Intermittency of CsPbBr$_3$ perovskite quantum dots analyzed by an unbiased statistical analysis

Isabelle M. Palstra, Ilse Maillette de Buy Wenniger, Biplab K. Patra, Erik C. Garnett, and A. Femius Koenderink

†Institute of Physics, University of Amsterdam, Science Park 904, 1098 XH Amsterdam, The Netherlands
‡Center for Nanophotonics, AMOLF, Science Park 104, 1098 XG, Amsterdam The Netherlands

E-mail: f.koenderink@amolf.nl

Abstract

We analyze intermittency in intensity and fluorescence lifetime of CsPbBr$_3$ perovskite quantum dots by applying unbiased Bayesian inference analysis methods. We apply changepoint analysis (CPA) and a Bayesian state clustering algorithm to determine the timing of switching events and the number of states between which switching occurs in a statistically unbiased manner, which we have benchmarked particularly to apply to highly multistate emitters. We conclude that perovskite quantum dots display a plethora of gray states in which brightness broadly speaking correlates inversely with decay rate, confirming the multiple recombination centers model. We leverage the CPA partitioning analysis to examine aging and memory effects. We find that dots tend to return to the bright state before jumping to a dim state, and that when choosing a dim state they tend to explore the entire set of states available.
Introduction

Cesium lead halide perovskite nanocrystals, introduced in a seminal paper by Protesescu et al.\textsuperscript{1} have emerged as highly attractive quantum dots, with advantageous properties in comparison to traditional colloidal II-VI semiconductor quantum dots. These include very large photon absorption cross sections,\textsuperscript{2} a wide degree of tunability by both size and halide (Br, I, Cl) composition,\textsuperscript{1} and reportedly a very high luminescence quantum yield without the need of protecting the nanodot core with epitaxial shells, as is required for CdSe quantum dots.\textsuperscript{3,4} Furthermore, inorganic halide perovskite materials generally show an exceptionally high tolerance to defects.\textsuperscript{5} Owing to these properties perovskite nanocrystals are intensively pursued as solar cell materials,\textsuperscript{6} as emitters for LEDs, display technologies and lasers,\textsuperscript{7–9} and could be interesting as single photon sources. For the purpose of single photon sources, emitters need to satisfy a variety of requirements beyond brightness, tuneability and high quantum efficiency, which includes single-photon purity, tight constraints on inhomogeneous spectral broadening, and stability in spectrum, decay rate and intensity.\textsuperscript{10}

Perovskite nanocrystals unfortunately follow the almost universally valid rule that solid-state single emitters at room temperature show intermittency.\textsuperscript{3,11–16} In the field of II-VI quantum dots, intermittency has been studied for over two decades, with the aim of identifying the nature of the usually two or three distinct bright, dark, and gray states, and the mechanism by which switching occurs, by analysis of the apparently discrete switching events between dark and bright states,\textsuperscript{17,18} and concomitant jumps in spectrum and lifetime. For instance, for II-VI quantum dots a popular model (reviewed in Ref.\textsuperscript{19}) is the charging/discharging model whereby quantum dots turn from bright to dark upon acquiring a single charge, with typically power-law distributed residence times for on and off states determined by the mechanism by which charges are exchanged with the environment.\textsuperscript{17,20–22} For perovskite nanocrystals several groups studied intermittency\textsuperscript{3,11–16} and found quite different physics. A set of works observe that perovskite quantum dots do not show bimodal behavior, like II-VI quantum dots do, but instead a continuous distribution of states between
which they switch.\textsuperscript{11,12,14} These observations are consistent with the so-called ‘multiple recombination centers (MRC) model’ proposed for CsPbBr\textsubscript{3} dots by Li et al.,\textsuperscript{12} wherein a single quantum dot does not immediately switch from bright to dark on introduction of a single charge or defect, but instead intrinsically has a multitude of recombination centers that may be activated to add nonradiative decay channels. Activation of individual recombination centers then shows as switching events, but with a wide distribution over intensity and rate. Other groups, in contrast, have analyzed intermittency on basis of changepoint analysis and cluster analysis, which are Bayesian inference tools for the unbiased estimate of the number of states. These reports claim that just of order 2-4 states are involved instead of a continuum. Finally, a recent study points at memory effects in intermittency, visible in that work as correlations between subsequent dwell times in the brightest state.\textsuperscript{16}

Intermittency analysis is a field known to be fraught by statistical bias in analysis methods,\textsuperscript{20} primarily due to binning of data prior to analysis. This is a recognized problem already for interpreting data from bimodal dots. These artefacts may be even more severe for multilevel dots. In this work we report a study of cesium-lead-bromide nanocrystal intermittency, analyzing the photon statistics of a large number of dots using unbiased Bayesian statistics analysis tools, tracing brightness and fluorescence lifetimes simultaneously, and screening for memory effects. These Bayesian statistics methods were first developed by Watkins and Yang\textsuperscript{23–25} and have since been applied in a small set of papers to two/few-level II-VI dots, and in one recent work to CsPbBr\textsubscript{3} dots.\textsuperscript{13} Our implementation is through a freely available Python-based analysis toolbox,\textsuperscript{26,27} which we have specifically benchmarked by Monte Carlo methods for application to highly multi-state, instead of bi-modal, systems. In this work, a first main purpose is to obtain statistically unbiased estimates, or at least lower bounds, for the number of dark/gray states of perovskite quantum dots from a large number of single dot measurements. Our conclusions solidly support the MRC model with a single well-defined bright state and a continuum — or at least over 10 — gray/darker states, in agreement with,\textsuperscript{11,12,14} but not Ref.\textsuperscript{13} Next, our purpose is to screen for memory effects in

residence times, intensity levels and decay rate sequences in data that has been separated in segments by unbiased changepoint analysis, thereby extending Ref. 16 which did not leverage the benefit of CPA analysis. We find no evidence for memory in residence times, but do find that a substantial fraction of dots tend to switch back and forth repeatedly between the quite uniquely defined bright state and the band of gray states, instead of jumping through all states in an uncorrelated random fashion.

Figure 1: Properties of a CsPbBr$_3$ quantum dot. (A) a SEM image showing a cluster of CsPbBr$_3$ quantum dots. The scale bar is 100 nm. A time trace of the (B) intensity and (C) lifetime of a typical qdot when split into bins of 10 ms (green). We find a single peak in both the intensities and lifetimes around 60 counts/ms and 0.05 ns$^{-1}$, respectively. For visualization purposes, we also show the photon events binned into 0.5 ms bins (purple). (D) The $g^2(\tau)$ of this qdot. The dots used in this analysis were selected for having $g^2(0) < 0.5 \cdot g^2(100\text{ns})$. (E) The spectrum of this qdot. We find a peak in the emission at 505 nm. (F) The decay trace of all the photon events combined. We have excluded an electronic artefact between 20 and 30 ns. We have a reasonable fit to a bi-exponential decay with rates of $\gamma_1$, $\gamma_2 = 0.43$, 0.03 ns$^{-1}$, respectively. (G) The FDID diagram of this dot. We see a main peak at $I, \gamma = 0.7 \times 10^5$ cts/s and 0.09 ns$^{-1}$.
Experimental methods

To introduce our measurement protocol and the photophysics of the CsPbBr$_3$ dots at hand we first present in Figure 1 the typical behavior of a CsPbBr$_3$ quantum dot, as analyzed with the standard approach of plotting time binned data.

**Preparation of cesium oleate.** We load 0.814 g of Cs$_2$CO$_3$ into a 100 mL 3-neck flask along with 40 ml of octadecene (ODE) and 2.5 ml of oleic acid (OA) and dry this for 1 hour at 120 °C. This is then heated under an N$_2$ atmosphere to 150 °C until all Cs$_2$CO$_3$ has reacted with OA. To prepare for the next step, we preheat the resulting cesium oleate to 100 °C before injection. This is necessary as it precipitates out from ODE at room temperature.

**Synthesis of CsPbBr$_3$ nanocubes.** We load 0.188 mmol of PbBr$_2$ in 5 ml of ODE, 0.5 ml of oleylamine and 0.5 ml of OA into a three-neck round bottom flask and dry this under vacuum at 120 °C for an hour, after which the reaction atmosphere is made inert by flushing the flask with N$_2$. After complete solubilization of PbBr$_2$, the temperature is raised to 200 °C and 0.4 ml of the preheated cesium oleate is injected into the three-neck flask. After the injection, the color of the solution turns from colorless to greenish yellow indicating the formation of perovskite cubes. Then we lower the temperature to 160 °C and anneal the solution at that temperature for 10 min to get uniform size dispersion of the cubes. After that we cool down the solution using ice water bath for further use.

**Isolation and purification of CsPbBr$_3$ cubes.** After the synthesis, we centrifuge our solution twice to collect the cubes. First, we take 1 ml from the stock solution just after the synthesis and centrifuge at 8000 rpm for 20 min to collect all CsPbBr$_3$ particles from the solution. We discard the supernatant, gently wash the inner wall of the tube using tissue paper and add 2 ml of toluene to disperse the CsPbBr$_3$ solid. The second step of centrifugation is run at 2000 rpm for 5 min to get rid of all the particles that are too large. In the supernatant, we have 2 ml of toluene containing CsPbBr$_3$ nanocubes having a size distribution around 10-15 nm. As the scanning electron micrograph in Figure 1(A) shows our quantum dots are essentially cubic in shape.
Before the measurement, about 400 µL of the solution is spin coated at 1000 rpm on glass coverslips that had been cleaned in a base piranha solution. In order to protect the quantum dots from moisture in the air, the quantum dots were covered by a layer of PMMA A8, by spincoating for 60 s at 4000 rpm.

**Single emitter microscope.** For optical characterisation and measurements, we use an inverted optical microscope to confocally pump the dots at 450 nm (LDH-P-C-450B pulsed laser, PicoQuant) at 10 MHz repetition rate of < 70 ps pulses, with 90 nW inserted into the microscope. An oil objective (Nikon Plan APO VC, NA=1.4) focuses the pump laser onto the sample and collects the fluorescence. The excitation provides similar pulse energy density as in\(^{13}\) at the lowest energy density \(\langle N \rangle \ll 1\) quoted in that work. With the estimated efficiency of our set up, the excitation probability per optical pulse is estimated at <0.1 from the count rate. The fluorescence from the sample is directed either to a camera (PCO.edge 4.2, PCO AG), a spectrometer (PI Acton SP2300) or two fiber-coupled avalanche photodiodes (APDs) (SPCM-AQRH-14, Excelitas) in a Hanbury-Brown & Twiss configuration. The APDs are coupled to a photon correlator (Becker & Hickl DPC-230) that measures the absolute photon arrival times.

**Measurement protocol.** Using the camera and wide-field pump illumination, we select an emitter that appears to be diffraction-limited. After driving it to the laser spot, we do a time-correlated single-photon counting (TCSPC) measurement to collect photon arrival times. To calculate the photon correlations we use a home-built TCSPC toolkit that utilizes the algorithm developed by Wahl et al.\(^{28}\) to calculate \(g^2(\tau)\) and the lifetimes for the different emitters and for the individual CPA segments. From \(g^2(\tau)\) we select the emitters with a strong anti-bunching signal (normalized \(g^2(\tau) < 0.5\)). Of the 75 dots measured, 40 passed this test. Our TCSPC measurements are taken over 120 seconds of acquisition time. We note that in our decay traces taken using a Becker-Hickl DPC 230 photon-counting and correlator card in reverse start-stop configuration, a small time interval centered at around 30 ns is subject to an electronic artefact which we attribute to a ringing in the DPC-230
TDS timing chip. Therefore we exclude this time interval for decay rate fitting.

**Initial characterization.** Figure 1 presents initial characterization of an exemplary single dot on basis of standard timebinned analysis, where the data is sliced in 10 ms long segments, to each of which intensity and decay rate is fitted. Throughout this work we consider photon counting data, in which absolute time-stamps are collected with 0.165 ns resolution for all collected photons and concomitant excitation laser pulses, on two avalanche photodiodes (APDs) in a Hanbury-Brown and Twiss configuration. This allows to construct a posteriori from one single data set the intensity, fluorescence decay rate, and the $g^{(2)}(\tau)$ photon-photon correlation. In our optical measurements we post select all single nanoparticles on basis of photon antibunching ($g^{(2)}(\tau = 0) < 0.5$). For the example at hand, the selected emitter shows clear intermittency in intensity and decay rate (panels Fig. 1(B,C) discussed further below), while Fig. 1(D) shows a marked anti-bunching at zero time delay in the $g^{(2)}(\tau)$ that is constructed from the full photon record. The quantum dot in Fig. 1(E) shows a time-averaged emission spectrum that peaks at around 5 nm and has a spectral FWHM of 20 nm, which is consistent with reports by Protesescu et al., and together with the antibunching photon statistics points at quantum confinement. The time-integrated fluorescence decay trace (Fig. 1(F)) is markedly non-single exponential. Fitted to a double exponential decay we find decay rates of $\gamma_1, \gamma_2 = 0.43, 0.03$ ns$^{-1}$. We must note, however, that a double exponential is often not sufficient to fit these emitters, and typical decay rates for our dots range from 0.05 to 0.9 ns$^{-1}$. At these decay rates, the fastest decay rate component of the quantum dots generally span at least 10 timing card bin widths.

Figure 1(B, C) shows just a fraction of the intensity and decay rate time trace, plotted according to the common practice of partitioning the single photon data stream in bins. The fluorescence decay rate for each bin is obtained by fitting data within each 10 ms bin to a single exponential decay law employing a maximum likelihood estimator method that is appropriate for Poissonian statistics. As expected from prior reports on single pervoskite nanocrystal blinking, the intensity and decay rate time trace show clear evidence for
intermittency. The intensity varies from essentially zero to 150 counts per ms. Figure 1(B, right panel) shows a histogram of intensities, binned over the entire time trace (for all dots in this work, 120 s, or till bleaching occurred). The histogram shows a broad distribution of intensities with most frequent intensities around 60 cts/ms. This is in contrast with the typical bi- or trimodal physics of II-VI quantum dots, which usually show distinct bright and dark states.\textsuperscript{12,13,21,22,30} However, the width of the peak well exceeds the Poisson variance expected at these count rates, suggesting that there are many intensity levels. The decay rate histogram also displays intermittent behavior, in step with the intensity blinking. The most frequent decay rate is around 0.07 ns\textsuperscript{−1}. Fig. 1(G) displays a \textit{Fluorescence Decay Rate Intensity Diagram} [FDID], a 2D histogram displaying the frequency of occurrence of intensity-decay rate combinations. This type of visualization was first introduced by\textsuperscript{31–33} to identify \textit{correlations} between intensity and fluorescence decay rate (FLIDs in those works, using lifetime instead of decay rate). For II-VI quantum dots, FDID diagrams typically separate out bright and slowly decaying states from dark, quickly decaying states.\textsuperscript{32,33} Instead, for the perovskite quantum dot at hand, the FDID diagram presents a broad distribution with a long tail towards dim states with a fast decay.

The picture that emerges from Fig. 1 is consistent with recent observations of several groups,\textsuperscript{11,12,14} and qualitatively supports the notion of a continuous distribution of dark, gray state, consistent with the MRC model. This should be contrasted to typical II-VI quantum dot behavior, and also the recent report by\textsuperscript{13} on very similar CsPbBr\textsubscript{3} dots, but taken under very low repetition rate excitation conditions (fs pulses at very low repetition rate, as opposed to picosecond pulses at $\geq 10$ MHz — at similar $\langle N \rangle < 0.1$).

\textbf{Computational methods}

Since extreme caution is warranted when scrutinizing photon counting statistics to determine quantitative intermittency metrics due to artefacts of binning,\textsuperscript{23,25–31} we proceed to
analyze the data of a large number of dots with state-of-the-art bias-free statistical analysis to determine a lower bound to the number of involved states, and the switching dynamics and memory effects therein. We apply tools of Bayesian statistics, specifically, changepoint analysis (CPA) to partition the data in segments separated by switching events, and level-clustering to determine (a lower bound to) the number of states, as a rigorous and bias-free approach to investigate the intermittency of quantum dots. These tools were first proposed by Watkins and Yang, and later also used and extended in the context of quantum dot intermittency by. We refer to Ref. for our freely available implementation and a detailed description of benchmarking of this tool set. Here we summarize just the salient outcomes relevant for this work, obtained by extensive Monte Carlo based benchmarking to determine the performance of CPA and clustering for highly multilevel emitters.

CPA performs segmentation of the time record of single photon counting events into intervals within which the count rate is most likely a constant value, delineated by switching events or ‘changepoints’ at which the count rate changes, in as far as can be judged given the shot noise in the data. Since CPA works on a full time series with many jumps by finding a single jump at a time, and successively subdividing the time stream until segments with no further jumps are found, the ultimate performance is ultimately set by how well CPA can pinpoint in the last stage of the subdivision single jumps in short fragments of the photon stream. For significant intensity contrasts CPA detects changepoint in very short fragments (e.g., to accurately resolve a jump with a 5-fold count rate contrast, a record of just 200 photons suffice), with single-photon event accuracy. Smaller jumps are missed unless fragments are longer (e.g., factors 1.5 contrast jumps require fragments of ca. $10^3$ photons for near sure ($> 90\%$) detection. At typical practical count rates of $10^5$ cts/s this means that switching events further apart than 10 ms are accurately identified as long as jump contrasts exceed a factor 1.5 (ca. 100 ms for contrasts as small as 1.2). Switching events that are even closer in time are missed by CPA. This is intrinsic to the photon budget, i.e., the ultimate information content in the discrete event time stream fundamentally does not
allow pinpointing even more closely spaced switching even.

After dividing the time trace into segments spaced by changepoints, one is left with sequences containing the residence times \( T_q \) for each segment, photon counts \( N_q \) and instantaneous segment intensities \( I_q = N_q / T_q \), as well as decay rates \( \gamma_q \), obtained by maximum likelihood fitting of the decay trace from each segment to a single exponential decay. The question how many actual intensity levels most likely underlie the measured noisy sequence \( I_{m,c} \) can be determined using Watkins & Yang’s clustering algorithm.\(^{23}\) While Watkins and Yang considered Poisson distributed noise, as in this work, we recommend also the work of Li and Yang\(^{41}\) as a very clear explanation of the method, though applied to Gaussian distributed noise. The idea is that expectation-maximization is used to group the most similar segments together into \( n_G \) intensity levels, where \( n_G = 1, 2, 3, \ldots \). After this, the most likely number of levels, \( n_G = m_r \), required to describe the data, given that photon counts are Poisson distributed, can be determined by a so-called Bayesian Information Criterion (BIC).\(^{23,41}\)

We have extensively verified by Monte Carlo simulations the performance of CPA and level clustering for dots with many assumed discrete intensity levels in a separate work.\(^{26,27}\) In brief, at small photon budgets in a total time series, only few levels can be detected, but conversely at the total photon budgets in this work, exceeding \( 5 \times 10^6 \) events, clustering has a > 95% success rate in pinpointing the exact number of levels in dots with at least 10 assumed intensity levels. Moreover, for photon budgets that are too small to pinpoint all levels exactly (e.g., at \( 10^4 \) counts in a total measurement record, only up to 4 levels can be accurately discerned), clustering always returns a lower bound for the actual number of intensity states.
Figure 2: (A) An example of the intensity time trace of a measured quantum dot (purple, binned in 0.5 ms bins for visualisation purposes), and the intensity segments found by CPA (green). In the lower panel the lifetime for the found CPA segments is shown. On the right are histograms of the occurrences of the intensities for both treatments with segments weighted by their duration. (B-D) Three FDID plots weighting each CPA segment (B) equally, (C) by their number of counts, (D) by their duration. The choice of weights puts emphasis on different parts of the intensity-decay rate diagram, as they report on differently defined probability density functions.
Results and discussion

Changepoint analysis and FDID diagrams

We have applied the unbiased CPA analysis and Bayesian inference tools to data from 40 single CsPbBr$_3$ quantum dots. We first discuss an exemplary single dot as example, and then discuss statistics over many single dots. The example dot is identical to the one considered in Figure 1 and refer to the supporting information for results on all dots. In Figure 2A we see that CPA is able to accurately follow the intensity trace of a typical CsPbBr$_3$ quantum dot. We show only a section of the total measurement for clarity, and strictly for plotting purposes only, binned the photon arrival times in 0.5 ms intervals. Note that this binning is only for visualization, and does not enter the CPA algorithm. Figure 2B displays the fitted decay rates for the same selected time interval, obtained by fitting each of the identified segments. The right-hand panels of Figures 2A and B show histograms of intensity and lifetime as accumulated over the full time trace. It should be noted that these histograms are intrinsically different from those in Figure 1 for two reasons. First, binned data has entries from bins containing jumps, leading to a smearing of the histogram. Second, since histogramming of segment values $I_q$ is agnostic to segment duration, events are differently weighted. Thus the histogram of intensities now shows a bimodal distribution. The histogram of the decay rates still exhibits only a single peak at ca. 0.05 ns$^{-1}$.

Next we construct correlation diagrams of fluorescent decay rate versus intensity (FDIDs) from CPA data. Customarily FDIDs are 2D histograms of intensity and decay rate as extracted from equally long time bins in binned data. As the length of segments found by CPA can vary over many orders of magnitude, an important questions is with what weight a given segment should contribute to a CPA-derived FDID. A first approach is to give all segments an equal contribution to the FDID, which emphasizes the probability for a dot to jump to a given intensity-decay rate combination. Alternatively, one could weight the contribution of each segment to the histogram by the amount of counts it contributes. This
histogram hence emphasizes those entries that contribute the most to the time-integrated observed photon flux. Lastly, if one uses the segment durations as weights for contribution of segments to the FDID one obtains an FDID closest in interpretation to the conventional FDID diagram, which presents the probability density for being in a certain state at a given time. Figures 2D, E and F provide all three visualizations. The data shows variations in intensity levels over approximately a factor 10, with concomitant decay rates also varying over an order of magnitude. Overall, all diagrams suggest an inverse dependence qualitatively consistent with the notion that the dots experience a fixed radiative rate, yet a dynamic variation in the number of available non-radiative decay channels, that make the dot both darker and faster emitting. The unweighted and photon count weighted FDIDs show a peak at similar intensity and decay rate at $\gamma, I = 0.06 \text{ ns}^{-1}, 12 \times 10^4 \text{ s}^{-1}$, indicative of the most frequently occurring intensity/rate combination that is simultaneously the apparent bright state. The different FDID weightings emphasize different aspect of the data. For instance, weighting by counts highlights mainly the emissive states and underrepresents the long tail of darker state, with respect to the other weighting approaches. This qualitative difference can result in a quantitative difference for extracted parameters, such as the apparently most frequently occurring combination of intensity and decay rate.

FDIDs for essentially all dots (see supporting information) are much like the example shown in Figure 2 showing a slow decaying bright state with a long tail towards both lower intensity and faster decay. In fact, we can collapse the FDIDs of all 40 dots onto each other by summing histograms (no weighting by, e.g., segment duration) of normalized intensity $I/\langle I \rangle$ versus $\gamma$, which further underlines this generic behavior, see Figure 3. An appealing explanation for the observed dynamics is if the perovskite dots are characterized by always emitting from one unique bright state that is efficient and has a slow rate of decay $\gamma_r$ [labelled as radiative decay rate], while suffering fluctuations in both brightness and rate through abrupt jumps in a nonradiative rate $\gamma_{nr}$, as in the MRC model. In this picture, one would expect the FDID feature to be parametrizable as $I \propto B + I_0 \gamma_r / (\gamma_r + \gamma_{nr})$. The feature
Figure 3: FDID of all 40 single dots, obtained by summing single dot FDIDs for which the segment intensities were normalized to the mean intensity. A simple histogramming was used (no specific weighting of entries by duration or counts). Overplotted is a parametric curve of the form \((\gamma_r + \gamma_{nr}, B + I_0\gamma_r/(\gamma_r + \gamma_{nr}))\) with as input a fixed value \(\gamma_r\), and a background \(B = 0.06I_0\), with \(I_0\) adjusted to match the peak in the FDID, and \(\gamma_{nr}\) scanned.
in the collapsed FDID plot can indeed be reasonably parametrized as such a hyperbola. This parametrization is consistent with Ref.\textsuperscript{12} in which a linear relation between intensity and fluorescence lifetime was reported. The required radiative decay rate for the bright state is $\gamma_r \sim 0.075 \text{ ns}^{-1}$, while the parametrization requires a residual background $B = 0.06I_0$. This residual background is not attributable to set up background or substrate fluorescence, suggesting a weak, slow luminescence component from the dots themselves. Moreover, we note that the FDID feature clearly has a somewhat stronger curvature than the hyperbolic parametrization (steepness of feature at $\gamma < 0.1 \text{ ns}^{-1}$, and $I/\langle I \rangle > 1.0$).

**Clustering analysis**

The FDID diagrams at hand qualitatively support the continuous distribution of states hypothesized in the MRC model, and observed by Refs.\textsuperscript{11,12,14} As quantification of the number of states involved we perform clustering analysis\textsuperscript{23,24,26} to estimate the most likely number of intensity states describing the data on basis of Bayesian inference. A plot of the Bayesian Information Criterion as function of the number of levels $n_G$ allowed for describing the data of the specific example dot at hand is shown in Figure 4A. Strikingly, the BIC does not exhibit any maximum in the range $n_G = 1 \ldots 5$, but at $n_G = 13$. Recalling that the BIC criterion in clustering analysis for multistate dots at finite budget generally report a lower bound, this finding indicates that the data for this dot requires at least as many levels to be accurately described, if a discrete level model is at all appropriate.

Similar conclusions can be drawn from Figure 4B. We have found in Monte Carlo simulations that if one allows the level clustering algorithm to find the best description of intensity traces in $n_G$ levels for dots that in fact have just $m < n_G$ levels, then the returned description of the data utilizes just $m$ levels, with the remaining levels having zero occupancy in the best description of the data returned by the algorithm. Figure 4B shows the occupancy assigned by the clustering algorithm for our measured quantum dot as function of the number of states offered to the algorithm for describing the segmented intensity trace. Each additional
Figure 4: (a) The BIC of a single CsPbBr$_3$ quantum dot. We see that the BIC of this dot peaks at $n_G = 13$. (b) The occupancy diagram of the same quantum dot. The number of occupied states keeps growing with the number of available states, somewhat saturating around $n_G = 15$. (c) A histogram showing the durations of the found CPA segments of the CsPbBr$_3$ quantum dots. For this dot, we find a powerlaw tail with an exponent of $\alpha = 2.9$. (d) The long-term correlations of a single CsPbBr$_3$ quantum dot, when the intensity is calculated over 1 ms bins. A truncated powerlaw is fitted to the data. We find an exponent of $C = 0.39$. Figures demonstrating the switching times, the long-term autocorrelations, the m-state likelihood, and the occupancy of a single dot.
state offered to the clustering algorithm is in fact used by the algorithm, whereas Monte Carlo simulations have shown that at the photon budgets involved (5.5 × 10^6 photons) the clustering algorithm generally does not assign occupancy to more than \( m \) levels to simulated \( m \)-levels dots. The occupancy diagram hence confirms the conclusion from the BIC criterion that the dot at hand requires many levels, or even a continuous set of levels, to be described.

For all 40 dots we extracted wavelength, brightness, and performed the same CPA and clustering analysis as for the example dot. Moreover, we examined segment duration statistics for power law exponents. The supporting information contains a detailed graphical overview of the CPA results for each of the 40 dots, while summarized results are shown in Figure 5. Figure 5A shows that the dots have a low dispersion in peak emission wavelength, with emission between 500 and 510 nm. All considered dots offered between 2 and 8×10^6 photon events (Figure 5B) for analysis (120 seconds collection time, or until photobleaching). The mean intensity per measured dot (histogram Figure 5C) is typically in the range from 15×10^3 and 80×10^3 cts/s, with one single dot as bright as 110×10^3 cts/s. According to the Monte Carlo analysis in, the total collected photon count for all dots therefore provides a sufficient photon budget to differentiate with high certainty at least up to 10 states. We can thus with confidence exclude that intermittency in these perovskite quantum dots involves switching between just two or three states as in usual quantum dots. Instead any physical picture that invokes a set of \( m \) discrete levels requires a description in upwards of \( m = 10 \) levels. In how far further distinctions between \( > 10 \) discrete levels, or instead a continuous band can be made on basis of data is fundamentally limited by the finite photon budget that can be extracted from a single emitter. This quantification matches the observation in Ref. (based on examining time-binned FDID diagrams), but is at variance with Gibson et al., who used CPA and clustering on similar CsPbBr\(_3\) quantum dots, but arrive at an estimate \( m_r = 2.6 \). This difference may be attributable to the different excitation conditions that are unique to Gibson et al. relative to all other works.
Residence times

In Figure 4C we show a histogram of the segment lengths found by CPA. In other works, on-states and off-states are often separated explicitly by thresholding following which on-times and off-times are separately analyzed, for instance to ascertain the almost universally observed power-law dependencies and their exponents. In the case of our CsPbBr$_3$ quantum dots a level assignment in on and off states is not obvious. Therefore we simply combine all segment lengths irrespective of intensity level in a single histogram. These switching times are power-law distributed, at least from minimum time durations of 10 ms onwards. The short-time roll off is consistent with the limitations of the information content of the discrete photon event data stream: for segments shorter than 50 photons or so, even if physically there would be a jump, the photon number would not suffice to resolve it. Thus the roll-off does not exclude that power-law behavior also occurs for shorter times, but instead signifies that the testability of such a hypothesis is fundamentally limited. Fitting the power law $t^{-\alpha}$ for time $> 10$ ms indicates a power-law exponent of $\alpha = 2.9 \pm 0.1$.

The peculiar segment-duration power law statistics with exponent $\alpha \sim 2.5$ of our example dot also extends to the full ensemble. Figure 5E shows the distribution of power-law exponents that were fitted to the tail of the switching time histograms. We find a broad distribution of power-law exponents ranging from 1.5 to 3.0, with the bulk of the dots showing exponents in the range of $\alpha = 2.0$ to 3.7. These values are significantly higher than the values found for many semiconductor quantum dots, which generally are close to 1.5.\[21\] Also, these values are significantly higher than the exponents reported for on-times of CsPbBr$_3$ dots extracted from intensity-thresholded time-binned data. We note that one can (somewhat arbitrarily) threshold CPA-segmented data in an attempt to isolate 'on-times' for the bright state from the ‘residence times’ associated with the long dark/gray tail of states. Doing so with thresholds $I/(I) > 1.3$ (on-state) and $I/(I) < 0.7$ (tail of gray/dark states) estimated from Fig. 3 resulted in residence-time histograms for on- and off-times with similarly high slopes as we obtain for the full set. We thus find no support in our data for power laws.
generally being close to 1.5 or even below, as reported in other recent reports.\textsuperscript{13,16} We note that apart from the methodological difference of not working with binned thresholded data but with CPA analysis, also the selection of dots reported on may matter. In this work we report on all dots identified as single photon emitters by their $g^{(2)}(0)$. Instead in Ref.\textsuperscript{16} dots are reported to have been selected as those for which inspection of binned time traces suggested the most apparent contrast between bright and dim states, qualitatively appearing closest to bimodal behavior. According to our analysis of FDIDs and in light of the MRC model, this post-selection may not single out the most representative dots.

An alternative approach to quantifying blinking statistics and power law exponents that requires neither thresholding binned data nor CPA is to simply determine intensity autocorrelation functions $g^{(2)}$ for time scales from ms to seconds, as proposed by Houel et al.\textsuperscript{42} According to Houel et al.\textsuperscript{42} the normalized autocorrelation minus 1 may be fit with the equation $At^{-C}\exp(-Bt)$. Figure 4(D) show such an analysis for the exemplary dot at hand, for which we find a reasonable fit with $C = 0.39$. As Figure 5F shows, across our collection of dots we generally fit exponents $C$ in the range 0.10 to 0.75 to intensity autocorrelation traces. We note that the relation $\alpha = 2 - C$ put forward by Houel et al.\textsuperscript{42} is only expected to hold for two-state quantum dot, and $C$ does not relate directly to $\alpha$ for quantum dots in which more than two states are at play.

Figure 5: Summary of the behavior of the 40 measured single quantum dots. We show the distribution of found (A) peak wavelengths, (B) total photon count, (C) intensities, (D) most likely number of states, (E) powerlaw exponents of the switching time $\alpha$, and (F) the power-law exponent of the autocorrelation $C$. 
Finally we examine the dots for aging and memory effects, leveraging the fact that CPA gives an unbiased data segmentation into segments $n = 1 \ldots N$ that are classified by segment duration $T_1, T_2, \ldots$, intensity in counts/sec $I_1, I_2, \ldots$, and decay rate $\gamma_1, \gamma_2, \ldots$ that is established without any distorting temporal binning. We present results again for the same example dot as in Fig. 1 in Figures 6 and 7. With regard to aging, one can ask if over the full measurement time in which a dot undergoes of order $10^8$ excitation cycles, the distribution of segment duration, intensity and decay rate show any sign of change. To this end, we subdivide the total measurement period (e.g., Fig. 6(A-C), total measurement time 60 s for this dot) in 100 slices that are equal length in terms of wall-clock time, and examine the evolution of histograms of $I_n, \gamma_n$ and $T_n$ for these short measurement intervals as function of their occurrence in the measurement time. As the residence times are very widely distributed, we plot histograms of $\log_{10}T_q$, with $q$ the index of the segments. There is no evidence that any of these observables change their statistical distribution over the time of the measurement. While Fig. 6(A-C) shows an example for just one dot, this conclusion holds for all dots in our
measurement sets, with the caveat that for some dots drifts in microscope focus caused a small gradual downward drift in intensity. We observed no photobrightening of dots during the experiment.

Clustering allows us to ask questions that are not accessible with simple binning of data, as we can examine the datasets for correlations between parameters and between subsequent segments. In terms of cross-correlating different observables, beyond FDIDs that correlate intensity and decay rate, one can also examine correlations between intensity and segment duration, and between decay rate and segment duration. Histogramming the clustered data to screen for such correlations (Fig.6(D,E)) show that both the distribution of intensities and of decay rates are uncorrelated, or only very weakly correlated, with the segment duration. In other words, we find no evidence that within the distribution of states between which the dot switches, some states have different residence time distributions than others.

Memory effects\[16\] should appear as correlations in the values for any given observable in subsequent segments, i.e. in conditional probabilities that quantify what the probability $P_{\Delta n}(A|B)$ is that a chosen observable to obtain a value $A$ is given that it had a value $B$ in the previous segment ($\Delta n = 1$), or generally counting $\Delta n$ events further back into the history of previous segments. Figure 7 shows such conditional probabilities for $\Delta n = 1$ (panels A-C) and $\Delta n = 2$ (panels D-F), for intensity (panels A,D), decay rate (panels B,E) and segment duration (panels C,F). These diagrams are obtained by applying a simple 2D histogramming approach, listing the value of $B$ as $x$–axis, the value of $A$ as $y$–axis, and normalizing the sum of each of the columns to obtain a conditional probability. We note that this approach means that at the extremes of the histograms (far left, and far right), there are few events to normalize to, leading to large uncertainty. When screening for memory in intensities, it’s important to consider that the CPA algorithm itself selects for intensity jumps. Due to this the intensity after one jump ($\Delta n = 1$) is a priori very unlikely to achieve a similar value, which leads to a near-zero conditional probability at the diagonal of Figure 7(a). Nonetheless, the distinct features in the diagram at $\Delta n = 2$ (Fig. 7(b)) do suggest that the dot generally
Figure 7: (A-F) Conditional probability of observing a value for intensity (A, D), rate (B, E) or segment duration (C, F), given the value of the same observable one or two steps earlier, respectively. (G,H,I) Normalized autocorrelation (difference from 1) of the sequence of intensity, decay rate and segment durations. This data is for the same single dot as considered in Fig. 6.
alternates repeatedly back and forth between a bright and a dark state. More telling than diagrams for intensity are those for decay rate. They show that if, in a given step the decay rate is low (slow, bright feature in FDID at \(< 0.1 \text{ ns}^{-1}\)), then in the subsequent step the decay rate is usually fast, yet widely distributed from 0.1 to 1 ns\(^{-1}\), and vice versa from any of the fast decaying states, the dot is likely to jump to the quite narrowly defined slow rate of the bright state. If one considers the conditional rate at \(\Delta n = 2\), the conclusion is that if the dot is in the bright state with its slow decay rate at a given step, then likely after two jumps it comes back to this bright, slowly decaying state. If, however, the rate was fast anywhere in the interval from 0.1 to 1 ns\(^{-1}\), then after an excursion to the slow rate at \(\Delta n = 1\), the dot likely in the second step again takes on a fast rate in the interval from 0.1 to 1 ns\(^{-1}\) but without a particularly clear preference for any value in that wide interval. Finally we note that there is no indication in our data that subsequent residence times \((P(T_{m+\Delta n}|T_{m})\) show any memory (Fig. 7(c,f), showing result for \(\log_{10} T_{m}\)). Similarly we found no memory effects for the sequence of on-times, as selected from CPA by thresholding at \(I/\langle I\rangle > 1.3\).

One could speculate that the information gleaned from such conditional probability diagrams could be advantageously condensed in autocorrelations of the traces \(I_q, \gamma_q\) and \(T_q\). We plot normalized autocorrelation traces \(G(\Delta q) - 1\) where for any sequence \(H \in I_q, \gamma_q, T_q\), one defines

\[
G(\Delta q) = \langle H(q)H(q + \Delta q) \rangle/\langle H(q) \rangle^2
\]

(where \(\langle . \rangle\) denotes the mean over \(q\) are all segment indices and \(\Delta q\) is \(1, 2, \ldots\)), so that at long times \(G_m - 1\) vanishes.

In Fig. 7(G-I) we plot \(G(\Delta q) - 1\) for intensity, rate and segment duration. Such segment-autocorrelations are distinct from, e.g., the usual intensity autocorrelation traces that one might examine to determine blinking power laws, since here one autocorrelates subsequent intensity segments without any regard for their time duration. For a conventional two-level dot, the autocorrelation trace \(I_q\) would oscillate with large contrast up to very large \(q\). Instead, we find that the dot at hand shows an oscillation with a distinct contrast in the
intensity segment autocorrelation contrast for up to 5—10 cycles. In the normalized autocorrelation for decay rates the memory is far less evident. We attribute this not to a lack of memory, but note that if a dot switches between a state of well defined slow rate, and an array of states with highly distributed fast rates, then upon averaging the wide distribution of fast rates washes out any autocorrelation signature. Finally, the residence times, which we already found to be uncorrelated between subsequent jumps, show no autocorrelation signature for $z \neq 0$. A similar behavior to that shown in Figure 7(G) was observed for circa 30% of the dots studied, with other dots showing no clear intensity autocorrelation.

Conclusion

To conclude, we have reported on intermittency properties of a large number of CsPbBr$_3$ quantum dots on basis of a Bayesian inference data analysis. This approach works with raw, unbinned, photon counting data streams and thereby avoids artefacts commonly associated with the analysis of timebinned data. Our data generally confirms the multiple recombination centers model. In particular, we find that dots have in addition to their bright emissive state a tail of gray states that qualitatively appears continuous in FDID diagrams, and that according to clustering analysis requires at least 10 to 20 levels to describe, if a discrete-level description would be appropriate. Thereby our work provides a confirmation of claims in earlier works$^{11,12,14}$ under similar excitation conditions, with the distinction that we do not use time binned data but rigorously exploit all the information in the data stream to the level that its intrinsic noise allows. We note that the same type of dots have displayed a different behavior, indicative of 2 to 3 levels, in Ref.$^{13}$ Since that work uses almost identical Bayesian inference methods, we conclude that this distinction is really due to the different physical realization. While differences in sample preparation can not be excluded, we note that Ref.$^{13}$ stands out from all other reports due to its very different excitation conditions. As in the first report proposing the MRC model for perovskite dots, we qualitatively find
that the tail of gray states display an inverse correlation between intensity and rate, suggesting that the dots have a unique bright state with a given decay rate, to which random activation of recombination centers add nonradiative decay channels. However, we note that this observation merits further refinement of models: while plotting intensity versus lifetime may point at strict proportionality, plotting rate instead of lifetime accentuates deviations, notably a deviation in curvature of our data relative to inverse proportional dependence.

Finally, we have analyzed correlations in the measured CPA-segmented sequences of intensity-levels, decay rates and segment lengths. We find no evidence for aging, i.e., gradual shifts in e.g., decay rate or blinking dynamics during photocycling of dots through $10^6$ to $10^7$ detected photons (i.e., well over $10^8$ cycles). Also, our data indicate that residence times are not correlated to the state that a dot is in. The residence times can be fiducially extracted for a limited time dynamic range from ca. 5-15 ms, limited intrinsically by count rate, to ca. 10 s, limited by the length of the photon record. We note that in residence time histograms determined by CPA, according to Monte Carlo simulations at long times the analysis fiducially reports on power laws without introducing artefacts, such as apparent long-time roll offs. The exponents that we find are in the range from 1.5 to 3.0, which appears high compared to the near-universal value of 1.5 observed for II-VI single photon sources. In the domain of CsPbBr$_3$ dots, reports have appeared of even lower exponents (down to 1.2)\textsuperscript{[13,16]} with exponential roll offs at times $\sim$ 0.1 s that cause a steepening of the residence time histogram at longer times. We note that although exponential roll off certainly steepens slopes in the residence time histogram, our histograms do not point at exponential, but at high-exponent power law behavior.

Regarding memory, we found a distinct memory effect in intensity and rate in the sense that dots appear to switch between a quite unique bright state with slow decay rate that is evident as the bright pocket in FDIDs, and the entire tail of dim states in the FDID. Moreover, the dots do not appear to return preferentially to this dim state, but explore the entire tail anew at each transition from the bright state. This is an important refinement...
on the MRC model which in itself leaves open if dots return at all to the bright state before choosing another dim state, and which does not specify if dots make repeated visits to the same dim state or not. In terms of analysis, this memory is only partially visible in autocorrelation traces of sequences of CPA intensity and rates, as the dim states are so widely distributed. The averaging involved in evaluating autocorrelations washes out some memory effects that do appear more clearly in conditional probability histograms reporting on subsequent jumps. Finally we found no evidence in our data set for the apparent memory in residence times (segment lengths $T_q$) reported by Hou et al.\textsuperscript{16} for on-times.

In our view, this rich data set will stimulate further theory development in the domain of inorganic quantum dot intermittency. The CPA-segmented data for all quantum dots is made available in digital form for all quantum dot single photon emitters that we report on in this work.

**Supporting Information Available**

PDF containing CPA-based intermittency analysis report for 40 single dots - appended to this preprint. For select dots, data is posted with the Python analysis code of Ref.\textsuperscript{26,27} at github.

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