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

Benjamin M. Lowe,a Kai Sun,a Ioannis Zeimpekis,a Chris-Kriton Skylarisb and Nicolas G. Green,c†

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.

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 architectures12–14 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...
analyte, this review provides the first comprehensive quantitative analysis of the FET-sensor response. Streptavidin-sensing was primarily chosen as a model system due to its general wide spread usage and well-understood (bio)chemistry. This is compared with pH sensing, which is a better understood application of FET sensors, and has already been well-characterised. This review also highlights factors that can influence the response yet have not always been fully appreciated, thereby resulting in sub-optimal experimental design.

One particular design aspect in pH sensing using FET-sensors\textsuperscript{18–20} is that the oxide material plays a dominant role in determining the magnitude of the response. An important motivation for this work was therefore to investigate whether similar trends holds for biomolecular-sensing BioFET experiments.

This review is divided into three main sections. First, the operating principle of FET-sensors and relevant physicochemical properties of streptavidin are introduced. Then relevant metrics for comparing device performance are critically reviewed, and finally, a quantitative analysis of streptavidin-sensing and pH-sensing literature is presented.

### Operation of field-effect sensors

FET-sensors are similar to Metal–Oxide–Semiconductor Field-effect Transistors (MOSFETs) wherein the gate is replaced by an electrolyte resulting in an oxide–electrolyte interface. An example setup is shown in Fig. 1.\textsuperscript{21} A voltage can be applied via a reference electrode in the liquid and operated similarly to the gate in a MOSFET, with gate voltage: \( V_g \). For an n-channel device, an increasingly positive gate voltage will result in the formation of the conductive channel beneath the interface between the semiconductor and the gate insulator (oxide). In contrast, for a p-channel device, an increasingly negative gate voltage will result in the formation of the conductive channel.

A typical MOSFET is a bulk transistor with a source/channel/drain configuration like n+/p/n+ or p+/n/p+. In contrast, the small thickness of nanowire-based or ultra-thin SOI FET-sensors facilitates complete depletion of the channel. This allows these devices to operate in junctionless (i.e. n/n/n, p/p/p) or even undoped (i.e. n/i/n or p/i/p) configurations. FET-sensors can also be operated in a MOSFET-like configuration (n/p/n, etc.).

---

Benjamin M. Lowe obtained a first class degree in Natural Sciences from the University of Bath in 2012. His degree programme included biological, chemical and mathematical sciences and was supported with a Year in Industry working in Computer Aided Drug Design at the Novartis Institute for Biomedical Research. As part of a four-year Doctoral Training Centre PhD programme in Complex Systems Simulation at the University of Southampton, he obtained his PhD in simulation of the interfacial physics of field-effect biosensors. He is currently continuing this work as part of a post-doctoral position at the University of Tokyo.

K. Sun received his BEng degree in Dalian University of Technology (China) in 2004. He was then awarded his MSc and PhD degrees in the University of Southampton (UK) in 2007 and 2011, respectively. He is currently a post-doctorate researcher in the University of Southampton. He has been actively working on thin-film transistor platform for biosensing applications since 2007.
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 nanoribbon18,27–12 and nanowire15,29,33–37,38 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.43 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-

---

**Ioannis Zeimpekis**

Ioannis Zeimpekis graduated from the University of Patras, Greece in 2005 with a degree in Physics. He then received his MSc and PhD from the University of Southampton, UK in 2007 and 2012. He is currently a senior research fellow in the University of Southampton. His research has focused on electrical biosensing systems and thin film transistors. More recently, he has been working with chalcogenide materials with both optical and electronic applications.

**Chris-Kriton Skylaris**

Chris-Kriton Skylaris is Professor of Computational Chemistry at the University of Southampton. Chris obtained a first class (apóter) Chemistry degree from the University of Athens in 1996 and a PhD in Quantum Chemistry from the University of Cambridge in 1999. He then carried out post doctoral research at the Physics Department in Cambridge and in 2004 was awarded a Royal Society University Research Fellowship. His work focuses on the development of theory, algorithms and codes for quantum mechanical calculations from first principles, and their applications. He is a principal and founding author of the ONETEP code for linear-scaling quantum chemistry calculations.
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. Miniaturisation of this system is problematic and results in reference electrodes with reduced lifetimes 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. Although Pt-pseudo reference electrodes have been used in various streptavidin-sensing experiments, 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’).\(^2\)\(^7\)\(^3\)\(^3\)\(^5\)\(^2\)\(^3\)

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. Rajan demonstrated experimentally that Ag/AgCl pseudo-reference electrodes are a suitable alternative to a conventional reference electrode,\(^5\)\(^2\)\(^3\)\(^4\) 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 Rajan\(^5\)\(^2\)\(^4\) and Rim et al.\(^3\)\(^4\) 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 suitable for biosensing. For a more detailed review of pseudo-reference electrode validation experiments, see ESI section 1.\(^1\)

**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.\(^5\)\(^8\)\(^3\)\(^5\)\(^5\)\(^5\)\(^5\)\(^6\) Nonetheless, it is not uncommon for devices to be fabricated without any reference electrode in the liquid,\(^2\)\(^5\)\(^3\)\(^6\)\(^3\)\(^7\)\(^7\)\(^7\) 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.\(^3\)\(^9\)). 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.\(^2\)\(^7\)\(^5\)\(^2\)

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.\(^5\)\(^8\)

---

Nicolas Green is currently a Reader (Associate Professor) at the University of Southampton, working in the Nano Group in the Department of Electronics and Computer Science, and has held a Marie Curie personal research fellowship at the University of Sevilla, Spain and a Royal Academy of Engineering personal research fellowship at the University of Glasgow and Southampton. Core research interests are in AC 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.

---

**Fig. 2** FET regions of operation. \(m\) is the body-effect coefficient (\(m \geq 1\)). Adapated from figure by Taur and Ning (2013).\(^5\)\(^8\)
The threshold voltage \(V_T\) is important in defining the region of operation and can qualitatively be described as the minimum gate voltage \(V_G\) for which the device conducts a significant current and is therefore ‘on’. More precisely, it describes the value of the gate voltage required to form an inverted channel, in which the induced carrier concentration in the inversion layer reaches the carrier concentration in the channel (bulk carrier concentration). A detailed description of the threshold voltage and its significance to sensing is provided in ESI section 2.†

In a traditional Metal–Oxide–Semiconductor FET (MOSFET), when the gate voltage is low (i.e. \(V_G < V_T\)), the drain current \(I\) is referred to as the ‘subthreshold current’ and the device is operated in the ‘subthreshold region’. This region is often used for FET-sensors because the response upon analyte addition can be enhanced with the drain current increasing exponentially rather than linearly dependent upon changes in the gate voltage. This is described by

\[
I \propto e^{\frac{qV_g - \varphi m}{kT}},
\]

where \(m\) is the empirical constant called the body-effect coefficient and \(q\) is the electronic charge. This expression is derived from MOSFET drift-diffusion equations shown in ESI section 3.†

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 \(\Delta V_S\) 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 \(\Delta V_T\) and the change in drain current \(\Delta I\) 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 response to analyte are the normalised change in current \(I_{norm}\) and the shift in threshold voltage \(\Delta V_T\). 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 \(I\) by one decade and is determined from measurements as the inverse of the slope of the linear region on a \(\log_{10} I\) versus \(V_{ds}\) graph at constant \(V_{ds}\):

\[
SS = -\frac{\partial V_S}{\partial \log_{10} I} |_{V_{ds}}
\]

When \(V_{ds}\) is greater than a few \(k_BT/q\) (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. The subthreshold slope can be modulated by a back-gate voltage, \(V_{g,back}\). The reciprocal of the subthreshold slope is also called the ‘gate voltage swing’ or ‘subthreshold swing’. The subthreshold slope can also be defined as:

\[
SS \approx \frac{2.3mV_{ds}}{q} \approx \frac{2.3k_BT}{q} \left(1 + \frac{C_{dl}}{C_{ox}}\right),
\]

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). Therefore, the subthreshold slope has a theoretical minimum value of approximately 59 mV dec\(^{-1}\) at room temperature. A lower value of the
subthreshold slope corresponds to a larger change in current for a given change in the gate voltage, and therefore a subthreshold slope of 59 mV dec\(^{-1}\) corresponds to the upper limit for the current response of conventional FET-sensors.\(^{44}\) The subthreshold slope characterises the ability of the transistor to transduce a change in gate voltage to a change in current and therefore for both biosensors and pH sensors, the maximal change in charge in current can only be obtained with a minimal value of the subthreshold slope. This will be discussed later.

The transconductance, \(g_m\), is constant in the linear region and is therefore particularly useful for describing the properties of the device in this region. Similar to the subthreshold slope, the transconductance measures the slope of the \(I_{\text{versus}} V_g\) plot:

\[
g_m = \left. \frac{\delta I}{\delta V_g} \right|_{V_{\text{ds}}} \tag{4}
\]

The sensor response metrics are related to the effect of analyte binding producing a horizontal shift in the \(I-V_g\) curve,\(^{45}\) corresponding to a change in \(\Delta V_T\) and a change in current, \(\Delta I\). The current-response (\(\Delta I\)) can be used as the performance metric, in which case it is common practice to normalise the current (\(I_{\text{norm}}\)) as the change in current divided by the initial current. Alternatively, \(\Delta V_T\) can be used as a direct measure. Assuming an electrostatic gating mechanism, \(\Delta V_T\) is equal to the change in surface potential, a direct indication that this parameter can be used for quantitative biosensing.\(^{45,61}\)

In some cases, however, the assumption of electrostatic gating is not valid. For example, if the metal contacts to the semiconductor are not well-passivated and therefore insufficiently protected from direct interaction with the analyte, analyte-induced changes to the metal–semiconductor work function produce a signal not originating from an electrostatic gating mechanism.\(^{20,45,62,63}\) Another important example is if the reference electrode potential is modified by the analyte. A detailed description of this can be found in the literature: Heller et al. describe how the \(I-V_g\) graph (transfer characteristics) can be used to diagnose the mechanism of FET response (e.g. electrostatic gating, work function change etc.).\(^{45}\)

**Streptavidin biochemistry**

Streptavidin is one of the most common model proteins used in biosensing studies\(^{64}\) and its detection is often used as the initial proof-of-concept for new biosensor designs.\(^{28,37,65}\) The relevance of streptavidin is much broader; for example, Gupta et al. have shown that streptavidin has direct clinical relevance in capturing biotinylated interferon \(\gamma\) (MIG), which is a biomarker for several inflammatory and autoimmune disease states.\(^{66}\) Streptavidin is also commonly used in enzyme-linked immunosorbent assays (ELISA), a standard biochemical assay. As BioFETs can, in principle, detect the electric field generated by a single elementary charge\(^{67,68}\) even small changes in the electrodynamic properties of streptavidin should be detectable, making an understanding of the charge properties of streptavidin important. Surprisingly, despite its common usage in BioFET experiments and most other biosensors,\(^{64}\) a rigorous description of its expected charge as a function of pH is not available. Furthermore, many publications do not report the commercial origin of their streptavidin samples under the assumption that it is not relevant to its charging-properties.\(^{49,69-71}\) In this section, based on analysis of published biochemical literature, this assumption is proposed to be false.

Streptavidin is a tetrameric protein composed of four identical subunits each with a high-affinity binding site for its ligand, biotin.\(^{72}\) Strong chaotropic agents (6 M urea) result in its dissociation into a dimeric form.\(^{73}\) It is commonly used as a model protein in sensing studies for a variety of reasons. First, streptavidin has an extremely high affinity for biotin which can be quantified by its affinity constant (\(K_a\)) of approximately \(10^{-15}\) M.\(^{24}\) Fast binding kinetics are evident from the association rate constant of greater than \(10^7\) M\(^{-1}\) s\(^{-1}\).\(^{35}\) A sensor surface can be given high streptavidin selectivity and affinity by functionalising it with biotin. Second, the ionic strength of the buffer affects the binding affinity and protein stability of many proteins but not streptavidin–biotin affinity.\(^{74}\) As discussed later, buffer dilution does still strongly affect sensor response due to screening effects. Third, it is well characterised, widely commercially available and there is an abundance of literature on chemical-functionalisation of biotin to sensor surfaces.\(^{42,72,75}\) Finally, the isoelectric point (pI) of streptavidin\(^{76-80}\) results in a near-neutral protein under physiological conditions, which reduces its propensity for non-specific binding. Despite these advantages, it should be noted that the unusually high affinity of streptavidin for biotin makes it unrepresentative of binding characteristics of most biomolecules.

The full sequence of streptavidin, as encoded by the native gene which is naturally expressed in the bacterium *Streptomyces avidini*, is shown in Fig. 4, with residues that are often charged shown as coloured and underlined. The blue highlighted region is unlikely to be present in any commercial sample of streptavidin used for biosensing experiments as it is a signalling region that is removed in vivo.\(^{81}\) The structure of streptavidin can vary between commercial preparations due to processing steps which result in artificial truncation of the protein, for example, in order to increase the protein solubility.\(^{82}\) In 1990, Green stated that most, but not all, commercial samples of streptavidin were truncated.\(^{78}\) Since 1990, residues 15–159 of the native gene were used to express a recombinant form (i.e. artificially expressed in a non-natural bacterium) of streptavidin with increased solubility in the bacterium *Escherichia coli*, and some modern commercial preparations of streptavidin are this recombinant form,‡ adding

\(^{‡}\text{E.g. Sigma-Aldrich (Product Id: S067, CAS number 9013-20-1 MDL number MFCD00082035).}\)
Analyst

Critical Review

Further possibilities for variability between different commercial preparations.

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 \(-7.20\) at pH 7.4 (ESI section 4†).

Examination of the sequence shows that, depending on where in the sequence streptavidin is truncated in different preparations, it is likely to differ in net charge by several elementary charge units. As high sensitivity sensors, BioFET's could in principle detect even small changes in the electrodynamic properties of streptavidin. \(^{67}\) A detailed analysis of streptavidin biochemistry literature (ESI section 4†) revealed that both the molecular weight and pH values can vary between different preparations with molecular weights in the range of between 53 and 66 kDa and pH values between 5 and 7.5. \(^{76-80}\) This variation in pH and molecular weight is likely due to both variation in the biomolecular structure, and di- ff erent preparations, it is likely to differ between the free protein. The pH of surface-bound streptavidin has been measured using Surface Force Apparatus (SFA) and Atomic Force Microscopy which demonstrated a pH of 5.0 ± 0.5. \(^{79,83}\) Experimental determination of streptavidin pH is discussed further in ESI section 4†.

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 pl 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 pl 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. \(^{87}\)

Buffer solution

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:

\[
pH = -\log_{10}(a_{H^+})
\]

(5)

Phosphate buffer is often used for biosensing experiments; a convention is to define 10 mM ionic strength phosphate buffer as 1× Na\(_2\)HPO\(_4\) (8.1 mM) and NaH\(_2\)PO\(_4\) (1.9 mM). The most common buffer for protein sensing is Phosphate-Buffered Saline, PBS, which is composed of 1× Na\(_2\)P\(_4\) mixed with 138 mM NaCl and 2.7 mM KCl. \(^{74}\) The ionic strength of
1× PBS is 162.7 mM. Some confusion can arise where authors use the acronym PBS for ‘Phosphate Buffered Solution’, as it becomes ambiguous whether they are reporting NaPi, or PBS. This distinction is often critical since, as discussed in ESI section 1, when a Ag/AgCl pseudo-reference electrode is used, then the chloride content of the buffer can significantly change the reference potential.

The dilution notation (×) is often used in this field due to the relevance of ionic strength and because dilution is often required for larger signals, as discussed later. Dilution can introduce a variety of issues, for example, at high dilutions (e.g. 0.01× PBS) the buffering capacity of the solution is reduced meaning that the system can become highly sensitive to small pH changes. For example, even the experimentalist breathing near the sample can result in elevated CO$_2$(g) dissolved into the sample and being detected as per:

$$\text{CO}_2(g) + \text{H}_2\text{O}(l) \Rightarrow \text{CO}_2(aq) + \text{H}_2\text{O}(l) = \text{HCO}_3^-(aq) + \text{H}^+(l)$$ (6)

This introduces a potential source of noise to the measurement.

A related issue arises upon addition of large amounts of biomolecule to a solution. Adsorption or release of protons to/from the biomolecules themselves can affect the pH of the bulk solution. Lloret et al. measured the bulk change in pH using a pH microelectrode in response to addition of streptavidin and found that for 1 µM streptavidin in 0.01× PBS (a common dilution), there was a change of 0.5 pH units. In contrast, at 1× PBS the shift in pH was negligible due to the high buffering capacity. The significance of this cannot be understated; for a typical PBS concentration of 0.01×, and a response of specific-binding, this should show no significant signal due to non-specifically bound streptavidin, however, this can cause ancillary issues such as a reduction in buffering capacity and instability of the biomolecule in solution.

The Debye length is the characteristic length scale for charge screening of electrolytes under the Debye–Hückel model. This concept can be useful when considering BioFET operation on large biomolecules which may contain charges located several Debye lengths away from the surface. In high ionic strength solutions, the Debye length is short so that distant charges would be screened and have no significant effect on the surface potential. Dilution reduces the ionic strength of the solution and extends the Debye length; however, this can cause ancillary issues such as a reduction in buffering capacity and instability of the biomolecule in solution, or reduced affinity of the biomolecule–target interaction, although the latter is not the case for streptavidin. For PBS dilutions of 1×, 0.1× and 0.01×, the calculated Debye lengths are 0.76 nm, 2.41 nm and 7.61 nm respectively. Stern et al. demonstrated a significant BioFET response for binding of streptavidin in solution to biotin on the sensor at 0.01× PBS, but no significant response at 1× PBS. Approximating streptavidin as ~5 nm from the surface, they found that this result agrees with the Debye–Hückel model.

Whilst the concept of Debye screening is useful in explaining reduced response in high ionic strength systems, caution must be taken in its use. For example, Bergveld used the Debye–Hückel model of the electrical double layer to argue that charged biomolecules cannot be detected at high ionic strengths, having written: ‘The resulting double layer, with a thickness of the Debye length, is of the order of 1 nm thick in moderate electrolyte concentrations. Beyond this distance no external electric field exists. Hence the idea that a layer of charged molecules at the surface of an ISFET modulates the electric field in the gate oxide should definitely be forgotten’ (emphasis added) – Bergveld (1996).

In the Debye–Hückel model, the Debye length is ~1 nm at ~160 mM ionic strength and yet streptavidin has been detected at this ionic strength by various authors and similar observations can be found for other biomolecules.

Even under the assumption that the Debye–Hückel model describes the system accurately, there are several misapprehensions which should be highlighted. First, the Debye length is not a hard cut-off beyond which no electrostatic effect is felt, because the screened Coulombic interactions reach to infinity. This can be seen from the expressions from the Debye–Hückel model at an electrode–electrolyte interface:

$$\psi_r = \psi_0 e^{-\frac{r}{\lambda_D}} \text{ and } E_r = \frac{\psi_0}{\lambda_D} e^{-\frac{r}{\lambda_D}},$$ (7)

where $\psi_r$ is the potential as a function of radial distance $r$, and $\lambda_D$ is the Debye length.
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 hydration104 and show reduced drift rates at pH 7.105

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 material101,105 which would show a corresponding change in the threshold voltage of the device.105 As an alternative material to SiO₂ which is impermeable to ions,95 AlGaN/GaN-based FETs have been demonstrated with significantly reduced drift in physiological buffers and various groups have successfully utilised AlGaN/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.117 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 IUPAC definition of sensitivity is not a measure of the ability to detect the minimal amount of an analyte, and therefore a second relevant metric is the limit of detection which IUPAC defines as the smallest measure that can be reasonably detected for a given analytical procedure, and is expressed as a concentration or quantity. More specifically:

‘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 $x_l = x_{bi} - k s_{bi}$, where $x_{bi}$ 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

Analysis of sensitivity

The IUPAC definition of sensitivity incorporates two elements: the measurement of a specific response, and its dependence upon analyte concentration. Some BioFET publications have quantified sensor performance using a definition of ‘sensitivity’ which differs from the IUPAC definition. For example, some authors refer to the absolute change in current ($\Delta I$) or, more frequently, the change in current divided by the initial current is referred to as the ‘sensitivity’. To avoid confusion and for comparison, a clear set of nomenclature is defined in this section for use in this review.

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

$$I_{\text{norm}} = \frac{I_f - I_0}{I_0} = \frac{\Delta I}{I_0},$$

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 ($\Delta I$) 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 ($\Delta I$) can either be positive or negative and consequently the normalised change in current, $I_{\text{norm}}$, 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

Response in the subthreshold region. A key factor when operating in the subthreshold region is the exponential relationship between threshold voltage and drain current (eqn (1)). As a result, a $+10$ mV change and a $-10$ mV change in threshold voltage do not produce an equal and opposite normalised change in current.

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

$$\Delta V_T = SS(\log_{10}(I_{\text{norm}} + 1)).$$

(9)

This equation can be used to calculate the expected magnitude of $I_{\text{norm}}$ 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 $I_{\text{norm}}$ is approximately $-33\%$. For the equivalent analyte and a p-channel device, the calculated change in $I_{\text{norm}}$ is approximately $+50\%$.

This important distinction between a negative $I_{\text{norm}}$ and a positive $I_{\text{norm}}$ 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 $I_{\text{norm}}$. 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 $I_{\text{low}}$ and $I_{\text{high}}$, respectively. This results in an $I_{\text{norm}}$ which is always positive and ‘unbounded’ with no mathematical upper limit. This ‘unbounded’ metric is herein referred to as $I_{\text{norm}}^+$ or the ‘unbounded normalised change in current’:

$$I_{\text{norm}}^+ = \frac{I_{\text{high}} - I_{\text{low}}}{I_{\text{low}}} = \frac{|\Delta I|}{I_{\text{low}}}. \quad (10)$$

The differences between the unbounded and bounded $I_{\text{norm}}$ are illustrated schematically in Fig. 5. For consistency with our previous quantitative analysis of the literature performed for pH sensing,18 $I_{\text{norm}}^+$ 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 $I_{\text{low}}$ would be the drain current after addition of streptavidin and $I_{\text{high}}$ 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 ($I_{\text{norm}}^+$) 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
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 \( I_{\text{norm}} \) 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 high concentration of bound 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 such an 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 equilibria 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

---

**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 \( I_{\text{low}} \) and the highest value as \( I_{\text{high}} \). The change in current in can be instead normalised by \( I_{\text{low}} \) instead of \( I_0 \), resulting in the ‘unbounded’ normalised change in current \( |I_{\text{norm}}| \). An alternative method would be to normalise by \( I_0 \left( I_{\text{norm}} \right) \). 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. \( I_{\text{norm}} \) cannot be directly compared to \( |I_{\text{norm}}| \) as different values are expected from the same stimulus.
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**

$I_{\text{norm}}$ is a measure of the amplification of the current response. By contrast, $\Delta V_T$ 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 supress 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 $I-V_T$ 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 $I_{\text{norm}}$. 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, $\Delta V_T$ can be extracted experimentally by simply dividing the current-response data ($\Delta I$) by the device transconductance, $g_m$:

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

(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 binding45 (i.e. electrostatic gating mechanism of FET operation43). Alternatively, the threshold voltage prior to analyte addition, $V_{T,0}$ and $I_{\text{norm}}$ 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 $\Delta I$ to the noise in the measurement ($\delta I$) 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 approach46,52,107,129,130 utilises the following expression:

$$\langle \delta I \rangle^2 = \int_{f_1}^{f_2} S(f = 1 \text{ Hz}) \, df = \ln \left( \frac{f_2}{f_1} \right) S_0 \quad (f = 1 \text{ Hz})$$  

(12)

where $S_0(f = 1 \text{ Hz})$ 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:

$$\text{SNR}_{\text{metric}} = \frac{\Delta I}{\delta I} = \frac{\Delta I}{\sqrt{S_0(f = 1 \text{ Hz})}}.$$  

(13)

Sometimes $\Delta I$ is converted into the voltage regime using $\Delta V = g_m \Delta V_T$, 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 $I_{\text{norm}}$ 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.33 Heller et al. performed charged-molecule sensing and salt concentration-change sensing experiments and remarked that for their...
with a bare SiO2 surface but saw that the noise was ionic regime of mobility fluctuation noise.131 Rajan also investigated in the literature was explained as the device having a different regime of mobility fluctuation noise.131 Rajan also investigated the SNR due to binding of streptavidin to biotin functionalised surfaces, over a range of gate voltages.26 A high concentration of ν-biotin in the bulk could displace surface-biotin-bound streptavidin so that the same sensor could be reused at various gate voltages. Increased SNR was then observed as the gate voltage was moved from the subthreshold region into the linear region.52 In summary, the evidence suggests that although operation in the subthreshold region often improves SNR, its ability to improve SNR may be system dependent.

A related question is therefore what region of operation is optimal for sensor Sensitivity (e.g. \( I_{\text{norm}} \) per pH). In the linear region (eqn (11)), \( I_{\text{norm}} \) increases as the gate voltage \( V_g \) approaches the threshold voltage \( V_T \), i.e. as the subthreshold region is approached. \( I_{\text{norm}} \) 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 (\( I_{\text{norm}}, \Delta I, I \) and \( g_n \)).

In summary, in the subthreshold region, Sensitivity is optimised as a result of the large normalised change in current, \( I_{\text{norm}} \), 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 \( (g_n) \) 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 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.126,129,134–137 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 \( V_{g,\text{back}} \). 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,20 the physical origin of the improved top-gate subthreshold slope
using a back-gate voltage is attributed to carrier depletion from back gate. The conducting channel is believed to be the top region of the channel whilst the bottom region is depleted under the applied back gate voltage. Thus, the back-gate reduces the channel thickness controlled by top gate and this reduces the depletion capacitance which in turn improves the subthreshold slope for the response measured through the top gate. The subthreshold slope is steeper with the back gate bias applied.

**Back-gate sweep/back-gated operation.** Alternatively to the aforementioned top-gate sensing, the back-gate voltage can be instead swept with the top-gate held fixed, thus resulting in a plot of $I$ versus $V_{g,\text{back}}$ similar to Fig. 3 but in which the dependent variable is $V_{g,\text{back}}$ instead of $V_{g,\text{top}}$. From this plot, $ΔV_{T,\text{back}}$ can be measured, which is related to $ΔV_{T,\text{top}}$ but amplified. Importantly, although the pH response of $ΔV_{T,\text{back}}$ can exceed the Nernst limit (~59 mV per pH), it is only an amplification of the value of $ΔV_{T,\text{top}}$ and does not necessarily correspond to any enhancement of the oxide–electrolyte bulk surface potential which is the causal factor generating the signal.

This amplification was described by Go et al. as the result of capacitive coupling within the system. The coupling amplified a change in top-gate voltage ($ΔV_{T,\text{top}}$) which consequently results in an amplified shift in bottom-gate voltage ($ΔV_{T,\text{back}}$) which depends upon the capacitance of the liquid-exposed top-gate dielectric ($C_{\text{top}}$), the back-gate dielectric ($C_{\text{back}}$) and a factor ($α_{SN}$) with a value between 0 and 1 depending on the extent of coupling due to the choice of biasing conditions:

$$ΔV_{T,\text{back}} = ΔV_{T,\text{top}} \frac{C_{\text{top}}}{C_{\text{back}}} α_{SN}$$

Further research is required in this field for the development of a general theory of back-gated operation. The $ΔV_{T,\text{back}}$ threshold voltage shift amplification strategy has been utilised for biosensing, such as the work of Duan et al. and Jayant et al.

While there is no doubt that double-gated FET operation can amplify the measured voltage shift, an important question that remains is whether it can provide an improved SNR. Go et al. concluded that double-gated FET operation can only improve the SNR if the noise from the measurement equipment sets the lower limit of performance. In the double-gated FET operation of Pu et al., noise spectroscopy revealed an improved SNR. Regardless, the use of a back-gate voltage offers the advantage of reducing the requirements for external circuitry to amplify the signal. Go et al. derived a quantitative model of double-gated FET operation which can explain the results of some systems; in their model the response is a result of capacitive coupling between the back-gate and the top-gate oxides, and under certain gate voltage conditions, the semiconductor body capacitance.

**Device geometry.** Finally, a great deal of literature is focused upon the investigation of new device-geometries aimed at improving sensitivity, SNR or limit of detection. The commonly stated argument that increased surface area-to-volume ratio increases the normalised change in current is the rationalisation for much of the focus in publications on creating devices with nanoscale dimensions. This argument is still subject to debate, with some authors suggesting that nanoscale dimensions offer increased response via a different mechanism or that the argument is not generally applicable to all structures. Operating in the linear region, the SNR was calculated and measured to increase with the square root of the device area for constant device thickness. This assumes that the entire surface contributes uniformly to the signal, and therefore is expected to apply to pH sensing and to biosensing experiments in which there is sufficient analyte to cover the surface receptors uniformly.

It has been demonstrated both experimentally and theoretically that increasing surface area-to-volume ratios increases normalised change in current. Operating in the subthreshold region, the improvement is attributed to the subthreshold slope improvement (i.e. decreased value) which results from the improved ratio of oxide capacitance to depletion-layer capacitance (eqn (3)). There is no evidence that nanowire or nanoribbon geometries achieve a better subthreshold slope than the ideal value of 59 mV dec$^{-1}$ through their superior surface area–volume ratio. Particularly, it is worth noting that scaling in channel length results in degradation of the subthreshold slope due to the Short Channel Effect (SCE). Planar devices are capable of providing comparable Sensitivity to nanoscale devices given that planar devices can be made with a subthreshold slope of ~80 mV dec$^{-1}$ for bulk or ~63 mV dec$^{-1}$ for fully-depleted SOI transistor or double-gated SOI transistor devices.

**Quantitative analysis**

**Methods**

This analysis compares FET-sensor results for pH-sensing with streptavidin-binding to oxide-APTES-(or the methoxysilane equivalent, APTMS)-biotin-functionalised surfaces. Where available, the parameters such as the subthreshold slope and normalised change in current were directly extracted from the
streptavidin literature, and where this was not reported explicitly it was extracted manually from I–V graphs.

As discussed, when comparing normalised change in current between devices of different types (i.e. p- or n-channel device), normalisation simply using the ‘initial’ drain current, prior to analyte addition, is insufficient. The change in drain current was normalised by the lower drain current to give the ‘unbounded’ normalised change in current, \( I_{\text{norm}} \), using eqn (10). As this was used consistently through this analysis, the symbol \( I_{\text{norm}} \) will be used to refer to the unbounded change in the rest of this review. As an example, for streptavidin on an n-channel device, the drain current is expected to decrease after addition of streptavidin and therefore \( I_{\text{norm}} \) will be the drain current after addition of streptavidin and \( I_{\text{high}} \) before addition. This is consistent with our previous quantitative analysis of the literature for pH sensing.\(^{18}\)

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 \( I_{\text{norm}} \) 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 (\( I_0 \)) can be calculated from the lowest current (\( I_0 \)) and the normalised change in current (\( I_{\text{norm}} \)):

\[
I_n = I_0 (I_{\text{norm}} + 1)^n. \tag{15}
\]

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

\[
\text{Sensitivity} = \left( \frac{I_n}{I_0} \right)^{\frac{1}{n}} - 1, \tag{16}
\]

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 \( I_{\text{norm}} \) 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 complete 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 (\( \Delta V_T \)) is related to the shift in surface potential at the oxide–electrolyte interface (\( \Delta \psi_0 \)) (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, \( \Delta Q \), with capacitance \( C_0 \) can give the voltage shift simply as:

\[
\Delta \psi_0 \approx \Delta V_T = \frac{\Delta Q}{C_0}. \tag{17}
\]

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:

\[
A + S \rightarrow AS. \tag{18}
\]

The resulting equilibrium constant, \( K_d = [AS]/([A][S]) \), bears many similarities to the well-known pK\(_a\) 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]\(_{\text{max}}\) are expected to be occupied (i.e. \([AS]/[S]_{\text{max}} \sim 100\%\) 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 \([AS]/[S]_{\text{max}}\) and showed that in low \( K_d \), small volume systems (i.e. low number of mole-
Comparison between Proton and Ligand Binding

| Low [H⁺] | Proton (H⁺) binding to Acid (A⁻) | High [H⁺] |
|---------|---------------------------------|-----------|
| High pH  | p[H⁺] = -log[H⁺] = pKₐ + log([A⁻]/[HA]) | Low pH    | p[H⁺] = -log[H⁺] = pKₐ - log([A⁻]/[HA]) |

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

| Low [A] | A + S ⇌ AS |
|---------|-----------|
| [A] ≤ 0.01 Kₐ | log([A]/[S]) = log(Kₐ) - 2 |
| [A] ≥ 100 Kₐ | log([A]) ≥ log(Kₐ) - 2 |

Fig. 7 Comparison between proton affinity and ligand affinity metrics.

The pKₐ describes the equilibrium constant of proton-binding, similarly the Kₐ describes the equilibrium constant of ligand-binding. The simple surface-analyte reaction modelled here (first-order langmuir isotherm) is the simplest biomolecular-surface binding scheme, and describes the fraction of bound analyte at equilibrium (AS) to the total binding site concentration (S)max. Various assumptions are made within this model, which are detailed within ESI 10.

A simple strategy for estimating the maximum density of bound molecules can be demonstrated. When measuring the response as a function of the analyte concentration, a concentration at which response is saturated (cmax) occurs. Assuming binding occurs specifically to receptors at the surface (i.e. no multilayers) then the saturation concentration corresponds to the upper limit of the density of specifically-bound biomolecule. To demonstrate this analysis, various examples from streptavidin-sensing publications are used.

From the work of Elfström et al., given a sample volume, V, of 200 µL solution and concentration, cmax of 0.5 nM, the total number of molecules in the solution (= cmaxV) can be obtained. Assuming all molecules bind to the surface (valid given Kd ≪ streptavidin concentration at the surface) and that they bind homogenously, then the bound density, ρ, will simply be cmaxV divided by the functionalised area, A, exposed to analyte (10 mm²):

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

This can be compared with the maximum streptavidin density, ρmax theoretically possible for streptavidin, assuming each molecule occupies ~25 nm².

$$\rho_{\text{max}} = 0.04 \text{ molecules per nm}^2.$$

This comparison suggests that the surface-bond density for the device of Elfström et al. is close to ideal. In contrast, if this is compared to Duan et al., who showed a cmax of 2 pM, taking into account the different sensor areas and amount of streptavidin, then a significantly lower surface coverage is calculated:

$$\rho_{\text{Duan}} = 0.000072 \text{ molecules per nm}^2.$$

This lower surface site density may well be explained by low levels of functionalisation of the surface. These calculations are useful as they provide an estimate of surface site density, i.e. functionalisation success rate, and ultimately dictates an upper limit for response. They can also inform experimental design when choosing sensible concentration ranges to perform experiments under. If a sensor has low functionalisation density, then it may also have a poor response time (a long time for proteins to diffuse to the binding sites).

Won Hee Lee et al. and Wen et al. 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. Fluorescence imaging confirmed that there was selective binding of streptavidin. Given the surface area and sample volume of their device, the calculated saturation concentration cmax 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.

---

4188 | Analyst, 2017, 142, 4173–4200

---

**Gate area 10 µm², cmax = 2 pM, analyte was exposed to the surface over two minutes, assuming all liquid came into contact with the surface (overestimate), at a flow rate of 300 µL min⁻¹ results in volume of 600 µL.**

**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 ~3 × 10⁷ nm⁻².**
If there were a linear relationship between analyte-binding and receptor-concentration, then it would be possible to extract the density of biotin on the surface from this simple analysis of sensor response as a function of analyte concentration. However, experiments have shown a non-linear relationship between biotin concentration at the surface and sensor response. Duan et al. showed a response saturation for polyelectrolyte films containing 30% biotin\(^{271}\) which agrees well with the Surface Plasma Resonance (SPR) experiments of Jung et al. who showed saturation with an alkylthiolate monolayer containing 34% biotin.\(^{172}\) The non-linearity in the concentration-dependence is likely a combination of steric-contraints of the protein and the fact that streptavidin has four binding sites. Jung et al. observed that streptavidin binds to one site for low concentrations of biotin (0.34% biotin), two sites for peak saturation (approximately 30% biotin), and that for high biotin content (>40% biotin) the biotin became less available within the monolayer due to reorientation ordering, resulting in less binding to streptavidin.\(^{172}\) If this result is transferable to other chemical functionalisation systems, it counter-intuitively suggests that extremely high-density biotin functionalisation would be detrimental to Sensitivity.

**Kinetic response.** For a system that is reaction-limited (reaction rate is slower than diffusion to the sensor surface from the bulk) the response is expected to increase with time but diminish exponentially. Streptavidin–biotin is unusual in that it has an exceptionally high affinity which results in diffusion-limited kinetics (the reaction is faster than diffusion to the surface). For this an initially linear response is expected, as has been shown by Duan et al. for streptavidin–biotin binding\(^{15}\) (ESI section 10.2).

For ultra-low concentrations of biomolecules, a lower value of the subthreshold slope can decrease the response time for a given signal change. The reason for this is that for a given change in signal, a lower subthreshold slope value requires a smaller change in surface potential. Assuming that the bound biomolecule concentration is proportional to surface potential change, a smaller number of bound molecules is required and therefore the time-to-detection is reduced.\(^{44,173}\)

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.\(^{47,148}\) 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.\(^{374,175}\) 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 choices should be considered when designing for optimal response.\(^{52,147,169,177}\)

**Non-specific response.** The concentration of biomolecular analyte available to bind to the surface may be lower than that expected based on the initially added bulk concentration due to binding to non-sensitive regions of the device, an issue which is often caused by non-specific binding but can also occur if biomolecular receptors have been functionalised on solvent-exposed regions of the device away from the sensor (e.g. if the passivation layer shown in Fig. 1 was functionalised). This is particularly problematic for nanoscale dimension devices in which the sensing region of the device can have a small surface area relative to the total exposed surface area.\(^{26}\)

### Streptavidin sensing data

In this section, current-response data for the detection of streptavidin is presented for devices with oxide–APTES–biotin surface functionalisation and streptavidin sensing.

**Linear region.** Measurements in this region of operation have been reported which show detection of streptavidin at the oxide–APTES–biotin interface.\(^{23,35,44,49,120}\) Data from these publications is tabulated in ESI sections 3 and 9. Three publications contained data with response as a function of streptavidin concentration, giving a Sensitivity of 72%\(^{162}\), 63%\(^{120}\) and 0.5%.\(^{49}\) As discussed previously, the linear region is not expected to optimise \(I_{norm}\) or Sensitivity and therefore measurements in this region are not the focus of this review.

**Subthreshold region.** Many publications only provide the change in response due to addition of an arbitrarily determined concentration of streptavidin. As a result, only the normalised change in current can be used for comparison and the Sensitivity cannot be calculated for such papers. In this section, \(I_{norm}\) from across the streptavidin-sensing literature is extracted and presented with the corresponding shift in threshold voltage for devices operated in the subthreshold region. A tabulated summary of papers identified can also be found in ESI section 5.

FET-sensor data was obtained from eight publications in which streptavidin binding to oxide–APTES–biotin coated surfaces was measured in the subthreshold region, and is shown in Fig. 8.\(^{26,28,32,37,49,70,107,178,179}\) The normalised change in current, \(I_{norm}\), is plotted against the device subthreshold slope. The data was obtained from experiments which were not all performed at the same concentration of streptavidin and at more than one ionic strength. Where experiments were performed at several concentrations of streptavidin, or at differing ionic strength, this is indicated in the figure. For experiments performed at a higher concentration of streptavidin, there is a higher density of bound analyte (until the sensor surface is saturated) and a corresponding increase in \(I_{norm}\).

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
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 $I_{\text{norm}}$ (from eqn (9)), and therefore it can be seen that most measurements correspond to a shift in the threshold voltage, $\Delta V_T$, 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 $I_{ds}$-$V_g$ 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. Assumption (b) has been shown to be approximately the case for pH sensing and assuming a consistent oxide material between devices. 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 that 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 SiO$_2$ surface typically has a pH-induced surface potential.
change of $33 \pm 6$ mV per pH.\textsuperscript{18} For this system, a maximum pH Sensitivity of between 190% to 360% is calculated assuming an optimal subthreshold slope of 59 mV dec\textsuperscript{−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\textsubscript{2} rather than SiO\textsubscript{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.\textsuperscript{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.\textsuperscript{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\textsuperscript{26,28,32,49,70,107} and the results are plotted against subthreshold slope in Fig. 9 and 10. In our previous work,\textsuperscript{18} we published a collation of pH sensing data\textsuperscript{18,38,44,183} 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\textsubscript{2} systems and $59$ mV per pH for HfO\textsubscript{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 (\textasciitilde33 mV per pH for SiO\textsubscript{2} with one anomaly, \textasciitilde59 mV per pH for HfO\textsubscript{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%\textsuperscript{32}) was lower than the maximum observed in pH sensing data (up to 600%\textsuperscript{44}). For some devices,\textsuperscript{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.\textsuperscript{26} and Buitrago et al.\textsuperscript{70} a significantly lower Sensitivity than pH sensing results was obtained, despite the low subthreshold slopes (112 mV dec\textsuperscript{−1} and 95 mV dec\textsuperscript{−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.\textsuperscript{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’.\textsuperscript{178}

The high Sensitivity of Won Hee Lee et al.\textsuperscript{12} 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
**Fig. 9** The relationship between Sensitivity and subthreshold slope, for measurements obtained within the subthreshold region. Both streptavidin sensing$^{26,28,32,49,70,107}$ (●) and pH sensing data (★)$^{18,38,44,183–190}$ are shown. For pH sensing, consistent surface potential changes are observed for a given oxide, resulting in the ability to predict Sensitivity accurately using eqn (9). In contrast, for biosensing, eqn (9) was insufficient to accurately predict Sensitivity, indicative of significant surface potential variation per 10-fold increase in streptavidin concentration. All biosensing and pH sensing measurements were performed using a liquid top-gate, with the exception of the biosensing measurements of Elfström et al. which utilised a fixed back-gate voltage with no liquid top-gate. For streptavidin-sensing data, the device geometry, liquid top-gate reference electrode material and authors, respectively, are shown in parenthesis within the legend. Dielectrics show a characteristic shift in threshold voltage per change in pH unit that is $\sim 33$ mV per pH for SiO$_2$ and $\sim 59$ mV per pH for HfO$_2$. Curves calculated using eqn (9) for a threshold voltage shift per 10 fold increase in analyte concentration of 33 mV and 59 mV are shown using solid lines. The dashed line indicates the theoretical lower limit of the subthreshold slope at room temperature of 59 mV dec$^{-1}$ available to classical FETs. Details of method used to obtain the data and a replot of the same data but on a linear x-axis can be found in ESI 7.6.$^†$

**Fig. 10** The relationship between the calculated shift in threshold voltage per 10-fold increase in analyte concentration and measured subthreshold slope, for measurements obtained within the subthreshold region. For streptavidin-sensing data, the device geometry and authors are shown in the legend. This is a replot of the experimental data shown in Fig. 9a, but in this figure, the shift in threshold voltage per 10-fold increase in analyte concentration is calculated by inserting both its subthreshold slope and Sensitivity into eqn (9). Both streptavidin sensing$^{26,28,32,49,70,107}$ (●) and pH sensing (★) results$^{18,38,44,183–190}$ are shown. Isocontours are drawn using eqn (9), in 1% increments, using values of the Sensitivity between 1% (red) and 150% (dark blue), with >150% being shown as a pale blue region. The theoretically optimal subthreshold slope of 59 mV dec$^{-1}$ for a classical transistor at room temperature is shown as a dashed line. Eqn (9) predicts that with a subthreshold slope of 59 mV dec$^{-1}$, it is impossible to obtain greater than $\sim 50$% Sensitivity with a shift in threshold voltage <10 mV per 10-fold increase in analyte concentration, and this can be seen graphically within the figure. The Nernst limit at 300 K (59 mV per pH) is drawn as a dotted line. Details of method used to obtain the data and a copy of the plot with a linear x-axis can be found in ESI 7.6.$^†$
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. 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 (SiO₂) and ~59 mV per pH (HfO₂). Most SiO₂ systems show a pH response of approximately 33 mV per pH, with some reports demonstrating a higher value of 43 mV per pH. One pH sensing result 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. and later refined by Healy et al. In these models, an increased density of hydroxyl groups on the surface corresponds to an increase in surface potential shift per unit pH. 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.

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 (pH = −log([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 per pH. 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:

\[
\Delta V_T \propto \Delta \text{Surface Potential}
\]

**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 (ΔVₜₚₚ). In contrast, for biomolecular systems, many more factors can affect the shift in threshold voltage per 10-fold increase in concentration of analyte (ΔVₜₚₚ). The net change of the biomolecular system (shown as green circles) can influence response. The debye length (\(\lambda\)) is compared to the distance of the analyte from the surface (Lₐ) on the figure, which can significantly affect BioFET signal. Here \(\tau_{eq}\) 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.
concentration is primarily limited by the density of receptors (and therefore bound-analyte) on the surface.

In order to investigate this, analysis of the concentration-dependent response curve was performed to infer the density of bound-biomolecule. As an example, the device of Elfström et al. in Fig. 10 demonstrated a surface potential shift of 52 mV per 10-fold increase in streptavidin, which is large compared to the majority of other streptavidin-sensing measurements and pH-sensing measurements. The density of streptavidin molecules upon the surface of the device of Elfström et al. was calculated to be nearly to the ideal density for close-packed streptavidin. It should be noted that their device was operated without a reference electrode at fixed back-gate voltage, which is generally believed to be required for a well-defined signal. This supports the hypothesis that the biosensing shift in threshold voltage per 10-fold increase in streptavidin concentration is largely determined by the density of surface receptors. Optimising the density of receptors and the sensor surface area is important in improving the device response.\cite{52,169,177}

Response times – non-equilibrium response (ii). In addition to receptor density as one factor that may affect the threshold voltage shift, there are many other potential factors which could explain the observed differences between pH-sensing and biosensing experiments. The shift in threshold voltage per 10-fold increase in analyte concentration for biosensing may be lower than that for a 10-fold increase in H$^+$ concentration in pH sensing simply because the biomolecular response measured may not be fully time-equilibrated. As discussed, transport of the biomolecule to the surface and the subsequent reaction can take a long time to equilibrate compared to the equilibration of the acid–base reactions relevant to pH sensing. The limitations imposed by mass transport, binding kinetics and device-geometry choices must be considered when designing for optimal biomolecular-sensing response.\cite{52,169,177}

Non-specific binding (iii). The concentration of biomolecular analyte available to bind to the surface may be lower than that expected based on the initially added bulk concentration due to binding to non-sensitive binding. If the concentration of analyte at the surface is overestimated because of non-specific binding, the effective threshold voltage shift per 10-fold increase in analyte is reduced.

Biomolecular charge (iv). As discussed in the streptavidin sequence analysis and literature review, variation between different experiments may occur due to different commercial preparations of the same biomolecule having different electrodynamic properties.

Biomolecular orientation (v). In principle, the biomolecular orientation can affect the response. Molecules such as antibodies are known to be oriented in as statistical manner rather than unidirectional as often indicated in schematics.\cite{55,150}

Buffering capacity (vi). The buffering capacity of the solution, is important in determining the stability, reliability and accuracy of the measurement. Low buffering capacities can result in additional noise from ambient fluctuations in p$\text{H}$, for 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 p$\text{H}$ of the solution can be modified by analyte and detected by the sensor.\cite{74} The practice of using minimal buffering capacity and high biomolecule concentration (added in a single aliquot) is not uncommon,\cite{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.\cite{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.\cite{49} and Won Hee Lee et al.\cite{74} achieved higher Sensitivity than the nanobelt arrays of Liu et al.\cite{26} and Cheng et al.\cite{25} or the nanowires of Elfström et al.\cite{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 p$\text{H}$) and the subthreshold slope.\cite{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...
increase in analyte concentration can provide a figure-of-merit for the device and by fitting the curve to an appropriate binding model, the binding affinity of the analyte to its receptor can be estimated. Simple calculations can be used to estimate the density of bound analyte based on the concentration at which the sensor response saturates, and both this value and measurement of the shift in threshold voltage per 10-fold increase in analyte concentration can be useful in determining whether sensitivity enhancement efforts are limited by surface chemistry or transistor performance.

Devices even of simple (microscopic) planar geometry were shown capable of obtaining comparable sensitivity to more elaborate nanodevice geometries (such as nanowires), suggesting nanoscale device design is not a requisite for high biomolecular sensitivity. Due to process control constraints, the subthreshold slope is limited to \( \sim 63 \text{ mV \, dec}^{-1} \), with a theoretical lower limit of 59 mV dec\(^{-1}\) at room temperature. Therefore, it is expected that no further sensitivity gain is obtained by further scaling of device geometry for equilibrated (with respect to the analyte binding reactions) pH sensing and biomolecular sensing experiments. This contradicts the commonly accepted claim that a high surface-to-volume ratio is a requisite for high biomolecular sensitivity. Nanoscale device geometries are however expected to offer improved response times for biomolecular detection, and therefore improve both the effective limit of detection given practical time-constraints for a sensing experiment.

State of the art devices operate by modulating the resistance of the semiconductor (with an upper limit of 59 mV of change in gate voltage to a decade of current change) by modulating the local carrier concentration in the semiconductor. For the published nanowire sensors, the nanowire dimensions are commonly around 100 nm in width and therefore far larger than typical biomolecule dimensions, resulting in the same sensing mechanism between nanowires and conventional sheet channels. Looking to the future of the field, an interesting possibility would be the fabrication of a novel BioFET (with semiconductor insulated from electrolyte with a dielectric, i.e. operated by field-effect mechanism) such that the width scale of the device is comparable to typical biomolecular dimensions (on the order of \( \sim 1-2 \text{ nm width and thickness} \)). In contrast to present devices, the channel of such a device could be completely inverted by individual biomolecules, which can result in different sensitivity and SNR characteristics. A competing challenge for such a design would be the challenge of fine dimensional control which is beyond the limit of top-down fabrication technology as well as the local variation in discrete dopant sites (discrete dopant fluctuations) within the semiconductor, which results in significant variation in device-to-device sensitivity and, in turn, presents a significant issue for reliable detection of single molecules without the use of additional circuitry.\(^{144}\)

Many of the current limitations which have been presented are similar to those that have been encountered for other emerging biosensing technologies. The current ‘gold standard’ for label-free biosensing is Surface Plasmon Resonance (SPR) biosensors. In a highly extensive review published in 2005, Rich and Myszka analysed 1113 articles (103 reviews, 1010 papers) and discussed common issues within the Surface Plasmon Resonance biosensor literature such as: authors only considering high concentrations of analyte, response being normalised inappropriately or the data was reported in insufficient detail.\(^{197}\) Surface chemistry optimisation has played an important part of Surface Plasmon Resonance biosensor development.\(^{198}\) In principle, BioFET devices have the potential to surpass Surface Plasmon Resonance biosensors due to their potential for low-cost fabrication, ability to detect low-mass analyte and the elimination of the requirement for additional optical equipment (which can be high cost and low throughput).

As highlighted in this review, current advances in the field of BioFET research are being obstructed by the lack of consensus upon which quantitative metrics (i.e. figure-of-merit) should be used to compare devices, with the result that most published studies can only be compared qualitatively. Despite this, BioFET research is a rapidly advancing field in which novel device design and operation methodologies are consistently being developed. Owing to the widespread capability for integration of this type of technology with the portable electronics used in Point-of-Care analysis and wearable technology, the potential for BioFET sensors to revolutionise approaches to biosensing in healthcare, security applications and sensing in the built and natural environments is prodigious.

**Conflicts of interest**

There are no conflicts to declare.

**Acknowledgements**

This work was supported by an EPSRC Doctoral Training Centre grant (EP/G03690X/1).

**Notes and references**

1. T. Vo-Dinh and B. Cullum, *Fresenius’ J. Anal. Chem.*, 2000, 366, 540–551.
2. L. Senesac and T. G. Thundat, *Mater. Today*, 2008, 11, 28–36.
3. P. Arora, A. Sindhu, N. Dilbaghi and A. Chaudhury, *Biosens. Bioelectron.*, 2011, 28, 1–12.
4. J. Riu, A. Maroto and F. Rius, *Talanta*, 2006, 69, 288–301.
5. A. Ramanavieius, F. W. Herberg, S. Hutschenreiter, B. Zimmermann, I. Lapėnaitė, A. Kaušaitė, A. Finkelsteinas and A. Ramanavičienė, *Acta Med. Litu.*, 2005, 12, 1–9.
6. S. G. Patching, *Biochim. Biophys. Acta, Biomembr.*, 2014, 1838, 43–55.
30 I. Zeimpekis, K. Sun, C. Hu, N. M. J. Ditshego, O. Thomas, M. R. R. de Planque, H. M. H. Chong, H. Morgan and P. Ashburn, Nanotechnology, 2016, 27, 165502.

31 A. Tarasov, W. Fu, O. Knopfmacher, J. Brunner, M. Calame and C. Schönenberger, Appl. Phys. Lett., 2011, 98, 012114.

32 W. H. Lee, J.-M. Lee, M. Uhm, J. Lee, K. R. Kim, S.-J. Choi, D. M. Kim, Y.-J. Jeong and D. H. Kim, IEEE Electron Device Lett., 2014, 35, 587–589.

33 S. Chen, L. Nyholm, N. Jokilaakso, A. E. Karlström, J. Linnros, U. Smith and S.-L. Zhang, Electrochem. Solid-State Lett., 2011, 14, J34–J37.

34 T. Kim, K. Kim, N. Hong, W. Ko, C.-K. Baek, S. Jeon, M. J. Deen, M. Meyyappan, Y.-H. Jeong and J.-S. Lee, RSC Adv., 2013, 3, 7963–7969.

35 X. Duan, Y. Li, N. K. Rajan, D. A. Routenberg, Y. Modis and M. A. Reed, Nat. Nanotechnol., 2012, 7, 401–407.

36 E. Stern, J. F. Klemic, D. A. Routenberg, P. N. Wyrembak, D. B. Turner-Evans, A. D. Hamilton, D. A. LaVan, T. M. Fahmy and M. A. Reed, Nature, 2007, 445, 519–522.

37 Y. Cui, Q. Wei, H. Park and C. M. Lieber, Science, 2001, 293, 1289–1292.

38 X. P. A. Gao, G. Zheng and C. M. Lieber, Nano Lett., 2010, 10, 547–552.

39 E. Buitrago, M. F.-B. Badia, Y. M. Georgiev, R. Yu, O. Lotty, J. D. Holmes, A. M. Nightingale, H. M. Guerin and A. M. Ionescu, Sens. Actuators, B, 2014, 199, 291–300.

40 J.-H. Ahn, S.-J. Choi, J.-W. Han, T. J. Park, S. Y. Lee and Y.-K. Choi, Nano Lett., 2010, 10, 2934–2938.

41 G. Shalev, G. Landman, I. Amit, Y. Rosenwaks and I. Levy, NPG Asia Mater., 2013, 5, e41.

42 M. O. Noor and U. J. Krull, Anal. Chim. Acta, 2014, 825, 1–25.

43 S. Liu and X. Guo, NPG Asia Mater., 2012, 4, e23.

44 D. Sarkar, W. Liu, X. Xie, A. C. Anselmo, S. Mitragoti and K. Banerjee, ACS Nano, 2014, 8, 3992–4003.

45 I. Heller, A. M. Janssens, J. Männik, E. D. Minot, S. G. Lemay and C. Dekker, Nano Lett., 2008, 8, 591–595.

46 I. Heller, J. Männik, S. G. Lemay and C. Dekker, Nano Lett., 2009, 9, 377–382.

47 T. E. Edmonds, in Chemical Sensors, Springer Science & Business Media, 2013, p. 223.

48 J. Janata, Analyst, 1994, 119, 2275.

49 X. Wen, S. Gupta, Y. Wang, T. R. N. Iii, S. C. Lee and W. Lu, Appl. Phys. Lett., 2011, 99, 043701.

50 X. Wen, S. Gupta, T. R. Nicholson, S. C. Lee and W. Lu, Phys. Status Solidi C, 2011, 8, 2489–2491.

51 F. N. Ishikawa, M. Curreli, C. A. Olson, H.-I. Liao, R. Sun, R. W. Roberts, R. J. Cote, M. E. Thompson and C. Zhou, ACS Nano, 2010, 4, 6914–6922.

52 N. K. Rajan, Limit of Detection of Silicon BioFETs, PhD thesis, Yale University, 2013.

53 E. D. Minot, A. M. Janssens, I. Heller, H. A. Heering, C. Dekker and S. G. Lemay, Appl. Phys. Lett., 2007, 91, 093507.

54 U. Gutz, F. Gerlach, M. Decker, W. Oelsen and W. Vonau, J. Solid State Electrochem., 2009, 13, 27–39.

55 M. Lambrechts and W. Sansen, in Biosensors: Microelectrochemical Devices, CRC Press, 1992, p. 69.

56 C. Toumazou and P. Georgiou, Electron. Lett., 2011, 47, S7–S12.
Critical Review

57 M. T. Martínez, Y.-C. Tseng, M. González and J. Bokor, J. Phys. Chem. C, 2012, 116, 22579–22586.
58 Y. Taur and T. H. Ning, Fundamentals of Modern VLSI Devices, Cambridge University Press, 2nd edn, 2013.
59 S. M. Sze, Semiconductor Devices: Physics and Technology, John Wiley & Sons, Inc., Hoboken, NJ, 2nd edn, 1985.
60 N. D. Arora, in MOSFET Models for VLSI Circuit Simulation: Theory and Practice, Springer Science & Business Media, 2012, p. 265.
61 A. Vacic and M. A. Reed, J. Exp. Nanosci., 2014, 9, 41–50.
62 R. J. Chen, H. C. Choi, S. Bangsaruntip, E. Yenilmez, X. Tang, Q. Wang, Y.-L. Chang and H. Dai, J. Am. Chem. Soc., 2004, 126, 1563–1568.
63 H. R. Byon and H. C. Choi, J. Am. Chem. Soc., 2006, 128, 2188–2189.
64 E. H. Williams, A. V. Davydov, A. Motayed, S. G. Sundaresan, P. Bocchini, L. J. Richter, G. Stan, K. Steffens, R. Zangmeister, J. A. Schriefels and M. V. Rao, Appl. Surf. Sci., 2012, 258, 6056–6063.
65 X. Duan, N. K. Rajan, M. H. Izadi and M. A. Reed, Nanomod., 2013, 8, 1839–1851.
66 S. Gupta, M. Elias, X. Wen, J. Shapiro, L. Brillson, W. Lu and S. C. Lee, Biosens. Bioelectron., 2008, 24, 505–511.
67 N. Clément, K. Nishiguchi, J. F. Dufreche, D. Guerin, A. Fujiwara and D. Vuillaume, Appl. Phys. Lett., 2011, 98, 014104.
68 A. Henning, M. Molotskii, N. Swaminathan, Y. Yaknin, A. Godkin, G. Shaley and Y. Rosenwaks, Small, 2015, 11, 4931–4937.
69 G. Shaley, Y. Rosenwaks and I. Levy, Biosens. Bioelectron., 2012, 31, 510–515.
70 E. Buitrago, M. Fernández-Bolaños, Y. M. Georgiev, R. Yu, O. Lotty, J. D. Holmes, A. M. Nightingale and A. M. Ionescu, in Proceedings of Technical Program - 2014 International Symposium on VLSI Technology, Systems and Application (VLSI-TSA), 2014, pp. 1–2.
71 S. Upadhyay, R. Frederiksen, N. Lloret, L. De Vico, P. Krogstrup, J. H. Jensen, K. L. Martinez and J. Nygård, Appl. Phys. Lett., 2014, 104, 203504.
72 R. A. Sperling and W. J. Parak, Philos. Trans. R. Soc. A, 2010, 368, 1333–1383.
73 T. Sano and C. R. Cantor, J. Biol. Chem., 1990, 265, 3369–3373.
74 N. Lloret, R. S. Frederiksen, T. C. Möller, N. I. Rieben, S. Upadhyay, L. D. Vico, J. H. Jensen, J. Nygård and K. L. Martinez, Nanotechnology, 2013, 24, 035501.
75 N. A. Lapin and Y. J. Chabal, J. Phys. Chem. B, 2009, 113, 8776–8783.
76 N. M. Green, in Methods in Enzymology, ed. M. Wilchek and E. A. Bayer, Academic Press, 1990, vol. 184, pp. 51–67.
77 L. Chaït and F. J. Wolf, Arch. Biochem. Biophys., 1964, 106, 1–5.
78 N. M. Green, in Advances in Protein Chemistry, ed. J. T. E. Anfinsen and F. M. R. C. B. Anfinsen, Academic Press, 1975, vol. 29, pp. 85–133.
79 S. Sivasankar, S. Subramaniam and D. Leckband, Proc. Natl. Acad. Sci. U. S. A., 1998, 95, 12961–12966.
80 Streptavidin Properties and Characterization, Rockland Inc.
81 C. E. Argarana, I. D. Kunz, S. Birken, R. Axel and C. R. Cantor, Nucleic Acids Res., 1986, 14, 1871–1882.
82 A. Pähler, W. A. Hendrickson, M. A. Kolks, C. E. Argarana and C. R. Cantor, J. Biol. Chem., 1987, 262, 13933–13937.
83 L. Almonte, E. Lopez-Elvira and A. M. Baró, ChemPhysChem, 2014, 15, 2768–2773.
84 R. C. Bruch and H. B. White, Biochemistry, 1982, 21, 5334–5341.
85 D. Wild, The Immunoassay Handbook, Gulf Professional Publishing, 2005.
86 M. D. Sonawane, S. B. Nimse, K.-S. Song and T. Kim, RSC Adv., 2016, 6, 7599–7609.
87 A. Poghosian, A. Cherstvy, S. Ingebrandt, A. Offenhäusser and M. J. Schöning, Sens. Actuators, B, 2005, 111–112, 470–480.
88 E. J. Fgot, D. F. Unterek, M. S. Noreenb and M. E. Meyerhoff, Anal. Chem., 1985, 57, 1995–1998.
89 J. O. Bockris and A. K. N. Reddy, Modern Electrochemistry 1, Kluwer Academic Publishers, Boston, 1998, vol. 1.
90 B. N. Dominy, D. Perl, F. X. Schmid and C. L. Brooks III, J. Mol. Biol., 2002, 319, 541–554.
91 E. Stern, R. Wagnerr, F. J. Sigworth, R. Breaker, T. M. Fahmy and M. A. Reed, Nano Lett., 2007, 7, 3405–3409.
92 P. Bergveld, Sens. Actuators, A, 1996, 56, 65–73.
93 H. H. Lee, M. Bae, S.-H. Jo, J.-K. Shin, D. H. Son, C.-H. Won and J.-H. Lee, Sens. Mater., 2015, 27, 575–583.
94 P. Estrela, D. Paul, Q. Song, L. K. Stadler, L. Wang, E. Huq, J. J. Davis, P. K. Ferrigno and P. Migliorato, Anal. Chem., 2010, 82, 3531–3536.
95 P. Casal, X. Wen, S. Gupta, T. Nicholson, Y. Wang, A. Theiss, B. Bhushan, L. Brillson, W. Lu and S. C. Lee, Philos. Trans. R. Soc., A, 2012, 370, 2474–2488.
96 A. Tarasov, M. Wipf, R. L. Stoop, K. Bedner, W. Fu, V. A. Guzenko, O. Knopfmacher, M. Calame and C. Schönengerber, ACS Nano, 2012, 6, 9291–9298.
97 Y. Maekawa, Y. Shibuta and T. Sakata, ChemElectroChem, 2014, 1, 1516–1524.
98 L. J. Criscenti, R. T. Cygan, A. S. Kooser and H. K. Moffat, Chem. Mater., 2008, 20, 4682–4693.
99 L. Bousse and P. Bergveld, Sens. Actuators, 1984, 6, 65–78.
100 S. Jamasb, IEEE Sens. J., 2004, 4, 795–801.
101 S. Jamasb, S. Collins and R. L. Smith, Sens. Actuators, B, 1998, 49, 146–155.
102 S. Jamasb, S. D. Collins and R. L. Smith, in 1997 International Conference on Solid State Sensors and Actuators, 1997. Transducers ’97 Chicago, 1997, vol. 2, pp. 1379–1382.
103 S. Jamasb, S. D. Collins and R. L. Smith, IEEE Trans. Electron Devices, 1998, 45, 1239–1245.
104 J.-L. Chiang, J.-C. Chou, Y.-C. Chen, G. S. Liu and C.-C. Cheng, Jpn. J. Appl. Phys., 2003, 42, 4973–4977.
105 S. Kim, D. W. Kwon, R. Lee, D. H. Kim and B.-G. Park, Jpn. J. Appl. Phys., 2016, 55, 06GG01.
106 B. S. Kang, F. Ren, L. Wang, C. Lofton, W. W. Tan, S. J. Pearston, A. Dabiran, A. Osinsky and P. P. Chow, Appl. Phys. Lett., 2005, 87, 023508.
107 Y. Wang, P. C. Sondergaard, A. Theiss, S. C. Lee and W. Lu, ECS Trans., 2014, 61, 139–146.
108 S. Jamasb, Circ. Syst. Signal. Pr., 2016, 10, 119–125.
109 R. B. Schasfoort, R. P. H. Kooyman, P. Bergveld and J. Greve, Biosens. Bioelectron., 1990, 5, 103–124.
110 P. Bergveld, Biosens. Bioelectron., 1991, 6, 55–72.
111 V. Krivitsky, M. Zverzhinetsky and F. Patolsky, Nano Lett., 2016, 16, 6272–6281.
112 S. Kim, J.-Y. Kim, J.-H. Ahn, T. J. Park, S. Y. Lee and Y.-K. Choi, Appl. Phys. Lett., 2010, 97, 073702.
113 N. Gao, W. Zhou, X. Jiang, T.-M. Fu and C. M. Lieber, Nano Lett., 2015, 15, 2143–2148.
114 K. Georgakopoulou, A. Birbas and C. Spathis, J. Appl. Phys., 2015, 117, 104505.
115 I. Jokić, M. Frantović, Z. Djurić, K. Radulović and Z. Jokić, Microelectron. Eng., 2015, 144, 32–36.
116 G. S. Kulkarni and Z. Zhong, Nano Lett., 2012, 12, 719–723.
117 G. S. Kulkarni, W. Zang and Z. Zhong, Acc. Chem. Res., 2016, 49, 2578–2586.
118 IUPAC Compendium of Chemical Terminology: Gold Book, ed. M. Nić, J. Jirát, B. Košata, A. Jenkins and A. McNaught, IUPAC, Research Triangle Park, NC, 2.1.0., 2009.
119 F. N. Ishikawa, M. Curreli, H.-K. Chang, P.-C. Chen, R. Zhang, R. J. Cote, M. E. Thompson and C. Zhou, ACS Nano, 2009, 3, 3969–3976.
120 H. Nam, B.-R. Oh, M. Chen, S. Wi, D. Li, K. Kurabayashi and X. Liang, J. Vac. Sci. Technol., B, 2015, 33, 06FG01.
121 N. Elfström, Silicon Nanowires for Biomolecule Detection, PhD thesis, Royal Institute of Technology, 2008.
122 R. L. Rich and D. G. Myszka, J. Mol. Recognit., 2008, 21, 355–400.
123 S. Chen, J. G. Bomer, E. T. Carlen and A. van den Berg, Nano Lett., 2011, 11, 2334–2341.
124 K. Shoorideh and C. O. Chui, Understanding and optimization of the sensitivity of nanoscale FET-based biosensors, Proceedings Volume 9174, Nanoepitaxy: Materials and Devices VI, 2014, vol. 9174, pp. 917413–917416.
125 K. Shoorideh and C. O. Chui, Proc. Natl. Acad. Sci. U. S. A., 2014, 111, 5111–5116.
126 J. Lee, J. Jang, B. Choi, J. Yoon, J.-Y. Kim, Y.-K. Choi, D. M. Kim, D. H. Kim and S.-J. Choi, Sci. Rep., 2015, 5, 12286.
127 A. Vacic, J. M. Criscione, E. Stern, N. K. Rajan, T. Fahmy and M. A. Reed, in 2014 International Conference on Microelectronic Test Structures (ICMTS), 2014, pp. 203–206.
128 A. Vacic, J. M. Criscione, E. Stern, N. K. Rajan, T. Fahmy and M. A. Reed, Biosens. Bioelectron., 2011, 28, 239–242.
129 N. K. Rajan, K. Brower, X. Duan and M. A. Reed, Appl. Phys. Lett., 2014, 104, 084106.
Critical Review

154 S. Baumgartner, C. Heitzinger, A. Vacic and M. A. Reed, Nanotechnology, 2013, 24, 225503.
155 T. Windbacher, V. Sverdlov, S. Selberherr, C. Heitzinger, N. Mauser and C. Ringhofer, AIP Conf. Proc., 2010, 1199, 507.
156 M. W. Shinwari, M. J. Deen and D. Landheer, Microelectron. Reliab., 2007, 47, 2025–2057.
157 D. Landheer, W. R. McKinnon, W. H. Jiang and G. Aers, Appl. Phys. Lett., 2008, 92, 253901.
158 M. H. Sørensen, N. A. Mortensen and M. Brandbyge, Appl. Phys. Lett., 2007, 91, 102105.
159 J. O. Bockris, A. K. N. Reddy and M. E. Gamboa-Aldeco, Modern Electrochemistry 2A, Kluwer Academic Publishers, Boston, 2002, vol. 2A.
160 A. Choi, K. Kim, H.-I. Jung and S. Y. Lee, Sens. Actuators, B, 2010, 148, 577–582.
161 P. Ginet, S. Akiyama, N. Takama, H. Fujita and B. Kim, J. Micromech. Microeng., 2011, 21, 065008.
162 C.-Y. Hsiao, C.-H. Lin, C.-H. Hung, C.-J. Su, Y.-R. Lo, C.-C. Lee, H.-C. Lin, F.-H. Ko, T.-Y. Huang and Y.-S. Yang, Biosens. Bioelectron., 2009, 24, 1223–1229.
163 P. Hu, J. Zhang, Z. Wen and C. Zhang, Nanotechnology, 2011, 22, 335502.
164 K. A. Jeon, H. J. Son, C. E. Kim, M. S. Shon, K. H. Yoo, A. M. Choi, H. I. Jung and S. Y. Lee, IEEE, 2006, pp. 1265–1268.
165 J. S. Kim, W. I. Park, C.-H. Lee and G.-C. Yi, Korean Phys. Soc., 2006, 49, 1635.
166 D. Khatayevich, T. Page, C. Gresswell, Y. Hayamizu, W. Grady and M. Sarikaya, Small, 2014, 10, 1505–1513.
167 I. R. Olmsted, A. Kussrow and D. J. Bornhop, Anal. Chem., 2012, 84, 10817–10822.
168 D. A. Hall and C. J. Langmead, Br. J. Pharmacol., 2010, 161, 1276–1290.
169 T. M. Squires, R. J. Messinger and S. R. Manalis, Nat. Biotechnol., 2008, 26, 417–426.
170 E. Stern, Label-free sensing with semiconducting nanowires, PhD thesis, Yale University, 2007.
171 X. Duan, L. Mu, S. D. Sawtelle, N. K. Rajan, Z. Han, Y. Wang, H. Qu and M. A. Reed, Adv. Funct. Mater., 2015, 25, 2279–2286.
172 L. S. Jung, K. E. Nelson, P. S. Stayton and C. T. Campbell, Langmuir, 2000, 16, 9421–9432.
173 D. Sarkar and K. Banerjee, Appl. Phys. Lett., 2012, 100, 143108.
174 M. L. Soudjin, Proton Transport in Aqueous Ionic Solutions, PhD thesis, University of Amsterdam, 2012.
175 C. J. T. de Grothhuss, Ann. Chim., 1806, 58, 54–73.
176 B. M. Lowe, C.-K. Skylaris and N. G. Green, J. Colloid Interface Sci., 2015, 451, 231–244.
177 P. Schuck and H. Zhao, Methods Mol. Biol., 2010, 627, 15–54.
178 E. Buitrago, École Polytechnique Fédérale de Lausanne, 2014.
179 G. Shalev, A. Cohen, A. Doron, A. Machauf, M. Horesh, U. Virobnik, D. Ullien and I. Levy, Sensors, 2009, 9, 4366–4379.
180 N. M. J. Ditshego, N. A. B. Ghazali, M. Ebert, K. Sun, I. Zeimpekis, P. Ashburn, M. R. R. de Planque and H. M. H. Chong, IEEE, 2015, pp. 801–804.
181 K. Sun, I. Zeimpekis, C. Hu, N. M. J. Ditshego, O. Thomas, M. R. R. de Planque, H. M. H. Chong, H. Morgan and P. Ashburn, Nanotechnology, 2016, 27, 285501.
182 M. A. Brown, Z. Abbas, A. Kleiber, R. G. Green, A. Goel, S. May and T. M. Squires, Phys. Rev. X, 2016, 6, 011007.
183 S. Kim, K. Kim, T. Rim, C. Park, D. Cho, C. K. Baek, Y. H. Jeong, M. Meyyappan and J. S. Lee, in 2011 IEEE International Conference on Nano/Micro Engineered and Molecular Systems (NEMS), 2011, pp. 1233–1236.
184 S. Kim, T. Rim, K. Kim, U. Lee, E. Baek, H. Lee, C.-K. Baek, M. Meyyappan, M. J. Deen and J.-S. Lee, Analyst, 2011, 136, 5012–5016.
185 X. T. Vu, R. Stockmann, B. Wolfrum, A. Offenhäuser and S. Ingebrandt, Phys. Status Solidi A, 2010, 207, 850–857.
186 C.-E. Lee, T.-C. Yu, C.-M. Yang, D. G. Pijanowska and C.-S. Lai, Sensors, 2011, 11, 4562–4571.
187 A. Tarasov, N. de Rooij, C. Schönberger and L. Linnros, Silicon nanowire field-effect transistors for sensing applications, PhD thesis, University of Basel, 2012.
188 S. Zafar, C. D’Emic, A. Afzali, B. Fletcher, Y. Zhu and T. Ning, Nanotechnology, 2011, 22, 405501.
189 S. Rigante, M. Wipf, A. Bazigos, K. Bedner, D. Bouvet and A. M. Ionescu, IEEE, San Francisco, CA, USA, 2014, pp. 1063–1066.
190 K. Bedner, V. A. Guzenko, A. Tarasov, M. Wipf, R. L. Stoop, D. Just, S. Rigante, W. Fu, R. A. Minamisawa, C. David, et al., Sens. Mater., 2013, 25, 567–576.
191 T. Akiyama, Y. Ujihira, Y. Okabe, T. Sugano and E. Niki, IEEE Trans. Electron Devices, 1982, 29, 1936–1941.
192 D. E. Yates, S. Levine and T. W. Healy, J. Chem. Soc., Faraday Trans., 1974, 70, 1807–1818.
193 T. W. Healy and L. R. White, Adv. Colloid Interface Sci., 1978, 9, 303–345.
194 A. Tarasov, M. Wipf, K. Bedner, J. Kurz, W. Fu, V. A. Guzenko, O. Knopfmacher, R. L. Stoop, M. Calame and C. Schönberger, Langmuir, 2012, 28, 9899–9905.
195 M.-N. Niu, X.-F. Ding and Q.-Yi Tong, IEEE, Penang, Malaysia, 1996, pp. 189–193.
196 A. Star, J.-C. P. Gabriel, K. Bradley and G. Grüner, Nano Lett., 2003, 3, 459–463.
197 R. L. Rich and D. G. Myszka, J. Mol. Recognit., 2006, 19, 478–534.
198 R. B. M. Schasfoort and A. J. Tudos, Handbook of Surface Plasmon Resonance, Royal Society of Chemistry, 2008.