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Field-effect sensors – from pH sensing to biosensing: sensitivity enhancement using streptavidin–biotin as a model system†

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Field-Effect Transistor sensors (FET-sensors) have been receiving increasing attention for biomolecular sensing over the last two decades due to their potential for ultra-high sensitivity sensing, label-free operation, cost reduction and miniaturisation. Whilst the commercial application of FET-sensors in pH sensing has been realised, their commercial application in biomolecular sensing (termed BioFETs) is hindered by poor understanding of how to optimise device design for highly reproducible operation and high sensitivity. In part, these problems stem from the highly interdisciplinary nature of the problems encountered in this field, in which knowledge of biomolecular-binding kinetics, surface chemistry, electrical double layer physics and electrical engineering is required. In this work, a quantitative analysis and critical review has been performed comparing literature FET-sensor data for pH-sensing with data for sensing of biomolecular streptavidin binding to surface-bound biotin systems. The aim is to provide the first systematic, quantitative comparison of BioFET results for a single biomolecular analyte, specifically streptavidin, which is the most commonly used model protein in biosensing experiments, and often used as an initial proof-of-concept for new biosensor designs. This novel quantitative and comparative analysis of the surface potential behaviour of a range of devices demonstrated a strong contrast between the trends observed in pH-sensing and those in biomolecule-sensing. Potential explanations are discussed in detail and surface-chemistry optimisation is shown to be a vital component in sensitivity-enhancement. Factors which can influence the response, yet which have not always been fully appreciated, are explored and practical suggestions are provided on how to improve experimental design.

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Introduction

Chemical sensors are important in a wide range of applications such as medical diagnostics,¹ explosives detection,² food safety³ and environmental monitoring.⁴ A promising class of chemical sensors are Field-Effect Transistor-sensors (FET-sensors). FET-sensors have been receiving increasing interest over the last two decades, motivated by the need for low-cost biosensors capable of direct and rapid detection of analyte molecules without the need for expensive and time-consuming labelling steps. As a medical diagnostic tool, this would increase survival rates of patients by reducing load on centralised-diagnostic facilities. FET-sensors have the potential to offer ultrahigh sensitivity, low-cost production, portability

and facile miniaturisation as part of a 'Lab-on-a-Chip'. FET-sensors present several advantages in comparison with currently available label-free biosensors which operate *via* mass-detection,^{5,6} the electrical detection offered by FET-sensors provides additional information on, for example, conformational changes⁷ or extracellular potentials.⁸ In contrast to many other biosensing methodologies, FET-sensors do not require bulky optical measurement equipment.

FET-based pH sensors were initially popularised as far back as the 1970s with the work of Bergveld,^{9,10} and have been successfully commercialised.¹¹ However, the extension of these devices to sensitive and reliable detection of biomolecular analytes ('BioFET' devices) has proved more difficult than initially expected. These problems have resulted in research providing a range of novel FET-sensor architectures^{12–14} and methods of operation,^{15–17} however advances in the field of BioFET research are obstructed by a lack of consensus on which quantitative metrics (*i.e.* figure-of-merit) should be used to compare devices. As a result, most published studies can only be compared qualitatively. By focusing on a single biomolecule

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p/n/p). FET-sensor designs vary, with the source–drain contacts not necessarily being highly doped. For a setup like that in Fig. 1 featuring an insulating oxide layer, FET-sensor response is driven by changes in the electric field at the oxide surface due to analyte molecules binding to the surface. In the case of pH sensors, the FET-sensor is usually termed an Ion-Sensitive FET (IS-FET), and the surface is an oxide material in which pH-induced changes in the protonation state of the surface hydroxyl groups result in a change in the surface charge and therefore the surface potential. For sensing molecular analytes such as biomolecules, the surface is usually functionalised to provide receptor sites which are specific to the analyte and the device is usually termed a BioFET. The change in the interfacial electric field on binding of analyte causes a change in the concentration of charge carriers in the channel region of the device, with a corresponding measurable change in the source–drain conductivity.

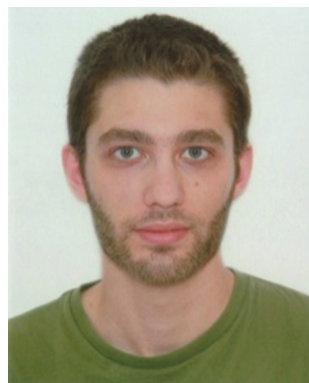
Whilst Fig. 1 shows a simple planar device geometry akin to a single-gate MOSFET, a plethora of device geometries have been developed such as nanogap,^{22–24} nanobelt,^{25,26} nanoribbon^{18,27–32} and nanowire^{15,29,33–37,38} structures. The application of additional lateral gates has been shown useful in tuning the carrier concentration of the device and pursuit of an optimal gate configuration is an active area of research.^{39–41} There is also significant interest in the development of carbon-based devices which do not have an oxide layer such as graphene or carbon-nanotube devices.^{14,42} The advantage of these devices is stated to be potential sensitivity enhancement *via* (i) direct contact with analyte (ii) the size of the material being comparable to the size of the analyte.⁴³ It should be noted that the lack of a band-gap in graphene produces a limitation in the ability of the transistor to transduce a change in surface potential to a change in current response.⁴⁴

The mechanism of operation of devices without an oxide layer, such as graphene and carbon nanotube devices, is expected to be different to those with an oxide layer due to the possibility of direct charge transfer from the analyte to the semiconducting carbon layer.^{45,46} Due to the different mechanism of gating, these carbon-based devices are not the focus of this review. Under certain conditions discussed later, the transistor component has the capability of transducing a change in surface potential to an exponential change in current. While, in principle, changes in surface potential due to analyte binding can be measured directly without a transistor similarly to a conventional glass pH sensing electrode,⁴⁷ this approach requires potentially large and expensive measurement equipment (high impedance amplifier). The FET-component facilitates miniaturisation of the device, providing smaller sample volumes and faster response times.²⁰

The reference electrode

Reference electrodes are used to provide a potential within the FET-sensor system to which other potentials can be referenced. Through the reference electrode, a gate voltage (V_g) can be applied in the sample and is sometimes referred to as a ‘top-gate’ voltage or ‘liquid-gate’ voltage in analogy to MOSFET operation. An ideal reference electrode ensures that the potential at the electrode–electrolyte interface is insensitive to changes in the electrolyte solution.³³ FET-sensors can be used to quantitatively measure binding of analyte as the response is a function of the electrolyte–oxide surface potential. The reference electrode provides a stable potential in the bulk electrolyte which is used to reference the measurement.

A stable reference potential requires a redox reaction in which there is constant thermodynamic activity of each participant in the reaction. In a conventional reference electrode, for example as used in electrochemistry or traditional glass-



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electrode pH sensors, this is achieved by placing a reference metal inside a compartment with a high concentration of salt solution connected to the analyte sample by a liquid junction which can only exchange ions.⁴⁸ Miniaturisation of this system is problematic and results in reference electrodes with reduced lifetimes⁴⁸ and a common alternative in the field of FET-sensing is the pseudo-reference electrode in which a bare metal or chlorinated silver wire is used. Unlike conventional reference electrodes, the interfacial potential is not known *a priori*, but under controlled conditions can still maintain a stable potential.

Noble metal pseudo-reference electrodes such as gold (Au) and platinum (Pt) are sometimes used but cause issues with current instability and pH dependency.³⁴ Although Pt-pseudo reference electrodes have been used in various streptavidin-sensing experiments,^{35,49–51} many research groups have reported that Pt pseudo-reference electrodes are unreliable and should not be used due to issues such as unstable potentials, transient noise and changes in electrical potential due to non-specific binding of biomolecules ('bio-fouling').^{27,33,52,53}

The most common pseudo-reference electrodes used in FET-sensors are silver-silver chloride (Ag/AgCl) electrodes which consist of a silver wire usually treated by either chemical or electrochemical chlorination.⁵⁴ Rajan demonstrated experimentally that Ag/AgCl pseudo-reference electrodes are a suitable alternative to a conventional reference electrode,⁵² with the proviso that, due to the strong interaction with chloride in the buffer, the chloride content of the buffer is kept constant throughout the sensing experiment. Both Rajan⁵² and Rim *et al.*³⁴ measured the open-circuit voltage between a Ag/AgCl pseudo-reference and conventional reference electrode to investigate bio-fouling and found the open-circuit voltage to be small, suggesting that pseudo-reference electrodes can be suit-

able for biosensing. For a more detailed review of pseudo-reference electrode validation experiments, see ESI section 1.†

The use of reference electrodes in FET sensors. It is often stated that a reference electrode (with corresponding liquid-gate voltage, V_g) is required for a reproducible and stable signal from FET-sensors.^{48,53,55,56} Nonetheless, it is not uncommon for devices to be fabricated without any reference electrode in the liquid^{25,28,36,37,57} which can reduce the possibility of dielectric breakdown of the device under applied gate voltage (e.g. as described in the ESI of Stern *et al.*³⁶). Such devices often have a gate connected to the substrate (back-gate) which is either (a) at a constant gate voltage, usually chosen to optimise the transconductance of the device at that gate voltage, or (b) swept across a range of gate voltages in a similar way to which a liquid-gate might be operated. In case (a), the gate is essentially acting as a pseudo-reference electrode and this setup can be unreliable. In some cases the device can be unresponsive when operated *via* back-gate with no liquid top-gate.^{27,52}

The reference electrode setups used in the literature are broadly diverse; based on 37 publications identified in this review, 11% used Pt pseudo-reference electrodes, 19% used Ag/AgCl pseudo-reference electrodes, 11% used Ag/AgCl conventional reference electrodes, 21% did not publish their setup or it was ambiguously presented and 27% utilised only a back-gate. Of these setups, only Ag/AgCl pseudo-reference electrodes have consistently been proven to be capable of providing a reliable reference potential *versus* a conventional reference electrode.

Regions of operation – linear, saturation and subthreshold

The current flowing through a FET is controlled by the gate voltage, V_g and the drain voltage, V_{ds} . Depending on the particular choice of these two parameters, the device is said to be operating in one of three main regions: sub-threshold, linear and saturation. These regions are summarised graphically in Fig. 2 and explained below.⁵⁸

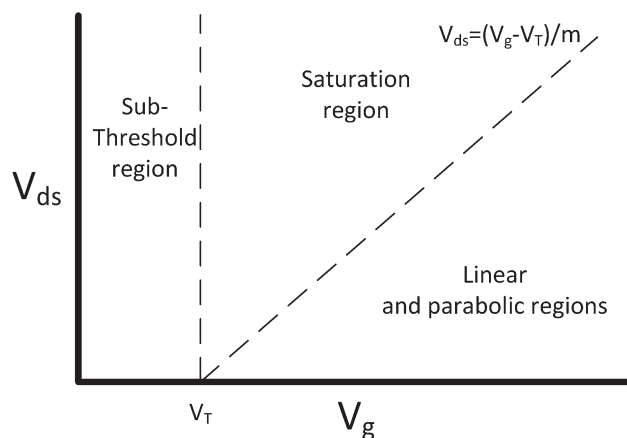


Fig. 2 FET regions of operation. m is the body-effect coefficient ($m \geq 1$). Adapted from figure by Taur and Ning (2013).⁵⁸



Nicolas Green

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Electrokinetics and nanoscale electronics device modelling. These research interests cover theoretical and experimental investigation into fundamental physical effects, as well as engineering applications such as in particle manipulation, and the development of advanced simulation models for design.



$$I \propto e^{\frac{q(V_g - V_T)}{mk_b T}}, \quad (1)$$

If a larger gate voltage is used and the drain voltage is low (*i.e.* $V_g > mV_{ds} + V_T$, where m is an empirical body-effect coefficient and $m \geq 1$ (ref. 58)) then the device is operating in the ‘linear’ region. If the drain current is high (*i.e.* $mV_{ds} + V_T > V_g$), then the saturation region is reached.

Device characterisation

FET-sensors operate on the principle that binding of analyte to the sensor surface results in a change in surface potential ($\Delta\psi_s$) *via* electrostatic gating. This induces a change in the device threshold voltage, which can be measured *via* the transistor as an amplified signal in the form of a change in the drain current. By measuring the variation in drain current (I) as a function of the reference electrode potential, the shift in threshold voltage (ΔV_T) and the change in drain current (ΔI) can be measured. Either of these properties are termed device ‘response’ in this work. From these measurements, metrics important for characterising FET-sensors can be calculated. Commonly used metrics to characterise the sensor are the Subthreshold Slope (SS) for the subthreshold region of operation and the transconductance (g_m) for the linear region of operation. Common metrics for characterising sensor response to analyte are the normalised change in current (I_{norm}) and the shift in threshold voltage (ΔV_T). These four metrics are illustrated in Fig. 3 and are defined as follows.

The subthreshold slope is constant in the subthreshold region and therefore provides a straightforward quantification of device response in this region. The value of the Subthreshold Slope is a measure of transistor quality in terms of its response to changes in the gate potential. It is defined as the change in gate voltage (V_g) needed to change the subthreshold current (I)

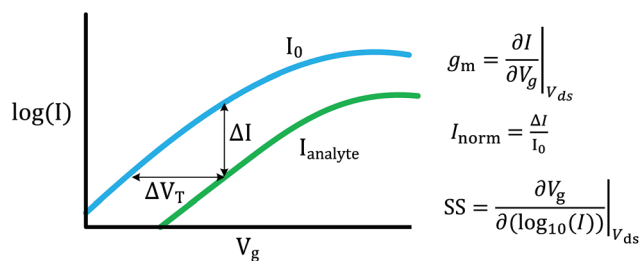


Fig. 3 Schematic graph showing the logarithmic drain current response as a function of gate voltage for an n-channel FET operating in the sub-threshold regime, $V_{ds} \neq 0$, and a liquid-gate voltage sweep before (blue line) and after (green line) addition of negatively charged analyte with currents I_0 and I_{analyte} respectively. The direction of response shown is typical of analytes such as streptavidin (at pH 7.4) or a change to more alkaline pH. On analyte binding, there is a parallel shift in the curve such that at constant current the corresponding shift in liquid-gate threshold voltage is equal to the change in threshold voltage (ΔV_T). Assuming electrostatic gating, and that the top-gate liquid electrode voltage is V_g , then ΔV_T is equal to the change in surface potential ($\Delta\psi_s$) and therefore the transistor can be used for quantitative sensing of analyte binding. If V_g is instead the back-gate voltage (with fixed liquid top-gate voltage), then a different I - V_g curve is expected which is qualitatively similar but quantitatively different. This approach has been used to enable amplified values of ΔV_T , but in this case the shift is not equal to the surface potential shift, and is instead amplified and related to the ratio of capacitances of the top-gate and bottom-gate oxide, as discussed later in this review. Unless otherwise stated in this review, V_g is used to refer to the top-gate liquid electrode voltage and thus SS and g_m are related to the top-gate response. At constant liquid-gate voltage there is a shift in current response (ΔI), which can be divided by the initial current to obtain the normalised change in current, I_{norm} . This current response is sometimes characterised in terms of the units decades, i.e. $\log(I_1/I_2)/\log(10) = x$ decades. The transconductance, g_m , is approximately constant in the linear region whereas the Subthreshold Slope (SS) is approximately constant in the subthreshold region.

by one decade⁶⁰ and is determined from measurements as the inverse of the slope of the linear region on a $(\log_{10} I)$ versus V_g graph at constant V_{ds} :

$$SS = \left. \frac{\partial V_g}{\partial (\log_{10} I)} \right|_{V_{ds}} \quad (2)$$

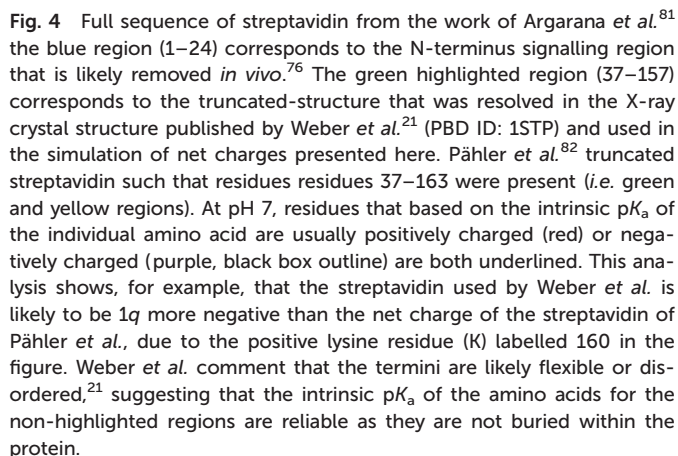
When V_{ds} is greater than a few $k_{\text{b}}T/q$ (where k_{b} is the Boltzmann constant and T is the temperature), the subthreshold slope is independent of V_{ds} because the current is diffusion-dominated.⁵⁸ The subthreshold slope can be modulated by a back-gate voltage, $V_{\text{g,back}}$.³⁰ The reciprocal of the subthreshold slope is also called the ‘gate voltage swing’ or ‘subthreshold swing’.⁶⁰

The subthreshold slope can also be defined as:

$$\text{SS} \approx \frac{2.3mk_{\text{b}}T}{q} \approx \frac{2.3k_{\text{b}}T}{q} \left(1 + \frac{C_{\text{dl}}}{C_{\text{ox}}}\right), \quad (3)$$

where m is the body-effect coefficient, C_{dl} is the depletion-layer specific capacitance (per unit area) and C_{ox} is the specific capacitance of the oxide (per unit area).⁵⁸ Therefore, the sub-threshold slope has a theoretical minimum value of approximately 59 mV dec⁻¹ at room temperature. A lower value of the

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In X-ray crystallography studies, Weber *et al.* reported that they were unable to crystallise non-truncated streptavidin, but successfully crystallised a truncated form of streptavidin,²¹ the sequence of which is highlighted in green in Fig. 4 together with the full sequence for the protein. The pH-dependent charging properties of this structure were modelled, with a calculated charge of $-7.20q$ at pH 7.4 (ESI section 4†).

In addition, surface-bound streptavidin has been measured with different properties to the free protein. The pI of surface-bound streptavidin has been measured using Surface Force

Other biomolecules. This review focuses on streptavidin because it represents the most prevalent and comparable model protein system available at the time of writing, both in the general literature and literature focussed on BioFET devices. Another commonly used protein in BioFET studies is Avidin,^{30,36,71} which is related to streptavidin, has an affinity for biotin but has a higher pI and is therefore oppositely charged at the physiological pH. This protein however is a poor choice for comparative study for several reasons, firstly, electrophoresis reveals distinctive charge heterogeneity even in purified avidin and both compositional and structural heterogeneity due to its carbohydrate content,⁸⁴ and secondly, its higher pI can result in higher non-specific binding due to a larger net-charge in physiological solutions.⁸⁵

Another biomolecule commonly used in BioFET experiments is prostate-specific antigen (PSA) due to biomedical applications as a prostate cancer biomarker, however the biomolecular capture step is more complex than streptavidin, requiring the use of a complex biomolecule such as an antibody or aptamer.⁸⁶ The most well-studied biomolecule in BioFETs is undoubtedly nucleic acid-based systems such as DNA. However, as summarised in a review by Poghossian *et al.*,⁸⁷ experiments are rarely performed using the same DNA-sequence or polynucleotide length, making their structural and electrical characteristics not easily comparable. Furthermore, in the DNA-sensing community there is high variability in experimental design, for example, the choice of buffer composition.⁸⁷

The buffer solution maintains the pH and the stability of the biomolecule, the system and the measurements. Its composition, the concentration and ratios of ions in solution, can have a significant effect on the bio-sensitivity of the sensor. Even in absence of biomolecules, ions in the buffer can modulate the surface chemistry or be selectively adsorbed to the surface, generating a detectable response. Accurate reporting of the composition of this solution is therefore essential to any biosensor experiment.

Buffering capacity. An important property of the buffer for biosensing experiments is the ‘buffering capacity’, which quantifies the capability of the buffer to maintain a constant pH, where the pH is defined as the negative logarithm of the hydrogen ion activity:

$$\text{pH} = -\log_{10}(a_{\text{H}^+}) \quad (5)$$

Phosphate buffer is often used for biosensing experiments; a convention is to define 10 mM ionic strength phosphate buffer as $1\times \text{NaPi} = \text{Na}_2\text{HPO}_4$ (8.1 mM) and NaH_2PO_4 (1.9 mM). The most common buffer for protein sensing is Phosphate-Buffered Saline, PBS, which is composed of $1\times \text{NaPi}$ mixed with 138 mM NaCl and 2.7 mM KCl.⁷⁴ The ionic strength of

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Second, it is also assumed that molecules are rigid and bound in a fixed orientation, whereas in reality they are flexible and dynamic and therefore some orientation of the analyte can bring the molecular charges closer to the surface than other orientations.⁹⁵

A further consideration is that the surface chemistry itself can change at different ionic strengths, resulting in alteration of the sensor response. For example, the FET-sensor of Tarasov *et al.* showed a shift in threshold voltage of 59 mV for every 10-fold increase in KCl concentration.⁹⁶ They attributed this effect to pH independent selective adsorption of anions. Similarly, a non-linear FET-sensor response to increasing NaCl concentration was measured by Maekawa *et al.*⁹⁷ This empirical observation has also been observed in molecular dynamics simulation studies.^{97,98}

The electronic device and signal drift

Physical and structural characteristics of the transistor device, particularly the choice of semiconducting material, can affect both the amplification ability of the transistor and its performance in terms of drift. Silicon is commonly used as the semiconductor material rather than more exotic materials, as its ubiquitous presence in the semiconductor industry means that fabrication is simple, well-understood and low cost. Silicon-based FET sensors can, however, suffer from long-term unidirectional drift in signal. Various mechanisms for this action have been proposed⁹⁹ yet this remains an important research question within the field as it is a key factor limiting commercialisation efforts and/or, in the case of Si₃N₄ films, *via* aqueous hydration of the surface to SiO₂ or an oxynitride.^{99–103} While SiO₂ surfaces are rapidly hydrated, Si₃N₄ surfaces and other common oxide materials such as Al₂O₃ and Ta₂O₅ are known to be more resistant to hydration¹⁰¹ and show reduced drift rates at pH 7.¹⁰⁴

Given the timescale of the drift often observed in SiO₂ surfaces (hours) device noise or drift could be due to ions from the aqueous solution diffusing into the insulating material^{101,105} which would show a corresponding change in the threshold voltage of the device.¹⁰¹ As an alternative material to SiO₂ which is impermeable to ions,⁹⁵ AlGaIn/GaN-based FETs have been demonstrated with significantly reduced drift in physiological buffers and various groups have successfully utilised AlGaIn/GaN-based devices for streptavidin sensing.^{49,50,66,106,107}

Understanding and modelling drift characteristics of FET-sensors remains a vital challenge in the field. As an illustration of this, there remains no physical model capable of accurate quantitative modelling of experimental signal drift for gas-based FET-sensors.¹⁰⁸

Recent developments

For successful commercialisation of BioFETs in biomedical applications, operation in high ionic strength environments remains a key engineering challenge. An alternative mode-of-operation for sensing FETs has been proposed, first by Schasfoort *et al.*^{109,110} and recently demonstrated for detection

of poly-L-lysine.¹⁷ This mode relies on making a sudden change in ionic strength change ('ion-step') and measuring the initial non-equilibrium response before the system fully equilibrates. More recently, Krivitsky *et al.* detected biomarkers from untreated blood samples using a similar strategy in which the transient response from specific-binding is compared to a non-specifically bound control.¹¹¹ By measuring the transient response of the system, biomolecules can potentially be detected in higher ionic strength buffers than conventional 'equilibrium' sensing experiments. This also potentially offers better sensitivity to the analyte and reduced sensitivity to the background drift in the drain current.

Other methods have shown promise at circumventing Debye screening limitations. For example, FET nanogap-devices operate on a different mechanism to conventional BioFETs, where the dielectric constant of the gate is modified by analyte binding and the device is operated after drying.^{22,23,112} Another example for conventional FET-sensors is that the surface layer could be modified with a biomolecule-permeable polymer layer which is proposed to operate by extending the effective distance over which charges are screened within the layer.¹¹³ A final example of particular interest is the use of frequency-mode detection,^{15,46,114–117} where the current response is measured in the frequency domain instead of the time-domain (details can be found in ESI 6†).

The lack of wide-spread adoption of alternative operating methods may be due to: complexity, with the additional required steps required making them less commercially appealing; a lack of awareness in a rapidly changing field; or lack of reproducibility. Given the variety of strategies which have shown some efficacy at circumventing the limitations arising from charge screening, it is likely that this will successfully be surmounted in a commercial device.

Device performance metrics

Given the lack of maturity of FET-sensor technology, it is understandable that this field has not yet converged upon a set of standards, both in terms of well-defined nomenclature or usable engineering parameters. Such nomenclature and parameters are required to rigorously compare the performance of BioFET sensors, particularly the metrics involved in the detection and quantification of analytes.

IUPAC definitions

IUPAC defines 'sensitivity' as a metric of the specific response as a function of the analyte concentration. More specifically, it is defined by IUPAC as 'the slope of the calibration curve'. IUPAC defines the calibration curve as:

'The functional (not statistical) relationship for the chemical measurement process, relating the expected value of the observed (gross) signal or response variable to the analyte amount. The corresponding graphical display for a single analyte is referred to as the calibration curve'.¹¹⁸



'The limit of detection, expressed as the concentration, c_L , or the quantity, q_L , is derived from the smallest measure, x_L , that can be detected with reasonable certainty for a given analytical procedure. The value of x_L is given by the equation $x_L = \bar{x}_{bi} - ks_{bi}$, where $\bar{x}_{bi}$ is the mean of the blank measures, s_{bi} is the standard deviation of the blank measures, and k is a numerical factor chosen according to the confidence level desired'.¹¹⁸

The symbol I_{norm} is used to refer to the normalised change in current, which is defined as:

where I_f and I_0 are final and initial values of the current, respectively. In this review, the term ‘normalised change in current’ is used to refer to this type of normalisation.

Some publications simply refer to the absolute measured change in current from the device (ΔI) as the metric however, this results in a large device-to-device variation. The normalised change in current has become common practise as it reduces this variation.¹¹⁹ The reason for this reduction in device-to-device ratio is that the absolute current is directly dependent upon geometric parameters and threshold voltage of the device, whereas the normalised change in current is only subject to device-to-device variation from differences in the device threshold voltage.¹¹⁹ Further, the change in current (ΔI) can either be positive or negative and consequently the normalised change in current, I_{norm} , also has an associated sign. Some authors have reported the normalised change in current without this sign which makes the polarity of the measured change ambiguous.⁴⁹

This can be explored using the MOSFET drift-diffusion equations and the shift in threshold voltage (due to analyte binding) can be related to the normalised change in current using the following equation:^{18,44}

This equation can be used to calculate the expected magnitude of I_{norm} for a given analyte. For example, considering the case of an ideal n-channel transistor with a subthreshold slope of 59 mV dec⁻¹ and an analyte that induces a change in threshold voltage of 10 mV, the calculated I_{norm} is approximately -33%. For the equivalent analyte and a p-channel device, the calculated change in I_{norm} is approximately +50%.

This important distinction between a negative I_{norm} and a positive I_{norm} can be emphasised as a qualitative difference; a current which decreases is ‘bounded’ by zero and therefore has a maximum possible value of -100% , whereas an current which increases has no mathematical upper limit for I_{norm} . In this paper, this is referred to as ‘unbounded’ to indicate that is always positive and can reach large positive values. In the sub-threshold region, a direct comparison between an unbounded signal from one device (negative molecule on an n-channel semiconductor) and the bounded signal of another (*i.e.* with a p-channel device) is uninformative. This is particularly relevant for ambipolar devices (*e.g.* the devices of Nam *et al.*¹²⁰); ambipolar devices can operate in either an n-branch or a p-branch depending on the choice of gate voltage, requiring a metric which can compare the response of the unlike branches.

In order to compare the signal between devices with different channel types, the change in current can be normalised by the current obtained either before the response (I_0) or after the response (I_f), whichever is lower. These lowest and highest values are referred to as I_{low} and I_{high} respectively. This results in an I_{norm} which is always positive and ‘unbounded’ with no mathematical upper limit. This ‘unbounded’ metric is herein referred to as I_{norm}^+ or the ‘unbounded normalised change in current’:

The differences between the unbounded and bounded I_{norm} are illustrated schematically in Fig. 5. For consistency with our previous quantitative analysis of the literature performed for pH sensing,¹⁸ I_{norm}^+ was used in this analysis. As an example, for streptavidin (negatively charged analyte at pH 7.4) on an n-channel device, the drain current would decrease and therefore I_{low} would be the drain current after addition of streptavidin and I_{high} high before addition. Alternatively, it would be possible to compare the ‘bounded’ change in current, where the current is normalised by the higher of the current (I_{norm}^-) before or after binding.

Concentration dependence. Normalised change in current is a function of the amount of bound analyte as well as the ability of the sensor to amplify response. This can be problematic from the perspective of biosensor design, because if an

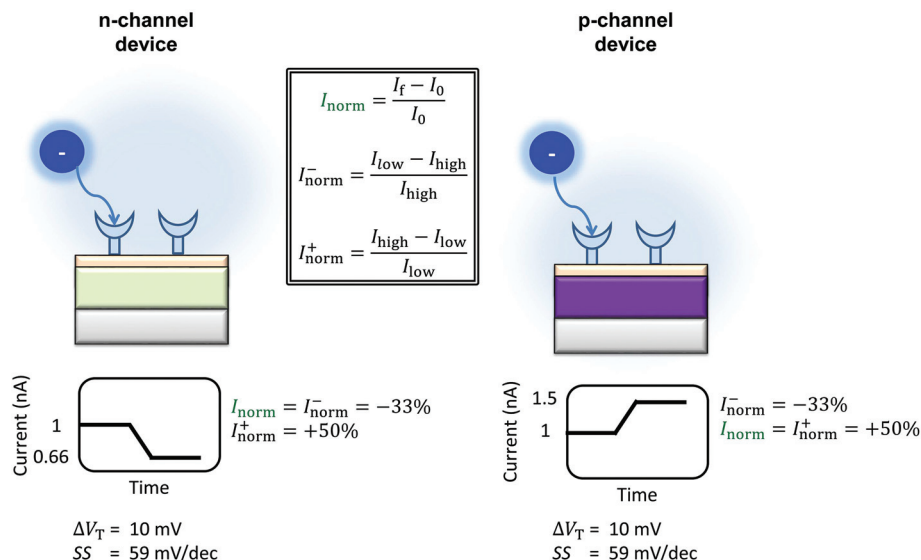


Fig. 5 Schematic of the change in current upon sensing a negatively charged analyte by an n-channel device (left) and p-channel device (right) is shown. The device is assumed to operate in the subthreshold region. The values shown are calculated using eqn (9), operating in the subthreshold region, assuming an ideal device ($SS = 59 \text{ mV dec}^{-1}$) and an analyte which can induce a change in the threshold voltage of 10.39 mV. The polarity of the response is reversed between the two types of device, and therefore the normalised change in current. The normalised change in current is obtained by $I_{\text{norm}} = \Delta I/I_0$, where I_0 is the initial current prior to analyte addition. In a sensing experiment, the lowest value of the current recorded is referred to here as I_{low} and the highest value as I_{high} . The change in current can be instead normalised by I_{low} instead of I_0 , resulting in the 'unbounded' normalised change in current (I_{norm}^+). An alternative method would be to normalise by I_{high} (I_{norm}^-). Inspection of the above figure illustrates that in order to compare the sensing results for a particular analyte between devices of different semiconductor, the normalised change in current must be calculated consistently *i.e.* I_{norm}^+ cannot be directly compared to $|I_{\text{norm}}^-|$ as different values are expected from the same stimulus.

experiment is performed on a very poor sensor (*i.e.* low ability to transduce chemical binding into measurable signal) with a high concentration of bound analyte, then the resulting I_{norm} value can be the same as that measured by an experiment on a superior sensor with a low concentration of bound analyte. The response per 10-fold increase in concentration of analyte is a more useful figure-of-merit for biosensors and is similar to the IUPAC definition of sensitivity. In BioFET literature, the normalised change in current is often calculated based on the change in drain current after the introduction of an arbitrary concentration of analyte. In contrast, in pH sensing literature, it is often defined over a single pH unit (*i.e.* per 10-fold increase in concentration of analyte).

Other than as a figure-of-merit, there are important scientific reasons for the measurement of the response as a function of analyte concentration. The affinity (*i.e.* equilibrium dissociation constant, K_d) of an analyte binding to a sensor can be estimated by fitting the concentration-dependent response data to an appropriate binding model.³⁵ Further, the concentration at which the response saturates can be used to estimate the density of bound molecules.^{28,121} Note that at concentrations much greater than the K_d of the binding reaction, sensor saturation occurs, along with increased levels of non-specific binding.¹²²

Performance metrics have also been discussed by Rajan *et al.*,^{29,52} who highlighted a few examples from the literature of concentration-response curves for BioFETs over different analytes and stressed the importance of considering, not only the IUPAC sensitivity, but also the Signal-to-Noise Ratio (SNR).

At this time, unfortunately, most BioFET data is reported without noise analysis or sufficient repeats to obtain an estimate of the statistical uncertainty of measurements, and so quantitative estimation of this SNR or IUPAC Limit of detection is rare. Given these limitations, the analysis presented in this analysis uses the normalised change in current per 10-fold increase in analyte concentration (*e.g.* % per unit pH or % per 10-fold increase in streptavidin concentration) as a performance metric to compare sensing results between different experiments. This metric will be referred to as 'Sensitivity' due to its similarity to the IUPAC definition and will be discussed in more detail later and in ESI 7.3.†

Sensitivity limits. Sensor response is improved with a larger shift in surface potential, and therefore knowledge of the upper and lower limits for the surface potential shift is an important aspect of sensor design. In pH sensing, a maximum shift in surface potential of approximately 59 mV per pH is observed, in good agreement with the theoretical prediction given by the Nernst model of equilibrium potentials of ions across semi-permeable membranes.¹²³ While it is reasonable to consider that a similar limit would apply to biomolecular surface-binding experiments (~59 mV per 10-fold increase in streptavidin concentration at room temperature), there are significant differences between proton equilibria and macromolecular biomolecule binding equilibria, which implies that the Nernst model is not applicable to protein binding. As a result, the existence of such a macromolecular 'surface-potential response limit' remains an open research question. Shoorideh



and Chui hypothesised that no such limit for biomolecule binding exists because the high affinity of biomolecular-ligand binding would result in the extent of binding being independent of the surface potential, in contrast to proton-binding where the binding is exponentially related to the potential of the surface.^{124,125}

The largest shift in threshold voltage for a pH sensor can be calculated based on the Nernstian response of 59 mV per pH. The steepest subthreshold slope for a classical transistor is approximately 59 mV dec⁻¹.⁴⁴ From eqn (9), the largest normalised change in current is obtained when the subthreshold slope is steepest and the shift in threshold voltage is largest, resulting in an upper theoretical Sensitivity limit of 900% pH. Lee *et al.* used additional circuitry intended to amplify the FET-signal beyond this value but did not provide quantitative evidence to support that it had improved the signal-to-noise ratio or limit of detection.¹²⁶

Relationship between surface potential and current response

I_{norm} is a measure of the amplification of the current response. By contrast, ΔV_T is a measurement of the changes in the electrostatic potential at the surface and therefore can be used both to provide quantitative detection of analyte and also to suppress device-to-device variation.^{119,127} The change in threshold voltage due to analyte binding can either be measured directly from the shift in the I - V_g curve, or calculated using MOSFET drift-diffusion equations (ESI sections 3 and 8†). The analytical relationship between current change and the change in threshold voltage is dependent upon the region of operation of the FET as follows.

For FET-sensors operated in the subthreshold region, this relationship is given by eqn (9)^{18,38,44} and has a logarithmic dependence on I_{norm} . In the subthreshold region, the transconductance is variable as a function of gate voltage and therefore cannot be treated as a constant.

For FET-sensors operated in the linear regime, ΔV_T can be extracted experimentally by simply dividing the current-response data (ΔI) by the device transconductance, g_m .^{29,61,119,128}

$$\Delta V_T = \frac{\Delta I}{g_m} = (V_g - V_{T,0})I_{\text{norm}}. \quad (11)$$

This expression is only valid if the analyte does not change the operating regime of the device⁶¹ and providing that g_m is the same before and after binding³⁵ (*i.e.* electrostatic gating mechanism of FET operation⁴⁵). Alternatively, the threshold voltage prior to analyte addition, $V_{T,0}$ and I_{norm} can be used to calculate the shift in threshold voltage.⁵² Derivation of these equations can be found in ESI section 8.†

Signal-to-noise ratio enhancement

This section presents common methodologies to obtain, optimise and understand the Signal-to-Noise Ratio (SNR) properties of FET-sensors as a figure-of-merit for biosensors.

Optimal region of operation. There are a number of noise sources such as semiconductor-device resistance, type of measurement equipment, electrolyte-interface chemistry and electrolyte-reference electrode chemistry. When comparing devices with similar noise levels and operating in the subthreshold region, an improved subthreshold slope is expected to improve the limit of detection. From eqn (9), for a minimum detectable normalised change in current, a device with smaller subthreshold slope results in a smaller required shift in surface potential, and therefore a lower required concentration of analyte for detection.⁴⁴

An important relevant question is: what gate voltage and region of operation is best for SNR and IUPAC limit of detection, given that the noise varies between devices and as a function of gate voltage? For the current response, the SNR is defined as the ratio of the change in current ΔI to the noise in the measurement (δi) response.

Empirically, there are various methods for determining the noise component. Given the low-frequency noise is of interest for biosensing, and that the most common noise source of FET-transistors is of the form $1/f$ (flicker noise), a common approach^{46,52,107,129,130} utilises the following expression:

$$(\delta i)^2 = \int_{f_1}^{f_2} \frac{S_I(f = 1 \text{ Hz})}{f} df = \ln\left(\frac{f_2}{f_1}\right) S_I \quad (f = 1 \text{ Hz}) \quad (12)$$

where $S_I(f = 1 \text{ Hz})$ is the drain current noise power spectral density at 1 Hz and the integral is between the largest and smallest frequencies sampled. Given that $\ln(f_2/f_1)$ is only weakly dependent upon the choice of bandwidth in most sensing experiments, this can be neglected to give a simplified calculated metric for SNR:

$$\text{SNR}_{\text{metric}} = \frac{\Delta I}{\delta i} = \frac{\Delta I}{\sqrt{S_I(f = 1 \text{ Hz})}}. \quad (13)$$

Sometimes ΔI is converted into the voltage regime using $\Delta I = g_m \Delta V_T$, however care must be taken with this equality as transconductance is constant only in the linear region (eqn (11)) or can only be accurately treated as such for small changes in threshold voltage.

There is experimental evidence from a number of authors demonstrating that the subthreshold region has better SNR. Operating in the subthreshold regime improved the SNR for the BioFET device of Gao *et al.* who measured prostate-specific antigen (PSA) binding and found the optimum I_{norm} agreed well with the maximum in SNR, but did not explain how SNR was calculated.³⁸ They explained their result giving resistive noise as the dominant noise source, which is proportional to carrier density and therefore greatly decreased in the subthreshold region. Tarasov *et al.*, who performed noise analysis of an ion-sensitive FET both in air and at pH 7 and defined the SNR as the reciprocal of the equivalent noise power of the threshold voltage, also demonstrated improved SNR in the subthreshold region.³¹ Heller *et al.* performed charged-molecule sensing and salt concentration-change sensing experiments and remarked that for their



A related question is therefore what region of operation is optimal for sensor Sensitivity (*e.g.* I_{norm} per pH). In the linear region (eqn (11)), I_{norm} increases as the gate voltage V_g approaches the threshold voltage V_T , *i.e.* as the subthreshold region is approached. I_{norm} is always optimised in the Subthreshold Region. Once the subthreshold region is reached, eqn (11) becomes invalid and an alternative expression (eqn (9)) for the normalised change in current is recovered, which is a function of the subthreshold slope and is independent of gate voltage. To demonstrate that these theoretical relations hold experimentally, Fig. 6 shows the results from the BioFET streptavidin binding experiment of Wen *et al.*⁴⁹ with all the relevant parameters plotted (I_{norm} , ΔI , I and g_m).

In summary, in the subthreshold region, Sensitivity is optimised as a result of the large normalised change in current, I_{norm} , which occurs when the initial drain current, I_0 , is small, however in this region of operation, the noise of the low drain current may become larger than the signal and therefore SNR or IUPAC limit of detection are not necessarily optimal.

Effect of electrolyte and surface chemistry. An important question with regards to the SNR is whether it is an intrinsic property of the device or whether the electrolyte solution has a significant effect. Rajan *et al.* defined the SNR as the transconductance (g_m) divided by the square root of the current noise power density (S_I) and measured this in pH sensing experiments.^{129,130} Focusing on the low-frequency range (around 1 Hz), they found that variation in ionic strength produced a negligible change in the SNR and concluded that the SNR an intrinsic property of the device.¹³⁰ They showed that the SNR can be maximised by tuning the gate voltages,¹³⁰ and increasing sensor area.¹²⁹ They also stated that their device had a surface functionalised with (3-aminopropyl)triethoxysilane (APTES) and demonstrated improved SNR and reduced current noise power compared with a bare SiO₂ surface,¹²⁹ hypothesising that the APTES lowered the effective density of trapped charges at the oxide–electrolyte interface.

Lu *et al.* investigated low frequency noise (~ 4 Hz) in devices with a bare SiO₂ surface but saw that the noise was ionic

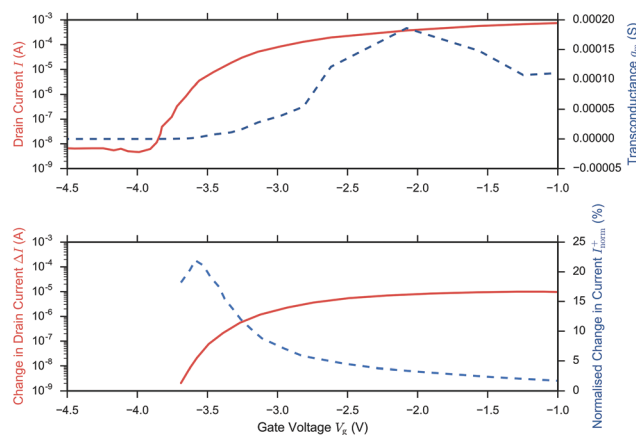


Fig. 6 $I-V_g$ device characterisation (upper figure) and corresponding streptavidin sensing data (lower figure) from Wen *et al.*⁴⁹ on addition of 4.73 pM streptavidin in 0.25X PBS and $V_d = 0.5$ V. The upper figure shows the drain current (solid) and because ΔV_T is small (0.019 V), the before/after streptavidin addition drain currents are overlaid at this scale. The drain current response was extracted from the paper and the transconductance (g_m , dashed line) was calculated in the figure, presented here from the drain current response using second order and first order central differences on the interior and boundaries respectively. Peak g_m and peak ΔI (dashed line, extracted from the paper) was observed in the linear region and optimum I_{norm}^+ (dashed, line extracted from the paper) in the subthreshold region, in agreement with theory. At low drain currents, the noise may become larger than the signal and therefore the SNR is not necessarily optimal.

strength dependent. From the paper, a 0.001 M PBS system showed a current noise power density which was 70% noisier than in 0.1 M PBS buffer, while the current noise power density showed only small pH dependence.¹³² Their results were partly rationalised with a number fluctuation-dominated current noise model but the authors acknowledged that the fluctuations cannot be fully explained in terms of modulation of the charge trapping probability with changes in ionic strength.

Top-gate sensing with a back-gate. The carrier concentration within the channel and its distribution between gates can be tuned by applying a fixed back-gate voltage,¹³³ although a back-gate is not inherently required in a FET-sensor. This back-gate voltage can be applied to optimise the transconductance/subthreshold slope with respect to a sweep of the ‘top-gate’ reference electrode in the liquid, resulting in a consequent improvement of the normalised change in current.^{30,39,134–137} This process is sometimes described as double-gated or dual-gate FET (DG-FET) operation.^{129,138} Modifying $V_{g,\text{back}}$ does not change the intrinsic properties of the electrolyte interface and therefore the top-gate shift in threshold voltage $\Delta V_{T,\text{top}}$ is constant as a function of sweeping $V_{g,\text{back}}$.³⁰ Pud *et al.* explain the physical origin of this improved transconductance as due to the back-gate voltage shifting the conducting channel from the oxide–semiconductor interface to the bulk semiconductor where less scattering occurs and a higher mobility is attained.¹⁶ In our published work,³⁰ the physical origin of the improved top-gate subthreshold slope



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For publications which present streptavidin-sensing data as a function of concentration of streptavidin, it is possible to calculate the ‘normalised change in current for a 10-fold increase in streptavidin concentration’, and equivalently for pH sensing, the response due to a change of 1 pH unit was used (pH is related to the proton concentration, eqn (5)). Assuming a linear relationship between a 10-fold change in analyte concentration (*e.g.* a pH unit for pH sensing) and a 10-fold change in current, then I_{norm} per 10-fold increase in analyte concentration is expected to be a constant. Therefore, for a set of n 10-fold changes in analyte concentration, the largest current after n 10-fold changes in concentration (I_n) can be calculated from the lowest current (I_0) and the normalised change in current (I_{norm}):

For an n-channel device with a negatively charged biomolecular analyte, I_n and I_0 correspond to the lowest and highest concentration of analyte, respectively. Rearranging for I_{norm} gives an expression for the representative normalised change in current per 10-fold increase in analyte concentration:

and will be referred to as ‘Sensitivity’ (capitalised). In this expression, I values for which the current response has saturated are excluded. The Sensitivity can also be calculated from a set of $I_{\text{norm},i}$ values (ESI section 7.3†). The Sensitivity facilitates comparison of current response between experiments performed at different analyte concentrations and is similar to the IUPAC definition of sensitivity (discussed previously). In this work the Sensitivity is calculated across a range of concentrations, the justification for this approach as opposed to, for example, using a single measurement at low analyte concentration is justified in ESI section 7.4.†

Note that not all published data are presented both unambiguously and with completeness. As a result, sometimes paper-specific assumptions had to be made in order to com-

plete this analysis. These assumptions are explained in detail in ESI section 7.5.[†]

Surface binding reactions

Equilibrium response. For quantitative biosensing experiments, the relationship between biomolecule concentration and response is required. Assuming electrostatic gating, the shift in the threshold voltage (ΔV_T) is related to the shift in surface potential at the oxide–electrolyte interface ($\Delta\psi_o$) (discussed in ESI section 2†). The precise relationship between the addition of biomolecular charge and the change in surface potential remains one of the most important modelling problems in this field,^{20,140–142,149–158} however an appealing model, due to its simplicity, is to approximate the oxide–electrolyte interface as a parallel plate capacitor (Helmholtz–Perrin Theory).¹⁵⁹ The change in charge on the surfaces of the capacitor, ΔQ , with capacitance C_0 can give the voltage shift simply as:

$$\Delta\psi_0 \approx \Delta V_T = \frac{\Delta Q}{C_0}. \quad (17)$$

From this simple model it can be seen that a concentration-dependent response with increased biomolecule binding to the surface is expected, a response demonstrated by many authors for streptavidin sensing.^{26,28,32,36,49,51,107,120,160–166}

The simplest model for biomolecule-binding reactions, is the first-order Langmuir model of surface-reactions (details in ESI section 10[†]). In this model the surface reaction of analyte (*A*) to the surface sites (*S*) is treated simply:



The resulting equilibrium constant, $K_d = [\text{AS}]/([\text{AS}][\text{S}])$, bears many similarities to the well-known $\text{p}K_a$ for the case of proton-binding, as illustrated in Fig. 7. Solving the rate equation for the above system and finding the steady-state solution analytically, results in the Langmuir equation, also shown in Fig. 7.

The equilibrium coverage, modelled by the Langmuir equation, is related to device response. As a result, the dynamic range of the sensor (*i.e.* the concentration range of analyte upon which it operates) can be seen to be fundamentally limited by K_d . At bulk concentrations $[A]_0 \gg K_d$, the sensor is expected to be saturated. For systems in which there is a low K_d , such as streptavidin–biotin binding, based on the Langmuir equation, the high affinity means that a large proportion of the total number of sites $[S]_{\max}$ are expected to be occupied (*i.e.* $[AS]/[S]_{\max} \sim 100\%$) even at low concentrations of bulk analyte. For surface-bound systems, there is evidence to suggest the K_d may be smaller than the solution measured K_d .¹⁶⁷

If a finite volume of analyte is used and the ligand has a high affinity, then the effects of finite numbers of sites can become important as the bulk concentration of ligand can be depleted by reactions to the surface.¹⁶⁸ Rajan utilised numerical simulation to demonstrate the consequences of ligand-depletion on the equilibrium coverage ($[AS]/[S]_{\max}$) and showed that in low K_d , small volume systems (*i.e.* low number of mole-

Proton (H⁺) binding to Acid (A₁)

Low [H⁺] High pH $\text{pH} \geq \text{p}K_a + 2$ $\approx 100\%$ deprotonated	$\text{H}^+ + \text{A}^- \rightleftharpoons \text{HA}$ $\text{pH} = -\log[\text{H}^+] = \text{p}K_a + \log\left(\frac{[\text{A}^-]}{[\text{HA}]}\right)$	High [H⁺] Low pH $\text{pH} \leq \text{p}K_a - 2$ $\approx 100\%$ protonated
--	--	--

Ligand Binding of Analyte (A) to Surface receptor (S)

Low [A] $[\text{A}] \leq 0.01 K_d$ $\log([\text{A}]) \leq \log(K_d) - 2$ $\approx 100\%$ unbound	$\text{A} + \text{S} \rightleftharpoons \text{AS}$ $\frac{[\text{AS}]_{\text{eq}}}{[\text{S}]_{\text{max}}} = \frac{[\text{A}]_0 / K_d}{1 + [\text{A}]_0 / K_d}$	High [A] $[\text{A}] \geq 100 K_d$ $\log([\text{A}]) \geq \log(K_d) + 2$ $\approx 100\%$ bound
--	--	--

The lower limit of detection of the sensor is restricted to single-molecule binding; the critical concentration at which only one target molecule binds the sensor at equilibrium ($c_{\min}$), to a sensor of area A with density of receptors $[S]_{\max}$ is:¹⁶⁹

$$c_{\min} = \frac{K_d}{[S]_{\max} A}. \quad (19)$$

From the work of Elfström *et al.*,²⁸ given a sample volume, V , of 200 μL solution and concentration, c_{max} , of 0.5 nM, the total number of molecules in the solution ($= c_{\text{max}}V$) can be obtained. Assuming all molecules bind to the surface (valid

$$\rho_{\text{Elfström}} = \frac{c_{\text{max}} V}{A} = 0.015 \text{ molecules per nm}^2.$$
$$\rho_{\max} = 0.04 \text{ molecules per nm}^2$$
$$\rho_{\text{Duan}} = 0.000072 \text{ molecules per nm}^2$$

Won Hee Lee *et al.*³² and Wen *et al.*⁴⁹ did not report the volume of analyte solution utilised, making this analysis impossible. However, using the data from Stern *et al.*, a similar calculation can be performed. Their nanowire-based FETs were selectively functionalised, in principle creating a device with very low active surface area.^{36,170} Fluorescence imaging confirmed that there was selective binding of streptavidin.¹⁷⁰ Given the surface area and sample volume of their device,[¶] the calculated saturation concentration c_{max} should be approximately 20–200 fM, however the measured signal increased up to 1 nM. A possible explanation for this disagreement would be that the high concentration response is due to longer-range interactions from streptavidin not directly bound to the nanowires (*e.g.* multilayer coverage) or that despite the selective-functionalisation, there is sufficient non-specifically bound streptavidin elsewhere on the device that the effective concentration at the surface of the nanowire is much lower than the initial bulk concentration.

A total of 100 μL sample volume exposed to the surface (with 10 μL exposed at any one point in time) which, assuming perfect selective functionalisation of just the nanowires, has an functionalised surface area of $\sim 3 \times 10^7 \text{ nm}^2$.

The transport of analyte to the surface of the sensor is important for setting the time-to-response and therefore the effective detection limit within a given time-frame. For example, using simple analytical arguments to describe the diffusive flux, detection of 1 fM biomolecular analyte using a 30 nm radius nanowire is expected to take an average detection time on the order of hours, compared to orders of magnitude longer using planar devices.^{147,148} In contrast to biomolecular sensing, pH sensing provides fast response times as the diffusion constant for H^+ is orders of magnitude higher due to Grotthuss transport, in which H^+ transport along hydrogen bonded networks.^{174,175} The theoretical detection time for an individual proton is non-trivial to calculate due to the complexity introduced by acid-base equilibria and Grotthuss transport, which cannot be described by simple Brownian motion.¹⁷⁶ In summary, for BioFET design the limitations imposed by the mass transport, binding kinetics and geometry

If the normalised change in current was measured between several different concentrations, the normalised change in current for each is shown on the graph and the relative concentration used is indicated *via* the size of the marker symbol. No

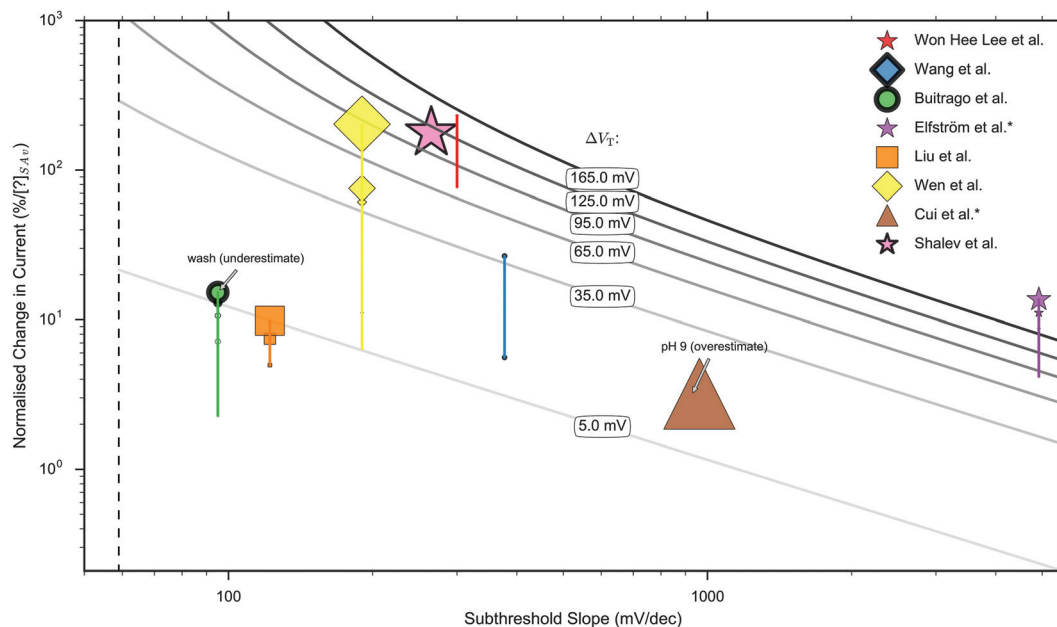


Fig. 8 Measured 'unbounded' normalised change in current (I_{norm}) versus measured subthreshold slope for sensing operating in the subthreshold region.^{26,28,32,37,49,70,107,178,179} Publications indicated by an asterisk (*) have no liquid top-gate thus the plotted properties correspond to the response measured by a back-gate. The remaining publications were performed with a liquid top-gate, with corresponding subthreshold slope measured from the current response upon sweeping the liquid top-gate voltage. The shape of marker symbol corresponds to the type of device: 3D nanowire stack = ●, nanoribbon = ★, nanobelt = ■, planar = ◆, nanowire = ▲. The size of the marker symbol is proportional to the log(concentration) of streptavidin used. The vertical lines joining some markers corresponds to a series of measurements under different concentrations. A thicker outline of a marker symbol indicates an experiment made under high ionic strength, where the maximum ionic strength was 1x PBS and the minimum 0.01x PBS. The dashed line shows the theoretical lower limit of the subthreshold slope of 59 mV dec⁻¹ available to classical FETs at room temperature. The solid grey lines are isocontours calculated by inserting the shift in threshold voltage, ΔV_T , into eqn (9), the value inserted is labelled upon each line. *i.e.* for a data point on top of a line, it represents its calculated shift in threshold voltage and therefore the calculated change in surface potential. The results of Buitrago *et al.*^{70,178} and Cui *et al.*³⁷ are likely to be an underestimate and overestimate respectively for the reasons given in ESI 7.5.† A more detailed explanation of the data extraction, the marker thickness, and a replot with a linear x-axis can be found in ESI 7.6.† A clear relationship between I_{norm} and subthreshold slope is not apparent.

clear relationship between the normalised change in current and subthreshold slope was observed. High ionic strength can reduce response and therefore increasing ionic strength is indicated in Fig. 8 with a thicker marker outline.

Each solid curve shown in Fig. 8 shows the calculated shift in threshold voltage for a given subthreshold slope and I_{norm} (from eqn (9)), and therefore it can be seen that most measurements correspond to a shift in the threshold voltage, ΔV_T , of between 5 mV and 165 mV.

Data analysis of responses in streptavidin and pH sensing

Introduction. The optimal response of a pH sensor can be estimated using eqn (9), which provides the relationship between changes in surface potential and FET current response in the subthreshold region. This equation shows that Sensitivity is expected to be a function of the transistor subthreshold slope and shift in threshold voltage. When operating at fixed gate voltage, a device with a lower value of subthreshold slope will by definition (eqn (2)) correspond to a larger change in drain current (*i.e.* higher Sensitivity) than a device with higher subthreshold slope, assuming that for both

devices (a) subthreshold slope is constant with analyte binding, *i.e.* parallel shift in $I_{\text{ds}}-V_g$ transfer and, (b) the surface potential shift per unit of analyte is constant between the different devices (eqn (9)). Assumption (a) has been widely experimentally evidenced for biosensing and pH sensing.^{30,35,70,119,180} Assumption (b) has been shown to be approximately the case for pH sensing and assuming a consistent oxide material between devices.^{181,182} As a result of this, pH sensing FET Sensitivity can be predicted *via* eqn (9) prior to pH sensing experiments simply by measuring the subthreshold slope and using knowledge of the oxide material. In this work we perform a similar analysis incorporating streptavidin-sensing data in order to understand whether a similar trend can be identified, and as a method of graphically rationalising differences in Sensitivity between biosensing experiments.

Other than the subthreshold slope, eqn (9) shows that the Sensitivity is a function of the surface potential variation induced by analyte binding. Oxide materials demonstrate a surface potential variation which is approximately linear as a function of pH (over a narrow range of pH values), which can therefore be represented as a single number, for example, a SiO₂ surface typically has a pH-induced surface potential



change of 33 ± 6 mV per pH.¹⁸ For this system, a maximum pH Sensitivity of between 190% to 360% is calculated assuming an optimal subthreshold slope of 59 mV dec^{-1} (ideal at room temperature). This is useful for the design of pH sensors as it indicates that optimal Sensitivity in the subthreshold region can be obtained by choosing a dielectric with high pH Sensitivity (e.g. HfO_2 rather than SiO_2) and engineering a good transistor (low subthreshold slope). An important research question is whether simple device-design strategies like these can also be developed for biosensing, which is a primary motivation for this analysis.

The precise relationship between pH Sensitivity and biomolecular Sensitivity remains an active area of research.^{69,74,142} Shalev *et al.* provided evidence to suggest that biomolecular sensing may be a result of biomolecules altering the chemical equilibria of titratable groups on the surface, and therefore for some systems pH-Sensitivity and biomolecular-Sensitivity are expected to be related.⁶⁹ In this analysis pH-sensing and biomolecular-sensing Sensitivity have been considered separately.

The analysis in this section focuses upon pH-sensing and streptavidin/biotin interactions on FET devices operating in the subthreshold region.

Results. Where sufficient data was available, Sensitivity was calculated for the previously presented streptavidin-sensing data^{26,28,32,49,70,107} and the results are plotted against subthreshold slope in Fig. 9 and 10. In our previous work,¹⁸ we published a collation of pH sensing data,^{18,38,44,183–190} using a metric equivalent to Sensitivity, and this data is also included in Fig. 9 and 10.

As shown in Fig. 9, the Sensitivity values ranged between 3% and 84% per 10 fold increase in streptavidin, whereas the pH-sensing Sensitivity varied between 1% and 600% per pH. Therefore, on the basis of available streptavidin-sensing data, a value for Sensitivity in streptavidin-sensing comparable to optimal pH sensing has not been obtained. The pH-sensing results are consistent with a response of 33 mV per pH for SiO_2 systems and 59 mV per pH for HfO_2 systems. Whilst Sensitivity is a function of subthreshold slope, it is also a function of surface potential variation, and therefore a correlation is only expected when the latter is constant. For pH sensing, for a given oxide material a constant potential shift per pH was identified, such that for a given oxide, subthreshold slope alone can be used to predict Sensitivity using eqn (9). In contrast, for streptavidin-sensing data, despite the same surface functionalisation and analyte between the experiments (oxide-APTES-biotin-streptavidin), the subthreshold slope alone was not sufficient to accurately predict Sensitivity. This result is indicative of large surface potential variation per 10-fold increase in streptavidin concentration.

In order to better visualise the calculated shift in threshold voltage, the subthreshold slope of each data point and corresponding Sensitivity value was inserted into eqn (9) to obtain the calculated shift in threshold voltage per 10-fold increase in analyte concentration for that measurement. This was then used to plot Fig. 10. In this figure, isocontours calculated using eqn (9) are drawn from red to blue for values of the

Sensitivity increasing in 1% increments in the range 1–150%. From Fig. 10, again it can be seen that the pH sensing data showed a shift in threshold voltage which is generally consistent for a particular oxide material (~ 33 mV per pH for SiO_2 with one anomaly, ~ 59 mV per pH for HfO_2 with one anomaly), in contrast to the streptavidin-sensing data (oxide-APTES-biotin surface chemistry) which showed no clear trend in threshold voltage shift.

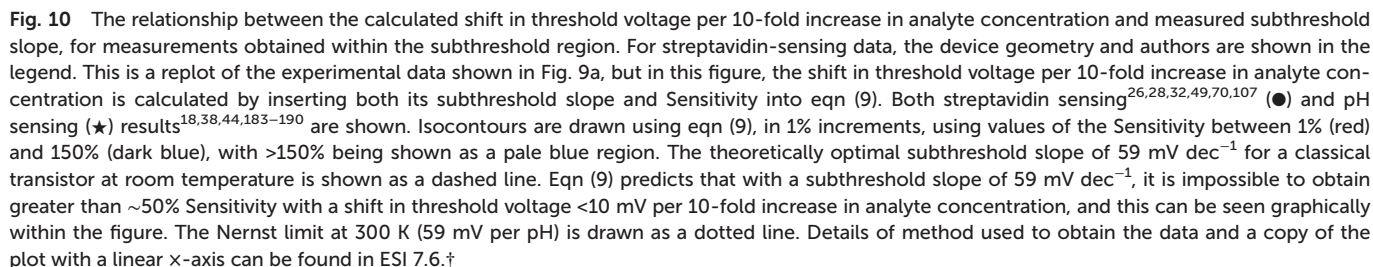
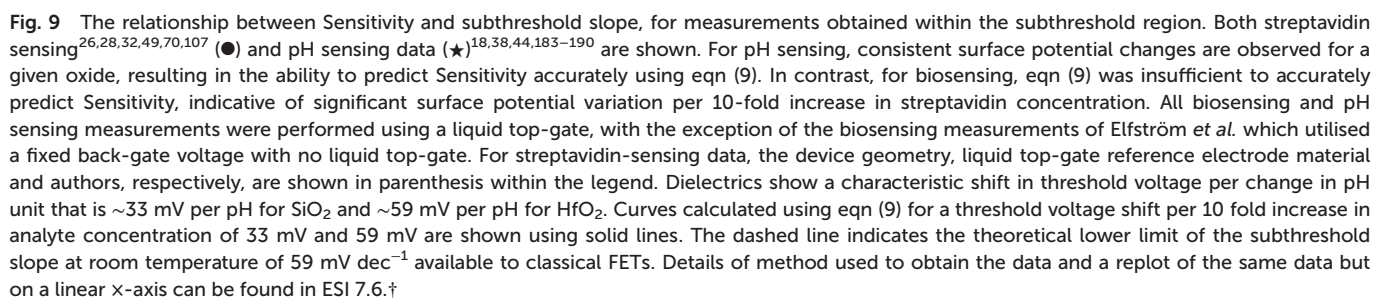
Discussion. A large degree of the variability in response illustrated in Fig. 8 is due to the variety of concentrations of analyte used between different experiments. In order to account for this variability, the Sensitivity was calculated and presented in Fig. 9. From the streptavidin-sensing data, the maximum Sensitivity (up to 84%³²) was lower than the maximum observed in pH sensing data (up to 600%⁴⁴). For some devices,^{28,32,49,107} the Sensitivity for biomolecular analyte was similar to that expected from a pH sensor with the same subthreshold slope, however for the devices of Liu *et al.*²⁶ and Buitrago *et al.*,⁷⁰ a significantly lower Sensitivity than pH sensing results was obtained, despite the low subthreshold slopes (112 mV dec^{-1} and 95 mV dec^{-1} respectively). This can be explained as a consequence of the fact that these biosensing experiments are not obtaining both the high shift in surface potential and low subthreshold slope that is required for a high Sensitivity (shown more clearly in Fig. 10). It should be noted that the low Sensitivity of 2% seen in the study of Buitrago *et al.*⁷⁰ may be simply a consequence of the response not being measured at equilibrium, as in Buitrago it was stated that analyte was ‘immediately washed away by PBS’.¹⁷⁸ The high Sensitivity of Won Hee Lee *et al.*³² of 84% may be unreliable given a key claim of the conference paper is that subthreshold slope is changing with increased analyte concentration and this is not clearly demonstrated in the paper (as discussed in ESI section 7.5†).

Eqn (9) shows that Sensitivity is a function of both the subthreshold slope and surface potential variation induced by analyte binding. Fig. 9 and 10 showed that a consistent surface potential change per pH is evident for pH sensing but not for streptavidin-sensing. This can be explained as for biosensing, even small variations between experiments in the surface chemistry or treatment can result in drastic changes in the amount of bound protein, and thus drastic changes in the surface potential variation. As a result of these variations, the relationship between subthreshold slope and Sensitivity is not clearly seen in Fig. 9 and 10 based on the current limited data. This result demonstrates the increased importance of effective surface treatment for biosensing experiments as compared to pH sensing experiments.

The theory (eqn (9)) states that it is unlikely to obtain high Sensitivity with a high subthreshold slope value as this would require extreme changes in surface potential. In agreement with this, neither pH sensing nor streptavidin sensing data was found in the high Sensitivity/high subthreshold slope region.

Regarding the calculated shift in threshold voltage per 10-fold increase in analyte concentration (Fig. 10), for pH





sensing, the results showed a shift in threshold voltage per pH which was consistent with the material of the oxide, as previously observed in the literature.¹⁸ Specifically, despite the fact that the pH sensing results are from disparate literature sources, they show highly consistent threshold voltage shifts of ~ 33 mV per pH (SiO_2) and ~ 59 mV per pH (HfO_2). Most SiO_2 systems show a pH response of approximately 33 mV per pH,^{18,38,183,184} with some reports demonstrating a higher value of 43 mV per pH,¹⁹¹ 46 mV per pH⁹⁷ and 42–50 mV per pH¹⁸⁵ which may be due to different surface preparation procedures. Other oxides demonstrate higher values, such as Ta_2O_5 at 52 mV per pH¹⁸⁶ and HfO_2 at approximately 59 mV per pH.^{44,187–190} One pH sensing result⁴⁴ in Fig. 10 can be seen to exceed the Nernst limit (59 mV per pH) with a calculated threshold voltage of 66 mV per pH. This result is likely due to small error in the extraction of the subthreshold slope: in the high Sensitivity, low subthreshold slope region of the plot, the effect of small errors in subthreshold slope is high. As an example, for this measurement, an error of only 8 mV dec^{-1} in the extraction of the subthreshold slope would place this result within the Nernst limit.

The physical origin for these material-consistent surface potential shifts can be explained in terms of surface complexation models, also known as site binding models, which were first introduced by Yates *et al.*¹⁹² and later refined by Healy *et al.*¹⁹³ In these models, an increased density of hydroxyl groups on the surface corresponds to an increase in surface potential shift per unit pH.^{123,191} This theory is supported by experimental work which has shown that blocking hydroxyl groups on the surface with organic functionality can reduce the surface potential shift per pH.^{123,194,195}

At the lower limit, Fig. 10 shows that most oxide surfaces demonstrate surface potential shifts as a function of pH which are above approximately 25 mV per pH. As this value is a shift per 10-fold increase in analyte ($\text{pH} = -\log([\text{H}^+])$), this can be contrasted to streptavidin sensing experiments, which showed both: (a) a greater variation in possible shifts in threshold voltage per 10-fold increase in streptavidin concentration and, (b) the possibility of very low shifts of less than 10 mV (Liu *et al.*²⁶ and Buitrago *et al.*⁷⁰). If a device has a shift in threshold voltage per 10-fold increase in streptavidin concentration that is <10 mV, it can be seen in Fig. 10 that it is theoretically impossible to obtain a Sensitivity value greater than approximately 50%. This form of analysis can be used to provide information for improved BioFET design; it suggests that if a device shows a low threshold voltage shift per 10-fold increase in analyte, then design focus should be put upon enhancing the change in surface potential by optimising the surface chemistry. Below we hypothesise an explanation for this contrast in pH and biosensing data, and posit that such considerations are needed in order to design BioFETs which are as sensitive to analyte as the already commercially-successful (pH) ion-sensitive FETs.

Comparison of streptavidin and pH sensors

There are many potential factors which could explain the observed differences in the threshold voltage shifts between

pH-sensing and biosensing experiments. Several of these factors relate to the surface and its chemistry. Fig. 11 shows a schematic which attempts to summarise possible differences in the surface chemistry, and therefore measured surface potential shifts, between pH and biosensing experiments. This section explores each of these factors in relation to the operation of a FET sensor system, highlighting the important implications for the understanding and design of such a sensor for biomolecules. Each section heading is labelled corresponding to the labels within Fig. 11.

Receptor density (i). Given the ability of Surface Complexation Models to explain the observed shifts in threshold voltage for pH sensing data, it is plausible a similar mechanism can explain much of the variation in biosensing data. In pH sensing, the density of analyte receptors is described by the density of hydroxyl groups at the oxide surface, whereas for streptavidin–biotin biosensing the density of analyte receptors is the density of biotin available at the sensor surface. Using analogous arguments to those used in Surface Complexation Models, an intuitive hypothesis is that the shift in threshold voltage per 10-fold increase in analyte

Sensitivity in Subthreshold Region:

$$= e^{\frac{\ln(10)\Delta V_T}{SS}} - 1$$

$$\Delta V_T \propto \Delta \text{Surface Potential}$$

SS – Subthreshold Slope
• Transistor transfer characteristic

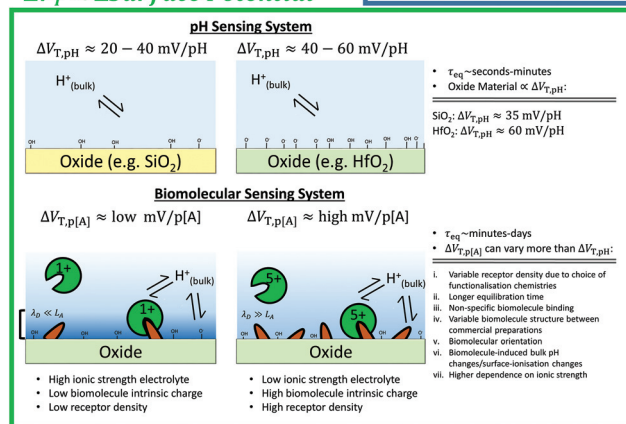


Fig. 11 Comparison of factors influencing the surface potential in a pH sensing experiment (above) with those in a biomolecular sensing experiment (below). The systems on the left and right have a low and high receptor density, respectively. Receptors for pH sensing are hydroxide groups, whereas for biomolecular detection specific receptors (shown as yellow wedges) are required. For pH sensing, the receptor density is the determining factor the change in surface potential, and therefore the shift in threshold voltage per pH ($\Delta V_{T,pH}$). In contrast, for biomolecular systems, many more factors can affect the shift in threshold voltage per 10-fold increase in concentration of analyte ($\Delta V_{T,p[A]}$). The net charge of the biomolecular system (shown as green circles) can influence response. The Debye length (λ_D) is compared to the distance of the analyte from the surface (L_A) on the figure, which can significantly affect BioFET signal. Here τ_{eq} is used to refer to the typical time-scale to equilibration. The equation shown is a rearranged form of eqn (9) from the main text.



example, due to reactions with carbon dioxide in the air changing the acidity of the solution.

As previously discussed, signal change being directly proportional to bound-analyte is a weaker assumption for biomolecule sensing, because the bulk pH of the solution can be modified by analyte and detected by the sensor.⁷⁴ The practice of using minimal buffering capacity and high biomolecule concentration (added in a single aliquot) is not uncommon^{26,44,66,106,119,196} leading to a solution which is not buffered adequately. Low concentrations of analyte are closer to the reality of applications in medical diagnostics in which biomolecules are often only present at low concentrations within blood samples. Alternatively a solution should be used which has the same buffering capacity as PBS at a lower ionic strength, for example, using just the sodium phosphate component of the PBS buffer.⁷⁴

Ionic strength (vii). As discussed, variability in ionic strength between experiments is expected to result in variability in the response due to changes in the extent of screening of the biomolecular charges.

Nanoscale device geometry (viii). The analysis in Fig. 9 shows that using a simple planar (macroscopic) device, Wen *et al.*⁴⁹ and Won Hee Lee *et al.*³² achieved higher Sensitivity than the nanobelt arrays of Liu *et al.*²⁶ and Cheng *et al.*²⁵ or the nanowires of Elfström *et al.*¹²¹ This shows that even simple planar geometries can obtain high Sensitivity for biomolecule detection compared to more complex nanoscale geometries. As introduced previously, whilst nanoscale devices can offer improved subthreshold slope compared to conventional devices, nanoscale dimensions are not required to obtain a near-ideal subthreshold slope for a device. For pH sensing, a consequence of this is that planar devices can provide similar Sensitivity to nanoscale devices, and that their response is accurately determined by the choice of oxide (*i.e.* shift in threshold voltage per pH) and the subthreshold slope.¹⁸ In contrast, for biomolecular sensing, the shift in threshold voltage per 10-fold increase in analyte is not primarily determined by the choice of oxide material, and can instead be explained as due to the aforementioned factors *e.g.* the receptor density.

Summary and future outlook

Biosensor design is a highly interdisciplinary field. Understandably, this has led to a diverse range of experimental designs and varying goals, resulting in many publications focusing on some aspects of sensor characterisation in more detail than others. It is important going forward that clear standards be developed and used for the purpose of developing reliable sensors with reproducible and translatable characteristics.

Reporting the drain current as a function of top-gate voltage is encouraged, as this provides information on the quality of the transistor, but also can be used to extract information regarding the change in surface potential at the sensor surface–electrolyte interface. This information can be used to quantitatively compare sensor capability.



At this stage in FET-sensor research, the focus is predominantly upon Sensitivity-enhancement. As the field moves toward commercial applications, it will become increasingly important to provide controls which evidence specificity as well as Sensitivity. In order to maximise Sensitivity, many BioFET experiments are performed at high concentration of biomolecules with low buffering capacity solution (diluted buffer), potentially resulting in a significant non-specific response due to biomolecule-induced changes in the bulk pH of the solution.

Metrics of reporting device performance have been discussed. The normalised change in current (I_{norm}) is currently almost ubiquitously used performance metric in the field and is useful for characterising the current response as it often reduces device-to-device variation compared to the absolute change in current. The normalised change in current is optimised in the subthreshold region, which is not necessarily the region of optimal SNR. The region of optimal SNR has been shown to be device dependent, with reports of optimum SNR both near to the region of maximum transconductance (which is usually in the linear region) and in the subthreshold region. An improved figure-of-merit that has been proposed in the literature is the SNR or IUPAC limit of detection, which involve identifying the noise levels in the signal. The importance of reporting the polarity of the normalised change in current should be emphasised, as the absolute value can lead to ambiguity in interpreting experimental results.

Reference electrodes are crucial for consistent and reliable sensing results, however, the difficulty in reference electrode miniaturisation has led to common usage of pseudo-reference electrodes in their place. Several publications suggest that Pt-based pseudo-reference electrodes are unsuitable for reliable sensing. In contrast, Ag/AgCl pseudo-reference electrodes have the potential to be a viable alternative to conventional reference electrodes. Using a back-gate as the pseudo-reference electrode without a top-gate liquid (pseudo-) reference electrode has been reported to provide unreliable results. The use of a top-gate liquid electrode together with an optimised back-gate electrode voltage which modifies the carrier concentration profile within the device can improve the subthreshold slope. By using a fixed top-gate liquid electrode voltage and sweeping the back-gate voltage, a measurement can be performed resulting in an amplified back-gate threshold voltage shift, however in this setup any SNR enhancement is expected only when the measurement equipment sets the lower limit of performance.

Streptavidin biochemistry was reviewed, and it was shown that there is a significant amount of variability for structural and electrical properties of streptavidin between different preparations of the biomolecule. An analysis of the sequence was presented which highlights that different commercial preparations are likely to have different structures, and as even small changes in electrodynamic properties of a biomolecule are in principle detectable by a BioFET, it is recommended that authors report the commercial origin of their biomolecule for reproducibility.

For pH sensing, it is known that the Sensitivity can be increased by optimising the transistor design such as to have a

low value of the subthreshold slope and by choosing an oxide material which has a large shift in surface threshold voltage per pH *e.g.* HfO_2 . In contrast to Sensitivity in pH sensing, the analysis of the literature presented in this review revealed that streptavidin-sensing Sensitivity showed no clear dependence on subthreshold slope. This observation can be attributed to streptavidin-sensing demonstrating a much greater variation in the threshold voltage shift per 10-fold increase in analyte concentration between devices, even when devices shared the same oxide material. This study shows that the pH sensor design-strategy in which focus is upon subthreshold slope optimisation and oxide material choice does not directly transfer to biosensing, because, for example, a poorly functionalised biosensor surface is expected to always have negligible Sensitivity regardless of the choice of oxide. This has addressed one of the original motivations of this work to investigate whether optimisation strategies that had been employed for pH sensors would be directly transferable to optimisation of biomolecular sensing. The analysis presented suggests that subthreshold slope improvement is not likely to be as beneficial for biosensor optimisation as it is for pH sensing because the limiting factor for Sensitivity is often the surface chemistry. In conclusion, the quantitative analysis presented here suggests that in order to optimise FET-sensor Sensitivity, the device should be operated in the subthreshold region. Device design optimisation should focus upon the simultaneous optimisation of both the subthreshold slope of the device and the electrolyte-oxide surface chemistry. For example, even using an ideal classical transistor (optimal subthreshold slope of approximately 59 mV dec^{-1}), if the surface is poorly functionalised with analyte receptors it will have a low Sensitivity.

The variability in the shift in threshold voltage per 10-fold increase in analyte concentration for biosensing responses was greater than that found for pH sensing responses. This higher variability can be explained by a variety of factors. pH models predict the most important factor in influencing this shift is the density of analyte-receptor (*e.g.* hydroxyl groups) at the surface, and therefore this could be an important factor determining the variability in biomolecular response. Further, the effects of ionic strength and buffering capacity in biosensing are much more pronounced than for pH sensing due to the biomolecule distance from the surface, and due to the ability of the biomolecule to change the pH of the buffer. Additional variability in biosensing threshold voltage shift per 10-fold increase in analyte concentration can originate from loss of biomolecule due to non-specific binding. This is particularly problematic for nanoscale dimension devices in which the bio-sensitive regions of the device can have a small surface area relative to the total exposed surface area. As discussed, a fundamental source of variability can originate from the biomolecule itself.

By measuring the response-curve as a function of analyte concentration, researchers are provided with important information for device design: the saturation point of the curve can indicate the density of bound analyte, the response per 10-fold



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