Review

The Challenges of Developing Biosensors for Clinical Assessment: A Review

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Abstract: Emerging research in biosensors has attracted much attention worldwide, particularly in response to the recent pandemic outbreak of coronavirus disease 2019 (COVID-19). Nevertheless, initiating research in biosensing applied to the diagnosis of diseases is still challenging for researchers, be it in the preferences of biosensor platforms, selection of biomarkers, detection strategies, or other aspects (e.g., cutoff values) to fulfill the clinical purpose. There are two sides to the development of a diagnostic tool: the biosensor development side and the clinical side. From the development side, the research engineers seek the typical characteristics of a biosensor: sensitivity, selectivity, linearity, stability, and reproducibility. On the other side are the physicians that expect a diagnostic tool that provides fast acquisition of patient information to obtain an early diagnosis or an efficient patient stratification, which consequently allows for making assertive and efficient clinical decisions. The development of diagnostic devices always involves assay developer researchers working as pivots to bridge both sides whose role is to find detection strategies suitable to the clinical needs by understanding (1) the intended use of the technology and its basic principle and (2) the preferable type of test: qualitative or quantitative, sample matrix challenges, biomarker(s) threshold (cutoff value), and if the system requires a mono- or multiplex assay format. This review highlights the challenges for the development of biosensors for clinical assessment and its broad application in multidisciplinary fields. This review paper highlights the following biosensor technologies: magnetoresistive (MR)-based, transistor-based, quartz crystal microbalance (QCM), and optical-based biosensors. Its working mechanisms are discussed with their pros and cons. The article also gives an overview of the most critical parameters that are optimized by developing a diagnostic tool.

Keywords: biosensors; biomarker; clinical; diagnostic; cutoff value; panel biomarkers; dynamic range; sensitivity

1. Introduction

Biosensors for clinical application is an emerging research field, particularly for the rapid detection and early screening of biomarkers in the case of an outbreak, such as coronavirus disease 2019 (COVID-19) [1,2]. More than hundreds of thousands of confirmed cases are reported from hundreds of countries around the globe. In this vast increment of the new cases, the paramedics in these countries are overwhelmed with screenings of suspected cases for quarantined patients. The facts brought by the recent pandemic caused by SARS-CoV-2 emphasize the need for early screening methods to interrupt the spread of the disease. Those screening methods can be relevant in airports, seaports, country borders, or among communities. The conventional methods for pathogen detection, such as cell culture systems or reverse transcript polymerase chain reactions (RT-PCRs), are laborious, costly, and time-consuming from sample preparation until signal interpretation.
The definition of a biosensor by the International Union of Pure and Applied Chemistry (IUPAC) is a device or platform that uses specific biochemical reactions mediated by isolated enzymes, immunosystems, tissues, organelles, or whole cells to detect specific biochemical compounds, usually by electrical, thermal, magnetic or optical signals [3]. By its definition, biosensors include three crucial parts: (1) the interesting biomarker to target, (2) the receptor (i.e., the biorecognition element to use in the sensing area that will specifically interact or react with the interested biomarker(s)), and (3) the method to generate the signal response of the biomarker–receptor interaction. Moreover, according to the World Health Organization (WHO), an emerging requirement for biosensor technology features is called ASSURED, which stands from affordable, sensitive, specific, user-friendly, rapid and robust, equipment-free, and deliverable to the end users [4].

This article discusses the state of the art, recent issues, and challenges of biosensor development for clinical applications. Aside from the several existing biosensor platforms available in the market for specific clinical applications, there is still no consensus by the health authorities regarding biosensor utilization, particularly during the outbreak or for screening purposes. The biosensor platform in this article focuses on magnetoresistive-based, field-effect transistor (FET)-based, quartz crystal microbalance (QCM), surface plasmon resonance (SPR)-based, and surface-enhanced Raman spectroscopy (SERS)-based biosensors. The preferred platform in this review represents different transducing mechanisms and working principles from various biosensor platforms in the literature.

Additionally, biomarker preferences are a critical issue from the clinical side. The biomarkers can be proteins, nucleic acids, volatile organic compounds, cells, or particular biochemical substances, either as single biomarkers or combinations of several biomarkers for the detection validity, accuracy, and disease stratification [5]. The preference of a multiplex detection approach can be very challenging because each biomarker has different characteristics in terms of biophysical properties, concentrations, and relevant clinical cut-off values. Aside from the importance of the limit of detection (LoD) and sensitivity performance within the different biosensor technologies, those aspects should be considered in developing biosensors for advanced clinical studies. Finally, this review article provides a comprehensive overview for either early scientists or experts and for engineers or biochemists for their research roadmaps in the clinical applications of the biosensors.

This paper is organized into six sections. Section 1, “Introduction”, discusses the background and challenges behind developing a biosensor for clinical applications. Section 2, “Biosensor Technologies”, focuses on the state of the art of the most common biosensor platforms. Section 3, “Biosensor Analytical Performance”, is related to the parameters that need to be tackled to obtain robust diagnostic devices. Section 4, “Biomarker Strategy,” explains the importance of using individual or multiple biomarkers. Section 5, “Biomarker Clinical Cut-off Value”, explains its role in the detection strategy. The last section, “Conclusion”, will cover the systematic key steps to consider in developing a clinical diagnostic tool in a workflow chart.

2. Biosensor Technologies

Several technologies are reporting their applications in biosensing, either as proofs of concept or as commercial platforms, with their pros and cons. In this section, the current technologies in biosensor development are categorized into the following clusters: (1) magnetoresistive (MR)-based, (2) transistor-based, (3) quartz crystal microbalance (QCM), and (4) optical-based biosensors. These technologies have specific working mechanisms that translate particular challenging physical properties of the biomarkers (e.g., the charge of biomarkers, molecular weight, and refractive indices). These will define which detection strategy to apply in a particular clinical application (e.g., sample matrix or required cut-off value). For example, a magnetoresistive-based biosensor relies on the magnetic field, while a transistor-based biosensor relies on charge transfer mechanisms between the
detection sample and the surface. QCM relies on the resonance frequency between piezoelectric materials, and finally, the optical-based biosensor relies on the photon interaction for the sensing region and target sample.

2.1. Magnetoresistive-Based Biosensors

MR-based biosensors make use of a device conventionally used for read heads in hard disk drives. This sensor measures the resistivity of the material or structures by the presence of magnetic fields [6]. In an MR-based biosensor, the magnetic nanoparticles (MNPs) are used to label the interest biomarkers that interact with the biorecognition elements over the magnetoresistive structures [7,8]. Consequently, this induces changes in the resistance value of the structure, which can be measured in real time (Figure 1).

Figure 1. The general concept of an MR-based biosensor for the detection of MNP-labeled biomarkers with (A) the current in plane (CIP) of the electrical measurement and (B) the current perpendicular to plane (CPP). (C) The typical real-time response curve of the biomarker binding on the MR-based biosensor.

Figure 1A depicts an MR-based sensor measured using the CIP, in which the electrical current flows parallel to the plane of the MR sensor. Moreover, several sensors are constructed using the current-to-perpendicular plane (CPP) as depicted in Figure 1B, where the electrical current flows perpendicular to the plane of the MR structures. Monitoring the voltage value of the MR structures results in a real-time signal (Figure 1C) when the MNP-labeled biomarkers are captured by the biorecognition elements immobilized in the sensing surface. The signal magnitude in the MR biosensor will be the difference between the reference signal (Vs) and the specific binding signal (Vb) such that \( \Delta V = V_s - V_b \).

Based on the structure of the materials, the MR-based biosensor can be configured mainly using giant magnetoresistance (GMR) and tunneling magnetoresistance (TMR) [7,9,10]. The main advantages of using an MR biosensor are that the magnetic field through the presence of MNPs is not sensitive to the charge and the mild temperature gradient of the sample. Therefore, the noise in the reference signal is independent of the temperature and charge effect from the sample or target markers [11,12].
2.2. Transistor-Based Biosensors

Transistor-based biosensors are dominated by the FET device that uses the gate structure to perform the biomolecular interaction. The detection principle of this biosensor is based on the accumulation of charged biomarkers at the biorecognition surface (e.g., gate oxide) that create changes in the surface potential, leading to variations in the drain current for a fixed gate voltage. This signal can be used to observe biomarker detection in the FET device, and the shift direction is related to the charge of the biomarker in the sample medium.

Figure 2A depicts a typical FET-based biosensor by using the gate of the device as a sensing membrane. The accumulation of the probes or targets will be represented by the charge accumulation on the gate and the shifting of the threshold voltage, as illustrated in Figure 2B. Assuming that the FET device in Figure 2A is an n-type device, the negatively charged samples result in the threshold voltage (V_{TH}) shifting to a lower value, as depicted by the solid lines of Figure 2B. In the case of positively charged targets, the V_{TH} shifts to the higher V_C (the dashed line in Figure 2B). The real-time measurement can be performed by monitoring the drain current (I_D) at a particular fixed gate voltage (V_G) value around the linear region of the I_D–V_G curve. Another strategy of signal acquisition will be monitoring the V_C at a constant threshold current (typically 1 µA in microelectronics). This signal acquisition in an FET biosensor that utilizes the feedback circuit is called constant voltage constant current (CVCC) [13].

The advantage of the FET biosensor device is that it is label-free. However, several FET technologies keep using labels to enhance the detection signal or to integrate the FET detection simultaneously with other methods, such as electrochemical or optical analyses [14–16]. Nevertheless, biomarkers with a neutral charge (e.g., IgG2 and hemoglobin) are challenging to detect using the FET biosensor device [17,18]. FET-based biosensors can be configured as portable and low-power platforms adapted from the integrated circuit (IC) industry, such as well-known FET technologies in microelectronics and microprocessors for digital devices [14,19,20]. Moreover, the use of the commercial FET device for this sensor configuration is also possible to apply by using the extended gate field effect transistor (EGFET) configuration. In the EGFET device, the sensing membrane is connected to the gate, and various treatments of the sensing area can be performed separately from the transistor circuit [21–23].

2.3. Quartz Crystal Microbalance Biosensors

A quartz crystal microbalance (QCM) device is a sensor for the detection of surface binding by monitoring the resonance frequency [24–26]. It has two conductive electrodes...
usually made from gold that are separated by a quartz crystal as a piezoelectric material. This technology is relatively mature because it is well known for monitoring material deposition in microelectronics fabrication. It works by supplying a frequency range from the kHz level to a few MHz to define the resonance state. The higher the thickness of the quartz substrate, the lower the resonance frequency measured. The QCM is well known for the high-quality factor (Q) of the sensor. However, due to the thick substrate (in mm), the only possible frequency operation is in the low range of frequencies. Consequently, this sensing platform has a high detection resolution and a wide dynamic range [27]. The advantage of the biosensing platform using a QCM is that the sample can be used either in the liquid or vapor phase [28].

The principle of signal acquisition on the QCM platform can be illustrated in Figure 3B. The resonance frequency can be shifted to the lower value when the substance or molecules bind to the electrode’s surface. Moreover, real-time signal acquisitions can be applied by monitoring the resonance frequency along with the time domain as depicted in the inset of Figure 3B.

![Figure 3. (A) Typical QCM electrodes. (B) Schematic circuit of a QCM with a frequency supply in the range from kHz to MHz (top). A frequency drop indicates the loaded electrode during the QCM measurement (bottom). The real-time measurement strategy in a QCM monitors the resonance frequency shifting (inset).](image)

2.4. Optical-Based Biosensors

Optical-based biosensors typically use light in the infrared or visible region as the excitation source for the sensing region. They are sensitive to the refractive index of the surface. In this article, this type of biosensors will be divided into two groups: plasmonic-based and surface-enhanced Raman scattering (SERS)-based platforms.

2.4.1. Plasmonic-Based Biosensors

A plasmonic is an electromagnetic field naturally generated in a noble metal fabricated in nanofilm or nanostructures. In a nanofilm metal structure, the plasmonic field is called surface plasmon resonance (SPR). The existing field is located at the interface of the metal film and the dielectric medium, while in the nanoparticle structures, it is called localized surface plasmon resonance (LSPR). Along with a massive trend in nanotechnology, this technology has gained much attention from scientists due to the plasmonic phenomena that do not exist in bulk structures [29]. The plasmonic phenomena in a metal, due to the sensitivity, have excellent potential for the application of single-molecule detection [30].

Figure 4A illustrates the SPR sensor using a prism coupler to excite the surface plasmon wave (SPW) in the interface of the thin metal film and sample medium. At a particular angular incident angle (θ) and wavelength (λ), the p-polarized light will be resonated
with the SPW. Therefore, a fraction of the incident light will be absorbed into the SPW. Consequently, in the reflectance spectra, the loss of reflected light can be observed as a dark band along with the angular or wavelength domain (Figure 4C,D). This dark band will be the reference point of the sensing. In case the medium shifts to a higher refractive index \( (n_2 > n_1) \) or the sensing metal adsorbs the biomolecules, the resonance condition will be shifting to the higher wavelength \( (\Delta n \equiv \Delta \lambda) \) or higher incident angle \( (\Delta n \equiv \Delta \theta) \). The shifting can be tracked as real-time signals from the intensity modulation, angular interrogation, or wavelength interrogation (Figure 4C,D). In Figure 4B, the sensing metal is replaced by the nanodisc array. In this configuration, the plasmonic field oscillates surrounding the discs and also at the interspace between the discs. Theoretically, the plasmonic field in the LSPR configuration is stronger than in the SPR. Thus, the sensing performance can be boosted to achieve smaller detection limits [31–35]. However, the fabrication process can be costly, such as with the nanostructure array with a precise arrangement. Aside from the nanodisc array, various metal nanostructures can be utilized, such as gold or silver nanoparticles, nanoholes, nanostars, nanocubes, nanorods, and other different shapes. The particular metal nanopatterns have specific resonance wavelength regions in the range of visible light, and the shifting of the resonance wavelength leads to the color change. This LSPR behavior can be utilized as a colorimetry sensor [36,37].
The plasmonic-based platforms are well known for the high sensitivity performance of the biosensors [29,38]. Nevertheless, most of the commercial platform in the market requires a very high investment cost [39–41]. The drawback of this type of platform is the high sensitivity to temperature gradients. This disadvantage is due to the transducing mechanism’s dependence on the solution’s refractive indices at the sensing interface. Because the refractive indices values of the solution sample entirely depend on the temperature, the accuracy of the measurement can therefore be misinterpreted.

2.4.2. SERS-Based Biosensors

The SERS-based biosensor uses Raman spectroscopy as the central platform. Raman spectroscopy measurements are based on inelastic scattering of light excitation when the light interacts with the vibration of chemical bonds. The scattered light shifts to different wavelengths compared with the incident light. Every chemical bond leads to different wavelength shifting. Therefore, Raman spectroscopy is one of the powerful techniques to observe the chemical fingerprint of materials [42–44].

The Raman system can be optimized as a SERS-based biosensor platform by using metallic nanostructures as the sensing region to detect the biomarker binding. The metallic nanostructures confine a strong plasmonic field in the LSPR mode. Next, it interacts with the incident photons to enhance the scattering intensity of the Raman signal [45]. When the immobilized biorecognition elements in the nanostructures capture the target biomarker, the unique Raman fingerprint signal can be obtained. The intensity modulation of the Raman signal can be correlated to the target concentrations. The illustration of the SERS measurement is depicted in Figure 5.

The photons of the laser’s incident light ($h\nu$) are exciting the abundance of electrons on the metal surface. The LSPR energy level by the metal nanostructures enhances the electron supply to the HOMO energy level at the Rhodamine 6G (R6G) molecules (Figure 5B). Consequently, a strong Raman signal is generated by the electron transition from the HOMO to the LUMO energy level. Therefore, the Raman signal from the R6G represents the existence of biomolecules on the surface of the nanostructures.

The SERS-based biosensor has been acknowledged for its very high sensitivity performance. This platform is promising for the future advanced study of single-cell detection [46,47]. The performance enhancement factor of SERS was reported to achieve several orders of magnitude [48–51]. Nevertheless, the SERS platform cannot perform real-time signal acquisition to observe the binding affinity of the biomolecules. In addition, the fabrication cost of the metal nanostructures is high, and electron beam lithography (EBL) in particular is required for high precision of the nanopatterns.
Figure 5. (A) Illustration of SERS-based biosensors using Rhodamine 6G (R6G) as the biomolecule label on the sensing structure with a gold nanosphere array. (B) The energy band diagram and electron transition illustration of SERS-based biosensors.

The assay developer researcher needs to identify the biophysical properties of the target sample before the investment of the specific biosensors. For example, for a target sample with a neutral charge (e.g., peptide, IgG2, and hemoglobin) [17,18,52], the decision of using a FET-based platform is not suitable. In addition, in the case of requiring the use of various solutions with significantly different refractive indices values or multiple temperature gradients, it can be challenging to be detected in the plasmonic-based platform (SPR and LSPR biosensors) [53].

Table 1 can be an essential list for potential biosensor users according to their sample characteristics and its biophysical properties. Therefore, the selection of a transducing mechanism can be adequately considered. In addition, the cost and simplicity of the operation are also important factors for the user, whether it requires advanced training for the user or not. In applications such as outbreak diseases, simple operation biosensors can be preferable for early users to perform enormous tests. Moreover, the label-free and real-time features can be another advantage for the user’s consideration. The label-free scheme leads to a straightforward and low-cost assay in practical use, while the real-time signal can be analyzed to see the binding affinity of the target molecules to the receptor, as well as the direct qualitative and quantitative interpretation along with the time domain.

| Platform Category | Transducing Mechanism | Cost of Equipment, Operation | Pros | Cons | Label-Free Option | Real-Time, Portability Options | Ref |
|-------------------|-----------------------|-----------------------------|------|------|------------------|-------------------------------|-----|
| Magneto-resistive | Detection of magnetic field fringe of MNP | High, High | Robust, not sensitive to charge and temperature; tailored labeling | Required magnetic labeling | No | Yes, Yes | [6–12] |
| FET               | Detection of sample charge | Low, Low | Mature technology | Difficulty in detecting neutral charge samples; noise | Yes | Yes, Yes | [13–16,19–23] |
3. Biosensor Analytical Performance

The analytical performance of a clinical biosensor allows us to understand the capabilities and limitations of the technology and essentially analyze if it addresses a specific application or not. Several general parameters indicate the performance of a biosensor, such as the sensitivity, limit of detection (LOD), specificity, reproducibility, and dynamic range (DR). Therefore, before starting to develop a biosensor, it is important to have an intended clinical purpose. For example, the diagnosis of some complex diseases requires the detection of multiple biomarkers with specific cutoff values, meaning that those biomarkers may exist in an ordinary person but at constant or low concentrations. This issue is one of the most challenging aspects to address in developing a diagnostic tool. The device needs to look for multiple rather than individual biomarkers and distinguish between certain levels of those biomarkers, rather than distinguishing a minimal concentration from zero. In this case, the biosensors’ sensitivity, dynamic ranges, and resolutions may be critical parameters to address. Therefore, the detection strategy should be optimized considering the complexity of the biomarker(s), the affinity of the biorecognition elements, the clinical cutoff values, the relevant dynamic range, and the sample matrix. In this section, the essential parameters related to the performance of a clinical biosensor will be described using simple illustrative curves for the detection of a small-to-high target concentration. Some examples of adjusting the parameters to the clinical challenge are also reported.

3.1. Sensitivity

The biosensor sensitivity is defined as the response signal for every unit of the target sample’s concentration. The typical standard curve of the biosensing response for the target detection with a dose–response fitting is depicted in Figure 6. In this curve, a pM unit of concentration is assumed. The slope of the linear region determines the sensitivity (s) in the fitting curve, which is the value of the signal magnitude (y) divided by the unit concentration (x) of the response slope. The higher y value in the response of the similar x value indicates the better sensitivity performance of the biosensor.
In principle, the sensitivity can be enhanced by improving the biosensors’ signal-to-noise ratio (SNR), such as by using sandwich assays [54–56] or by utilization of nanoparticle labels in the analytical system [57–59]. Other attempts include downscaling the sensing area into two-dimensional (2D) or one-dimensional (1D) structures [60–63]. The 2D materials for the advanced sensing structure can be graphene [38,64,65], which enhances the detection down to attomolar and femtomolar ranges [66–68]. Other 2D materials, such as molybdenum disulfide (MoS2), were reported to be able to boost the detection down to the femtomolar range for protein [69] and DNA detection [70]. Furthermore, the advanced 2D material Mxene was reported to enhance the biosensor detection performance down to 330 fM for the breast cancer marker [71]. While the popular 1D material for biosensing structures can be carbon nanotubes [72–74], it was reported to be able to perform the detection down to 10 fM for the trimethylamine target [72], while metal oxide-decorated carbon nanofiber was reported to be able to improve the detection down to 5 fM for the binding of platelet-derived growth factor-BB (PDGF-BB) [75]. Nevertheless, the significant challenges for 1D material sensing are the high fabrication cost for the scaled-up production and the device-to-device variations.

3.2. Dynamic Range

The definition of the dynamic range is the concentration window between the maximum and minimum usable indication of the signal response. Technically, the dynamic range is calculated from the detection limit to the saturation level of the biosensor signal (Figure 7A). Nevertheless, in the dynamic range, the signal response does not always increase proportionally to the target concentration, particularly in the extremely low or high concentration target. Therefore, another term that biosensor experts use is the linear dynamic range to determine the biosensor signal response in the linear region of the fitting curve (Figure 7B). Consequently, the linear range is a trade-off parameter compared with the sensitivity (i.e., when the sensitivity value is high, the detection range will be small, since the saturation level will be reached at a lower target concentration). On the other hand, if the application requires a large linear range performance, the sensitivity of the biosensor will be lower [76,77].
Figure 7. (A) The illustrated dynamic range in a biosensor standard curve with the dose–response fitting curve. (B) Linear dynamic range.

The dynamic range can be enhanced by increasing the surface area of the sensing. The larger surface area can be engineered by using three-dimensional (3D) sensing structures, such as nanopillar or nanoflower arrays, to improve the area for the occupancy of the biorecognition elements [78–80], thereby improving the probability to detect high concentrations of the target. However, when developing detection strategies for clinical biomarkers, what will determine the valid dynamic range is the cut-off value (threshold) of the interested biomarkers (explained in detail in Section 4). Technology may allow a log-linear range, but depending on the device’s intended use, the researcher may need to adjust several parameters (if possible) to obtain a narrower dynamic range. The use of different structures of sensing [77,81], diluted and label tailoring [82], varying the concentration of biorecognition elements [83], and the use of antibodies with different affinities [84] are some of the parameters that can be manipulated to achieve the pretended sensitivity and dynamic range. Several strategies for dynamic range enhancements are listed in Table 2.

Table 2. Summary of the dynamic range enhancement.

| Platform       | Sample            | Technical Remark                  | Dynamic Range            | Ref.  |
|----------------|-------------------|-----------------------------------|--------------------------|-------|
| Dual cavity FPI| N/a               | Utilizing weak composite          | 1.359–1.452              | [76]  |
| SPR            | Sucrose water     | Al/Au sensing structure           | 1.33–1.45 RIU            | [77]  |
| SPR            | Antibody loading  | Zwitterionic hydrogel             | Antibody capacity up to 693 ng/cm² | [78]  |
| SPR            | N/a               | TiO2/SiO2                         | 1.7742–1.9542 RIU        | [81]  |
| SERS           | Bilirubin         | Au-MoS2 NFs                       | 10⁻¹² to 10⁻⁴ M          | [79]  |
| Electrochemical| *Staphylococcus epidermidis* | ZnO Nanoglass                  | 10⁻¹¹–10⁻⁴ M              | [80]  |
| Magnetoresistive| cFn               | Dilution MNP                     | 1–4 µg/mL                | [82]  |
| Octet Biosensor| Human mAb         | Varying biorecognition concentration | 0.15–40 nM              | [83]  |

FPI: Fabry–Perot interferometer; SPR: surface plasmon resonance; SERS: surface-enhanced Raman spectroscopy; cFn: cellular fibronectin; RIU: refractive index unit; Au-MoS2 NFs: gold molybdenum disulfide nanoflowers; MNP: magnetic nanoparticles; mAb: monoclonal antibody.

3.3. Limit of Detection

Based on the IUPAC, the definition of the LOD is the value expressed by the smallest concentration (x) that can be detected with a reasonable certainty confidence level [85]. Figure 8 shows a common LOD concept in biosensor application. The lowest response signal is obtained by three times the standard deviation (3SD) from the mean value of a
blank measurement \((y_0)\). However, another article reported using different confidence level values instead of three times the standard deviation, such as 3.29 times the SD, to obtain a 95% confidence interval of the smallest signal from the blank measurement [86]. For the simplicity of the illustration, in Figure 8, a confidence level of 3 is applied. The magnitude of \(3SD\) from the reference signal can be plotted to find the concentration value of the LOD. Therefore, to define a valid LOD value, the measurement of the zero-concentration target as the reference signal is mandatory. The accurate value of the LOD \((x)\), can be calculated by the known value \((y_0 + 3SD)\) on the fitting curve’s (red) formula. Based on this definition, the SD of the blank measurement is the most critical factor in obtaining the LOD value. The measurement uncertainty will determine the accuracy of the LOD. Moreover, if the linear sensitivity slope is higher, a lower LOD can be obtained (in the case of an identical SD value). Therefore, for early detection purposes, the high sensitivity value is preferable to achieve a lower LOD.

![Figure 8. The basic concept for determination of the LOD in a fitting curve of the biosensor response.](image)

3.4. Specificity

The general concept of sensing is distinguishing a particular substance from others (selective) and capturing a particular target (specific). Therefore, the biosensing system should be configured to detect a specific target by binding with the specific biorecognition element through strong and selective affinity. The quality of the biological components will determine the specificity and, consequently, the sensitivity of the developed detection strategy. Moreover, complex samples may require two labels (dual labeling) to increase the assay specificity [87]. For the evaluation of the specificity performance, the measurement of the non-complementary samples should be compared to the specific target. A functional biosensor should be able to distinguish these different samples producing significantly different response signals. Ideally, the non-specific target’s resulting signal level is similar to the zero-concentration measurement of the biosensor. Nevertheless, in the high concentration of non-specific targets, the false positive or drift signal can appear due to non-specific bindings and interferences. The illustration of the specificity performance in biosensors can be depicted in Figure 9.
Several attempts have been reported to improve the specificity and low-fouling binding during the biosensing measurements, either using a competitive assay or complex media such as serum or whole blood, as well as blocking assays [88–90], a polymer brush, and a zwitterionic functionalization surface [91–94], or by using label and sandwich assays to obtain a specific signal [95–97]. In immunoassays, monoclonal antibodies are also recommended to obtain high specificity and lower cross-reactivity events [98]. The summary of specificity strategies is listed in Table 3.

Table 3. Summary of the specificity enhancement.

| Strategy                | Platform | Sample                                                                 | Technical Remark                                                                 | Performance                | Ref.  |
|-------------------------|----------|------------------------------------------------------------------------|----------------------------------------------------------------------------------|----------------------------|-------|
| Blocking                | SPR      | ssODNs of MTBC and *M. gordonae*                                       | Spacer arms on probe-ssODNs, MCH for blocking and probe orientation              | Regenerative sensing, up to 12 weeks, RT storage | [88]  |
| Blocking                | Electro-chemical | cTnI                                     | nMoSe embedded on rGO, BSA blocking                                           | LDR: 10⁻⁶–100 ng/mL; shelf life 35 days | [89]  |
| Blocking                | Electro-chemical | PDGF                                     | GOD blocking, P-Gra-GNPs                                                      | LOD 1.7 pM; LDR: 0.005–60 nM  | [90]  |
| Anti-fouling membrane   | Fluorescence | DENV and ZIKV                                                            | 3D Cu- zwitterionic MOF, simultaneous detection                                 | LOD (in pM): 192 (DENV) 332 (ZIKV) | [91]  |
| Anti-fouling membrane   | SPR      | Undiluted human plasma and serum, mammalian cells                       | polyAAEE brushes on sensing                                                      | Undetectable protein adsorption (<0.3 ng/cm²) | [93]  |
| Anti-fouling membrane   | SPR      | Epstein-Barr virus in serum                                              | Label-free, regenerative sensor                                                 | Detection clinical samples in serum | [94]  |
| Sandwich assay          | SPR      | CEA in plasma blood                                                      | bio-AuNPs                                                                       | LOD: 0.1 ng/mL               | [95]  |
| Sandwich assay          | Lateral flow | PDGF-BB and thrombin                                                    | Au-labeled aptamer probe                                                        | LOD (in nM): 1.0 (PDGF-BB) 1.5 (thrombin) | [96]  |
3.5. Reproducibility

Reproducibility is characterized by the accuracy and precision of the biosensor. It is essential to determine the degree of agreement between independent measurements under slightly different circumstances (or different users) for similar samples and concentrations (a mean value close to the true value). Meanwhile, for precision performance for identical circumstances (or same user), such as experiments on the same day, this can be considered repeatability [99]. Therefore, measuring each sample several times is essential to obtain the SD value in the standard curve, such as in Figure 10. Later, the accuracy and precision of the measurement can be determined. The illustrations of accuracy and precision are depicted in Figure 10.

The first inset illustration (1) represents the excellent accuracy and precision performance of a biosensor. The repetition of the measurements results in an accurate average value and low SD in the signal response. The second illustration (2) of low-accuracy and low-precision measurements leads to a valid average value but a high SD value. For the next illustration (3) for the low-accuracy and high-precision measurement, an inaccurate average value with a small SD is obtained in the standard curve. In this case of the biosensor performance, a calibration protocol is required before measuring the target. The last illustration (4) of the low-accuracy and low-precision measurement is an example of the biosensor’s poor repeatability or reproducibility performance. A high-accuracy and high-precision platform typically requires complicated, bulky, and non-portable instruments [10,100,101]. Still, another method to improve the accuracy and precision of the biosensor signal is advanced signal processing [102–106]. One aspect that is sometimes undervalued is the operator. The detection system should avoid operator interference as much as possible during the measurements. Therefore, robust, portable, and autonomous technology will benefit the reproducibility of the technology for border applications. The main analytical parameters of a biosensor that should be considered when developing a diagnostic device are summarized in Table 4.
Figure 10. Illustration of the accuracy and precision concepts in the biosensor repeatability and its representation in the standard curve.

Table 4. Analytical performance of a biosensor.

| Analytical Performance | Definition | Significant Factor | Improvement Strategy | Ref |
|------------------------|------------|--------------------|----------------------|-----|
| Sensitivity            | The signal response in every target concentration | SNR, slope | Downscaling sensing area, sandwich assays, and use of labels (e.g., nanoparticles) | [54–63] |
| Dynamic range          | The ratio between the maximum and minimum usable indication of the signal response | Linear range, saturation level | Surface area enhancement, 3D sensing structure, label tailoring, antibodies affinity, concentration | [35,76–80] |
| LOD                    | The value expressed by the smallest concentration that can be detected with a reasonable certainty and confidence level | Sensitivity, SD of reference | Sensitivity improvement, SNR enhancement, signal stability, instrument calibrations, signal processing | [10,54–63,98–104] |
| Specificity            | The sensing ability for capturing a specific target from the other substance | Biomarker preference, low-fouling binding | Blocking assays, advanced surface functionalization, dual-labeling, molecule affinity | [88–97] |
| Reproducibility        | The accuracy and precision of the measurements by different users using identical samples and concentrations | SD, instrumentation | Instrument configuration, signal processing, reduce the number of variables (e.g., operator interference) | [10,100–106] |

SNR: signal-to-noise ratio; 3D: three-dimensional; LOD: limit of detection; SD: standard deviation.

4. Biomarker Strategy

A biological marker or biomarker is an objective indicator or parameter of biological conditions to assess the normal biological processes, pathogenic processes, or pharmacologic responses to a therapeutic intervention [107,108]. Generally, the biomarkers are categorized as molecular biomarkers, cellular biomarkers, or imaging biomarkers. The molecular biomarkers commonly used in biosensors are proteins (such as IgG, IgM, antibodies, and proteins) and nucleic acid-based markers (such as RNA, DNA, or KRAS genes) [109–114]. The biomarker identification methods based on protein biomarkers and nucleic acids are called proteomics and transcriptomics, respectively [115]. Cellular-based biomarkers are very popular for cancer studies, such as detecting circulating tumor cells (CTCs) [116]. An imaging biomarker is a biological characteristic that is detectable on an
image, such as bone, tissue, and tumor cells in diagnostics using, for example, magnetic resonance image (MRI) or computed tomography (CT) scans [117–119]. In the biosensors field, the biomarker characteristics lead to biorecognition preferences, such as the immobilization method on the sensing area, to configure the detection strategy. In this section, biosensors for the detection of individual and multiple biomarkers are discussed.

4.1. Single Biomarker

Sequence-based biomarkers mainly dominate the single (or simplex)-biomarker strategy in biosensing detection due to their specific correlation with pathogen species or cells. The detection method in biosensing studies can either monitor DNA hybridization directly [23,34,120] or in combination with the DNA amplification method, such as by polymerase chain reactions (PCRs) and loop-mediated isothermal amplification (LAMP) [102,121]. Recently, potential biomarkers have been proposed using microRNA (miRNA) for several diseases, where miRNA consists of 19–23 nucleotide-long, noncoding, ribonucleic acid (RNA) molecules that are highly conserved across species [115]. In addition, the use of aptamers (synthetically processed single-stranded DNA or RNA) in the biosensing field has also gained much attention from researchers. The simplicity of the aptamer is that the required sequence can be customized to the specific purpose of the sensing method, either to direct detection of the complementary sequence or to utilize it as the biorecognition element for other biomolecules or toxins [122–125].

Although sequence-based biomarkers are widely available and suitable for multiple diseases, protein biomarkers are of paramount importance in clinical settings [126]. A well-known protein biomarker for the early screening of prostate cancer is the prostate-specific antigen (PSA). It is a typical glycoprotein produced by the prostate epithelium, which is secreted in the disruption of the ductal lumen and basement membrane disorder [127]. Several biosensor applications and methods have been reported utilizing PSA as a biomarker, such as in the nanowire FET-based devices [61,128], SERS platform [129], piezoelectric-based device [130], and SPR sensors [131,132].

Another acknowledged simplex biomarker in trophoblastic tumors is human chorionic gonadotropin (hCG) [133]. Recent studies found the hCG can be used as a biomarker for multiple diseases [134]. The hCG is a hormone present in healthy pregnant women. Nevertheless, in the patient with the trophoblastic tumor, the hCG secretion can increase up to 10,000 mIU/mL at 5 weeks, 1000 mIU/mL at 8 weeks, or detectable at 24 weeks after the evacuation of the mole, at which point is recommended for the patient to undergo chemotherapy [133,135]. Several studies reported the detection of hCG in biosensor platforms, such as in magnetometer-based biosensor [8,136–138], plasmonic-based biosensor [139,140], QCM platform [141], and FET-based device [142].

Cardiac troponins (cTn) that are part of the contractile apparatus proteins are specific to the heart, indicating they can be applied as biomarkers for cardiovascular diseases [143,144]. Various detection techniques and biosensor platforms have been proposed for the detection of cTn biomarkers, such as in piezoelectric platforms [145], plasmonic sensor platforms [146–151], and FET-based biosensors [152–155]. A resume of different biomarkers detected with the single biomarker strategy is presented in Table 5.

Table 5. List of biosensor studies utilizing single-biomarker detection strategy.

| Biomarker                  | Platform | Technical Remark                  | Performance     | Ref. |
|----------------------------|----------|-----------------------------------|-----------------|-----|
| 16S rRNA gene *S. aureus*  | EGFET sensor | Direct hybridization, Nanopatterned gold sensing | LOD 1 pM [23]   |     |
| ZIKV RNA                   | LSPR mediated fluorescence | Hybrid probes NP-qDots-MB | LOD 1.7-7.6 copies/mL [34] |     |
| DNA                       | FET      | PNA probe, RGO sensing            | LOD 100 fM [156] |     |
| IS6110 DNA of              | SPR      | Amplified- and labeled-DNA by PCR | LOD 63 pg/mL [102] |     |
4.2. Multiple Biomarkers

In the case of complex diseases in medical science, analyzing a single biomarker is not enough to evaluate the biological or pathogenic process. Therefore, assessing a panel of biomarkers is required to improve the specificity and accuracy of the diagnosis. For example, Duffy 2020[157] noted that PSA single biomarker detection is not enough to assess the prostate cancer state. Their study proposed additional biomarkers such as (Prostate Health Index) PHI, 4K score, and prostate cancer gene 3 (PCA3) as the panel biomarkers. The 4K score itself contains the total PSA, free PSA, intact PSA (a form of free PSA), and human kallikrein 2 (hK2)[157]. Another panel of biomarkers has been proposed for the evaluation of colorectal cancer (CRC). In clinical practice, the RAS gene family (KRAS and NRAS), BRAF, HER2, and microsatellite instability (MSI) have been reported as candidates for the panel of biomarkers[158]. A panel of biomarkers has also been pro-
posed for ovarian cancer assessment[159]. These findings in medical sciences can be a valuable milestone for the clinical application of biosensors to perform panel biomarkers simultaneously.

An example of the biosensing application of the multiplex strategy was reported by Katchman 2016[160] using the flexible display of organic-light emitting diodes (OLED). This study demonstrated an application for multiplex detection of the HPV16 proteins E2, E6, and E7, which are known as biomarkers for cervical, head, and neck cancer. A detection limit of 10 pg/mL was achieved[160]. Qureshi et al. proposed a sensing array using a capacitor-based biosensor for high-throughput measurements. Gold interdigitated electrodes (GID) are configured to detect a panel of inflammation and cardiovascular risk biomarkers, including C-reactive protein (CRP), TNFα, and IL6. A dynamic range performance was presented from 25 pg/ml to 25 ng/ml [161]. Another proof-of-concept of multiplex biosensor for the detection of cardiovascular diseases was reported by Shanmugam et al. The troponin markers group (cTnl and cTnT) were detected simultaneously in human serum, using Electrochemical Impedance Spectroscopy (EIS) and Mott-Schottky methods[162]. A GMR-based biosensor was reported to detect a panel of cardiovascular biomarkers: PAPP-A, PCSK9, and ST2. A LOD of 40 pg/mL for the ST2 antigen was achieved, which the dynamic range is going up four orders of magnitude[7]. A magnetoresistive-based biosensor was demonstrated for the detection of rare cell biomarkers: epithelial cell adhesion molecule (EpCAM), human epidermal growth factor receptor 2 (HER2), and epidermal growth factor receptor (EGFR) on individual cells. The accuracy was claimed to be 96%, compared to the 15% obtained with the Cell-Search method, while the throughput sample achieved was up to ~107 cells/min[163].

Multi-analyte detection to assess Salmonella infection was demonstrated by Ewald et al. using an optical biosensor reflectometer. Anti-Salmonella antibodies and CRP were detected simultaneously, with a dynamic range from 5.74 to 122.52 mg/L[164]. Fountoglou et al. reported an LSPR sensor using colorimetric-induced gold nanoparticles as a multi-allele DNA biosensor to detect thrombosis biomarkers. DNA amplification using PCR was performed before the colorimetric detection, and 100% accuracy was reported with this methodology[165]. A nanoparticle-based LSPR sensor was also proposed by Lee et al. to detect a panel of cancer biomarkers, such as α-fetoprotein (AFP), carcinoembryonic antigen (CEA), and PSA. The reported detection limit performances from the analytes in patient-mimicked sera were 91 fM for AFP, 94 fM for CEA, and 10 fM for PSA[166]. Another colorimetric sensor utilizing AuNP and aptamer probes was reported to detect multiple proteins on the exosome surface[167]. A photonic crystal biosensor achieved the multiplex detection of exosome vesicles (EV) released by macrophages. A panel of seven biomarkers was used, including CD9, CD63, CD68, CD80, CD81, CD86, and I-A/I-E. A dynamic range of EVs detection from 2×1011 particles/mL to 2×109 particles/mL was achieved[168].

Multiplex detection of sepsis biomarkers was reported in a modular electrochemical biosensor. The sensing layer used a nanoporous nylon membrane integrated onto the microelectrode. The LOD was of 0.1 ng/mL for procalcitonin (PCT), and 1 µg/mL for lipoteichoic acid (LTA) and lipopolysaccharide (LPS) and LTA, while the dynamic ranges achieved were from 0.1 ng/mL to 10 µg/mL for PCT and from 1 µg/mL to 1000 µg/mL for LPS and LTA biomarkers[169]. Multiplex measurement of physiological body fluid biomarkers was demonstrated in a wearable biosensor by Yokus et al. This research reported a wireless, real-time, and high-throughput measurement of glucose, lactate, pH, and temperature. The wearable sensors demonstrated sensitivity performance of 26.31 µA/mM-cm2 for glucose, 1.49 µA/mM-cm2 for lactate, 54 mV/pH for pH, and 0.002 °C-1 for temperature[170]. Some biosensors studies for multiplex biomarkers assessment are summarized in Table 6.
Table 6. Summary of biosensor studies for multiplex detection of biomarkers.

| Disease    | Diagnostic | Biomarkers                  | Platform                        | Technical Remark                                                   | Performance                              | Ref. |
|------------|------------|-----------------------------|---------------------------------|-------------------------------------------------------------------|------------------------------------------|------|
| Cardiovascular |            | CRP; TNFα; IL6              | GID capacitor arrays            | Co-immobilized chip with equimolar mixtures of antibodies        | LDR: 0.025–25 ng/ml                     | [161]|
| Cardiovascular |            | cTnI; cTnT                  | Disposable electrochemical sensor | ZnO vertical rod sensing                                         | LDR: 0.1–1×10^8 pg/mL; LOD (in pg/mL): 1 (cTnI); 0.1 (cTnT) | [162]|
| Cardiovascular |            | PAPP-A; PCSK9; ST2          | GMR                             | 8×8 sensor array                                                  | LDR: 0.04–400 ng/mL                     | [7]  |
| Cancer      |            | EpCAM; HER2; EGFR; CTC      | Micro-Hall detector             | 2×4 sensor array                                                  | LDR: 10–10^6 cells                      | [163]|
| Cancer      |            | CD9; CD63; CD68; I-A/i-E.   | Photonic crystal                | 12 channels                                                      | LDR: 2×10^11–2×10^9 particles/mL        | [168]|
| Cancer      |            | AFP; CEA; PSA CD63;         | LSPR                            | Biomarkers detected in the human serum sample                    | LOD (in fM): 91 (AFP), 94 (CEA) and 10 (PSA) | [166]|
| Cancer      |            | EpCAM; PDGF; PSMA; PTK7     | LSPR colorimetric               | Aptamer probes on AuNP                                            | Profiling exosomal proteins in minutes  | [167]|
| Salmonella  | infection  | CRP; anti-Salmonella (AS)   | 1-lambda-reflectometry          | Label-free, portable platform                                    | LDR (in ng/mL): 5.74–122.52 (AS); 1.26–29.56 (CRP) | [164]|
|            |            | DNA sequences from H1299R;  |                                | PCA amplification, dipstick sensing                               | Accuracy: 100% for 15 blind clinical samples | [165]|
|            |            | A1298C; V34L; 4G/5G polymorphism |                                |                                                                   |                                          |      |
| Thrombosis  |            | PCT; LTA; LPS               | Nanochannel electrochemical     | Detection in the whole blood sample                              | LDR (in µg/mL): 0.0001–10 (PCT); 1–1000 (LPS and LTA) | [169]|
| Sepsis      |            | Glucose, lactate, pH, and temperature from sweat | Electrochemical | Wearable and wireless device                                      | Sensitivity (in µA/mM·cm²) 26.31 (glucose); 1.49 (lactate); 54 mV/pH (pH); 0.002 °C⁻¹ | [170]|

CRP: C-reactive protein; TNFα: tumor necrosis factor-α; IL6: interleukin 6; GID: gold interdigitated electrodes; cTnI: cardiac troponin I; cTnT: cardiac troponin T; ZNO: zinc oxide; LDR: linear dynamic range; PAPP-A: pregnancy-associated plasma protein A; PCSK9: proprotein convertase subtilisin/kexin type 9; ST2: suppression of tumorigenicity 2; GMR: giant magnetoresistive; EpCAM: epithelial cell adhesion molecule; HER2: human epidermal growth factor receptor 2; EGFR: estimated glomerular filtration rate; CTC: circulated tumor cell; AFP: alpha-fetoprotein; CEA: carcinoembryonic antigen; PSA: prostate-specific antigen; PDGF: platelet-derived growth factor; PSMA: prostate-specific membrane antigen; PTK7: protein tyrosine kinase 7; AuNP: gold nanoparticles; DNA: deoxyribonucleic acid; PCR: polymerase chain reaction; PCT: procalcitonin; LTA: lipoteichoic acid; LPS: lipopolysaccharide.

5. Biomarker Clinical Cut-Off Value

The cutoff value is an important element to be considered in the development of a diagnostic tool. In principle, the clinical cutoff is a threshold value of the biomarker concentration that distinguishes the clinical condition from the healthy patient. Therefore, the precision and accuracy of quantifying this type of value play a significant role in biosensor research. The typical illustration of the biomarker clinical cutoff value is depicted in Figure
The use of cutoff value can help to provide the stage or diagnosis of the disease. However, a fraction of the population fits in the false negative or false positive region, and to overcome this issue, the use of a panel of biomarkers can enhance the analytical performance of the biosensor. Therefore, a comprehensive analysis of the biomarkers in the context of the clinical diagnostic is critical for the development of a biosensing method able to improve clinical decisions.

For example, in the case of prostate cancer, a study suggested a clinical cutoff value for the PSA biomarker of 4 ng/ml in blood. In patients, the PSA concentration can reach up to 104 ng/mL, while it is usually below 0.1 ng/mL in healthy people. As such, patients with a PSA concentration higher than 4 ng/mL are required to perform a biopsy analysis. Another study recommended that the range of 2-4 ng/mL be considered the grey area, although it will decrease the TP rate [127]. Although only one biomarker is detected, if the biosensor is sensitive enough (LoD below 4 ng/mL), it can increase the chance for diagnostic of prostate cancer at an earlier stage of the disease, improving the chances of a positive outcome for the patient. Another option could be the inclusion of multiple biomarkers in the initial analysis as suggested by Duffy 2020[157], which proposed the use of additional biomarkers such as prostate cancer gene 3 (PCA3), and the Prostate Health Index (PHI), 4K score. These biomarkers must be thoughtfully studied to determine a cutoff value and the entire panel for clinical analysis.

For ovarian cancer, the biomarker CA-125 is a peptide epitope with a molecular weight of 3–5 MDa mucin, MUC16. This biomarker is present in healthy individuals in the range of 0-35 U/mL, while it is overexpressed in ovarian cancer patients. As such, some studies suggest a clinical cutoff level around 30-35 U/mL[171,172]. In addition, CRP is a well-known biomarker for cardiovascular disease and general inflammation, including bacterial and viral infections, such as Salmonella and COVID-19[161,164,173,174]. The WHO reported this biomarker also for the nutrition state assessment. The CRP clinical cutoff concentration in blood is around 10 ng/mL. This concentration rapidly increases several hours after the inflammatory disease onset and can reach up to 350–400 ng/mL in
2 days[175]. Consequently, for this biomarker, a biosensor platform with a dynamic range from less than 10 ng/mL to 400 ng/mL, would be necessary for inflammatory disease assessment. Nevertheless, CRP is a general inflammatory marker; a biosensor for single biomarker CRP detection cannot distinguish the origin of the disease. Therefore this scenario will be challenging for a clinical decision.

To obtain a full picture of complex diseases, the cutoff value for multiple biomarkers must be determined to achieve precision and accuracy in the clinical diagnostic tools. One such example can be seen for stratification ischemic stroke patients for tPA (tissue Plasminogen Activator) therapy. A group of researchers from clinical neurosciences proposed a panel of biomarkers that reflect patients undergoing ischemic stroke[176]. It was proposed that for levels of cellular-Fibronectin above 3.6 µg/mL[177,178], Matrix Metalloproteinase-9 above 140 ng/mL[177,178], Platelet-derived Growth Factor-CC above 175 ng/mL[179], Neuroserpin below 70 ng/mL[180], and the calcium-binding protein S100β above 230 ng/mL[177,181], the patients can be at a higher risk on developing a hemorrhagic transformation (HT) after tPA treatment. Although the detection of each biomarker is not specific for HT risk, the combination of at least two biomarkers (c-Fn and MMP9) can provide a detection specificity of 87%, increasing the chances of a safe diagnosis and improving clinical decision[178].

It is also essential to notice that the same biomarkers could be used to diagnose other diseases as long as the cutoff values and biomarker combination considered are adjusted to the clinical condition at study. Wood et al.[182] also report optimizing multiples parameters in a particle-based platform to achieve detection of 3 different analytes (multiplex assay) with different ranges of relevant concentrations. Parameters such as particle concentration[82,182], capture antibody affinity[82,182], measuring conditions[82], and sample preparation (e.g., dilutions) [82,182] affect the outcome of the developed assay and can be tuned to fit the necessary cutoff value for the clinical application being targeted.

As such, biosensor platform application to the clinical environment depends on adjusting its dynamic range to cover the clinical cutoff concentration of the biomarkers and the ability to provide multiple biomarker detection to provide specific diagnostics.

6. Conclusions

In conclusion, to initiate multidisciplinary research in the biosensors field, several steps should be considered systematically, which are illustrated in Figure 12.

![Figure 12](image)

Figure 12. The main steps to be considered in the development of a diagnostic tool from biomarker selection, analysis of the biophysical properties, clinical cut-off overview, biosensor preferences, and detection strategy to data analysis.

The sequential concept from biomarker preferences allows the assay developer to choose the platform and detection methods with broader options.

Moreover, the future outlook and challenges of the development biosensor platform will be for high-throughput samples that will be an essential feature for panel biomarker study. Nevertheless, if the biomarkers in the panel contain different structures or physical properties, a different dynamic range for the biosensor platform could be required, such as if the panel biomarkers consist of nucleic acid sequences, proteins, and cells. Another case is if the panel biomarkers consist of sequences of nucleic acid with different lengths.
Therefore, each marker may require different clinical cut-off values, detection methods, and dynamic ranges of biosensor performance. In these cases, each channel of the high-throughput biosensors will be excellent if it can be adjustable for the different dynamic range performances to comply with the biomarker properties. In addition, in the high-throughput biosensors for panel biomarker analysis, advanced signal processing and integration with a machine learning algorithm to improve the detection accuracy will be a remarkable trend in the future.

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