in case of multi-dispersed particle distributions. This becomes evident, e.g., when considering a suspension with a bimodal distribution of monodispersed spherical particle sizes, which obviously are physically governed by two different diffusion coefficients as smaller particles diffuse faster than larger ones. Therefore, further methods are used to determine diffusion coefficients, and thus hydrodynamic radii, from the measured autocorrelation function of an DLS experiment in the case of multi-dispersed samples.

3.2. DLS Measurement of Multi-Dispersed Particle Suspensions and Analysis of the Autocorrelation Function

For multi-dispersed particle suspensions, such as mixtures of nanoparticles of several sizes or e.g., mixtures of nanoparticles and proteins, particles have different velocity leading to different time scales of the correlation decay. As a consequence, the resulting decay of the autocorrelation function cannot any longer be modeled by a single exponential. If a suspension contains fast and slow moving particles, the experimental autocorrelation function of the scattered intensity, $g^{(2)}(q,t)$, can e.g., be modeled by two exponential decay functions as expressed by Equation (2):

$$g^{(2)}(q,t) = 1 + \beta^2[A_f \exp(-\Gamma_f t) + A_s \exp(-\Gamma_s t)]^2 \quad (2)$$

where $\Gamma_f$ and $\Gamma_s$ are the decay rates related to the fast and slow moving particles, and $A_f$ and $A_s$ the corresponding amplitudes [15].

Therefore, further algorithms exist to take into account a composition of several decay time scales in the analysis of the autocorrelation function [16]. The CONTIN algorithm is a widely used example [4].
A comparative study of several algorithms has recently been performed showing the non-equivalence between methods as well as the robustness of the Cumulants method [16]. For polydisperse particle distributions, the CONTIN algorithm is performant and widely used, however, it has also limitations, as it is e.g., known to be sensitive to signal noise at large lag times, and can in some cases also lead to phantom peaks. Methods have thus been developed to improve its performance [5]. A very recent algorithm, which is based on a machine learning approach in order to circumvent those drawbacks of the CONTIN algorithm, is provided by the CORENN method, which has been developed and implemented by LS Instruments [6]. The acronym CORENN stands for COstrain weighted, Regularized Non-linNear, least-squares fit with free baseline for the integral inversion. According to the developer LS Instruments, Fribourg, FR, Switzerland, this algorithm uses a similar approach as the CONTIN algorithm and uses for the solution of the non-linear inverse integral a novel robust and fast theoretical estimate of the correlation function noise to remove noisy data. In particular, the standard deviation of the noise of the cross-correlation function is estimated from the experiment and used in the algorithm to weight data points depending on their reliability as assessed by the noise estimation itself [6]. Recently, we have used the CORENN method to analyse the motion of gold nanoparticles and gold nanorods, as well as the one of model proteins (BSA) and their interaction with nanoparticles [9].

3.3. DLS Measurement of Rotational and Translational Contributions in Particle Motions: Nanorods

For non-spherical particles translational and rotational motion takes place, both affect the observed decay of the autocorrelation function. For monodispersed nanorods, the autocorrelation function is thus governed by two exponential decay parameters, one is proportional to the translational diffusion coefficient $D_t$, the other one to the rotational diffusion coefficient $D_r$. One distinguishes in this case a hydrodynamic radius related to translational motion, $R_{ht}$, which is calculated according to Equation (1), and a hydrodynamic radius related to rotational motion, $D_{hr}$, which is expressed by Equation (3) [17]:

$$D_r = kT/8\pi\eta(R_{hr})^3,$$

3.4. DLS Measurement of Rotational and Translational Contributions in Particle Motions: Nanorods

The analysis of nanorods in DLS leads to two relaxation modes in the intensity correlation function, as observed and discussed by Rodríguez-Fernández et al. [18]. From angular dependent DLS experiments on gold nanorods, recorded in different polarization geometries (vertical-vertical vs vertical-horizontal), they have shown that it is possible to determine translational and rotational diffusion coefficients of nanorods from the decay of the intensity correlation function, and attributed this to anisotropy of the polarizability caused by the shape of the particles [18]. Methods for the analysis of non-spherical nanoparticles based electrical birefringence methods are discussed in [19].

Using 3D-ccDLS measurements and the CONTIN and CORENN methods, has allowed us, to determine the rotational and translational diffusion coefficients for gold nanoparticles (spheres and rods, aspect ratios varying from 1 to 5.6), and to calculate the corresponding hydrodynamic radii [10]. A comparison of Cumulants, which are stable in the range of investigated concentrations from 5 to 16 µg/mL [9], with the CORENN analysis has recently been performed by analyzing the modes of the CORENN size distributions. The values obtained from Cumulants and CORENN analysis were compared by an extrapolation to zero concentration limit by linear regression. We could show that the CORENN method provides precise values of the hydrodynamic radius for rotational and translational motion of the nanorods [9]. The obtained values for the rotational motion (as listed in Table 2) are smaller than the ones earlier obtained using the CONTIN analysis which led to $R_H$ values around 2·nm for the rotational motion and around 19 nm for the translational motion [10].
Table 2. Hydrodynamic radii of gold nanospheres and gold nanorods, as obtained from the analysis of intensity size distributions: extrapolations for the zero concentration limit by linear regression, based on Cumulants and CORENN method. For nanorods the hydrodynamic radii of translation, $R_{0t}$, and rotation, $R_{0r}$, were calculated with the CORENN method [9].

| Nanoparticles | Cumulants $R_0$ (nm) | CORENN $R_{0t}$ | CORENN $R_{0r}$ |
|---------------|----------------------|-----------------|-----------------|
| Au-NP1.0      | 28.50 ± 0.17         | 24.2 ± 0.7      | N.A.            |
| Au-NP2.6      | 1.15 ± 0.09          | 8.7 ± 0.3       | 0.423 ± 0.007   |
| Au-NP3.6      | 0.86 ± 0.05          | 8.4 ± 1.5       | 0.401 ± 0.010   |
| Au-NP5.6      | 2.61 ± 0.04          | 13.6 ± 1.4      | 1.55 ± 0.08     |

1 Numbers refer to the aspect ratio of the gold nanoparticles.

For example, for the gold nanospheres a hydrodynamic radius of (24.2 ± 0.7) nm was found from the analysis of the mode by the CORENN method, compared to an average of 27.6 nm by the CONTIN algorithm [10] and (28.50 ± 0.16) from Cumulants [9].

The fact that the hydrodynamic radius obtained for nanospheres in DLS experiments exceeds the nominal particle radius of 20 nm (40 nm diameter) as indicated by the supplier based on TEM analysis, is observed in all three cases and could potentially be explained by interaction of the particle with molecules from the liquid [20]. We note, that the above listed values correspond to the hydrodynamic radii as obtained from the analysis of intensity size-distributions. For the nanospheres, by analyzing the number-size distributions, we obtain values around 19 nm, which is in good agreement with the nominal particle radius. An example of an intensity correlation function as measured for the Au NPs at a concentration of 0.0016 mg/mL, as well as the intensity and number distributions obtained from CORENN analysis are shown in Figure 1.

Figure 1. Gold nanospheres, an example of a single measurement and CORENN analysis is shown (concentration 0.0016 mg/mL): (a) Measured intensity correlation function with CORENN-fit (red circles: measured signal; green: CORENN-fit, blue vertical lines: lower and upper limit of the time used for the fit); (b) intensity distribution, (c) number distribution of the hydrodynamic radii (green curves). Average values from the distributions (b, c) are indicated as vertical red line, the dashed blue lines indicate the interval of the standard deviation.
In summary, we find that the hydrodynamic radius obtained from the CORENN method is the closest to the nominal particle radius when analyzing the mode of the intensity distributions. The average hydrodynamic radius as obtained from the number distributions (about 19 nm), which can reasonably be analyzed for monodispersed spherical particles, corresponds well to the nominal particle radius of 20 nm (40 nm diameter).

Methods to relate the hydrodynamic radii of translation and rotation of nanorods and rod-like macromolecules to the particle dimensions have been proposed and demonstrated in [17,21].

3.5. Time-Dependent Monitoring of Protein-Nanoparticle Interactions

Time dependent monitoring of protein-nanoparticle interactions by DLS is of interest for the investigation of the initial phases of interaction and the study of corona formation, which dominates the further interaction with other biological fluids, see e.g., [12] and references therein.

Recently we have used 3D-ccDLS and the CORENN method, for the study of interaction and corona formation between gold nanoparticles (spheres and rods as listed in Table 1) and the model protein bovine serum albumin (BSA). By comparing the scattering intensities of rotational and translational modes from CORENN analysis over time, we could show that BSA preferentially interacts at the cylindrical side of the gold nanorods. 3D-ccDLS combined with the CORENN method provides thus a powerful tool for monitoring of protein-nanoparticle interactions [9].

Time dependent investigations of the influence of other parameters, such as the influence of the size, capping layer, or zeta potential [22], could be further studied in order to gain a deeper understanding of the corona formation on nanoparticle with proteins and in biological fluids.

4. Analysis of Mouse Serum and Interaction with Nanoparticles

This section presents an application example of 3D-ccDLS and CORENN method to study the interaction of nanoparticles with a more complex biofluid. Recently we have shown that CONTIN method can be used to distinguish translational and rotational motion of the components of mouse-serum and to investigate changes of their hydrodynamic radii upon interaction with gold nanoparticles of different shape [10] at a mouse serum-concentration of 33%. As a further application example for particle monitoring with 3D-ccDLS and CORENN method, we present here recent results from a study of the interaction of gold nanospheres and a low concentration of mouse serum.

The chosen mouse serum is the same product of sterile, pre-filtered mouse serum (BioWest, Nuaillé, France) as used in [10]. We filtered the purchased serum through a 1.2 µm micropore membrane in order to remove large agglomerates and thus to improve the signal stability for DLS measurements [10]. Figure 2 shows the result of the additional filtering.

![Autocorrelation Function of Mouse Serum](image1)

![Particle Size Distribution](image2)

**Figure 2.** DLS data of mouse serum before and after filtering through a 1.2 µm micropore membrane: (a) Correlation functions; (b) Example of relative scattered intensity as a function of hydrodynamic radius based on CONTIN analysis.
Upon filtering the correlation function decays faster, and shows less noise at large lag times, which is consistent with the removal of larger particles. The analyzed PDI decreased from 2.72 to 1.61. Figure 2b furthermore illustrates an example that the presence of large agglomerates before filtering might affect calculated results of hydrodynamic radii also for smaller particles. Here, an example is shown based on CONTIN analysis. The observation of a shift towards smaller apparent hydrodynamic radii may indicate a change in particle-particle interactions or a possible change of the effective viscosity upon removal of bigger agglomerates, however, it also illustrates the high sensitivity of the CONTIN algorithm to noise at large lag time, when the autocorrelation function approaches zero.

Using the CORENN method, we have performed further analysis of our 3D-ccDLS data recorded at various volume concentration between 9% and 100% (with respect to the filtered mouse serum), using DNase-free distilled water for dilution. Figure 3 shows the Cumulant and CORENN analysis, as well as optical absorbance of the filtered mouse serum at various concentration.

The optical UV-vis-NIR absorbance spectra of mouse serum (Figure 3c), show absorbance in the ultraviolet range, which is characteristic for proteins. In agreement with the Beer-Lambert law, the absorption peak intensity at 410 nm, which can be indicative for enzymatic reactions or other components of the mouse serum, is proportional to the serum concentration (Figure 3c) [10]. As main components we consider proteins in the mouse serum.

Mouse serum was recently studied by CONTIN analysis [10]. Rotational and translational motion were distinguished as well as the presence of agglomerates [10]. Figure 3b shows results from CORENN analysis of mouse-serum DLS data, samples were prepared and data recorded as described in [10]. The analysis using the CORENN method shows consistency with the CONTIN analysis presented in [10]. It clearly distinguishes the rotational and translation motion of proteins in the mouse-serum, as well as the formation of agglomerates.

The analyzed mode of the rotational peak, with a hydrodynamic radius between 2 and 5 nm, stays stable over the full range of concentration (Figure 2b), and can be related to the particles lengths [23].

The mode with a hydrodynamic radius around 14 nm (Figure 3b) is related to translation; its obtained positions show similar values over the entire range of concentrations within the error (Figure 3b). The value obtained at a low concentration of only 9% of the mouse serum, as analyzed...
using the CORENN algorithm is well in agreement with the one obtained at higher concentration, 
e.g., 33% (Figure 3b).

Larger hydrodynamic radii of around 200 nm are associated to agglomerates (Figure 3b) as 
discussed in [10], the hydrodynamic radius is stable for concentrations of 33% and below (9%). 
The result stays in agreement with the CONTIN analysis within the errors [10]. As a tendency the 
CONTIN analysis [10] shows for agglomerates values between 100 and 140 nm which are also obtained 
by CORENN analysis at concentrations above 33% (Figure 3b); further investigations would thus be 
needed in order to best quantify the hydrodynamic radii of large particle agglomerates.

Essentially, we find that the CORENN analysis is able to provide conclusive results for concentration 
down to 9% mouse serum concentration (Figure 3b) and which are in agreement with the values 
obtained at 33%, we note that the CONTIN algorithm used for the data analysis in [10] did not allow 
for a stable analysis of the DLS data at the very low concentration of 9%.

5. Detection and Monitoring of Nanoparticles Interaction in Mouse-Serum

The detection of nanoparticles in complex biological fluids is of relevance to study 
nanoparticle-protein interactions in a biologically representative environment. We have recently shown 
that the interaction of gold nanospheres and nanorods can be successfully detected in mouse-serum 
and have compared their interaction for different aspect ratios of the nanoparticles using the CONTIN 
algorithm [10]. Figure 4 shows the results based on the analysis using the CORENN algorithm.

![Figure 4](image_url)

**Figure 4.** Hydrodynamic radii of filtered mouse serum upon interaction of mouse serum (33%) with 
gold nanoparticles analyzed using the CORENN algorithm. The hydrodynamic radii of the pure 
filtered 33% mouse-serum are shown for the convenience of comparison in the same graph.

While for nanospheres (aspect ratio 1.0) and for nanorods with an aspect ratio of 2.6 DLS results 
show only small changes of the hydrodynamic radii of the mouse serum (within the error bars), 
nanorods with aspect ratios of 3.6 and 5.6 lead to an increase of the hydrodynamic radius related to 
translational motion. These observations are indicative of interaction between the nanoparticles and 
the serum. The CORENN analysis confirms the observations based on CONTIN analysis [10].

In order to investigate whether proteins of the mouse serum interact however with the nanospheres, 
we monitor citrate stabilized gold nanospheres, nominal diameter of 40 nm, (NanoComposix, San Diego, 
CA, USA, 0.05 mg/mL dispersed in water) together with a low concentration of the mouse serum over 
time. For incubation experiments, the mixture of the gold nanoparticle suspension and the mouse 
serum was prepared from the already diluted mouse serum of 33% concentration, which we used at a 
concentration of 7% leading to a mouse serum concentration of 2.3% for incubation.

We monitor the hydrodynamic radii. Figure 5 shows results from cumulants and CORENN 
analysis of the gold nanospheres together with mouse serum over time. At these low mouse serum 
concentrations and high nanoparticle concentration, the translational peaks of the nanospheres are
clearly identified. The hydrodynamic radius of the nanospheres initially decreases, which might be due to an increase in viscosity and/or signal contributions from translations of the mouse-serum.

![Figure 5](image.png)

**Figure 5.** Gold nanospheres together with mouse serum (2.3%). Cumulants and CORENN results.

On a timescale over the first minutes we then observe that the hydrodynamic radius of the nanospheres slightly increases. Cumulants and CORENN results show essentially similar behavior over time, with cumulants values superior to the CORENN values of the hydrodynamic radius (Figure 5). The difference between the hydrodynamic radii obtained from cumulants and CORENN analysis can be attributed to the polydispersity of the sample.

In summary, at low serum concentrations the motion of nanoparticles can be reliably detected, despite the complexity of the composition of the added biofluid, while at high concentration the characteristics of the mouse-serum, including a distinction of rotation, translation and agglomeration are detected, in the presence of nanoparticles [10]. In both cases, the CORENN analysis of 3D-DLS data shows to be powerful for the DLS data analysis.

6. Conclusions

In summary, we have reviewed recent applications of 3D-ccDLS and CORENN methods for applications in optical biosensing and time-dependent monitoring of nanoparticle-biofluid interactions. The methods allow to differentiate rotational and translational motion, as well as agglomerate formation. Time dependent monitoring of protein-nanorod interactions and monitoring of serum together with nanoparticles allows to analyse either changes of hydrodynamic radii of the components of the particles or of monitoring nanoparticle motion in the presence of serum. The CORENN method shows robust results over a large range of concentrations of serum or protein concentrations, which is expected to contribute to the developments of new standards in dynamic light scattering.

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