Implantable photonic probes are of increasing interest to the field of bio-photonics and in particular, optogenetic neural stimulation. Active probes with onboard light emissive elements allow for electronic multiplexing and can be manufactured through existing micro-electronics methods. However, as the optogenetics field moves towards clinical practice, an important question arises as to whether such probes will cause excessive thermal heating of the surrounding tissue. Light emitting diodes typically produce more heat than light. The resultant temperature rise of the probe surface therefore needs to be maintained under the regulatory limit of 2°C. This work combines optical and thermal modelling, which have been experimental verified. Analysis has been performed on the effect of probe/emitter geometries, emitter, and radiance requirements. Finally, the effective illumination volume has been calculated within thermal limits for different probe emitter types and required thresholds.

**KEYWORDS**
brain implant, neuroprosthesis, optoelectronics, optogenetics, optrode, prosthesis

**1 | INTRODUCTION**

Optogenetics is a gene therapy technique to incorporate light-sensitive proteins into cells. The field began with demonstration of channelrhodopsin-2 (ChR2) proteins ectopically expressed in frog oocytes in 2003 [1]. ChR2 is a light-sensitive cation channel and therefore allows optical depolarisation (stimulation) of nerve cells. Complementary to light-sensitive ion channels are light sensitive pumps such as the chloride pump—halorhodopsin (NpHR) [2]. NpHR acts to pump hyperpolarising (inhibiting) chloride ions into the nerve cell, thus reducing or preventing nerve cell activity. These techniques have been demonstrated in many neuroscience studies, for example, by Busskamp et al. [3]. Perhaps more significantly to the biomedical engineering community, the technique is now moving towards clinical practice, with interest in retinal prosthesis [4, 5], visual brain prosthesis, and brain pacemaking activity for conditions such as epilepsy. At the time of writing, there are trials underway for optogenetic retinal prostheses, and various groups are starting the process of developing clinically approved devices for more invasive brain prostheses for conditions such as epilepsy.
For typical laboratory experiments, light delivery can be achieved via laser-coupled optical fibres. Being external, power consumption and thermal dissipation is not an issue to such devices if the intensities are kept to moderate levels above the stimulus threshold. In contrast, technology for chronic long-term implants in freely moving animals or medical neuroprostheses is still relatively immature. Two primary sources of “optrodes”—the optical equivalent of the electrode—have been explored: Waveguiding [6, 7] and local photogeneration [8, 9]. The former generates light far from the tissue and then delivers it to the target via a delivery system with multiple light guides. The latter aims to generate light on the implantable structure.

The key caveat of optogenetics from the perspective of light delivery systems is that the illumination required to stimulate neural tissue is very high. Cultured cells encoded with wild-type ChR2 typically require photon fluxes of 1017 to 1019 photons/s/cm² (0.4-40 mW/mm²) to achieve a full dynamic range of photocurrent response [10–12]. Classically, reliable neural firing at 20 Hz has required 0.7 mW/mm² [11]. There has also been some evidence that in vivo, the requirement is somewhat less. For example, Bi et al. [4] noted an S-curve dynamic range of photocurrent with the logarithm of irradiance between 1016 and 1019 photons/s/cm² (40 μW/mm²-40 mW/mm²) in retinal ganglion cells. Lagali et al. [13] noted a range of 1014 to 1017 photons/s/cm² (4-400 μW/mm²) in retinal bipolar cells.

The resultant emission requirement can put a strain on the probe photonics. Thus, there has been a considerable effort in exploring the biophysics of ChR2 [10, 12] and developing more efficient variants. One example is the CatCh [11] variant of ChR2, which utilises a Ca2+ feedback mechanism to bring sensitivity down to 1016 to 1017 photons/s/cm². Another example is melanopsin which is connected to a G-protein cascade giving amplification and resulting in a sensitivity of 1012 photons/s/cm² [14]. Neither of these is ideal—the Ca2+ permeability of CatCh may interfere with cell metabolism and the response time of melanopsin is very long (~10 seconds). However, they demonstrate that we can expect to see a range of required thresholds in the future depending on the opsin used and the target tissue type.

Probes with light emissive elements have an advantage over light guided systems in that they can perform onboard multiplexing. In particular, probes with in-built control circuits [15, 16] hold the potential for large-scale multi-emissive systems. However, from an engineering perspective, no light emitter is 100% efficient. In fact, typical efficiencies range from about 5% [9, 17, 18] for micro-light emitting diodes (μLEDs) to 80% [19] for best in class LEDs. There is, therefore, a danger that such probes will form surface hot spots that exceed the 2°C regulatory limit if driven too hard. An alternative perspective of the same issue is that for the given thermal limit, their radiance may not achieve sufficient stimulation at significant penetration depths.

The key question, therefore, arises as to how far light can penetrate into tissues for a given thermal limit. McAlinden et al. [9] discussed the optical and thermal characterisation of their respective monolithic gallium nitride optrode. Similarly, Wu et al. [18] demonstrated and characterised a silicon probe with integrated micro-light emissive elements. However, a comprehensive analysis of thermal effects for different substrate configurations and dimensions, and opsin sensitivities has not yet been performed.

In this work, we combine an optical scattering model of light propagation in tissue [20, 21] with a finite element model of tissue heating effects. We calibrate these with experimental data from real probes and then simulate a concept probe to explore key questions in the field:

1. What are the light penetration depth and volume for a given emission power and different LED geometries and emission profiles?
2. What is the maximum energy dissipation that may be allowed while keeping surface hotspots within acceptable limits?
3. Given the 2 parameters above, what volume of tissue can be illuminated for different levels of gliosis around long-term implants?

We hope this work can provide the basis for future optrode designs based on active light emissive modules.

2 | METHODOLOGY

Our objective is to use optical and thermal modelling to explore the interplay between illumination and thermal profiles for different emitter geometries and probe architectures. In particular, we aim to determine the maximum optical penetration depth and volume for a given thermal limit. To achieve this, we model the effect of light scattering in tissue with a 4-state biophysical model of the photoresponse of the ChR.

2.1 | Tissue thermal limit

In 1989, Lamanna et al. [22] performed acute experiments on anaesthetized rats and suggested that temperature should not increase more than 1°C above ambient. This threshold has since been used in over 60 other studies [9, 23]. However, the 1°C limit refers to the threshold where the authors started to see abnormal signalling. They did not observe actual tissue damage while transient increases were kept below 5°C. Similarly, Matsumi et al. [24] and Fuji and Ibata [25], respectively, inserted heated probes into non-human primate brains and independently determined the thermal limit of 43°C, that is, ΔT = 6°C. More recently in 2012, Opie et al. [26] looked at ex vivo tissue and determined 38.7°C as the threshold at which damage becomes apparent in the retina. But, perhaps the retina is less prone to thermal
damage than brain tissue due to its various mechanisms to protect itself from phototoxity. Goldstein et al. [27] observed cell death in neuronal cells of the central nervous system after 60 minutes at temperatures as low as 40.5°C.

The specific condition of these studies is that they are acute, with sacrifice and histology a few days after the experiment. There are very few chronic long-term studies of the effect of raised temperatures. Seese et al. [28] and Okazaki et al. [29] demonstrated that ΔT = +2°C is tolerated in muscle. However, the long-term effects in the up-regulation of heat shock proteins, and glial response in the central nervous system has not been published. This is a challenge that needs to be explored in more detail by the biology community.

It is therefore perhaps not surprising that the regulatory guidance is also limited. Directive 93/42/EEC simply states: “Devices must be designed and manufactured in such a way as to remove or minimise as far as is possible: <temperature rise>”. The American Association of Medical Instrumentation (AAMI) recommend a limit of ΔT = +2°C, which seems in keeping with the review above, that is, T ≤ 39°C.

We, therefore, suggest 2 limits: (1) for basic experimental neurobiology, the evidence suggests that a 1°C limit would prevent thermally induced activity. (2) For translational efforts, the regulatory limit is 2°C for any surface hotspot.

2.2 Thermal transmission modelling

The flow of heat through different media can be represented by a series of linear equations which can be solved using finite element analysis. Assuming the whole optrode is floating in the tissue solution, the temperature rise of the optrode can be predicted by solving the time-resolved heat equation:

\[ ρ C \frac{∂T}{∂t} = ∇ \cdot (k ∇ T) + Q_0, \]

where \( T \) is the temperature, \( t \) is the time, \( ρ \) is the medium density, \( C_p \) is the specific heat capacity, \( k \) is the thermal conductivity (assumed to be a constant across a single medium), and \( Q_0 \) is the heat source.

There are a number of potential heat sources in an optrode. If it contains active electronic components, then these can dissipate heat wherever there is significant power consumption. However, for this work, we consider the μLED as the primary heat source with defined thermal emissions for given time periods and repetition rates. This is because its power consumption is in the milliwatt range, whereas implantable circuitry typically has significantly lower power consumption.

We consider the μLED as an even surface heat source so \( Q_0 = Q \). The Neumann boundary condition can then be applied to the surface heat source:

\[ -n \cdot (-k ∇ T) = Q_0, \]

where \( n \) is the normal directional vector to the surface of the emitter. Working with such models, in 2007, Kim et al. [23] demonstrated modelling of thermal dissipation through Utah arrays (8 × 8 array of penetrating silicon probes) in both phantom media and cat cortex. They found that 13 mW of power dissipation (≈81 μW/mm²) resulted in a temperature rise of 0.43°C. McAlinden et al. [9] also explored the thermal dissipation relative to optical output for an all-gallium nitride-based optrode. Their model produced a temperature rise of 1°C over 10 milliseconds at an output radiance of 600 mW/mm² and a final (saturated) temperature rise of 1.5°C.

There are, however, some further complications. The first is that in addition to passive flow, there is also an active heatsinking effect from vasculature in the cortical tissue. This removes thermal energy at a higher rate than would otherwise be expected. For example, Kim et al. [23] found the temperature rise to be 22% lower in cortical tissue than in a passive phantom construct. Opie et al. [26] found the temperature rise to be up to 6 times lower.

The second complication is that it was recently shown by Stujenske [30] that light absorbed by the tissue could raise its temperature. We thus modify the Pennes’ model [31] of bio-heat transfer to include the optical absorption effect:

\[ ρ C \frac{∂T}{∂t} = ∇ \cdot (k ∇ T) + ρ_b C_b (T - T_b) + Q_m + Q_a(x, y, z, t), \]

where the second, third and fourth term on the right of the equation denotes heat dissipated through the blood vessels, metabolic heat (\( Q_m \)) and heat generated from light absorption (\( Q_a \)), respectively. \( P_b, C_b, w_b \) and \( T_b \) denote the density, heat capacity, average perfusion velocity and temperature of the blood in the vessels, respectively. To solve [3], we utilise COMSOL Multiphysics (V5 with the [bio]heat transfer module) to perform the thermal analysis.

We developed 2 probe models shown in Figure 1A,B. Panel (A) is a model of a custom-made probe we have developed to calibrate the simulations. Panel (B) is a model of a general purpose multi-LED concept probe which we would foresee being utilised in clinical practice.

We envisage the concept probe either as a solitary device or as part of an array, typically having a spacing of around 1 mm. The probe head would need a height of 2 mm or less to fit in the skull. As such, we chose exemplar head dimensions of 1 x 2 mm. Such an area is sufficient to accommodate multiplexing electronics, amplifiers and digital to analog conversion [15, 16]. The shaft would typically have both LEDs and recording electrodes. An exemplar of a comparable recording-only probe is the Neuropixels probe [32]. As the human cortex is between 2 and 3.5 mm in length [33], we determine the length of this probe to be 5.5 mm to allow for gliosis effects pushing the probe up slightly from the cortex and the thickness of any mechanical baseplate.
For simplicity, we consider 8 LEDs and recording sites for this probe to interact with each of the 6 cortical layers with some redundancy. It should be noted that for comparison, some high-density probes, for example, by Lopez et al. [34], and Angotzi et al. [35] have hundreds of recording sites. We also simulate each probe as an individual entity. It may well be that probes are arranged in a 4 × 4 or 8 × 8 penetrating matrix. In such a scenario, there would be a larger surface area from which to dissipate heat, and possibilities of heat transfer between probes. However, we feel this would complicate the analysis. Furthermore, from our simulations, the most important hotspots are local to the area around individual LEDs, and the probe shaft. Thus, we focus on the individual optrode configuration shown in Figure 1B.

The dimensions of the exemplar probe, plus a custom test probe used in the experimental validation are given in Table 1. Where dimensions are defined as a constant, we have used a specific value. Where dimensions are defined as a variable, we have explored the effect of different dimensions on the thermal outcome. In these simulations, we use the same value for width and thickness, which means that the optrode shaft has a square shape. It should be noted that this may be different for future probes which may be thinner with respect to their width. We only simulate down to thicknesses of 100 μm. However, for even thinner probes, we would expect the greater surface area to bulk ratio to support higher dissipation for long-term heat transfer. Simultaneously, for short pulses, a reduced bulk would have a reduced capacity to temporarily store (smooth) spikes in heat generation.

The layer structure of the optrode is given in Figure 1C. We assume that active emitting optrodes will have a silicon core with metal control lines and emitter (typically a gallium nitride μLED). The metal control lines are typically coated with a silicon dioxide dielectric. For CMOS (complementary metal oxide semiconductor) devices, there would also be a thin silicon oxy-nitride top layer. But as it is thin, and its thermal conductivity is much higher than silicon dioxide or the polymer encapsulation layer, its effect will be negligible compared to its neighbouring layers. We thus ignore this layer.

The whole structure will be coated with a polymer encapsulant which acts as a biocompatible interface. Polydimethylsiloxanes (PDMSs/silicone) and parylene-C, epoxies and urethanes are typical choices for encapsulation. These are chemically different and vary considerably in glass transition temperature. But for the operational range of these devices, their thermal properties are very similar. We

---

**Table 1** Probe dimensions for the thermal modelling

| Optrode element          | Dimension (μm) | Type |
|--------------------------|----------------|------|
| Head width               | 3000           | 1000 | Const |
| Head length              | 4000           | 2000 | Const |
| Head substrate thickness | 200            | 100-400 | Var |
| Shaft length             | 4000           | 5500 | Const |
| Shaft width              | 400            | 100-400 | Var |
| Shaft substrate thickness| 200            | 100-400 | Var |
| Emitter diameter         | 240 × 320      | 20-120 | Var |
| Electrode diameter       | 50             | 50   | Cons |
have thus used representative values, for “polymer” in Table 3. For the test probe, we used PDMS as the primary encapsulant.

The dimensions of the individual layers simulated are given in Table 2.

The material parameters of the constituent layers used in the thermal analyses are shown in Table 3.

For optogenetics, short optical pulses (<100 milliseconds) are generally used rather than continuous illumination. So, the primary concern is the transient temperature change from the device. \( Q_0 \) (the heat source) is determined by the inefficiencies in the optical generation by the LED. This can be given by

\[
Q_0 = (1 - \eta_{\text{LED}}) \cdot V_{\text{LED}}I_{\text{LED}},
\]

where \( \eta_{\text{LED}} \) is the efficiency of the LED which denotes the ratio between optical power output and electrical power input; \( V_{\text{LED}} \) and \( I_{\text{LED}} \) are the bias voltage and bias current, respectively.

A fixed 37°C is used as the boundary condition, ignoring metabolic heat generation. For the test probe described in Figures 1 and 9A, we used mini LEDs with an efficiency of around 30% [36], that is, 70% of the applied electrical energy would convert to heat.

We set continuous boundary conditions for inner boundaries, while for outer boundaries we set a heat convection boundary condition (Eq. (5)):

\[
-n \cdot (-k \nabla T) = h \cdot (T_{\text{ext}} - T),
\]

where \( T_{\text{ext}} \) is the temperature usually being constant for the external environment and \( h \) is the heat transfer coefficient of the boundary material. For the tissue boundary in the model, we set \( h \) at 1000 W/(m² K) as this would be a typical value [37].

2.3 Optical systems

Opsin absorption can be tuned from the near ultraviolet across to the red region of the visual spectrum. Peak

![Related Emission Phase](image)

**TABLE 2** Probe layer thicknesses for the thermal modelling

| Optrode element          | Thickness (μm) | Type     |
|---------------------------|----------------|----------|
| Polymer encapsulation     | 10-100         | Variable |
| Silicon core              | 100-300        | Variable |
| GaN μLED                  | 1-100          | Variable |
| Al driving track          | 1              | Constant |
| SiO2 passivation          | 5              | Constant |
| Au electrode              | 1              | Constant |

Abbreviation: μLED, micro-light emitting diode.

**TABLE 3** Simulation parameters for the thermal modelling

| Medium    | \( k \), W/(m K) | \( \rho \), kg/m³ | \( C_p \), J/(kg K) |
|-----------|------------------|------------------|-------------------|
| Tissue    | 0.6              | 1057             | 3600              |
| GaN       | 200              | 6150             | 485               |
| Sapphire  | 42               | 400              | 854.8             |
| SiO2      | 1.5              | 2500             | 966               |
| Silicon   | 149              | 2329             | 706.75            |
| Gold      | 315              | 1930             | 129               |
| Polymer   | 0.15             | 965              | 1460              |
absorption varies between 470 and 590 nm. A full review of the wavelength selectivity of different opsins can be found in the review by Kurihara and Sudo [38]. In this range, micro-lasers are still relatively immature, so we primarily focus on LEDs. In particular, we consider gallium nitride (typical peak ~470 nm)—the primary technology used in modern high radiance lighting.

The illumination penetration profile into tissue is largely dependent on the total amount of light being emitted. As light is generated by electron-hole recombination, this means higher drive current. Smaller LEDs will, therefore, require higher drive current densities to achieve the same emission profile. However, this means that they are more susceptible to “droop.” Droop is a phenomenon well-documented within the LED community whereby the efficiency has an inverse relationship with the current density. It was recently attributed to Auger recombination by Iveland et al. [39].

Figure 2A combines data from our previously presented μLEDs [17] with the efficiency profile of commercial mini-LEDs (Cree Corp DA2432 [36]). The latter is both (currently) the smallest commercially available LED and the largest (size: 320 × 240 μm, emission surface 280 × 210 μm) conceivable LED that could be placed on an optrode. We also include the profile taken from the best efficiency we have seen in the published literature (Narukawa et al. [19]).

The micro-LED efficiency (3%-5%) is comparable to that presented by McAlinden et al. (3%-5%) [9] but significantly higher than that presented by Wu et al. [18] (0.8%). It should be noted that none of these studies used integrating spheres for their measuring technique. So, the efficiency may be an underestimation. From the micro/mini LEDs, we created a model for the full range of current densities on the basis of a curve fit. The output of this model for different LED sizes and input powers can be seen in Figure 2B.

LEDs, however, cannot be used in isolation. There is a small overhead due to the transistor driving electronics, that is, the drive transistors typically require a combined source-drain voltage drop of 0.5 to 1 V, compared to 2.5 to 3.5 V for the operation of the LED. If there are significant contact and/or line resistances, these will take additional voltage drops. As such, we assume LED system efficiencies (ie, including the driving transistors) to be around 75% of the LED only efficiency. If the driving electronics are elsewhere, this perhaps is only interesting to the battery life. However, if the driving circuitry is also incorporated into the silicon of the optrode—that is, an active optrode, then it needs to be included.

2.4 Emitter emission profile

The emission profile—that is, how the light spreads from the LED upon emission—is an important determinant in the light penetration profile. We consider 3 types of emission profile, as shown in Figure 2C.
determined by the scattering phase with the updated weight until the weight decreases below a reasonable extinction threshold.

Therefore, the light diffusion profile in the tissue for the emitter can be obtained from the statistic of numerous photon behaviours.

2.5.3 Combined scattering

The scattering mismatch in the brain tissue results from a heterogeneous arrangement of cellular and sub-cellular structures of various shapes and sizes. For simplicity we assume the various scattering structures to be spherical particles of different sizes floating in the host medium. The total scattering results from a mixture of Mie scattering and Rayleigh scattering [20].

We assume the refractive index of the host medium and the scattering particles are \( n_e \) and \( n_p \), respectively. We assume \( n_p \) to be complex, \( n_p = n_p' + i n_p'' \), and \( n_e \) to be real. We use \( \rho_e \) and \( \lambda \) to denote the volume density of the scattering particles, and the operating wavelength in the vacuum, respectively. The main refractive index contrast \( m = n_p/n_e \).

We also use \( d_p \) as the diameter of the scattering particle, but it is diverse for Mie and Rayleigh scattering. We can define the size contrast \( \alpha \) in the brain tissue as:

\[
\alpha = \pi n_e d_p/\lambda.
\] (8)

\( \alpha \ll 1 \) for Rayleigh scattering and \( \alpha \sim 1 \) for Mie scattering. As the size distribution in the brain tissue is complicated, for simplicity, we assume \( d_p = d_{\text{Mie}} \) for Mie scattering and \( d_p = d_{\text{Rayleigh}} \) for Rayleigh scattering where \( d_{\text{Mie}} \) and \( d_{\text{Rayleigh}} \) are the average diameters for Mie and Rayleigh particles, respectively.

We assume \( 0 \leq f_{\text{Mie}} \leq 1 \) is the fraction of Mie scattering so that \( 0 \leq f_{\text{Rayleigh}} \leq 1 \) is the fraction of Rayleigh scattering and the fractions satisfy: \( f_{\text{Mie}} + f_{\text{Rayleigh}} = 1 \) [20].

Thus, the scattering and absorption coefficients are weighted summations of Mie and Rayleigh coefficients:

\[
\mu_s = f_{\text{Mie}} \mu_s^{(\text{Mie})} + f_{\text{Rayleigh}} \mu_s^{(\text{Rayleigh})},
\] (9)

\[
\mu_a = f_{\text{Mie}} \mu_a^{(\text{Mie})} + f_{\text{Rayleigh}} \mu_a^{(\text{Rayleigh})}.
\] (10)

The total phase function of the tissue is also the combination of Mie and Rayleigh scattering:

\[
p(\theta) = f_{\text{Mie}} \cdot p_{\text{Mie}}(\theta) + f_{\text{Rayleigh}} \cdot p_{\text{Rayleigh}}(\theta).
\] (11)

2.5.4 Mie and Rayleigh scattering

The Mie scattering coefficient can be derived from the Mie-Lorentz theory [21]:

\[
\mu_s^{(\text{Mie})} = \frac{\rho_e d_p^2}{2\pi} \sum_{n=1}^{\infty} (2n+1) \left( |a_n(m,\alpha_{\text{Mie}})|^2 + |b_n(m,\alpha_{\text{Mie}})|^2 \right),
\] (12)

where \( a_n \) and \( b_n \) are the Mie coefficients, and in practice, we usually use \( \alpha_{\max} = \alpha + 4\alpha/3 + 2 \) as the maximum limit of the summation. As defined before, \( m \) is the main refractive index contrast and \( \alpha_{\text{Mie}} = \pi n_e d_{\text{Mie}}/\lambda \) is the main size contrast in the Mie scattering.

For Rayleigh scattering, we have \( \alpha_{\text{Rayleigh}} = \pi n_p d_{\text{Rayleigh}}/\lambda \ll 1 \), where \( d_{\text{Rayleigh}} \) is the diameter of the Rayleigh particle. So, the Rayleigh cross-section can be calculated by the following equation:

\[
\mu_s^{(\text{Rayleigh})} = \frac{2\pi^5 d_p^6}{3\lambda^4} \left( \frac{n_e^2-1}{n_p^2+2} \right)^2 \lesssim \frac{2\pi}{3} \left( \frac{n_e^2-1}{n_p^2+2} \right)^2 \frac{\lambda^2}{n_p^5 \pi^2}. \] (13)

The Rayleigh scattering coefficient is usually 2 to 3 orders of magnitude lower than the Mie scattering coefficient in tissue. The Mie extinction coefficient can then be calculated using Mie-Lorentz theory [21]:

\[
\mu_e^{(\text{Mie})} = \frac{\rho_e d_p^2}{2\pi} \sum_{n=1}^{\infty} (2n+1) \left( |a_n(m,\alpha)| + |b_n(m,\alpha)| \right). \] (14)

Thus, we have \( \mu_a = \mu_a^{(\text{Mie})} - \mu_a^{(\text{Rayleigh})} \). Assuming the absorption cross-section is proportional to the area of the particle cross-section, we have: \( \mu_a^{(\text{Mie})}/\mu_a^{(\text{Rayleigh})} = (d_{\text{Mie}}/d_{\text{Rayleigh}})^2 \). So, we obtain the total absorption coefficient \( \mu_a^{(\text{Mie})} \):

\[
\mu_a = \mu_a^{(\text{Mie})} \cdot \left( f_{\text{Mie}} + f_{\text{Rayleigh}} \cdot \left( \frac{d_{\text{Rayleigh}}}{d_{\text{Mie}}} \right)^2 \right) \cdot \left( f_{\text{Mie}} f_{\text{Rayleigh}} \right). \] (15)

The Mie phase function \( p_{\text{Mie}}(\theta) \) can be obtained in the Mie-Lorentz scattering theory [21]:

\[
p_{\text{Mie}}(\theta) = \frac{|S_1(\theta)|^2 + |S_2(\theta)|^2}{4\pi \cdot \sum_{n=1}^{\infty} (2n+1) \left( |a_n|^2 + |b_n|^2 \right)}, \] (16)

where \( S_1(\theta) \) and \( S_2(\theta) \) are the scattering amplitudes of a certain scattering angle \( \theta \). More details of the derivation of these intermediate parameters with the involvement of spherical Bessel functions can be found in the paper by Bohren and Huffman [21]. The Rayleigh phase function \( p_{\text{Rayleigh}}(\theta) \) is proportional to \( 1 + \cos^2 \theta \).

2.5.5 Parameters

Table 4 gives the main simulation parameters in the optical analysis. Among these parameters the operating wavelength \( \lambda \) is usually determined by the opsin we use, typically 470 nm for ChR2.

The main refractive index contrast of the brain tissue can be the difference between the cytoplasmic solvent and unsolvable lipid particles [44]. Drezek et al. [44] have reported that the refractive index of tissue varies from 1.36 of the extracellular fluid to 1.7. The measured refractive index of the scattering particles in human tissue is from 1.4 to 1.48 for visible light [45]. So, we set \( n_e \) and \( n_p \) to 1.36 and 1.48, respectively.
The Mie particle diameter $d_{\text{Mie}}$ should vary from 0.1 to 1 μm [50]. For Rayleigh particle the size contrast $\alpha \ll 1$, so we assume $d_{\text{Rayleigh}} = 0.01$ μm.

The experimentally measured optical properties of the brain tissue including $\mu_s$, $\mu_a$, $\mu_e$ and $g$ are reported by Yaroslavsky et al. [43] and Taddeucci et al. [51] before. In the grey matter, for the typical operating wavelength, the measured scattering coefficient $\mu_s$ varies from 7.8 to 12.5/mm; the measured asymmetry factor $g$ varies from 0.87 to 0.9. The absorption coefficient $\mu_a$ is 2 to 3 orders of magnitude smaller than $\mu_s$ so light diffusion profile in the tissue is not sensitive to $\mu_a$. The particle volume density $\rho_p$ should be between 1 and $10 \times 10^{-10}$ mol/L. Given these statements we estimated the input parameters.

The scattering-absorption ratio $\mu_t/\mu_s$ is no more than 0.01, so we assume this to be 0.01 around the responding peak of the opsins. Thus, the imaginary part of the refractive index $n_p''$ can be derived. The specific $\rho_p$, $f_{\text{Mie}}$, and $f_{\text{Rayleigh}}$ are also derived from experimentally reported values for the tissue.

2.6 | Tissue deadzone

Reactive gliosis is a common pathological feature which commonly encapsulates surgical implants in neural tissue, for example, deep brain stimulators [52]. Specifically, a thin capsule of glial cells containing inflammatory cells such as macrophages and lymphocytes forms around the device [52]. Szarowski et al. demonstrated that when neural prosthetic devices were implanted in rats, glial encapsulation has been shown to extend up to 100 μm around the device, 12 weeks post-insertion [53]. Similar patterns of long-term gliosis have also been reported in non-human primates, 3 years post-implantation of electrodes [54].

The issue with such gliosis regions is that any neurons contained within are effectively dead or inactive, and therefore no longer contributes to the surrounding neural circuits. As such any stimulus to be effective must be effective beyond this region.

Such effects are not seen in acute experiments. So, for example, although Wu et al. [18] demonstrated light penetration of only 50 μm (beyond the ChR2 threshold), they could still see strong neural responses. However, for chronic long-term implants, this needs to be considered. For example, if the light penetration above the ChR2 threshold was only 100 μm, and the gliosis thickness was 100 μm, no net stimulus could be expected.

So, to calculate the volume of stimulus, we define the glial region as the “dead zone,” that is, no effect from the light. We then look at an optimistic case (100 μm), and pessimistic case (200 μm) for dead zone thickness.

3 | EXPERIMENTAL VALIDATION

3.1 | Optothermal measurement of optrodes

We developed a test optrode according to the dimensions in Table 1. It consists of a simple silicon shaft and head. At its tip is a single CREE DA2432 mini LED [36]. Titanium-gold lines connect the LED to the wire-bonded connectors in the head. For experimentation, the optrode was placed on a printed circuit board (PCB) ceramic base.

The LED on the optrode was operated with large current ranges from 10 to 25 mA. This is around 20 times the range we would expect in practice (ie, 0.5-1.5 mA). However, such a large range allows for accurate measurement with our experimental equipment. Measurements of the optrode surface temperature in air were taken with an Optris micromodule thermal camera (Optris GmbH, Berlin, Germany). The measurements were taken in an enclosed space with the baseline air temperature measured at a constant 28°C. Both continuous wave (CW) (up to 15 minutes) and pulses were tested to show a broad range of the thermal effect on the optrode.

The optrode devices can be seen in Figure 3. Figure 3A shows LED emission from a bare LED optrode (left) and an encapsulated optrode (right). Figure 3B,C shows the optrode from the perspective of a thermal camera at 2 different temperatures. The apparently warmer head part of the optrode is a reflectance artefact. The actual temperature at the tip could be determined by the change in temperature of the specific pixels with LED activity.

3.2 | Light penetration measurement of tissue

To ensure the validity of the optical modelling we performed experiments to compare with the modelling results. We, therefore, created an experiment with a light guided cannula which was inserted laterally, but near the surface of brain tissue. The transmission profile could be monitored via the microscope imaging system.
To achieve this, brains were extracted from 6-week old C57BL6 mice to represent typical brain tissue. The brain was quickly removed after cervical dislocation and dissected in 2 hemispheres along the sagittal line. Each hemisphere was subsequently cut along the dorsal surface forming a 90° angle between the dorsal and sagittal planes. The brain hemisphere was then placed in a microscope (Olympus BX 61, Olympus, Tokyo, Japan, upright) stage chamber submerged in artificial cerebrospinal fluid (ACSF) (Figure 4B) at room temperature with the dorsal surface (ii) facing upwards for observation through the microscope lens (Olympus ×2.5 NA). (vi) The sagittal plane (ix) was placed facing a 400 μm diameter cannula tip (viii) connected to a LED via a fibre optic (v).

Light stimulation was performed using CoolLEDs precise Excite LEDs (470 nm, 585 nm). Light penetration measurements were taken with an Andor iXon DV887 back-illuminated EMCCD camera (Andor Technology Ltd., Belfast, UK) and the images of the brain dorsal surface acquired with Andor Solis software (Andor Technology Ltd., Belfast, UK). The LEDs normalised light intensity was set immediately under the maximum brightness level the Andor camera could accept (14 bits) and the brightness intensity curve measured on the X and Y planes (X plane correspond to the...
4 | RESULTS AND DISCUSSION

4.1 | Thermal modelling results

To demonstrate the validity of the thermal modelling analysis, we performed a comparison of experimental results from the optrode shown in Figure 3 and its COMSOL model described in Figure 1A and Table 1. Both experiment and model were in air and results have been presented in Figure 5.

Figure 5A,B shows the effect of long-term heating over 15 minutes for both (A) CW LED illumination and (B) pulsed stimuli situation, respectively. In the pulsed stimuli situation, the temporal profile of the pulsed stimuli is pulse series with the pulse width of 0.7 seconds and the duty

coronal brain section; Y plane correspond to the sagittal one).
ratio of 50%, temperature in the first minute is zoomed in to see the oscillation of the temperature change along the pulsed stimuli. The quantization error that can be seen is due to the 0.1°C accuracy of the thermal camera.

Figure 5C-E shows the effect of heating from short 3.5 second pulses from a CREE LED optrode in air. The shaded regions define the 3.5 second period with the LED on, and the white background defines the period with the LED off. The x-axis is plotted on a logarithmic time period to show the full range of activity. The y-axis is plotted for both linear (left) and logarithmic (right) time period. The simulation results (red lines) and the experiment results (black lines) are comparable during both the ascent and the decay periods. The experiment was repeated multiple times with similar effect. The data presented are therefore an average for which the quantization error is not as apparent as for (A,B). (C) Represents pulsed stimulus with an LED current of 2.44 mA (6.5 V LED drive voltage). (D) Represents 7.1 mA (15 V LED drive voltage) (E) represents (8 mA) (7.75 V LED drive voltage).

Having built confidence that our model matches the experimental results in air, we utilise the model in water to explore heating effects on the exemplar probe shown in Figure 1B and Table 1. The specific sub-questions we aim to answer are:

1. Will uncoated electrodes act as a thermal hotspot?
2. How does the encapsulation affect the surface temperature?
3. How do specific surface hotspots vary with emission intensity?

These questions cannot be easily answered experimentally, particularly in water and thus can benefit from the use of modelling.

Figure 6A-C shows a typical result of the modelling for the exemplar optrode, in this case, reaching a steady state for 8 LEDs with a thermal emission of 2.5 mW. Figure 6A is the top-view of the total optrode, (B) and (C) are the cross-section-view and side-section-view at a single LED, respectively.
To answer question (1) the primary hotspots are concentrated around the LED emitters. However, additional spots can be seen at the electrode sites. To see the specific details of the hotspot temperature distribution, we set 3 lines I, II and III in Figure 6A,B. Temperature distribution along these 3 lines is shown in Figure 6D-F, respectively.

To answer question (2): we compare the optrode with or without a 20-μm thick polymer encapsulation. The on-surface temperature distribution of non-passivated optrode is given in the green curves; while for passivated optrode, we present the temperature distribution 0, 25 and 100 μm to the optrode surface in black, red and blue curves, respectively. The encapsulation we chose can reduce the hot spot temperature by up to 55% (from 2 to 0.9°C). The temperature increase falls to about 50% at 25 μm to the optrode surface and about 20% at 100 μm.

To explore this further, Figure 6G shows the dependence of the optrode temperature on the thickness of the substrate (100, 200 and 300 μm). Clearly, thicker substrates provide a temporary heat buffer and thus act to distribute the thermal energy, thus reducing hot spots.

Figure 6H shows the dependence of the optrode temperature on the Emitter-electrode distance (50, 100 and 200 μm). There is very little effect with distance from the primary LED heat source. We attribute this to heat traversing along the aluminium wires. Figure 6I shows the dependence of the optrode temperature on the encapsulation coating thickness (10-100 μm). Interestingly, beyond 20 μm, the heat from the LEDs begin to primarily escape via the electrodes, thus forming local hot spots at that point.

The final question (3): centres on the timing of the temperature rise of the maximum local hotspot as a result of the thermal energy pulse transmitted by a μLED. Note, we refer to the hotspot—not the average surface temperature as would be measured by a thermal camera.

Figure 7A displays the temperature change for a single LED with the dimensions given in Table 1 and layer thicknesses given in Table 2. Thermal pulses between 1 millisecond and 10 seconds, between 2.5 and 10 mW, are displayed. Figure 7B shows the same profile if all 8 LEDs are illuminated.

Figure 7C shows the maximum allowed pulse width for a given thermal power and a temperature increase limit of 2°C.

4.2 | Optical diffusion in the tissue

Figure 8A shows the extracted light profile taken from Figure 5C. Figure 8B shows a quantitative comparison between the experiment profile (lines) and the simulation profile (dots). This demonstrates that for a given emission profile we are broadly able to mimic the transmission profile.

It could be argued that there may be some deviation between rodent and human brain due to the differences in cellular density. However, we feel these results demonstrate...
applicability for both. The results also match previous efforts in the literature such as by Wu et al. [18]. With confidence in the model, we can then explore the key sub-questions relating to light penetration into tissue:

1. Is it better to have a smaller lower efficiency LED emitting light at a high radiant density or vice versa?
2. What is the effect of light collimation?
3. What is the effective volume of illumination for different types of emitter profile, taking probe gliosis into account?

Figure 9A-C shows the simulation results of the light penetration profile for Lambertian, optical-shaped and pseudo-collimated emitters, respectively. The data in these cases are plotted as the log(normalised intensity). Figure 9D-F shows the optical power vs penetration depth along the tangential line from the centre of the LED. The penetration depth is defined as the depth which can be attained while still presenting the target irradiance (in this case 1 mW/mm²). The plot is shown for each of the different emission profiles and LED diameters from 40 to 120 μm.

To answer question (1): Each of the LEDs in Figure 9D-F follow the penetration distance for absolute radianc, not radiant density. Thus, for a constant radiant power, the radiant density will increase, and the LED size decreases. However, as can be seen, this only has significance on the penetration depth of a few tens of microns for Lambertian emitters. There is a very limited difference in penetration depth for different sized LEDs with the same absolute radiance. As such, given the droop (efficiency) profile in Figure 2A, it is better to use larger LEDs within limits set by the probe dimensions. The caveat is that arrays of pseudo-collimated micro-LEDs, even if not particularly efficient, could be used to individually stimulate local neurons acutely at high resolution, and without crosstalk.

To answer question (2): Currently, most presented μLEDs have a Lambertian profile, with some having a tighter emission profile which we call here “optical shaped.” Pseudo-collimated LEDs do not currently exist in the literature for optogenetics. So, it is interesting to predict their relative performance. Figure 9G-I shows the optical power vs penetration depth for different ChR thresholds (1, 0.1, 0.01 mW/mm²). In addition, even with scattering effects, there is a strong correlation between emitter collimation and penetration depth. As an example, it requires 20, 8 and 1.5 mW for Lambertian, optical-shaped and pseudo-collimated emitters, respectively, to reach a 1 mm penetration depth into the tissue at the threshold of 1 mW/mm². A further advantage of increasing collimation (not shown here) is reduced crosstalk, which could allow the shown probe scale to high-density stimulation.

Typically, chronically implanted probes will have a tissue reaction whereby there is a zone of glial cells which displace or kill neurons in a region around an implant. From discussions with neuroscientists, such regions of dead...
neurons can be up to 200 μm in depth from the probe. This effectively creates a “deadzone” whereby light stimulus has no impact.

As such, we wanted to explore this question with 3 scenarios: (0, 100, 200) μm of deadzone representing (acute implant, moderate gliosis, severe gliosis). This concept is described in Figure 10A.

Figure 10B shows the illumination volume as a function of optical power for the acute case, that is, no deadzone. For comparison, the graphs are plotted for different opsin thresholds and emitter types. The results indicate that there is a slight advantage in illumination volume for more collimated emitters for equal radiance.

![Figure 10](image)

**FIGURE 10** (A) Illustration of the effective illuminated volume in tissue for the given optrode considering dead zone effect; (B) optical power vs effective illuminated volume for different emitter types and thresholds; (C) optical power vs effective illuminated volume for Lambertian emitter, different threshold and dead zones; (D) optical power vs effective illuminated volume for optical-shaped emitter, different threshold and dead zones; (E) optical power vs effective illuminated volume for pseudo-collimated emitter, different threshold and dead zones

Figure 10C-E shows the effect of deadzone size on illumination volume for different emitter types: Lambertian (C), optical-shaped (D), pseudo-collimated (E). Deadzones of (0, 10, 200) μm are presented, as are the 3 different possible opsin thresholds of (1, 0.1, 0.01) mW/mm².

From Figure 7 for a typical 10% duty cycle and a stimulus pulse of 100 milliseconds. The maximum thermal emission for a single LED is ~5.2 mW. Under these conditions, 3 lines (A, B, C) have been drawn to allow comparison. These represent the maximum optical emission for LED efficiencies of (0.9%, 1.8%, 3.6%) using the relation (16). These efficiencies represent typical efficiencies presented for micro-LEDs in the literature. Corresponding illumination volumes can then be compared for different emitter profiles, ChR thresholds and gliosis effects.

### 4.3 | Discussion

A number of important conclusions need to be drawn from this work. First, are LED-optrodes viable? Clearly, a number of past papers have demonstrated their utility in acute experiments. But what about chronic experiments, where long-term thermal effects and tissue gliosis around the probe become more apparent?

To answer this question, let us consider a typical case of a Lambertian emitter, an opsin threshold of 1 mW/mm², and a gliosis layer that will eventually surround the probe to a

TABLE 5  Thermal modelling parameters optimization

| Optrode dimension (μm) | Variable range | Optimised |
|------------------------|---------------|-----------|
| Encapsulation thickness| 10-100        | 20        |
| Silicon thickness      | 100-300       | 200       |
| Emitter diameter       | 20-200        | 50        |
| Emitter-electrode distance | 50-200     | 100       |
| μLED thickness         | 1-100         | 1         |
| Head thickness         | 100-400       | 200       |
| Shaft width             | 100-400       | 300       |
| Shaft thickness        | 100-400       | 200       |

Abbreviation: μLED, micro-light emitting diode.
thickness of 200 μm. Figure 9 shows that ~0.5 mW is required to achieve a penetration threshold of 300 μm, that is, 100 μm of effective penetration beyond the gliosis region. We can then further assume a typical 10% duty cycle and a stimulus pulse of 100 milliseconds. From Figure 7, under these conditions, the maximum allowed thermal emission for a single LED is ~5.2 mW, and for 8 LEDs it is 10.4 mW.

We can then relate the optical power requirement with the thermal limit using the following relation, where ηLED is the LED efficiency:

\[ P_{\text{thermal}} = P_{\text{optical}} \left( \frac{1 - \eta_{\text{LED}}}{\eta_{\text{LED}}} \right). \]  

(17)

The result is that for a single LED optrode, the LED efficiency would need to exceed 4.6%, and for a multiLED optrode, the efficiency would need to exceed 8.5%.

Clearly, with different parameters, these efficiency requirements could be reduced. However, as a general figure of merit, they significantly exceed the optimistic case presented in prior papers. The primary reason for this more pessimistic analysis is that previous work by McAlinden et al. [9] and Wu et al. [18] who did not explore the effects of hotspots or the effects of gliosis.

The caveat to this analysis is that even if a small hotspot forms on the surface—there is a rapid decay with distance from the probe. And such a decay would be presumably through dead tissue. Such consideration would perhaps be acceptable for neuroscience experiments. However, for clinical devices, the current regulations are clear: the surface temperature of hotspots or the effects of gliosis.

The second point to consider is the size of the LED. Figure 9 shows that size has little impact on the irradiance density for a given radiance (in mW). This is certainly the case beyond the 200 μm target discussed above. As such, given the effect of droop shown in Figure 2 and the high efficiency requirement described above, it makes more sense to distribute the light over a larger LED than to utilise smaller LEDs. For example, the commercial miniLED we utilised in our test probe had an efficiency of 30%, compared to only 1%-5% for μLEDs. There are 2 clarifications to this. First, there are applications where high-density optical information transfer is necessary. In such cases, smaller, more collimated emitters are required. Second, a cluster of smaller LEDs can have similar aggregate properties as a single larger LED, that is, for the same total light emission, the overall current density per LED within such a cluster can be low.

The third point to consider is the effect of collimation. All LED-optrodes to date have utilised Lambertian emitters. Our results show that both the penetration depth and penetration volume can be improved with collimation. Importantly, this effect is large when considering the effect of the gliosis deadzone. For a 1 mW/mm² threshold, a 200 μm deadzone, and an emitter radiance of 0.1 mW, a pseudo-collimated emitter will achieve an effective illumination volume of 0.01 mm³ (107 μm³ or 0.2 × 0.2 × 0.2 mm), whereas a Lambertian emitter would fail to illuminate any significant volume.

The fourth point relates to encapsulation, which has a low thermal conductivity and thus acts to block heat transmission. This is beneficial, as it acts to spread the heat from a local hot spot around an LED over the whole shaft. However, if too thick, it can result in local hot spots at the next opening—the electrode, which is clearly undesirable. To balance the main μLED hotspot with this secondary electrode hotspot, we find the encapsulation thickness of around 20 μm to be ideal. We do not notice this to be affected considerably by LED—electrode separation.

The final consideration relates to the optrode dimensions. The bulk of the silicon substrate acts as a primary heat-buffer, thus significantly smoothing spikes in heating. As such, a larger shaft is beneficial from a thermal perspective. However, from a tissue interaction perspective, a smaller shaft would be preferable. As such, we attempt to describe such trade-offs in Table 5.

For brevity, we have not considered the effects of circuit heat dissipation on active probes. Such dissipation would typically be in the optrode head and could traverse down the shaft having a small chronic effect on the local temperature. This would need to be considered for high levels of dissipation. However, our recent work with such systems [15, 16] indicates dissipation between 0.1 and 1 mW in the head, for which we would expect a relatively small effect.

As a final note, in this work, we have primarily considered the surface temperature of the LEDs. The internal junction temperature will be higher giving rise to questions about effects on longevity. In other work on this topic [17, 55], we have explored such effects which typically amount to 15°C above ambient (ie, 52°C). This is actually much less than the 80°C operating temperature for lighting LEDs which have operational lifetimes of 20 000 hours.

5 | CONCLUSION

We believe this work to be the most detailed study of LED-optrode opto-thermal design to date. It makes a number of conclusions which can be drawn from Figures 7–10 and Table 5. For brevity, we will summarise with 4 primary conclusions:

1. Our data suggest that LED-optrodes can viably perform chronic stimulation tissue without excessive heating, as long as they are reasonably efficient. We suggest >4.6% for single LED probes and >8.5% for multi-LED probes.
2. The size of the LED has little effect on the illumination profile, so given efficiency is a function of current density, it is best to have as large an LED as is viable for a given probe.

3. Collimated emitters are superior to equivalent Lambertian emitters with the same efficiency. This is particularly the case for chronic use where the presence of a “deadzone” will negate irradiation near to the emitter.

4. We note an optimal encapsulation thickness of around 20 μm, which balances broad dissipation through the bulk vs local hotspot formation on the electrode.

We believe in the long term, continuous improvements in both μLED efficiency and the sensitivity of the opsins will be possible, and with the correct stimulation protocols to provide optical stimulus without significant tissue heating.

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AUTHOR BIOGRAPHIES

Please see Supporting Information online.

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