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Dual channel detection of ultra low concentration of bacteria in real time by scanning fluorescence correlation spectroscopy

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Abstract

We describe a novel method to detect very low concentrations of bacteria in water. Our device consists of a portable horizontal geometry small confocal microscope with large pinhole and a holder for cylindrical cuvettes containing the sample. Two motors provide fast rotational and slow vertical motion of the cuvette so the device looks like a simplified flow cytometer without flow. To achieve high sensitivity, the design has two detection channels. Bacteria are stained by two different nucleic acid dyes and excited with two different lasers. Data are analyzed with a correlation filter based on particle passage pattern recognition. The passage of a particle through the illumination volume is compared with a Gaussian pattern in both channels. The width of the Gaussian correlates with the time of passage of the particle so one particle is counted when the algorithm finds a match with a Gaussian in both channels. The concentration of particles in the sample is deduced from the total number of coincident hits and the total volume scanned. This portable setup provides higher sensitivity, low-cost advantage, and it can have a wide use ranging from clinical applications to pollution monitors and water and air quality control.

Keywords: bacteria detection, ultra-low concentration, pattern recognition algorithm

(Some figures may appear in colour only in the online journal)

1. Introduction

The rapid detection of infectious disease agents is of primary importance for medical diagnosis, public health, food safety and environment monitoring. Traditional culture-based methods for the detection and enumeration of bacteria are limited to viable and cultivable cells and are typically time consuming due to the several days of incubation necessary for the visualization of colonies (Reynolds and Fricker 1999).

In past years, considerable effort has been devoted toward the development of culture-independent techniques for the

rapid and sensitive detection of pathogens, including non-viable or non-cultivable cells (Yang *et al* 2010). In recent years different techniques have been developed such as DNA microarray (Ahn and Walt 2005, Kim *et al* 2005, Cho and Tiedje 2001), PT-PCR (Malorny *et al* 2008, Kramer *et al* 2009, Mafu *et al* 2009), enzyme-linked immunosorbent assay (ELISA) (Ferguson *et al* 2000, Tamminen *et al* 2004), surface plasmon resonance (SPR) (Maalouf *et al* 2007, Oh *et al* 2005, Dudak and Boyaci 2009), magnetic nanoparticle-based immunoassay (Malorny *et al* 2008, Gu *et al* 2003), various formats of biosensors (Gehring *et al* 1998, Liao and Ho 2009, Lu *et al* 2009, Wong *et al* 2002) and flow cytometry (Busam

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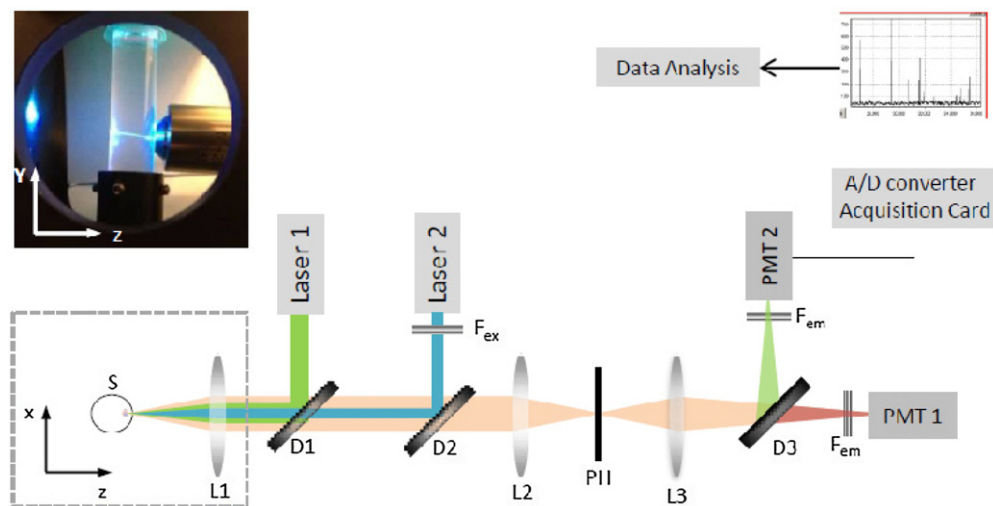


Figure 1. Schematic diagram of the instrument. Excitation light from the laser sources (Laser1 and Laser2) is combined through the dichroic mirrors (D1 and D2) and focused on the sample (S) through an objective lens (L1) (see photo). Emission light collected from the same objective and transmitted through the dichroic filters is focused via a lens (L2) into a confocal pinhole (PH). The light beam is further collimated by another lens (L3) toward the detection unit. A dichroic filter (D3) splits the emission beam before reaching the emission filters (F_{em}) placed in front of the two photomultiplier tubes (PMT1 and PMT2). The analog signals from the PMTs are converted and acquired through a card on a computer for the data analysis.

et al 2007, Hahn *et al* 2008, Hibi *et al* 2007, Qin *et al* 2008, Moragues *et al* 2004, Ferrari *et al* 2004).

Among these techniques, flow cytometry has emerged as a powerful analytical tool in the hand of molecular biologists and toxicologists for quantizing bacterial amounts in food or water samples, assessing cell surface antigens with immunofluorescence, analyzing cell cycles and testing cell viability (Hahn *et al* 2008, Holmes *et al* 2001, June and Moore 2004, Steen 2000). Compared to epifluorescence microscopy, which is time consuming and requires highly trained personnel, flow cytometry provides a faster and automatic way to detect bacteria labeled with fluorescent markers such as viability substrates, antibodies and nucleic acid probes (Reynolds and Fricker 1999).

In flow cytometry, micro-organisms are detected and enumerated as they move in a fluid stream through an optical detection region where the particles interact one at a time. Measurements on cells are made by exciting them, most commonly with a laser, and an optical sensor measures the light scattered or emitted by the cells. Labeled probes and fluorescent dyes aid the detection of a variety of cell types and cell components. Flow cytometry has proved to be an efficient and rapid method, it is a powerful and versatile technique that can be applied to many biological problems, but there are limitations and drawbacks to this technology.

The measured cell concentrations are typically on the order of 10^4 – 10^6 particles/ml unless the sample is pre-concentrated. The sample needs to be carefully filtered to prevent obstructions of the flow. Other limitations are portability and relatively high cost and for these reasons alternative methods have been proposed (Sakamoto *et al* 2007).

The purpose of this work is to present an alternative to optical flow cytometry with a portable, high-sensitive, dual channel fluorescence detection setup. The device is based on the principles of scanning fluorescence correlation

spectroscopy (FCS), a technique that has been extensively applied for detecting the motion of single molecules and cells (Berland *et al* 1996). The setup is similar to a small confocal microscope with the observation volume located inside a cylindrical cuvette containing the sample. Two motors generate fast rotation and slower vertical motion of the cuvette so that particles are detected as they cross the observation volume. In this sense the device looks like a simplified flow cytometer without flow. The device is able to measure the concentrations of fluorescent particles in water even at extremely low concentration (few particles/ml) and in very short sample scanning periods (about 1 min). Our method does not require sample filtration and does not result in loss or degradation of the sample during the analysis. Another advantage of our device is that it is highly portable and easily operated by technicians without extensive training. Samples could be held in sealed cuvettes and the measurement produces no contamination of the instrument or of the environment. Therefore the method could be applied to highly toxic biomaterials or samples that need to be preserved for further analysis.

2. Experimental section

2.1. Instrument design

The instrument design is largely based on an original version of the device which had only a single detection channel (Gratton *et al* 2009). The device has been developed into a two-channel setup to allow simultaneous red and green fluorescence detection for the rapid quantification of the total number of bacterial cells. Figure 1 shows the schematic diagram of the setup.

The apparatus consists of a small microscope that has a horizontal geometry and a mechanical part that holds a

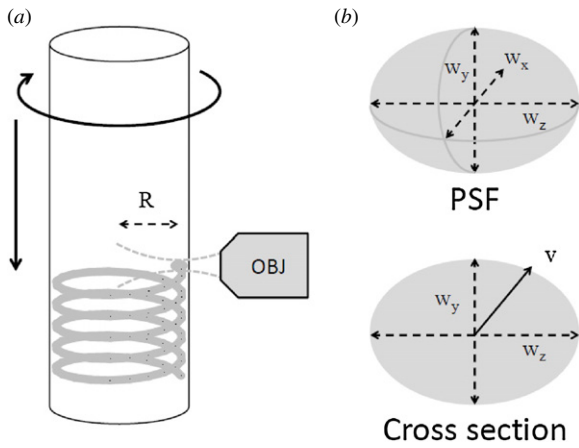


Figure 2. Generation of the scanned volume in the setup. (a) As the cuvette is rotated and translated, the observation volume, kept at distance R from the rotation axis, scans a helicoid pattern. (b) The static observation volume is estimated using a 3D Gaussian PSF with dimensions w_x , w_y and w_z . The scanned volume per unit time is estimated multiplying the cross-section and the velocity v in the x direction.

cylindrical cuvette whose diameter is 1 cm. Two motors provide rotational and vertical up-and-down motion of the cuvette. The software allows the rotational speed to be varied in the range 10–1100 rpm and the vertical speed in the range 1–15 mm s⁻¹. The vertical and rotational motions are produced respectively by the Linear Actuator series3500-Haydon (Haydon Kerk, Waterbury, CT, USA) and a VEXTA stepping motor model PK233PB (Oriental Motor USA Corp., Torrance, CA, USA). These motors are connected to a stage holding the transparent cuvette containing the sample. The excitation light generated by two lasers is focused at the volume of observation. The excitation focus is positioned inside the cuvette and relatively close to the wall of the cuvette, at a distance of about 1 mm (figure 2(a)). This distance can be adjusted so that detection of particles and analysis could be done even in highly scattering media.

The excitation sources are a diode laser emitting at 469 nm (ISS, Champaign, IL, USA) and a laser pointer emitting at 532 nm (Aquaplan, Busto Arsizio, Italy). When a particle is in the volume of excitation, it fluoresces. Use of a confocal microscope in combination with a simple mechanical motion of the sample container in front of the objective provides a means for transporting the sample containing particles through an observation region without requiring a complex optical system comprising moveable optical components, such as translating optical sources, mirrors or photodetectors. The excitation light from the two lasers is combined in one path through a set of dichroic filters ZT532nbdc and Z470rdc (Chroma Technology Corporation, Rockingham, VT, USA) and directed through a 20 × 0.4 NA air objective (Newport, Newport Beach, CA, USA) to the same volume of excitation.

Fluorescence emitted from the sample is collected by the same objective, transmitted through the set of dichroic filters, focused by a lens into a large pinhole (diameter = 2 mm) and then collimated by a second lens to the detectors. A dichroic beam splitter T550lpxr-25mmNR (Chroma Technology Corporation, Rockingham, VT, USA)

separates the emission into two light paths prior to its detection by two photomultiplier tubes (PMT) (Hamamatsu HC120-08, Hamamatsu, Japan). Two emission filters (FF01-HQ 500/24-25 and LP5600) (Semrock, Rochester, NY, USA) are located in front of each PMT. The signal from the PMT is sent to the analog to digital converter (ADC) and to the acquisition card (IOtech, Cleveland, OH, USA). The sampling frequency was set to 100 000 Hz, corresponding to a time resolution of 10 μs.

The confocal aperture provided between the sample and photodetectors ensures that only fluorescence originating from the observation volume is detected by the PMTs. The photodetectors measure the fluorescence signal originating from the observation volume and generate a temporal profile of the fluorescence. Given the typical size of the observation volume ($w \sim 10 \mu\text{m}$), it is clear that diffusion alone is not sufficient to generate fluctuations on a short time scale. For instance, a particle with a diffusion coefficient $D \sim 0.1 \mu\text{m}^2 \text{s}^{-1}$ will take hundreds of seconds to diffuse across the detection volume. The rotational motion of the cuvette is used to transport the particles across the observation volume and generate the temporal fluctuations in the intensity on a much faster time scale (typically hundreds of microseconds). The vertical motion ensures that we explore different regions inside the cuvette by scanning along the y direction.

The total volume analyzed is proportional to the total sample scanning period and depends on the rotational velocity of the cuvette and size of the observation volume. In our device the total volume explored during the scanning period can be in principle as high as 1 ml min⁻¹ by proper adjustment of the setup parameters.

The estimated volume of the point spread function (PSF) in the confocal microscope can be defined by the expression

$$V_{\text{PSF}} = \left(\frac{\pi}{2}\right)^{3/2} w_x w_y w_z, \quad (1)$$

where we have assumed a Gaussian shape of the PSF and w_x , w_y and w_z are the dimensions of the PSF along the x , y and z directions, respectively (see figure 2).

The observation volume is scanned at a velocity v given by

$$v = 2\pi fR, \quad (2)$$

where R is the distance of the observation volume from the rotation axis (equal to about half the cuvette size) and f is the rotational frequency of the cuvette. The total length L of the trajectory for a movement involving both rotation and vertical inversion can be calculated using the expression

$$L = 2\pi fRt, \quad (3)$$

where t is the sample scanning period.

The total volume (V) explored during a sample scanning period is provided by the expression

$$V = \frac{\pi}{2} w_y w_z L, \quad (4)$$

where we have taken into account the area of the cross section ($A = \pi w_y w_z / 2$) of the observation volume orthogonal to the direction of motion. In a typical configuration of our apparatus, for a scanning period of 60 s, a cuvette diameter of 1 cm, a frequency of five revolutions per second, and dimensions of the observation volume $w_y \sim 10 \mu\text{m}$ and $w_z \sim 500 \mu\text{m}$, we obtain a total volume of observation of about $V \sim 0.1 \text{ ml min}^{-1}$.

2.2. From molecular FCS to large volume scanning

The basis for single molecule detection in FCS is that most of the time the observation volume is free of fluorescent particles and only noise is recorded, but when a particle moves into the observation volume it emits a burst of fluorescence photons. These bursts can be identified by autocorrelation of the time series. In the FCS method, one extracts information about the sample by analyzing the fluorescence signal fluctuations arising from a microscopic sub-volume. Random fluctuations of fluorescence intensity from single molecules provide information on important molecular properties such as translational and rotational diffusion or chemical kinetics (Magde *et al* 1974).

The signal $I(t)$ generated by particles traversing the observation volume will fluctuate around a mean value $\langle I(t) \rangle$. The autocorrelation function (ACF) is defined as

$$G(\tau) = \frac{\langle \delta I(t) \cdot \delta I(t + \tau) \rangle}{\langle I \rangle^2}. \quad (5)$$

The functional form of $G(\tau)$ depends on the specific kind of motion of the particles. The autocorrelation curve for the particles in our system is expected to contain both a flow and a diffusion component:

$$G(\tau) = G(0) e^{-\frac{v^2 \tau^2}{w_0^2}} \frac{1}{1 + \frac{4D\tau}{w_0^2}}. \quad (6)$$

In this equation the first term is due to a flow along the x direction characterized by a velocity v given by equation (2) and a second term is due to random motion with diffusion constant D . We have ignored the diffusion component along z , assuming that $w_z^2 \gg w_0^2$. For large enough values of the rotational frequency, the flow term is predominant over the diffusion term, which becomes relevant only for very slow motion of the cuvette.

In a typical FCS experiment, the amplitude of the fluctuations $G(0)$ is inversely proportional to the number of particles in the observation volume, and therefore to the concentration of particles in the sample:

$$G(0) = \frac{\gamma}{N}, \quad (7)$$

where γ is a dimensionless shape factor which depends only on the functional form of the excitation profile (Berland *et al* 1996). For our setup, correlated noise inherent in the mechanical scanning device was an obstacle and prevented us from using the ACF for particle-counting purposes (i.e. using equation (7)). Indeed, in the limit of very low concentrations, where fluctuations due to the particles may be only a few, the value of $G(0)$ is affected by fluctuations due to mechanical scanning, so that conventional FCS cannot be applied to very dilute samples, i.e. samples with concentration of fluorescent particles well below the picomolar range. In our sample the particles exhibit very bright fluorescence, much brighter than normally dealt with in conventional FCS experiments, and the presence of particles in the laser focus was very rare.

To apply the principles of scanning FCS in our setup, we employ a different data analysis technique that would provide a much better way of discriminating the passage of a particle from the noises in the system. Examples of noise in the

system are the intensity fluctuations that arise from mechanical scanning or from fluorescent impurities in the cuvette glass. In this technique we make specific use of the shape of the burst and we exploit the relatively large brightness of the particle in relation to the quasi constant fluorescence background. If we know the shape of the fluorescence burst we can analyze the fluorescence for the presence of specific burst patterns to identify the passage of a particle in the volume of excitation. Combination of the large volume of exploration, the regular motion of the sample and the pattern recognition algorithm, results in the sensitive detection of a small number of particles in a relatively large volume.

2.3. Data acquisition and processing

In order to extract the measurement of the concentration and/or brightness of the particles in the sample, the temporal profile generated by the photodetectors has been analyzed with a pattern recognition filter implemented in the software SimFCS (Laboratory for Fluorescence Dynamics, Irvine, CA, available at www.lfd.uci.edu/globals/). This pattern recognition program analyzes the temporal profile of the digitized signal from the two detectors. The pattern recognition algorithm matches features in the temporal profile to a predetermined pattern that is characteristic of the time-dependent fluorescence intensity of particles passing through the observation volume.

The predetermined patterns used in the recognition algorithm include distributions (e.g. Gaussian) of intensities as a function of time, which may be determined empirically taking into account the shape of the PSF of the optical setup, the speed of the particle and the rate of signal sampling. Discrete particle detection events are identified by establishing a match between the amplitude and shape of a feature in the temporal profile and the predetermined pattern. The number of detected particles is determined by calculating the number of predetermined patterns matched to features in the temporal profile for a given sample scanning period. The operational principle of the pattern recognition algorithm is shown schematically in figure 3.

The algorithm proceeds first by subtracting the background level, which generally is not constant due to the noise inherent in the mechanical scanning. The local background level is determined performing a moving average of the intensity signal which removes only the high frequency fluctuations but preserves the low frequency variations in the baseline (figure 3(a)). After background subtraction, the filter algorithm scans the signal performing a least-square calculation at each point of the file representing the fluorescence detected as a function of time. The program calculates the best amplitude A of the filter that minimizes the local chi square, the intensity amplitude not being restricted. The chi square is calculated by weighting the difference between the raw data and the fit to the predetermined pattern with the sum of (1) a noise level specified by the user σ_n and (2) an estimate of the standard deviation σ_d that accounts for

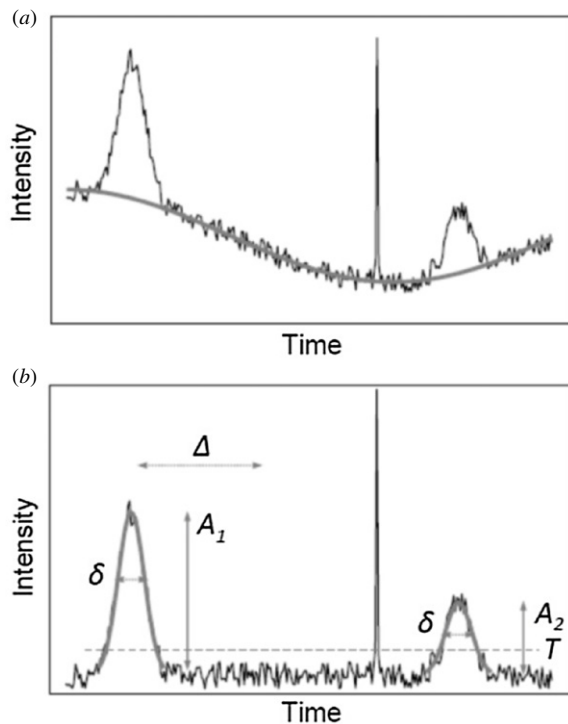


Figure 3. Principle of the pattern recognition algorithm. (a) The slowly varying background level is determined performing a moving average of the intensity signal and subtracted. (b) At each point of the time series a fit to a pattern of predetermined shape is performed. When the chi square of the fit is below a preset value and the amplitude of the fitted pattern is above a given threshold T , a hit is counted. After a hit, a number of points Δ must be skipped to ensure a peak is not counted twice. Then analysis of the time series continues until another hit is detected. Detected peaks share the same width δ but can have different amplitudes A_1 and A_2 . A spike of large amplitude as shown in the figure is not counted because it has a shape different from the characteristic shape of the particles.

the square root of the number of counts at the particular point of the file

$$\chi^2 = \frac{1}{N-2} \sum_i \frac{(I_{\text{exp},i} - I_{\text{pattern},i})^2}{\sigma_n^2 + \sigma_{d,i}^2}. \quad (8)$$

Whenever the chi square is below a given value and the amplitude is above a given threshold value T , one hit is counted in that channel (figure 3(b)). After a hit has been detected, the software skips a given number of points Δ in the file in order to count each peak only once, then it scans again until the next match. Spikes are rejected because they are too narrow with respect to the filter, even if their intensity is above the threshold T . Note that the shape of the filter is kept constant while the amplitude value can be different between different hits.

For the enumeration of coincidence hits, the algorithm examines one channel first and then the trace of the other channel. If there is a peak at the same time where a hit has been found in the first channel, one coincident hit is counted. The threshold value can be set independently for the peak amplitude in both channels, in order to reject the noise in each channel or for instance to discard false coincidences due to bleed-through effects.

One advantage of using a predetermined pattern is that it is not dependent on particle composition or the composition

of the medium the particles are dispersed in. The use of pattern recognition in a two-channel setup is advantageous in preventing detection of false hits originating from noise or spikes since only particles generating a cross-correlated intensity profile are detected.

Concentration for particles detected in each channel or in both channels is extracted from the analyzed temporal profile by dividing the number of matches by the volume V of the sample analyzed during a selected sample scanning period, which can be accurately calculated with knowledge of the size of the observation volume, rate of movement of the cuvette (rotational) and the duration of the sample scanning period, using equations (3) and (4). The size of the observation volume is controlled by the size of the confocal aperture employed in the confocal microscope.

2.4. Reagent and chemicals

The membrane-permeable nucleic acid stains Syto9 and Syto25 were purchased from Molecular Probes (Eugene, OR). They have different spectral characteristics and both stain all cells. Syto9 excitation and emission peaks are $\lambda_{\text{ex}} = 485$ nm and $\lambda_{\text{em}} = 498$ nm, respectively, whereas Syto25 excitation and emission peaks are $\lambda_{\text{ex}} = 521$ nm and $\lambda_{\text{em}} = 556$ nm, respectively.

E. coli MG1655 bacteria were grown overnight in 5 ml of Luria Broth (LB) at a temperature of 37 °C in an incubator shaker I2400 (New Brunswick Scientific Inc., Enfield, CT, USA). A PBS buffer solution (Life Technologies, Carlsbad, CA, USA) was used for sample dilutions. Petri plates were used to determine the concentration of bacteria by the conventional plate-counting method. The concentration of bacteria is expressed in colony-forming units (CFU). This is a measure of the number of viable cells in a culture, capable of producing new colonies when seeded on the plates. Petri plates were prepared using LB powder and agar powder.

Glass cuvettes of 1 cm in diameter from Abbot (Abbott Park, IL, USA) were used to hold the sample. Yellow-green and Nile-red fluorescent microspheres (1 μm size) were purchased from Invitrogen (Carlsbad, CA, USA) to calibrate the instrument. Suspensions of fluorescent beads in water at different concentrations were prepared according to manufacturer's instructions. Suspensions of fluorescent beads in highly scattering media were prepared using a solution of Intralipid (Intralipid 20%, Fresenius Kabi, Uppsala, Sweden) at a final concentration of 0.2%.

2.5. Fluorescence staining of bacteria cells

All experiments were performed using the *E. coli* strain MG1655. Cells were rod shaped and between 0.7 and 1.5 μm long. A glycerol stock was used to prepare our samples. The exponential bacteria growth was performed overnight at 37 °C. A bacteria suspension was prepared at a starting concentration corresponding to a value $OD_{600} = 0.4$, as measured by a spectrophotometer (Perkin-Elmer Lambda 40, Waltham, MA, USA). The bacteria were centrifuged for 10 min at 5000 rpm and then resuspended in phosphate buffered saline (PBS). An aliquot of the bacteria suspension (100 μl) was then added

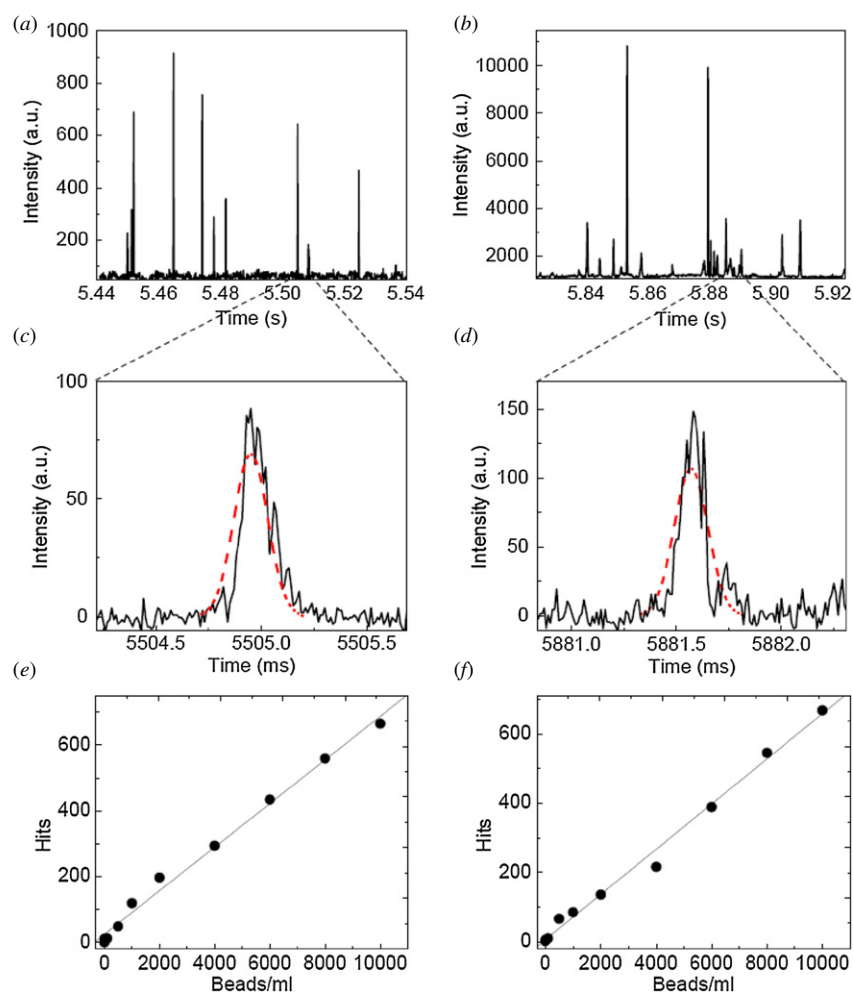


Figure 4. Calibration with fluorescent spheres. Typical intensity traces from NR (a, c) and YG (b, d) fluorescent beads. In the expanded regions (c, d) the intensity profile corresponding to the passage of a particle (solid) is compared with a Gaussian distribution of predetermined width (dashed). The number of hits registered during one scan is reported as a function of several concentrations of NR beads in channel 1 (e) (linear fit: $y = 0.066x + 24.3$, $R^2 = 0.996$) and YG beads in channel 2 (f) (linear fit: $y = 0.065x + 6.9$, $R^2 = 0.992$).

to 3ml of a PBS solution of the dyes Syto9 and Syto25 at a concentration of $1.5 \mu\text{M}$ each. The sample was vortexed and incubated for 30 min in the dark at room temperature ($\sim 25^\circ\text{C}$) before measurement. We chose this incubation time because the number of total Syto-stained cells is not expected to increase after 30 min incubation (Khan *et al* 2010).

After measurement with our setup, 1 ml of a dilution 1:10, 1:100, 1:1000, 1:10000 of our solution was plated on antibiotic free LB agar dishes. After 24 h the colonies on the plates were counted and the concentration was determined. The concentration derived from the counting of the colonies was then compared with the one obtained from the measurement with our device using the same sample.

3. Results and discussion

Before performing dual channel measurements to detect low concentrations of bacteria, the detection capabilities of the device on both channels independently were tested using fluorescent spheres of different colors. Suspensions of beads were prepared at different concentrations ranging from 10^5 beads/ml down to 10 beads/ml. To evaluate the

effectiveness of the pattern recognition algorithms in detecting particles of size comparable to that of the bacteria, instrument calibration was performed using $1 \mu\text{m}$ diameter yellow-green (YG) and Nile-red (NR) fluorescent beads. The NR and YG beads have been used for calibration of the red (Ch1) and green (Ch2) channels, respectively. NR beads were excited at 532 nm whereas YG beads were excited at 469 nm.

Figure 4 shows a typical time trace obtained from single channel measurements on NR (a) and YG (b) fluorescent beads at a concentration of 10^4 ml^{-1} . The total scanning time was 1 min and the cuvette was rotated at a speed of 400 rpm while translated up and down at a speed of 10 mm s^{-1} . The bursts above the background level are due to particles passing through the observation volume and generating a characteristic intensity profile. In figures 4(c) and (d) we can show a detail of the temporal profiles in a region corresponding to a hit. The intensity profile is matched by the software to a predetermined Gaussian filter. The software allows us to display individual matches.

Figures 4(e) and (f) show the results of a concentration-dilution study obtained with the present device and employing pattern recognition data analysis. There is excellent linear

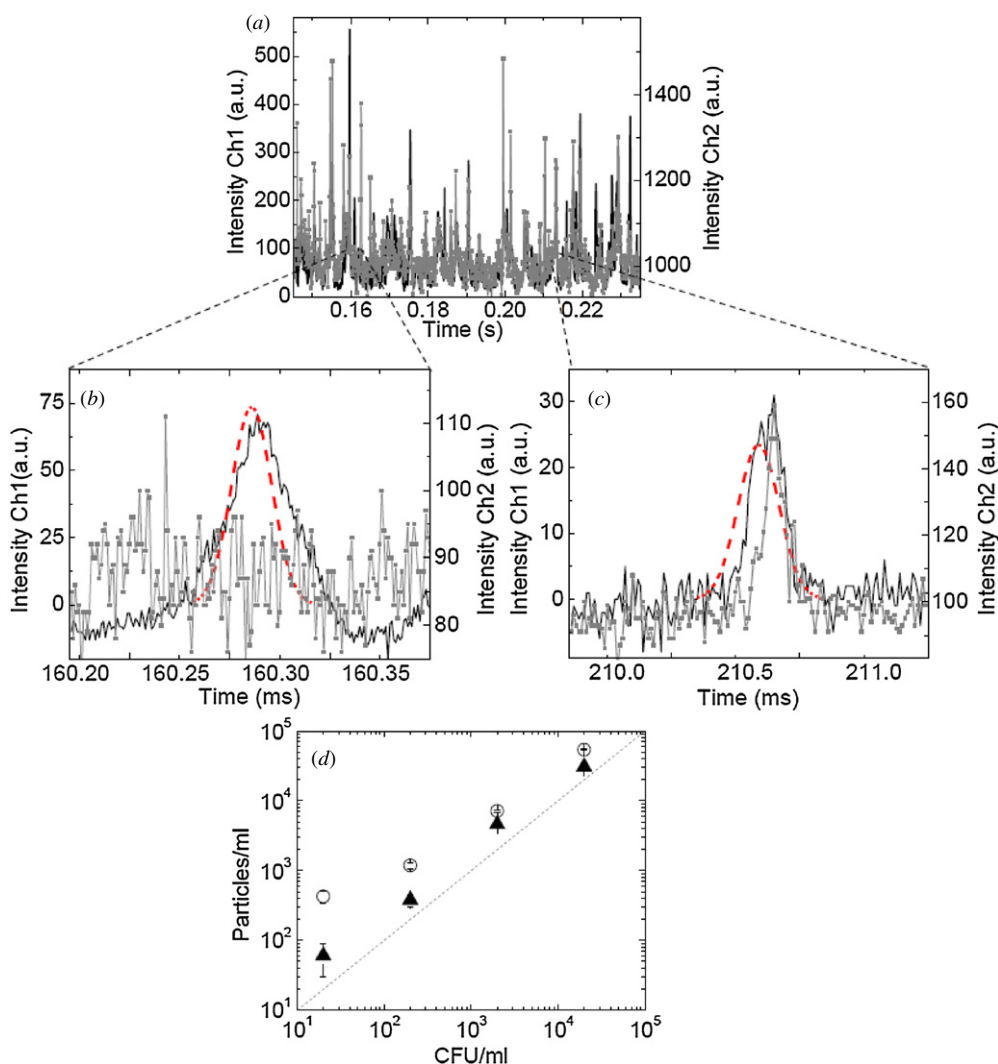


Figure 5. Dual channel detection of *E. coli* bacteria stained with Syto25 and Syto9. (a) Intensity traces generated in channel 1 (solid) and channel 2 (solid with squares) by the passage of the fluorescent bacteria through the observation volume. In the expanded regions are reported the examples of a bacterium detected only in one channel (b) and a bacterium detected in both channels (c). (d) Concentration of fluorescent bacteria as determined by single (circles) or dual (triangles) channel detection as a function of the concentration of viable bacteria determined by the colony counting method (CFU = colony forming units).

correlation between the total number of events as detected by the digital correlation filter and the actual concentration of fluorospheres in the sample for both channels. From the slope of the linear fit we get an effective scanned volume $V = 0.07$ ml.

After calibration with beads, the instrument was used to detect labeled bacteria at low concentration. *E. coli* bacteria were fluorescently tagged with the nucleic acid probes Syto25 and Syto9 to be detected simultaneously in both channels. The actual concentration of bacteria in the sample was estimated independently through the standard plate counting method. The measurements were performed employing 1 min scanning time, rotational frequency of 400 rpm and a vertical travel speed of 10 mm s^{-1} . Both channels were analyzed with the filter algorithm and the occurrence of coincident hits was quantified.

Representative data for detection of *E. coli* in both channels are shown in figure 5(a). Bursts generated from bacteria crossing the observation volume are easily distinguished from the background in each channel. The

software extracts the number of hits in each channel and the cross-correlated hits by examining the intensity trace in both channels using a predetermined Gaussian filter. Not every hit detected in one channel corresponds to a cross-correlated hit in the other channel. One example of a non-coincident hit is reported in figure 5(b), whereas figure 5(c) shows a detail of the overlapping temporal profiles in a region corresponding to a coincident hit.

The number of detected particles has been converted into a value of concentration using the calibration factor obtained previously with the fluorescent spheres. As we decrease the concentration of bacteria in the suspension, the instrument detects a linearly decreasing concentration of particles (figure 5(d)). The cross-correlated detection method yields values for the concentration of bacteria close to the values of colony-forming units (CFU), especially at lower concentration values where the false hits due to noise become more relevant. The concentration of viable bacteria in the sample has been determined for the highest concentration

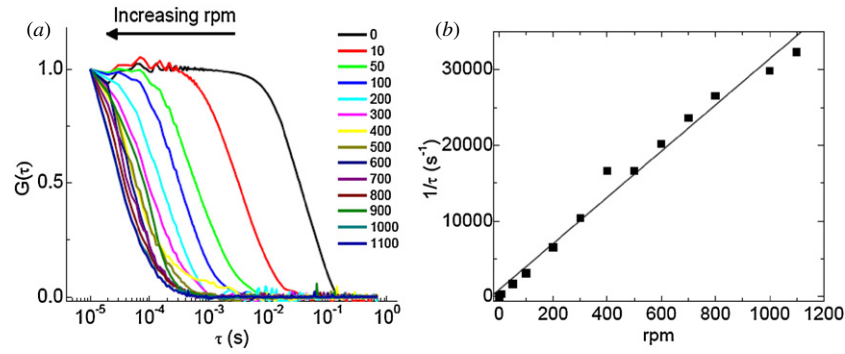


Figure 6. Effect of the cuvette rotation on the characteristic time of transit through the observation volume. (a) The normalized intensity ACF shrinks toward shorter time as the value of the rotational frequency is increased. (b) The reciprocal of the characteristic time τ derived from the ACF is plotted as a function of the rpm. The straight line is a linear fit to the data.

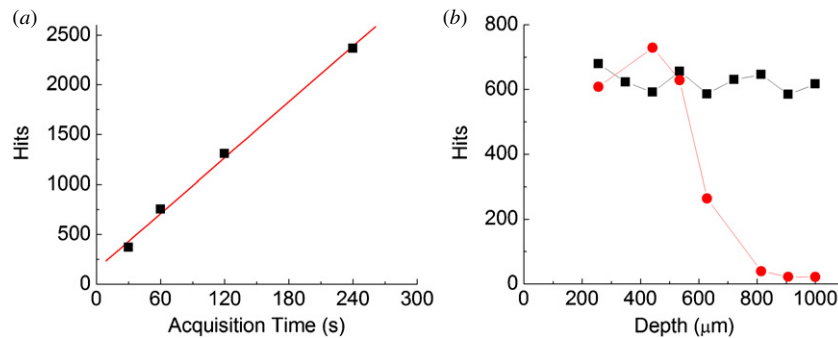


Figure 7. (a) For a given concentration of fluorescent microspheres (10000/ml in the example), the number of particles detected in the scanned volume (solid squares) increases linearly with the acquisition time. (b) Particle detection at different depths of the focal point with respect to the wall of the cuvette. For increasing depth the number of particles detectable in a highly scattering sample (red dots) is strongly reduced with respect to a transparent sample (black squares).

(20 000 bacteria/ml) using the standard plate counting method. We can see good linearity down to 20 CFU/ml and in particular at low concentrations. The staining protocol does not discriminate between dead and viable bacteria, so it is possible that we detect a higher concentration of particles with respect to the concentration of viable bacteria in the sample as determined from the enumeration of the colonies.

The width of the predetermined Gaussian distribution used by the filter algorithm should match the width of the intensity fluctuations generated by the passage of the particles. If we change the rotational frequency of the cuvettes we also change the velocity of the particles that cross the observation volume. In figure 6 we plotted the normalized autocorrelation curves of the intensity traces obtained measuring the same sample (20 000 bacteria/ml) at different values of the rotational speed of the cuvette. As we increase the rpm, the autocorrelation curve decays at a shorter time. Plotting the inverse of the characteristic time τ (determined as the value at which $G(\tau) = 1/2$) as a function of the rotational frequency confirms the linearity between $1/\tau$ and f , as expected from equations (2) and (6) indicating that the fluctuations are mainly due to the ‘flow’ generated by the rotational motion and that the diffusion term can be neglected. From a linear fit of the data we find an effective size of the observation volume along the x direction, $w_0 = 16 \mu\text{m}$. A smaller size of the PSF (down to $1 \mu\text{m}$) could be obtained in principle by expanding the beam before the back aperture of

the objective. When the cuvette is not rotated we get a residual correlation ($\tau = 33 \text{ ms}$) which could be explained as arising from rare fluctuations due to bacteria crossing the observation volume while diffusing or settling down.

The size of the PSF or observation volume V_{PSF} can be modified (for instance using a different objective, varying the size of the beam at the back aperture of the objective or changing the size of the pinhole) in order to optimize the particle detection at different ranges of concentration. At very low concentrations, too small a PSF could be inconvenient due to the longer time needed to explore a relatively large portion V of the sample. In general, to increase the number of particles detected at any given concentration C during a given amount of time, we can increase the size of the PSF up to a point where saturation is reached because the condition of observing one particle at the time, $C \ll 1/V_{\text{PSF}}$, is no longer fulfilled. On the other hand, increasing the size of the confocal volume can limit the capability of detecting the particles due to an increasing level of background signal generated, for instance, by the presence of free dye in the sample. In order to discriminate the particles from the background and to achieve the same S/N ratio, the brightness of the particles needs to be larger for larger observation volumes. Therefore, for particles of given brightness and concentration, there must be a compromise between the increase in S/N ratio and increase in measuring speed that can be optimized by changing the size of the confocal volume.

Another parameter that can be adjusted in order to increase the number of detected particles is the duration of the measurement. As shown in figure 7(a), the number of hits increases linearly with the acquisition time. Longer acquisition times are needed, especially for samples of low concentration or for highly scattering samples in which detection can be more difficult. In the latter case particle detection can be optimized by changing the position of the focal spot inside the cuvette. For particles immersed in a scattering medium (water solution of Intralipid 0.2%) the number of hits decreases with increasing depth of the focal spot into the sample (figure 7(b)) whereas the number of hits in a transparent medium does not vary significantly. Thus, particle detection in highly scattering media is still possible provided that the focus is positioned closer to the wall of the cuvette.

4. Conclusions

In this work we described a device capable of measuring very low concentrations of bacteria labeled with fluorescent probes. The rotational motion of a cuvette is used to make particles flow through a small observation volume. The small observation volume is obtained using laser excitation and confocal optics. The detection of the particles can be performed in one or two channels by analyzing the intensity profile generated by the passage of the particles.

We showed that through the use of a pattern recognition algorithm it is possible to filter and count the intensity fluctuations due to the bacteria passing through the confocal volume. The use of coincidence detection in two channels provides an additional filter to count only the fluctuations generated by bacteria labeled with both fluorescent probes.

The system is not limited to detection of bacteria and in principle can be applied to the detection of other kinds of pathogens. In this sense we showed that the size of the observation volume can be adjusted in order to optimize the performance of the device for samples of different concentration and brightness. Given its portability, high sensitivity, speed and low cost, we suggest the use of this device for clinical applications and for water and air quality control.

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