Impact of Noise and Background on Measurement Uncertainties in Luminescence Thermometry

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ABSTRACT: Materials with temperature-dependent luminescence can be used as local thermometers when incorporated in, for example, a biological environment or chemical reactor. Researchers have continuously developed new materials aiming for the highest sensitivity of luminescence to temperature. Although the comparison of luminescent materials based on their temperature sensitivity is convenient, this parameter gives an incomplete description of the potential performance of the materials in applications. Here, we demonstrate how the precision of a temperature measurement with luminescent nanocrystals depends not only on the temperature sensitivity of the nanocrystals but also on their luminescence strength compared to measurement noise and background signal. After first determining the noise characteristics of our instrumentation, we show how the uncertainty of a temperature measurement can be predicted quantitatively. Our predictions match the temperature uncertainties that we extract from repeated measurements, over a wide temperature range (303–473 K), for different CCD readout settings, and for different background levels. The work presented here is the first study that incorporates all of these practical issues to accurately calculate the uncertainty of luminescent nanothermometers. This method will be important for the optimization and development of luminescent nanothermometers.

KEYWORDS: luminescence thermometry, temperature uncertainty, statistics, (EM)CCD, background, absorption cross section

Nanomaterials with temperature-dependent luminescence are one of the most versatile thermometers on the microscopic scale with applications in biology, electronics, and catalysis.1–3 The temperature of a nanothermometer is determined by recording its emission spectrum or its luminescence lifetime. The intensity ratio between two emission bands is most frequently considered because this parameter is insensitive to fluctuations in the excitation intensity, changes in alignment, and scattering of the luminescence. The relative sensitivity, $S_r$ (in % per K), expresses how strongly the intensity ratio changes with temperature and is thus a measure for the measurement accuracy. $S_r$ is an intrinsic property of a thermometer material (which however depends on temperature) and is easily determined by measuring emission spectra over a range of temperatures. Newly developed thermometer materials are therefore often characterized and compared in terms of this parameter.4,5

In practice, the reliability of temperature readout depends not only on the relative sensitivity but also on the signal-to-noise ratio of a measurement. These parameters together determine the temperature uncertainty, $\sigma_T$. Current methods to determine $\sigma_T$ are diverse. The most direct method is experimentally recording a series of luminescence spectra and calculating the standard deviation of the extracted temperatures.6 Alternatively, the noise level on a single spectrum may be estimated from fluctuations in the baseline. The latter method underestimates the temperature uncertainty because it fails to take the noise on detected photons into account. More importantly, both methods are often used in idealized circumstances where background signal is minimal, a large amount of thermometer material is measured, luminescence is efficiently collected, and/or long measurement times are used. The extracted values of $\sigma_T$ depend strongly on these circumstances. In contrast to $S_r$, $\sigma_T$ is not an intrinsic property of a (nano)thermometer material. Consequently, user-to-user differences have caused variations in reported uncertainties of several orders of magnitude for the same thermometer, while some measurement conditions such as the environment of the thermometer were similar.4 It is not clear to what extent these reported values of $\sigma_T$ are relevant for actual applications of the (nano)thermometers, which may put restrictions on the measurement procedure and/or introduce background fluorescence and blackbody radiation.6 In addition, undesired
emissions from the thermometer itself, for example, from higher-excited levels, can interfere with temperature measurements. Although subtracting a reference spectrum of any background signal removes the systematic error, the influence on the temperature uncertainty remains. It is currently unclear how these practical complications affect the performance of luminescent (nano)thermometers. This makes a fair comparison of potential thermometer materials impossible.

In this article, we use the statistics of photon detection to quantify how noise and background signal affect the temperature uncertainty of a luminescence (nano)thermometry experiment. Not only the properties of the temperature material are important but also the characteristics of the detector and the sample. We first measure the background-free upconversion luminescence of NaYF₄:Er³⁺(2%),Yb³⁺(18%) nanocrystals using a conventional CCD and characterize the different types of detector noise. Using error propagation, we quantitatively explain the temperature uncertainties determined by recording a series of spectra, which increases with the set temperature. We study the impact of detector noise by recording upconversion luminescence of nanocrystals with electron multiplication gain using an electron-multiplying CCD (EMCCD). An increase in gain boosts the signal to overcome readout noise and thus reduces the temperature uncertainty. Finally, we examine the effect of background signal. Even if we subtract the background signal, the experimental temperature uncertainty increases with higher background levels as predicted from the larger error on the number of detected photons. These new methods to calculate the uncertainty show that not only the relative sensitivity \( S \) of a thermometer determines its performance but also the achievable signal-to-noise ratio. The temperature uncertainty \( \sigma_T \) depends strongly on measurement conditions and is therefore a poor parameter to compare the potential of thermometer materials. We propose alternative metrics that could be considered.

## RESULTS

We first discuss the uncertainty achieved with a model thermometer based on two emissive excited levels A and B (Figure 1a). An increase in temperature affects the relative intensities emitted by these levels, resulting in a change in the intensity ratio in the emission spectrum. In a typical experiment, these emissions are spectrally separated by a grating and captured by a CCD, photomultiplier tube, or photodiode array. The photosensitive material in the detector converts the incident photons to photoelectrons. The number of photoelectrons \( k \) recorded in one exposure will follow the Poisson distribution

\[
p_{\text{Poisson}}(k, \bar{k}) = \frac{\bar{k}^k e^{-\bar{k}}}{k!}
\]

where \( \bar{k} \) is the expected number of photoelectrons, which is proportional to the product of the photon flux and the acquisition time and is in general different for levels A and B. An interesting property of the Poisson distribution is that the standard deviation is equal to the square root of the expected value (Figure 1b). The next step in the detection process is the translation of photoelectrons to digital counts for each pixel, which enables the construction of the emission spectrum. For a luminescence thermometry experiment, the observables of interest in the spectrum are the integrated counts of the emissions from levels A and B: \( n_A \) and \( n_B \), respectively. As \( n_A \) and \( n_B \) are independent random variables, the measurement error on the intensity ratio \( R = n_B/n_A \) follows from error propagation

\[
\sigma_R = \sqrt{\left(\frac{\partial R}{\partial n_A}\sigma_A\right)^2 + \left(\frac{\partial R}{\partial n_B}\sigma_B\right)^2} = \frac{n_B}{n_A} \sqrt{\left(\frac{\sigma_A}{n_A}\right)^2 + \left(\frac{\sigma_B}{n_B}\right)^2}
\]

Here, \( \sigma_A \) and \( \sigma_B \) are the expected counts of A and B with corresponding variances \( \sigma_A^2 \) and \( \sigma_B^2 \), which are the signal-to-noise ratios on \( n_A \) and \( n_B \). Assuming that the errors and expected counts are related as described by the Poisson distribution, we expect lower \( \sigma_R \) for higher counts.

Conversion of \( R \) to a temperature value requires knowledge of the relative sensitivity. This is often obtained by calibrating the spectral response of the thermometer over a range of temperatures. Any error in the calibrated relative sensitivity leads to a systematic difference between the measured and physical temperature. However, the random error \( \sigma_T \) on the measured temperature only depends on the probability distribution function of the measured \( R \) and on the relative sensitivity of the thermometer

\[
\sigma_T = \sqrt{\left(\frac{\partial T}{\partial R}\sigma_R\right)^2} = \frac{1}{S_T} \sqrt{\left(\frac{\sigma_A}{n_A}\right)^2 + \left(\frac{\sigma_B}{n_B}\right)^2}
\]

This shows that, as expected, the temperature uncertainty decreases with the increasing signal-to-noise ratio. An
alternative approach of luminescence thermometry relies on the shift of an emission band at varying temperatures. In the Supporting Information, we use a similar analysis as above to determine the temperature uncertainty based on a spectral shift. Finally, we verify eq 3 by simulating luminescence spectra with two emissions bands and determine the temperature from the ratio of the simulated counts (Figure 1c). These simulated temperatures follow a normal distribution with a standard deviation that matches eq 3 (Figure 1d). We thus understand quantitatively how experiments with higher counts, performed with, for example, longer acquisition times or brighter thermometers, have a lower temperature uncertainty.

In Figure 2, we experimentally study the temperature uncertainty of thermometry measurements at elevated temperatures. We acquired spectra with a CCD camera because this is the most frequently used detector in the luminescence thermometry community. The CCD camera conveniently records an entire spectrum within one capture. In contrast, step-wise acquisition of a spectrum with a scanning monochromator and single-point detector such as a photomultiplier tube leads to additional temperature errors if the excitation intensity fluctuates during the measurement. We use NaYF₄:Er³⁺(2%),Yb³⁺(18%) nanocrystals as thermometers because green upconversion emission from this popular thermometer material can be excited with 980 nm light, preventing background fluorescence. Figure 2a shows the luminescence spectra, in which the two emission bands at 540 nm (green dots), respectively. The black line is a fit of the experimental data to the normal distribution ($\bar{K} = 1629, \sigma^2 = 2112$), and the dashed line shows the Poisson distribution with $\bar{K} = 1629$. (f) Plot of the variance against the mean (purple dots) measured via the procedure in (e) for different intensities of the white lamp. The black line is a fit of the experimental data to model $\sigma^2 = \frac{k}{f} + \sigma_e^2$, where $f = 0.78$ is the analog-to-digital conversion factor and $\sigma_e^2 = 57$ is the readout variance of one pixel. (g) Temperature uncertainties at various physical temperatures obtained via the procedure in (c) (red dots). The black line is the temperature uncertainty calculated using eq 3, with no fit parameters.

Figure 2. Experimental temperature uncertainty. (a) Upconversion luminescence of dried NaYF₄:Er³⁺(2%),Yb³⁺(18%) nanocrystals upon 980 nm excitation at 302 K (blue) and 473 K (red). (b) Logarithm of the ratio between the $^2H_{11/2}$ and $^4S_{3/2}$ emission with integration ranges 516–534 and 538–545 nm (green dots), respectively. The black line is a fit of the experimental ratios to Boltzmann’s distribution (eq 4), yielding values of 746 cm⁻¹ for $\Delta E$ and 15.2 for C. (c) Temperatures extracted from 200 experimental spectra using the calibration in (b). The thermocouple in our heating stage measured a temperature of 302 K during the acquisition of spectra, showing a small deviation with the mean of the temperatures extracted from the spectra, likely caused by a systematic error in the calibration. The gray shaded area covers the temperature range of the mean ± standard deviation. (d) Correlation between the measured temperature and the total counts within the integration ranges of the $^2H_{11/2}$ and $^4S_{3/2}$ emissions. The correlation coefficient ($\rho$) is close to zero, indicating that measured temperature and total counts are uncorrelated. (e) Distribution of counts per 1000 ms frame, for pixels on our CCD camera showing an average of 1629 counts/1000 ms when measured over 200 frames. The camera recorded the reflection of a white lamp on a microscopy slide. The solid line is a fit of the experimental data to the normal distribution ($\bar{K} = 1629, \sigma^2 = 2112$), and the dashed line shows the Poisson distribution with $\bar{K} = 1629$. (f) Plot of the variance against the mean (purple dots) measured via the procedure in (e) for different intensities of the white lamp. The black line is a fit of the experimental data to model $\sigma^2 = \frac{k}{f} + \sigma_e^2$, where $f = 0.78$ is the analog-to-digital conversion factor and $\sigma_e^2 = 57$ is the readout variance of one pixel. (g) Temperature uncertainties at various physical temperatures obtained via the procedure in (c) (red dots). The black line is the temperature uncertainty calculated using eq 3, with no fit parameters.
intensity ratios from spectra with different levels of noise to apparent temperatures.

Figure 2c shows the temperatures that we extracted from a series of spectra using the calibration of Figure 2b.14 These values are evenly distributed around the mean, which is a sign of a stable physical temperature during the measurement. Correlations between the extracted temperature and the total green luminescence counts could indicate that the laser heats the sample as variations in laser intensity would result in higher count rates coinciding with more laser heating. Experiments at higher laser powers do show such correlations (Figure S2), which indicates that increasing the excitation intensity to reduce the uncertainty can induce a systematic error on temperature readout. This is not observed in our experiments shown in Figure 2d. We therefore used the standard deviation of the measured temperatures as the experimental temperature uncertainty at a fixed sample temperature. Fluctuations in excitation intensity below the heating threshold do not affect the intensity ratio nor the signal-to-noise ratio (Figure S3).

To understand the magnitude of the variations in measured temperature (Figure 2c), we must consider the noise generated by our detector. The main noise sources in a CCD measurement are counting noise due to the statistics of incident photons and readout noise due to the translation of photoelectrons to digital counts by the analog-to-digital converter (ADC).15 We characterize these by acquiring a large set of 200 reference images on our CCD camera, illuminating it with a white lamp, and histogramming the digital counts of pixels with the same mean (Figure 2e). The distribution of digital counts approximates a normal distribution with a variance that is, in our case, slightly larger than the corresponding Poisson distribution would have, taking the single-pixel readout variance into account. This difference is due to the conversion of photoelectrons to digital counts, which changes the variance on the output counts by the ADC factor, f.16,17 Figure 2f shows a fit of the experimental variances to a model that includes the ADC factor and the readout noise

\[
\sigma^2 = \frac{\bar{n}}{f} + \sigma_r^2
\]

(5)

where \(\bar{n}\) is the average number of counts per pixel, \(\sigma_r\) is the Poisson noise, \(\sigma_f\) is the ADC factor, and \(\sigma_r\) is the readout noise. We find a value of 0.78 for the ADC factor, which is specific to our camera in the used setup. If an emission band is integrated over N camera pixels, the variance of the total readout noise on this band is N times larger than the single-pixel readout variance \(\sigma_f^2\) = 57. We can now insert this expression and the relative sensitivity of a Boltzmann thermometer, \(\Delta E/k_bT^2\), into eq 3 to calculate the expected temperature uncertainty from a luminescence spectrum.

Figure 2g shows the temperature dependence of the uncertainty by comparing experiments with the theoretical trend (eqs 3–5). Using the method presented in Figure 2c, we determine the temperature uncertainty at various physical temperatures and find values of 0.2 K at room temperature increasing to more than 0.4 K at 473 K.19 Between the different physical temperatures of the experiment, the total counts within the spectrum varied slightly—a decrease is likely due to thermal quenching and an increase could indicate water desorption from the surface of the dried nanocrystals.21 This affects the signal-to-noise ratio and thus obscures the impact of the intensity ratio on the temperature uncertainty. We therefore kept the sum of \(\bar{n}_g\) and \(\bar{n}_n\) roughly constant at 2 × 10^6 counts, which allows us to separately calculate \(\bar{n}_n\) and \(\bar{n}_g\) using eq 4 in a range of physical temperatures. We can therefore use eq 4 to calculate \(\bar{n}_g\) and \(\bar{n}_n\) separately, depending on the physical temperature. Inserting these values, along with the detector characteristics (eq 5), into eq 3 yields the theoretical uncertainty (black line in Figure 2g). The calculated uncertainties agree well with the experimental values, without any fit parameters. We therefore conclude that after proper characterization of the photodetector, error propagation correctly predicts the experimental uncertainty and its temperature dependence.

As a further illustration of the effect of the detector noise characteristics on a temperature measurement, we consider the effect of electron multiplication (EM) in an EMCCD. Emerging applications of luminescent nanomaterials, such as single-particle thermometry, require photodetectors that are able to record extremely weak signals.22,23 EMCCDs could offer a solution as they enhance the signal by orders of magnitude, compared to conventional CCDs, but the electron multiplication process causes additional noise.15,16 We start by considering the detection of photons and generation of photoelectrons, which in both a conventional CCD and an EMCCD follows Poisson statistics. Both types of cameras then transfer the photoelectrons to the ADC via the serial readout register and convert them to digital counts. In an EMCCD camera, the readout register is extended with additional registers that, depending on the applied voltage, multiply the number of photoelectrons and thus boost the signal. In practice, the output electrons pass through hundreds of multiplication registers, resulting in a total EM gain G. The number of output electrons \(n_{out}\) follows the gamma distribution

\[
p_{\text{gamma}}(n_{out}, k, G) = \frac{k^k n_{out}^{k-1} e^{-kG}}{G^k (k-1)!}
\]

(6)

Here, k is the number of photoelectrons generated by a CCD pixel, which enter the multiplication registers. Figure 3a shows the probability distribution of \(n_{out}\) as a function of the expected number of input photoelectrons \(\bar{k}\) and the EM gain G. EM produces an expected number of counts of \(\bar{n}_{out} = kG\) with a variance that approximates \(\sigma_{out}^2 = 2kG^2/\bar{k}\) (eq S1 and Figure S4). The signal-to-EM-counting-noise ratio \(\bar{n}_{out}/\sigma_{out}\) is \(\sqrt{kG}/2\), where readout noise is excluded, is thus independent of the gain factor. This derivation shows that EMCCD measurements have an additional counting noise of \(\sqrt{2}\), commonly referred to as the excess noise factor. Therefore, EM gain can only improve a temperature measurement if \(n_{out}\) is small with respect to other noise sources.

In Figure 3b, we compare the experimental temperature uncertainty at various levels of EM gain with theoretical predictions. First, we acquired 200 experimental spectra with an EM gain of only a factor 2 while keeping the number of incident photons per pixel low. This resulted in an extremely high uncertainty of 30 K. Increasing the EM gain to values of 25 causes a sharp decline of the uncertainty to 3 K. The effect of even higher EM gains is weak. We again explain this trend using eqs 3 and 5 (solid line) by realizing that the variance of \(\bar{n}_{out}\) is \(\bar{s}_{out}^2 = 2kG^2/\bar{k}\) and the \(\bar{n}_{out}\) due to the Poissonian counting noise is multiplied by the ADC factor f and the EM gain factor G and readout noise. (For photon fluxes relevant for luminescence thermometry, we can neglect noise due to spurious electrons created during shifting of charges through the multiplication register (Figure S5).) EM gain increases the signal (as well as the counting noise) with respect to the readout noise.
The contour lines correspond to uncertainties were normalized to the minimum value within the map. The temperature uncertainties in the color map were calculated via eqs 3 and 7, with no effect of EM gain for a range of different input photoelectrons \( \bar{k} \) per pixel and the EM gain. The average numbers of photoelectrons generated by a pixel are 5 and 30 for the \( ^2\text{H}_{11/2} \) and \( ^4\text{S}_{3/2} \) emissions, respectively. The solid black line is the temperature uncertainty calculated via eqs 3 and 7, with no fit parameters. (c) Color map of the temperature uncertainty as a function of the expected photoelectrons \( \bar{k} \) per pixel and the EM gain. The temperature uncertainties in the color map were calculated via the expected value of the output electrons, excluding spurious electrons, and the variance of all output electrons (eq S1). All uncertainties were normalized to the minimum value within the map. The contour lines correspond to \( \sigma_i \) values of 1.2, 2, and 8 K. We obtain the simulated spectra in (c) by drawing random numbers of different input photoelectrons \( \bar{k} \) per pixel and the EM gain.

The measured uncertainties as a function of physical temperature match the predicted values for a range of different background levels (Figure 4d). We observe higher absolute values of the uncertainty with an increasing background. This effect is as large as a factor of 3 for the background level in Figure 4a, even though we could subtract the background using a reference measurement. We further observe that the minimum of the temperature uncertainty (dashed line) shifts to higher set temperatures with increasing background counts. In practice, low levels of background are challenging to completely avoid, especially at elevated temperatures where blackbody radiation becomes an issue. Figure 4 shows how this affects the absolute value of temperature uncertainty as well as the optimal operating temperature of a thermometer compared to idealized measurement conditions without background signal.

Our work clearly demonstrates how the precision of a temperature measurement depends not only on the relative sensitivity of the thermometer but also on the measurement conditions. As these measurement conditions will be different for different applications, this raises the question of how to define a relevant metric to compare thermometers. Currently, the achieved temperature uncertainty is frequently reported in

\[
\frac{\bar{n}_{\text{out}}}{\sigma_i} = \frac{\bar{k}G}{\sqrt{2EG^2/f} + \sigma_i^2}
\]  

where \( \bar{n}_{\text{out}} \) is the expected number of background counts removed. Additional counts from dark current in the photodetector have an equivalent impact on the temperature uncertainty as background emissions. Again, inserting this expression into eq 3 gives the theoretical temperature uncertainty after background removal.

Figure 3c illustrates the effect of EM gain for a range of expected photoelectrons. We observe the highest uncertainty in the bottom left of the map, where both the EM gain and the number of expected photoelectrons are low—the corresponding simulated spectrum barely shows emission. For all numbers of expected photoelectrons, we observe a rapid decrease of the uncertainty with increasing gain, although this effect becomes negligible once readout noise is overcome (Figure S6). Indeed, the signal-to-total-noise ratio (eq 7) approaches a constant value of \( \sqrt{\bar{k}f/2} \) for large \( G \) and EM cannot improve it further (Figure S7). In practice, EM gain is thus useful if and only if the signal is weak compared to readout noise.

Besides the photodetector, background emission by the surroundings of the thermometer can be another source of uncertainty, which is relevant when the thermometer is used in realistic experimental conditions. We discuss how such a distortion of the spectrum affects the temperature uncertainty even after subtraction of the background. Figure 4a shows how we have mimicked this experimental issue: we have measured the upconversion luminescence with and without an additional broadband background signal from a white lamp. Subtracting a reference measurement of the lamp recovers a clean thermometer spectrum from the experiment with background. However, the noise on the background signal cannot be removed. The corrected spectrum therefore contains more noise compared to the background-free upconversion emission spectrum (Figure 4b). This translates to an increased temperature uncertainty (Figure 4c). We can further understand this from the expression for the variance \( \sigma_{b,i}^2 \) in the counts from emitting state \( i \) after background removal

\[
\sigma_{b,i}^2 = \frac{\bar{n}_{i} + \bar{n}_{b,i}}{f} + \sigma_i^2
\]
the literature. However, our results show that the temperature uncertainty is not a fundamental property of the thermometer. Temperature uncertainties measured under idealized experimental conditions are difficult to compare and may not be relevant for applications. In particular, different experimental settings can yield wildly different contributions of various noise sources. However, whatever the specific experimental noise contributions are, the temperature uncertainty is always minimal for a high relative sensitivity $S$, and for a strong luminescence signal (eq 3). Although $S$, of newly developed thermometer materials is commonly reported, the potential signal strength is hardly considered.

The realization that signal strength is essential for precise temperature measurement makes it possible to identify relevant parameters, in addition to $S$, that define a good thermometer. The expected counts on emissions A and B can be written as

$$\bar{n}_{AB} = N_C \sigma_{abs} \eta_{PL} \phi_{AB} G$$

(9)

Here, absorption of excitation light is determined by the number of luminescent centers in the excitation volume ($N_C$) and the absorption cross section $\sigma_{abs}$ per luminescent center. The luminescence further scales with the photoluminescence quantum yield $\eta_{PL}$ of the relevant thermometer emission lines and the temperature-dependent populations that define the fractions $\phi_{AB}$ of emission coming from A or B. The spectroscopic equipment sets the detection efficiency $\eta_{det}$. The integration time $t$, the excitation power used $P_{exc}$, and the EM gain $G$ can be chosen by the experimentalist.

In eq 9, we can distinguish the experimental factors ($\eta_{det}$, $P_{exc}$, $N_C$, and $G$) from the thermometer properties ($\sigma_{abs}$, $\eta_{PL}$, and $\phi_{AB}$). The experimental factors will depend on the available equipment and the type of sample. The freedom to choose a long $t$, high $N_C$, or high $P_{exc}$ may be restricted if the sample is not static, if the sample volume is small, or if strong excitation induces laser heating (which also depends on heat dissipation in the sample). The values of these parameters are not intrinsic thermometer properties but depend strongly on the application. Another factor that affects the temperature uncertainty is the emission wavelength of the thermometer because it determines the required type of detector and therefore the amount of dark current. Infrared detectors typically have a high dark current due to the small band gap of the photosensitive material, resulting in a relatively high uncertainty for infrared-emitting thermometers. Equation 9 also contains some intrinsic properties that can vary by orders of magnitude between different materials. We propose that a fair comparison between potential thermometer materials should consider these intrinsic properties. For practical applications, high $\sigma_{abs}$ and $\eta_{PL}$ for a given doping content $N_C$ and excitation power $P_{exc}$ are as important as a high thermal sensitivity of $\phi_{AB}$. These intrinsic properties of the thermometer determine the achievable signal compared to various application-related issues, like background fluorescence or blackbody radiation.

Studies of new thermometers often include measurements of the sensitivity, but experimental values of $\eta_{det}$ and $\sigma_{abs}$ are rare. Depending on the doping concentration, the synthesis procedure, and the excitation power, $\eta_{det}$ can vary over a few orders of magnitude, and it is therefore an important parameter to report. A well-established method to measure $\eta_{det}$ is to determine the number of absorbed and emitted photons of a sample using an integrating sphere. This has already improved the design and synthesis of thermometer materials. For example, recent studies on NaYF$_4$:Er$^{3+}$,Yb$^{3+}$ nanocrystals have optimized the quantum yield of NaYF$_4$:Er$^{3+}$,Yb$^{3+}$ in a range of excitation powers, reaching values comparable to bulk material.

Characterizing $\sigma_{abs}$ can be more challenging, especially for microcrystalline samples where strong light scattering prevents measuring optical absorption over a well-defined path length. We study a clear dispersion of Yb$^{3+}$-doped NaYF$_4$ nanocrystals with absorption spectroscopy to show that for a specific ion–host combination $\sigma_{abs}$ simply follows Lambert–Beer’s law (Figure 5a). We find a maximum value of $7.5 \times 10^{-3}$ cm$^{-1}$ at 977 nm, which matches literature values obtained from absorption measurements of single crystals and from the kinetics of upconversion luminescence. These values are an
Figure 5. Experimental methods to determine the absorption cross section. (a) Absorption spectrum of NaYF₄:Yb³⁺(18%) nanocrystals dispersed in cyclohexane. The spectrum shows a clear absorption band due to the ⁴F₉/₂ → ⁴F₅/₂ transition of Yb³⁺. Rayleigh scattering caused a background of roughly 0.0015 absorbance units, which was subtracted from the spectrum. The concentration of the bare nanocrystals without ligands is 10 mg/mL, which corresponds to a concentration of Yb³⁺ ions of 5.4 × 10¹⁸ cm⁻³. Taking the inhomogeneous dielectric surroundings of the Yb³⁺ ions inside the dispersed nanocrystal into account, this translates into an absorption cross section of σ₁₂ = 7.5 × 10⁻²¹ cm² at 977 nm. (b) Intensity of the Yb³⁺ luminescence measured on microcrystalline NaYF₄:Yb³⁺(18%) for various excitation intensities Iexc of 980 nm light (green dots). The solid black line is a fit of the experimental data to eq 10, which yields a σ₁₂ value of 7.8 × 10⁻²¹ cm² at 980 nm. The dashed line is a linear fit to the low-excitation-intensity data to clearly visualize the nonlinear trend of the high-excitation-intensity data.

In conclusion, we have characterized how experimental conditions affect the uncertainty of temperature measurements through (nano)thermometry based on luminescence intensity ratios. We first measured all noise sources associated with photon detection and developed statistical models to quantitatively predict the temperature uncertainty in a wide range of temperatures and for various experimental settings. We observed that enhancement of the luminescence signal by applying EM gain significantly reduces the uncertainty until readout noise is overcome. In addition, we studied the impact of background emissions, which is a realistic practical issue. Background increases the uncertainty of a temperature measurement even if it is properly subtracted from the measurement. Our work demonstrates that the temperature uncertainty is not an intrinsic property of a luminescent (nano)thermometer but instead strongly depends on the photodetector and measurement conditions. We propose a guideline of how to compare different thermometers in a way that is relevant irrespective of the spectroscopic equipment used or of the sample under consideration. Such new ways of comparing luminescent (nano)thermometers are essential to develop and choose the ideal thermometer for the desired application.

■ ASSOCIATED CONTENT

Supporting Information

The Supporting Information is available free of charge at https://pubs.acs.org/doi/10.1021/acsphotonics.2c00039.

Experimental procedure and additional results; experimental uncertainties at high excitation intensities; analytical expressions for the probability distributions of electron multiplication; and simulations of the uncertainties introduced by electron multiplication (PDF)

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Notes

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REFERENCES

(1) Zhu, X.; Li, J.; Qu, X.; Liu, Y.; Feng, W.; Li, F. Upconversion Nanocomposite for Programming Combination Cancer Therapy by Precise Control of Microscopic Temperature. Nat. Commun. 2018, 9, No. 2176.

(2) van Swieten, T. P.; van Omme, T.; van den Heuvel, D. J.; Vonk, S. J. W.; Spruit, R. G.; Meier, F.; Garza, H. H. P.; Weckhuysen, B. M.; Meijerink, A.; Rabouw, F. T.; Geitenbeek, R. G. Mapping Elevated Temperatures with a Micrometer Resolution Using the Luminescence of Chemically Stable Upconversion Nanoparticles. ACS Appl. Nano Mater. 2021, 4, 4208–4215.

(3) Hartman, T.; Geitenbeek, R. G.; Wondergem, C. S.; van der Stam, W.; Weckhuysen, B. M. Operando Nanoscale Sensors in Catalysis: All Eyes on Catalyst Particles. ACS Nano 2020, 14, 3725–3735.

(4) Brites, C. D. S.; Balabhadra, S.; Carlos, L. D. Lanthanide-Based Thermometers: At the Cutting-Edge of Luminescence Thermometry. Adv. Opt. Mater. 2019, 7, No. 1801239.

(5) Maturi, F. E.; Brites, C. D. S.; Ximendes, E. C. Mills, C.; Olsen, B.; Jaque, D.; Ribeiro, S. J. L.; Carlos, L. D. Going Above and Beyond: A Tenfold Gain in the Performance of Luminescence Thermometers Joining Multiparametric Sensing and Multiple Regression. Laser Photonics Rev. 2021, 15, No. 2100301.

(6) Brites, C. D. S.; Fuertes, M. C.; Angelome, P. C.; Martínez, E. D.; Lima, P. P.; Soler-Illia, G. J. A.; Carlos, L. D. Tethering Luminescence Thermometry and Plasmonics: Light Manipulation to Assess Real-Time Thermal Flow in Nanoarchitectures. Nano Lett. 2017, 17, 4746–4752.

(7) Back, M.; Trave, E.; Ueda, J.; Tanabe, S. Ratiometric Optical Thermometer Based on Dual Near-Infrared Emission in Cr3+-Doped Bismuth-Based Gallate Host. Chem. Mater. 2016, 28, 8347–8356.

(8) Shen, Y.; Lifante, J.; Fernandez, N.; Jaque, D.; Ximendes, E. In Vivo Spectral Distortions of Infrared Luminescence Nanothermometers Compromise Their Reliability. ACS Nano 2020, 14, 4122–4133.

(9) Pickel, A. D.; Teitelboim, A.; Chan, E. M.; Borys, N. J.; Schuck, P. J.; Dames, C. Appetant Self-Heating of Individual Upconverting Nanoparticles (Yb3+/Tm3+). ACS Nano 2020, 15, No. 4847–4854.

(10) Ru, X.; Akerboom, S.; Jong, M.; Mutikainen, I.; Tanase, S.; Meijerink, A.; Bouwman, E. Mixed-Lanthanoid Metal-Organo Framework for Ratiometric Cryogenic Temperature Imaging. Inorg. Chem. 2015, 54, 11323–11329.

(11) McMullin, E. J.; Braddock, L. R.; Gamelin, D. R. Dual-Emitting Nanoscale Temperature Sensors. Chem. Mater. 2013, 25, 1283–1292.

(12) Homann, C.; Kruikwett, L.; Frenzel, F.; Grauel, B.; Würth, C.; Resch-Genger, U.; Haase, M. NaYF4:Yb/Er/NaF Core/Shell Nanocrystals with High Upconversion Luminescence Quantum Yield. Angew. Chem., Int. Ed. 2018, 57, 8765–8769.

(13) Berghede, C. A.; Hoek, W.-K.; Mikulec, F. V.; Bawendi, M. G. On the Absorption Cross Section of CdSe Nanocrystal Quantum Dots. J. Phys. Chem. B 2002, 106, 7619–7622.

(14) De Roo, J.; Ibáñez, M.; Geiregat, P.; Pedentucu, G.; Walravens, W.; Maex, J.; Martins, J. C.; Van Driessche, I.; Kovalenko, M. V.; Heeney, M. Highly Ligated Binding and Light Absorption Coefficient of Cesium Lead Bromide Perovskite Nanocrystals. ACS Nano 2016, 10, 2071–2081.

(15) Xia, C.; Wu, W.; Yu, T.; Xie, X.; Van Oversteeg, C.; Gerritsen, H. C.; de Mello Donega, C. Size-Dependent Band-Gap and Molar Absorption Coefficients of Colloidal CunSn5 Quantum Dots. ACS Nano 2018, 12, 8350–8361.

(16) Saleta Reig, D.; Grauel, B.; Konysshkin, V. A.; Nakladov, A. N.; Fedorov, P. P.; Busko, D.; Howard, I. A.; Richards, B. S.; Resch-Genger, U.; Kuznetsov, S. V.; Turanov, A.; Würth, C. Upconversion Properties of SrF2:Yb3+,Er3+ Single Crystals. J. Mater. Chem. C 2020, 8, 4093–4101.

(17) Page, R. H.; Schaffers, K. I.; Waide, P. A.; Tassano, J. B.; Payne, S. A.; Kruipke, W. F.; Bawendi, M. G. Upconversion-Pumped Luminescence Efficiency of Rare-Earth-Doped Hosts Sensitized with Trivalent Yterbium. JOSA B 1998, 15, 996–1008.

(18) Hassan, M. Y.; Hor, A.; Luo, Q.; Smith, S. J.; May, P. S.; Berry, M. T. Explaining the Nanoscale Effect in the Upconversion Dynamics of β-NaYF4:Yb3+,Er3+ Core and Core–Shell Nanocrystals. J. Phys. Chem. C 2017, 121, 16592–16606.

(19) Sui, G.; Chen, B.; Zhang, X.; Li, X.; Zhang, J.; Xu, S.; Sun, J.; Cao, Y.; Wang, X.; Zhang, Y.; Zhang, Y. Radiative Transition Properties of Yb3+ in Er3+/Yb3+ Co-Doped NaYF4 Phosphor. J. Alloys Compd. 2020, 834, No. 155242.

(20) Zhang, Y.; Chen, B.; Xu, S.; Li, X.; Zhang, J.; Sun, J.; Zhang, X.; Xia, H.; Hua, R. A Universal Approach for Calculating the Judd–Ofelt Parameters of RE3+ in Powdered Phosphors and Its Application for...
the $\beta$-NaYF$_4$:Er$^{3+}$/Yb$^{3+}$ Phosphor Derived from Auto-Combustion-Assisted Fluoridation. *Phys. Chem. Chem. Phys.* 2018, 20, 15876–15883.

(37) Nanda, J.; Ivanov, S. A.; Htoon, H.; Bezel, I.; Piryatinski, A.; Tretiak, S.; Klimov, V. I. Absorption Cross Sections and Auger Recombination Lifetimes in Inverted Core-Shell Nanocrystals: Implications for Lasing Performance. *J. Appl. Phys.* 2006, 99, No. 034309.

(38) van de Haar, M. A.; Tachikirt, M.; Berends, A. C.; Krames, M. R.; Meijerink, A.; Rabouw, F. T. Saturation Mechanisms in Common LED Phosphors. *ACS Photonics* 2021, 8, 1784–1793.

(39) Krämer, K. W.; Biner, D.; Frei, G.; Güdel, H. U.; Hehlen, M. P.; Lüthi, S. R. Hexagonal Sodium Yttrium Fluoride Based Green and Blue Emitting Upconversion Phosphors. *Chem. Mater.* 2004, 16, 1244–1251.

(40) Rabouw, F. T.; Prins, P. T.; Villanueva-Delgado, P.; Castelijns, M.; Geitenbeek, R. G.; Meijerink, A. Quenching Pathways in NaYF$_4$:Er$^{3+}$,Yb$^{3+}$ Upconversion Nanocrystals. *ACS Nano* 2018, 12, 4812–4823.