



## Review

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# Blood–brain barrier permeability assessment: from biomimetic models to multimodal imaging

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**Abstract:** The blood–brain barrier (BBB) is a vital physiological structure that maintains the microenvironmental homeostasis in the central nervous system (CNS). Imbalances in its permeability play a key role in various neurological disorders, including stroke, neurodegenerative diseases, and brain tumors. The development of precise techniques for assessing BBB permeability is therefore paramount for elucidating the mechanisms of neurological diseases, overcoming drug development challenges, and achieving precise diagnosis and treatment of CNS disorders. This review systematically summarizes the latest advances in the assessment of BBB permeability. Regarding in vitro models, platforms have evolved from the traditional transwell system to microfluidic chips incorporating fluid shear forces and subsequently to highly biomimetic brain organoids, with continuous improvements in the ability to simulate the neurovascular unit (NVU) microenvironment. For in vivo assessment, we detail the principles and applications of imaging techniques, including dynamic contrast-enhanced magnetic resonance imaging (DCE-MRI), positron emission tomography (PET), near-infrared II (NIR-II) fluorescence imaging (FI), and two-photon microscopy (TPM), highlighting their complementary strengths in macroscopic quantification, molecular targeting, and microscopic dynamic observation. The integration of multi-modal technologies and precise quantitative assessment is a prominent trend. Future investigations will focus on artificial intelligence (AI)-driven personalized permeability assessment, the development of novel intelligent probes, and the dynamic real-time monitoring of the BBB, thereby providing powerful methodological support for neurological disease research.

**Key words:** Blood–brain barrier (BBB); Permeability assessment; In vitro BBB model; In vivo imaging; Central nervous system (CNS)

## 1 Introduction

The blood–brain barrier (BBB) is a highly selective, semipermeable structure that separates the blood circulatory system and the central nervous system (CNS). By regulating the exchange of substances, it helps maintain the homeostasis of the cerebral microenvironment (He et al., 2018; Wu et al., 2023; Liu et al., 2025). The BBB primarily comprises endothelial cells,

astrocytes, pericytes, and the basement membrane (Fig. 1). Brain microvascular endothelial cells (BMECs) constitute the core component of the BBB. They form a continuous physical barrier through intercellular connections, mainly tight junctions and adherens junctions. Tight junctions consist of transmembrane proteins that are linked to the actin cytoskeleton via cytoplasmic scaffolding proteins. They are responsible for maintaining selective permeability and controlling tissue homeostasis (Anderson and van Itallie, 2009; Luissint et al., 2012; Dong, 2018; Ahlawat et al., 2020). Adherens junctions, mainly constituted by vascular endothelial (VE)-cadherin and its intracellular binding partners such as  $\beta$ -catenin, are instrumental in regulating intercellular adhesion strength and barrier integrity (Vorbodt and Dobrogowska, 2003; Turowski and Kenny, 2015; Stamatovic et al., 2016; Malinova

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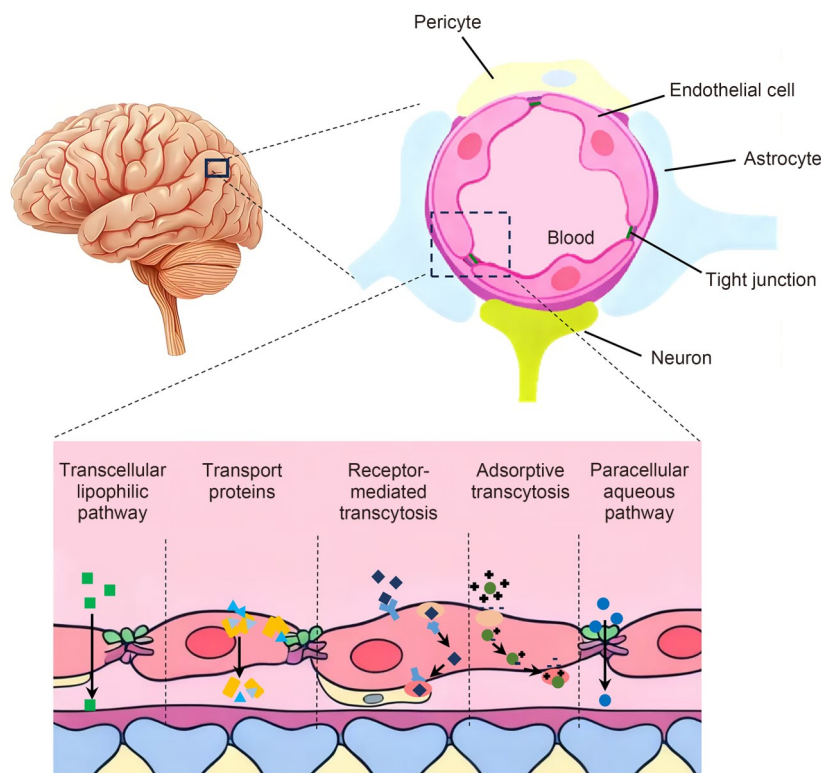
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and Huveneers, 2018). Studies have shown that when these junctional proteins are downregulated or dysfunctional, BBB permeability becomes abnormal, which can contribute to various neurological disorders (Krol et al., 2013). Astrocytes envelop the cerebral microvasculature with their endfeet. By secreting diverse signaling molecules, they induce tight junction formation among endothelial cells and maintain barrier properties (Zhao et al., 2015; Schurhoff and Toborek, 2023). Pericytes are embedded in the capillary basement membrane and envelop the endothelial cells. They can modulate capillary blood flow by contracting, and they also participate in the development, homeostatic maintenance, and repair of barrier function (Bell et al., 2010; Kisler et al., 2017; Kadry et al., 2020). The basement membrane, a dense extracellular matrix network, provides structural support for the BBB's cellular components and mediates cell–matrix interactions to regulate its permeability (Bayir et al., 2019).

As a physiological interface, the BBB allows the passage of water and certain small molecules while

effectively excluding the majority of large-molecular-weight substances. This selective impermeability protects the CNS from toxins, pathogens, and metabolic waste present in the systemic circulation, while also restricting the nonspecific permeation of neurotransmitters and hormones. As a result, it provides a highly stable internal environment for neuronal activity (Bhattacharya et al., 2020; Segarra et al., 2021). For instance, specific ion channels and transport proteins at the BBB meticulously regulate ion flux between the blood and brain tissue (Harilal et al., 2020; Yang et al., 2020). This precise regulation maintains ionic balance and pH homeostasis between the blood and cerebrospinal fluid, ensuring an optimal cerebral microenvironment. The BBB also actively delivers essential nutrients to the brain via specialized transport proteins, meeting its high metabolic demands. For example, the glucose transporter type 1 (GLUT1) efficiently moves glucose—the brain's primary energy source—from the blood into the CNS. Similarly, the L-type amino acid transporter 1 (LAT1) mediates the uptake of large neutral amino acids (de Vivo et al.,



**Fig. 1** Schematic of the blood–brain barrier (BBB) structure and key transport pathways. The BBB comprises endothelial cells interconnected with tight junctions, and is supported by pericytes and astrocytic endfeet. The schematic highlights the principal pathways of substance traversal: the transcellular route (including lipophilic diffusion and transcytosis) and the restricted paracellular aqueous pathway.

1991). In terms of active defense, the BBB expresses various efflux transporters, such as P-glycoprotein (P-gp) and breast cancer resistance protein (BCRP). These pump xenobiotics and metabolic byproducts back into the bloodstream, preventing harmful substances from accumulating in the brain (Balzer et al., 2022). Recent studies have further elucidated the BBB's role in the glymphatic system, a paravascular pathway responsible for cerebrospinal and interstitial fluid circulation. This process relies on aquaporin-4 (AQP4) water channels located on astrocytic endfeet, which promote the convective flow of cerebrospinal fluid (CSF) through the brain parenchyma and facilitate interstitial solute clearance. This mechanism is recognized as essential for maintaining a clean internal cerebral environment (Iliff et al., 2012).

The selective permeability of the BBB is not only essential for CNS homeostasis but also plays a key role in the pathogenesis and treatment of various CNS disorders. Many studies have demonstrated that BBB dysfunction is a central pathological event in several neurological diseases. For instance, amyloid- $\beta$  (A $\beta$ ) is a primary pathological protein in Alzheimer's disease (AD). The production and deposition of A $\beta$  strongly influence the progression and prognosis of AD, and this process is closely related to the BBB's diminished capacity to clear A $\beta$ . Furthermore, A $\beta$  deposition can damage the structure of the BBB, establishing a vicious cycle of "A $\beta$  deposition–BBB impairment–attenuated clearance ability–further A $\beta$  accumulation," thereby accelerating disease progression (Wang et al., 2021). Consequently, in addition to their role in the pathogenesis of CNS diseases, changes in BBB permeability can serve as diagnostic indicators and prognostic biomarkers for associated disorders. BBB permeability is also a factor in drug development, as it affects the therapeutic efficacy of interventions for CNS diseases. Although an intact BBB effectively prevents most blood-borne substances from entering the brain, it also blocks most therapeutic agents. It is estimated that over 98% of small-molecule drugs and nearly all large-molecule biologics fail to cross the BBB and reach brain tissue at clinically relevant concentrations, presenting a major bottleneck in the treatment of diseases such as stroke, neurodegenerative diseases, brain tumors, meningitis, and multiple sclerosis (Banks, 2016; Pandit et al., 2020). Hence, advanced techniques for assessing BBB permeability are essential

for the early screening of candidate compounds with favorable brain penetration potential, thereby enhancing the success rate and efficiency of CNS-targeted drug development programs.

Developing reliable methods to assess BBB permeability is essential for understanding CNS pathologies and discovering new therapies (Ahlawat et al., 2020; Zheng et al., 2024). Major breakthroughs in detection technologies have been made in recent years, ranging from in vitro models to in vivo imaging. This review summarizes recent progress, compares the advantages and limitations of various technologies, and offers methodological guidance and conceptual frameworks for research on BBB structure, CNS disorders, and therapeutic strategies.

## 2 In vitro assessments of BBB permeability

### 2.1 In vitro BBB model

The construction of in vitro BBB models that accurately simulate the physiological parameters and molecular characteristics of the neurovascular unit (NVU) provides a key tool for studying the pathological mechanisms and screening potential drugs for CNS diseases. Driven by microfluidics and organoid technologies, the field of in vitro BBB models has seen major advances in recent years. This has led to a transition from traditional monolayer, two-dimensional (2D) co-culture systems toward three-dimensional (3D) biomimetic systems that more closely simulate the in vivo microenvironment (Sivandzade and Cucullo, 2018). Presently, two main types of in vitro models are used to assess BBB permeability: static models based on transwell chambers and dynamic models (Table 1).

#### 2.1.1 Static BBB model

The static model is the most widely used in vitro BBB model. It is constructed by incubating BMECs on the porous membrane of a transwell insert. To assess barrier permeability, the apparent permeability coefficient ( $P_{app}$ ) is calculated by measuring the translocation of test substances from the apical (luminal) compartment to the basolateral compartment. Based on the cellular composition, static BBB models are further classified into two principal categories: monoculture and co-culture systems (Fig. 2) (Bagchi et al., 2019). The simplest monoculture model initially used primary

Table 1 Comparison of in vitro BBB models

| In vitro model                | Cell composition                                                                                               | Physiological functions and characteristics                                                                                                             | Physiological fidelity | Primary application scenarios                                                                       |
|-------------------------------|----------------------------------------------------------------------------------------------------------------|---------------------------------------------------------------------------------------------------------------------------------------------------------|------------------------|-----------------------------------------------------------------------------------------------------|
| Static model                  |                                                                                                                |                                                                                                                                                         |                        |                                                                                                     |
| Monoculture                   | Endothelial cells                                                                                              | Reflects basic barrier properties; low TEER                                                                                                             | Low                    | Study of signaling pathways, transporter kinetics, binding affinity, and high-throughput screenings |
| Co-culture                    | Endothelial cells, astrocytes, and pericytes                                                                   | Enables partial biochemical signal responses through fluid exchange; improved TEER                                                                      | Medium                 | Standard permeability assays and mechanistic studies                                                |
| Dynamic model                 |                                                                                                                |                                                                                                                                                         |                        |                                                                                                     |
| Cone-plate BBB apparatus      | Endothelial cells (expandable to co-culture)                                                                   | Applies controlled shear stress; TEER can be modulated                                                                                                  | Medium                 | Hemodynamic study                                                                                   |
| Hollow fiber bioreactors      | Endothelial cells and astrocytes/pericytes (3D co-culture)                                                     | Dynamic perfusion provides shear stress, supports sustained high barrier function, and allows for long-term responsiveness to circulating factors/drugs | High                   | Validation and optimization of lead compounds in drug discovery                                     |
| Microfluidics (BBB-on-a-Chip) | Endothelial cells, astrocytes, pericytes, and other NVU cells (controllable 3D co-culture)                     | Highly controllable system capable of integrating shear force, chemical gradients, and real-time monitoring                                             | High                   | High spatiotemporal resolution research on mechanisms, drug evaluation, and disease modeling        |
| Brain organoids               | Endothelial cells, pericytes, astrocytes, neurons, and microglia (self-organized or engineered a complete NVU) | Designed to replicate complex cell–cell interactions and pathophysiological responses                                                                   | High                   | Research on complex brain diseases and exploration of individualized treatment                      |

BBB: blood–brain barrier; TEER: transendothelial electrical resistance; 3D: three-dimensional; NVU: neurovascular unit.

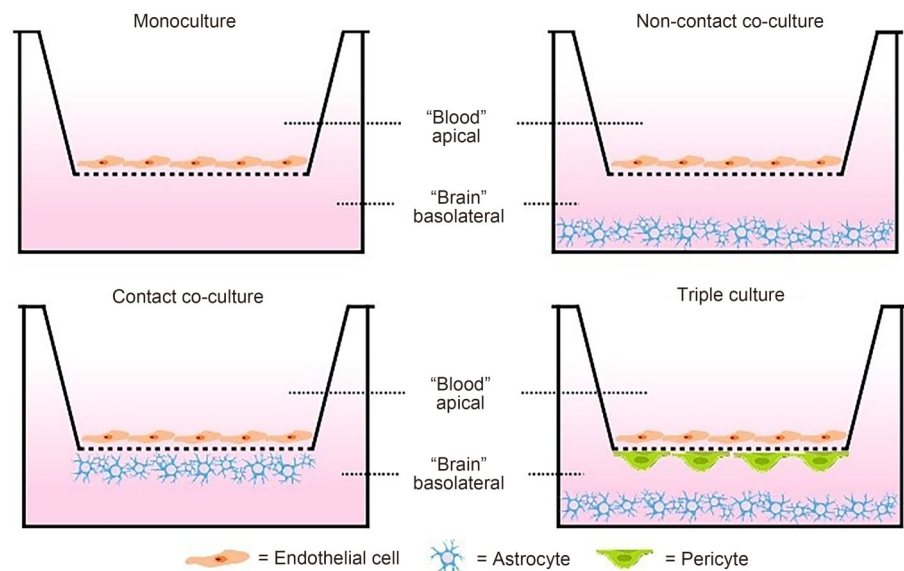


Fig. 2 Schematic of four common static in vitro models of the blood–brain barrier (BBB). These transwell-based models vary in cellular complexity. Monoculture consists of only brain microvascular endothelial cells (BMECs). Co-culture models consist of BMECs and astrocytes. In the non-contact co-culture model, BMECs are seeded on the transwell membrane while astrocytes are cultured in the basolateral compartment, allowing paracrine signaling. In the contact co-culture model, BMECs are grown in direct physical contact with an astrocyte layer. In the triple culture model, BMECs are co-cultured with both astrocytes and pericytes, thereby best recapitulating the native neurovascular unit. The apical chamber represents the “blood” side, and the basolateral chamber represents the “brain” side.

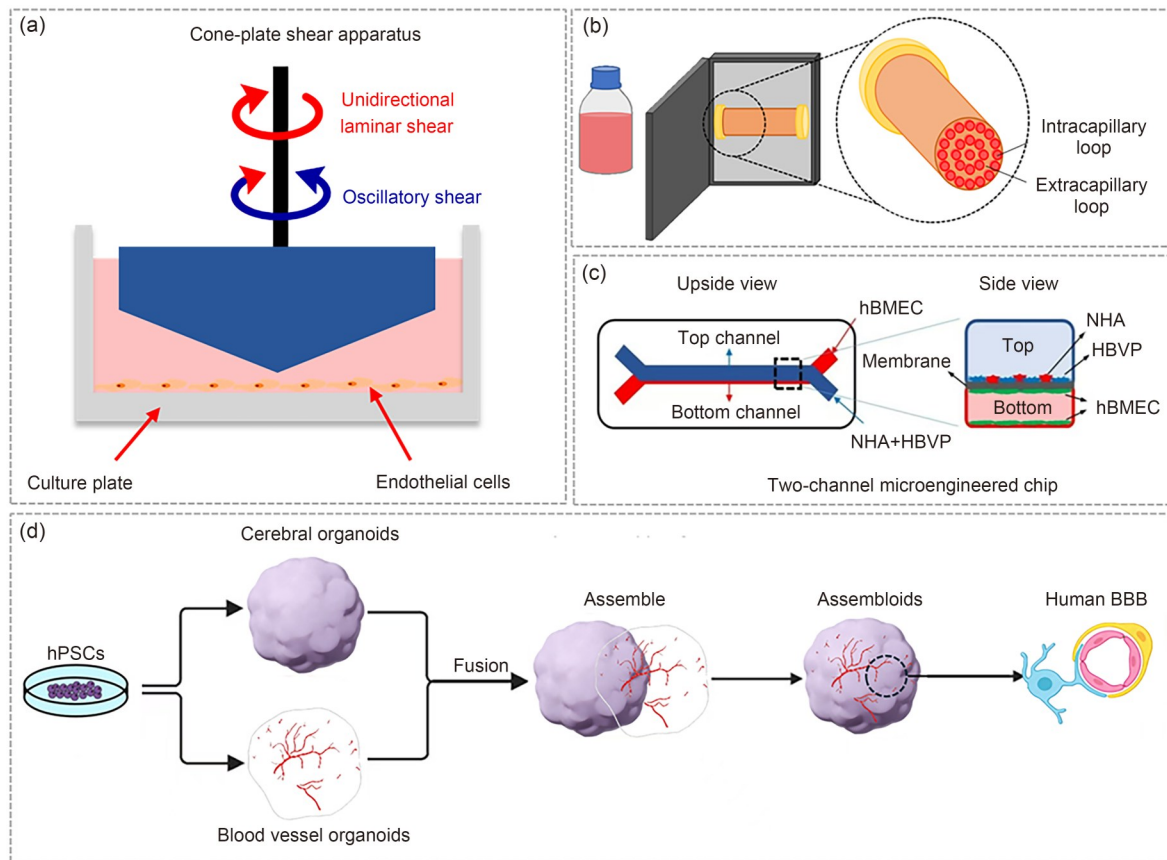
BMECs isolated from rats, cows, or pigs. A key advantage is that these primary cells retain essential BBB characteristics, such as well-defined tight junctions and the polarized expression of various transporters (Cecchelli et al., 2007). To circumvent the limitations of primary cells—namely, low yield and susceptibility to contamination—immortalized cell lines (e.g., human cortical microvessel endothelial cell/D3 [hCMEC/D3]) have been widely adopted (Muruganandam et al., 1997; Weksler et al., 2013). However, these cell lines often exhibit an incomplete barrier phenotype, frequently necessitating functional enhancement with compounds such as cyclic adenosine monophosphate (cAMP) or glucocorticoids to improve barrier properties (Hurst and Fritz, 1996; Franke et al., 1999; Naik and Cucullo, 2012; Daniels et al., 2013). The monoculture model is simple to operate and well suited for high-throughput drug screening and basic mechanistic studies. Its main limitations are the lack of synergistic interactions with other NVU cells and the inability to simulate hemodynamic shear stress. Consequently, it can not fully mimic the complex functions of the *in vivo* BBB (Mantecón-Oria et al., 2022b). Co-culture models address the shortcomings of monoculture systems. Co-culturing BMECs with astrocytes promotes tight junction maturation and reduces pinocytosis activity (Rubin et al., 1991; Dohgu et al., 2005). Further introduction of pericytes to form a triple-culture model enables the secretion of factors such as transforming growth factor- $\beta$  (TGF- $\beta$ ), which synergistically increase transendothelial electrical resistance (TEER), lower passive permeability, and enhance efflux transporter activity. It is currently regarded as the static model that most closely mimics the *in vivo* BBB (Nakagawa et al., 2009).

### 2.1.2 Dynamic BBB model

The dynamic model can better mimic the mechanical microenvironment of the BBB *in vivo* by introducing fluid shear stress (Fig. 3). Studies have shown that applying shear stress upregulates the expression of tight junction proteins, such as zonula occludens protein 1 (ZO-1), and thus enhances barrier integrity (Bagchi et al., 2019). Representative dynamic BBB platforms include the cone-plate BBB apparatus (Naik and Cucullo, 2012), hollow fiber bioreactors (Stanness et al., 1997; Janigro et al., 1999; Cucullo et al., 2008; Nogueira et al., 2021), microfluidics (Jiang et al.,

2019; Oddo et al., 2019; Wang et al., 2020; Hajal et al., 2022), and organoids (Bergmann et al., 2018; Logan et al., 2019; Dao et al., 2024). The cone-plate BBB apparatus uses a rotating cone to impose a defined shear stress on endothelial cells, facilitating the study of their mechanoresponse. However, its utility in modeling the BBB is limited by its inherent inability to incorporate the multicellular crosstalk essential for a fully functional NVU. In contrast, hollow fiber bioreactors use a 3D tubular design. BMECs are seeded on the luminal surface of semipermeable hollow fibers. Culture medium is perfused through the fiber lumen to generate physiological shear stress, while astrocytes or pericytes can be co-cultured within the extracapillary space (Cucullo et al., 2011). This 3D tubular architecture extends the viable culture period of the model (from days to months) and induces cells to form *in vivo*-like morphologies and more complex connections (Neuhaus et al., 2006).

In recent years, dynamic models have become more complex and physiologically relevant to better reflect human physiology and improve predictive value (Banerjee et al., 2016; Mantecón-Oria et al., 2020). For instance, polycaprolactone/graphene (PCL/G) composite materials have been explored to promote astrocyte differentiation (Mantecón-Oria et al., 2022a), and transparent polyvinylidene fluoride (PVDF) hollow fibers now enable real-time, non-invasive cell observation (Mantecón-Oria et al., 2025). However, hollow fiber models still face challenges in mimicking *in vivo* tissue complexity and fluid dynamics, making it difficult to accurately reproduce physiological shear forces and pulsatile flow fields. In contrast, microfluidic technology offers high integration, controlled fluid environment, and 3D architectures, making it a research focus (Mármol et al., 2023; Ceccarelli et al., 2024; Sun and Song, 2024; Sonninen et al., 2025). A common design incorporates parallel channels for the co-culture of endothelial cells, pericytes, and astrocytes to investigate the trans-barrier transport mechanisms of nanoparticles or antibodies (e.g., transferrin receptor targeting) (Wevers et al., 2018). More advanced systems enable the fabrication of 3D tubular vascular structures and have been employed to quantitatively assess the permeability of peptides such as apolipoprotein E (ApoE) and to determine drug permeability by liquid chromatography-tandem mass spectrometry (LC-MS/MS), with data consistent with *in vivo* studies.



**Fig. 3** Dynamic blood–brain barrier (BBB) model. (a) Cone-plate BBB apparatus. (b) Schematics of a hollow fiber bioreactor system and a close-up of the hollow fiber module, comprising an outer sheath and internal fibers. Cells can adhere to the fiber surface, enabling mass transfer through the semipermeable membrane. Reprinted from Nogueira et al. (2021), Copyright (2021), with permission from MDPI. (c) A two-channel microengineered chip had human brain microvascular endothelial cells (hBMECs, green) on all surfaces of the bottom channel, and normal human astrocytes (NHAs, red) and human brain vascular pericytes (HBVPs, blue) on the surface of the top channel. Reprinted from Yang et al. (2024), Copyright (2024), with permission from MDPI. (d) Schematic protocol for generating three-dimensional (3D) human BBB assembloids from cerebral organoids and blood vessel organoids. hPSCs: human pluripotent stem cells.

This demonstrates their potential for substituting animal experiments during early-stage drug development (Chung et al., 2020; Yang et al., 2024). However, microfluidic models have yet to fully replicate the multicellular interactions, dynamic blood flow, and long-term stability of the native BBB due to limited structural complexity and physiological realism.

Organoids are another rapidly developing type of BBB model. They are widely applied in drug delivery and disease modeling due to their highly biomimetic characteristics. For instance, Nzou et al. (2018) developed a 3D human cortical spheroid BBB model with six major brain cell types (endothelial cells, pericytes, astrocytes, etc.), showing tight junctions and charge selectivity. This model is useful for neurotoxicity screening and hypoxic disease modeling, providing a more

human-relevant in vitro platform for drug development. To improve throughput and reproducibility, Simonneau et al. (2021) engineered a high-throughput BBB organoid platform based on a hydrogel microwell array. Co-culturing the three core cell types generates thousands of uniform organoids in a single experiment, with semi-automated imaging. Notably, this study first applied clustered regularly interspaced short palindromic repeats (CRISPR)/CRISPR-associated protein 9 (Cas9) system on this platform to reveal the molecular mechanisms governing the BBB penetration of specific antibodies. This approach provides an efficient tool for the large-scale screening of CNS-targeting antibodies and for analyzing their transport mechanisms.

Although 3D models better reflect the human BBB, challenges remain: lack of standardized protocols,

limited maturity, and poor long-term stability of the barriers. Overcoming these obstacles will likely require a synergistic strategy that combines the controllability of microfluidics with the biological authenticity of organoids, paving the way for advances in CNS drug development and disease modeling.

## 2.2 Techniques for evaluating the permeability of in vitro BBB models

In studies of in vitro BBB models, accurate assessment of the barrier permeability is essential. Currently, the two most common and accepted methods for functional validation are  $P_{app}$  and the measurement of TEER (Fig. 4).

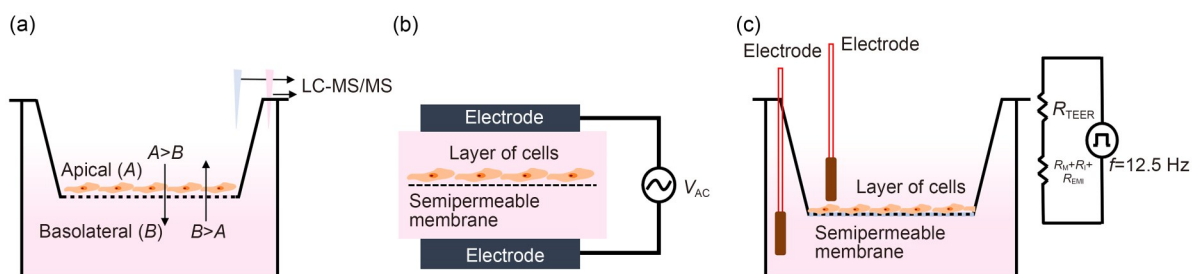
### 2.2.1 $P_{app}$ determination of in vitro BBB models

The principle of  $P_{app}$  determination involves quantitatively measuring the transport rate of a specific substance across a cellular monolayer under controlled conditions to assess barrier function. Its advantages include operational simplicity, low cost, and ease of implementation for high-throughput screening. It also allows simultaneous use of tracer molecules with different molecular weights to evaluate the BBB's size-dependent selectivity. For instance, Ito et al. (2019) co-cultured human BMECs with astrocytes and pericytes in a transwell system to build a functional in vitro BBB model. Using unidirectional transport assays, they found that  $P_{app}$  values for highly permeable drugs were substantially higher than those for low-permeability markers, spanning a 287-fold range. Compared to monoculture, the triple-culture model exhibited a tighter barrier and better function of efflux transporters such as P-gp. This model provides a reliable tool for assessing the brain penetration in

CNS drug discovery. Appelt-Menzel et al. (2017) determined  $P_{app}$  values to characterize the barrier properties of a stem cell-derived BBB model. Hydrophilic markers of different molecular weights revealed a molecular-weight-dependent permeation profile:  $P_{app}$  values decreased from approximately 1.5  $\mu\text{m}/\text{min}$  for small molecules to 0.0030  $\mu\text{m}/\text{min}$  for 40 kDa dextran. Transcellular permeability and active transport function of reference drugs were evaluated with  $P_{app}$  values normalized to that of diazepam. The quadruple-culture model had the tightest barrier and lower permeability for large molecules than monocultures. Adding verapamil (a P-gp inhibitor) increased the  $P_{app}$  value of its substrate rhodamine 123, confirming functional, correctly polarized efflux transporters. These systematic  $P_{app}$  determinations demonstrate that the quadruple-co-culture BBB model exhibits excellent barrier integrity and holds potential for predicting drug permeability.

### 2.2.2 TEER measurement of in vitro BBB models

TEER measurement is a key method for assessing the integrity and permeability of cellular monolayers (Deli et al., 2005; Benson et al., 2013). It quantifies electrical resistance across a cell layer, offering high sensitivity and real-time monitoring. It can detect early-stage, subtle alterations in barrier function induced by pharmaceuticals, cytokines, or pathological conditions. Furthermore, TEER enables longitudinal, dynamic tracking of barrier function in the same samples during drug screening or disease modeling. It directly reflects the barrier's physiological functional state rather than merely reporting protein expression levels. TEER measurement techniques are broadly categorized into two approaches: the Ohm's Law



**Fig. 4** Permeability assessment of the in vitro blood–brain barrier (BBB) model. (a) Measurement of the apparent permeability coefficient ( $P_{app}$ ). (b) Measurement of transendothelial electrical resistance (TEER) using impedance spectroscopy. (c) Measurement of TEER using Ohm's Law. The total electrical resistance includes ohmic resistance of the cell layer ( $R_{TEER}$ ), cell culture medium ( $R_M$ ), semipermeable membrane insert ( $R_s$ ), and electrode medium interface ( $R_{EMI}$ ).  $V_{AC}$ : alternating current voltage;  $f$ : frequency; LC-MS/MS: liquid chromatography-tandem mass spectrometry.

method and impedance spectroscopy (Srinivasan et al., 2015; Nazari et al., 2023). The Ohm's Law method applies an alternating current across electrodes on either side of the cell layer, measures the resistance, and multiplies it by the membrane surface area to obtain the TEER value, which is then normalized to that of a blank control. A higher resistance corresponds to a greater TEER value, indicating a more intact barrier. The Ohm's Law method is simple, but its measurements are affected by electrode position and current density uniformity. Conversely, impedance spectroscopy applies a small-amplitude alternating current excitation signal with frequency sweeps and measures the resulting current's amplitude and phase response. By scanning frequencies and fitting the data to equivalent circuit models, it captures more comprehensive electrical properties and provides a more accurate TEER value (Douville et al., 2010).

TEER measurement technology is advancing toward standardization, integration, and high-throughput capabilities. For example, van der Helm et al. (2016) developed a four-electrode TEER system for direct quantification in organ-on-a-chip platforms. Six distinct electrode configurations are employed for measurements, and a derived formula is used to directly calculate the resistance of the cellular barrier and the membrane, effectively eliminating background noise from microchannel resistance. In an hCMEC/D3 cell-based BBB-on-a-chip model, stable, reliable TEER measurements ( $22 \Omega \cdot \text{cm}^2$ ) were achieved, consistent with traditional transwell results (Malik et al., 2025). This approach does not interfere with microscopic observation, enhancing the reliability of quantitative barrier function assessment on organ-on-a-chip platforms. Separately, Badiola-Mateos et al. (2021) introduced a multi-frequency TEER sensor array for real-time monitoring of BBB integrity. This technology integrates a microfluidic chip with embedded gold electrodes and couples it with broadband impedance spectroscopy, acquiring impedance data across a frequency range of 100 Hz to 10 MHz. By incorporating a machine learning algorithm, the system enables high-precision identification of different physiological stages, including barrier formation, maturation, injury, and recovery. This system possesses multi-position, non-invasive dynamic monitoring capabilities, and its measurements correlate strongly with changes in tight junction proteins as revealed by immunofluorescence

staining. This technology provides a more powerful and reliable in vitro evaluation tool for drug delivery research and barrier-related pathology studies.

### 3 Imaging techniques for BBB permeability assessment

In vitro BBB models partly recapitulate the structure and function of the barrier, and corresponding permeability assessment in these models is essential for CNS research and early-stage drug screening. However, if subtle regional changes in BBB integrity need to be detected, multiple in vivo imaging techniques are required, such as computed tomography (CT), magnetic resonance imaging (MRI), positron emission tomography (PET), and optical imaging (Table 2). These methods can dynamically monitor changes in BBB permeability, molecular transport processes, and intercellular interactions, providing direct visual evidence for understanding the pathological mechanisms of CNS disorders and evaluating treatment efficacy.

#### 3.1 MRI

##### 3.1.1 Non-contrast MRI

Water molecules freely traverse the BBB and exhibit high sensitivity to early pathological changes in the BBB (Cornford and Hyman, 1999). As the main source of MRI signal, water provides sufficient sensitivity without exogenous contrast agents, making it an ideal endogenous tracer for assessing BBB function. Arterial spin labeling (ASL) is widely employed for the quantitative evaluation of cerebral perfusion and other hemodynamic parameters through the manipulation and modeling of the water signal in blood (St. Lawrence et al., 2000, 2012; Parkes and Tofts, 2002; Wang et al., 2007; Liu et al., 2011). Based on ASL, Lin et al. (2018) developed a non-contrast MRI technique that quantifies the water-extraction fraction and permeability parameters by detecting the signal from venous blood flowing into the superior sagittal sinus, which enables the quantitative assessment of BBB water permeability. Under mild hypercapnic conditions, this technique becomes much more sensitive, opening a new avenue for the non-invasive, safe monitoring of BBB function.

In recent years, this technique has shown promise in various neurological disease models. Mouchtouris

**Table 2 In vivo imaging technologies for BBB assessment**

| Imaging technology | Application in BBB assessment                                                                                                                                             | Sensitivity                                                           | Resolution                                          | Penetration depth                               |
|--------------------|---------------------------------------------------------------------------------------------------------------------------------------------------------------------------|-----------------------------------------------------------------------|-----------------------------------------------------|-------------------------------------------------|
| <b>MRI</b>         |                                                                                                                                                                           |                                                                       |                                                     |                                                 |
| NC-MRI             | Assessment of the BBB integrity through water MRI signals                                                                                                                 | Relatively low (based on water proton signal)                         | 10–100 $\mu\text{m}$                                |                                                 |
| DCE-MRI            | Evaluation of BBB leakage based on contrast agent (gadolinium) injection                                                                                                  | Medium (dependent on contrast agent)                                  | 1–2 mm                                              |                                                 |
| <b>CT</b>          |                                                                                                                                                                           |                                                                       |                                                     |                                                 |
| Conventional CT    | Evaluation of the BBB permeability based on X-ray signal of contrast agents (iodine)                                                                                      | Relatively low ( $10^{-6}$ mol/L)                                     | 50 $\mu\text{m}$                                    |                                                 |
| CTP                | Quantitative assessment of BBB damage by analyzing the time–density curve of contrast agent                                                                               | Moderate (dependent on iodine concentration)                          | 0.5–1.0 mm                                          | Equivalent to CT (limited by detector coverage) |
| Spectral CT        | Precise analysis of BBB leakage of contrast agent based on X-ray energy spectrum                                                                                          | Higher for specific substances (iodine) than those of conventional CT | 50–200 $\mu\text{m}$                                |                                                 |
| PET                | Dynamic quantitative analysis of the BBB permeability of radionuclide-labeled substances (drug, protein, and peptide)                                                     | High ( $10^{-12}$ – $10^{-11}$ mol/L)                                 | 1–2 mm                                              |                                                 |
| <b>FI</b>          |                                                                                                                                                                           |                                                                       |                                                     |                                                 |
| NIR-II             | Real-time, dynamic visualization of the entire process of BBB based on NIR-II fluorescence imaging                                                                        | High ( $10^{-12}$ – $10^{-9}$ mol/L)                                  | 2–3 mm (deep dependency)                            | 3 cm                                            |
| TPM                | High-resolution assessment of molecular crossing of the BBB based on in vivo or cellular-level FI                                                                         | High                                                                  | Submicron scale (approximately 0.65 $\mu\text{m}$ ) | 1 mm                                            |
| PAI                | A comprehensive assessment of BBB permeability by combining high-resolution vascular structure and function information obtained from fluorescence and ultrasound imaging | High                                                                  | 0.4–0.6 mm                                          | Approximately 1 cm                              |

BBB: blood–brain barrier; MRI: magnetic resonance imaging; CT: computed tomography; PET: positron emission tomography; FI: fluorescence imaging; PAI: photoacoustic imaging; NC-MRI: non-contrast MRI; DCE-MRI: dynamic contrast-enhanced MRI; CTP: CT perfusion; NIR-II: near-infrared II; TPM: two-photon microscopy.

et al. (2024) used non-contrast MRI to assess BBB permeability in patients with acute ischemic stroke. They employed a diffusion-prepared pseudo-continuous ASL (DP-pCASL) sequence to measure the water exchange rate and combined it with neurite orientation dispersion and density imaging (NODDI) sequence analysis to characterize tissue microstructure. The results revealed that the water exchange rate was correlated with post-stroke time and tissue parameters, and was also influenced by mechanical thrombectomy. Xiong et al. (2025) used the non-contrast MRI technique to conduct a longitudinal study in an AD mouse model. They found that as the disease progressed, both BBB water permeability and trans-membrane water exchange in the hippocampal region

were increased, accompanied by abnormal polarization of the AQP4 water channel protein. These impairments in key components of the glymphatic system are associated with the deposition of A $\beta$  and Tau proteins, revealing the role of waste clearance failure in AD pathogenesis.

### 3.1.2 Dynamic contrast-enhanced MRI

Dynamic contrast-enhanced MRI (DCE-MRI) is another MRI method that requires intravenous injection of gadolinium-based contrast agents. Through repeated scanning, it captures signal changes caused by trans-BBB extravasation of contrast agents and quantifies its leakage rate, thereby assessing the degree of BBB opening. Over the past two decades, DCE-MRI

has become a well-established technique for evaluating BBB leakage in vivo (Heye et al., 2014; Raja et al., 2018).

DCE-MRI has been used to study cerebral small vessel disease (cSVD). Zhang et al. (2019) used a dual-temporal-resolution DCE-MRI protocol combined with the Patlak pharmacokinetic model, which quantified both the BBB leakage rate and leakage volume. Their results indicated that within white matter hyperintensity (WMH) regions of cSVD patients, the spatial extent of leakage was expanded, yet the leakage rate was reduced, suggesting a chronic, diffuse leakage process. In contrast, during healthy aging, the leakage rate was correlated with cognitive decline. Further elucidating this, Verstappen et al. (2025) utilized high-resolution DCE-MRI to reveal that BBB impairment in cSVD patients exhibits spatial heterogeneity: leakage was more pronounced in the normal-appearing white matter (NAWM) distal to WMH borders, forming a proximal-to-distal gradient pattern. Thus, BBB damage in cSVD is not uniform but is closely associated with microvascular architecture and local blood flow regulation. These findings provide crucial imaging evidence for understanding the spatial heterogeneity of BBB impairment and highlight the potential of BBB leakage as a biomarker for early diagnosis and therapeutic evaluation in cSVD.

DCE-MRI has also provided important evidence on BBB dysfunction in the early stages of AD. Van de Haar et al. (2016) found no significant difference in leakage rate between global NAWM and cortical gray matter (GA) in early AD patients. BBB dysfunction in early AD mainly appears as diffuse, widespread, subtle leakage rather than a marked local increase. This study indicates that widespread BBB disruption is a principal pathological feature of early AD. Nation et al. (2019) further investigated the relationship between BBB disruption and early cognitive impairment through a combined analysis of DCE-MRI and CSF biomarkers. Their results demonstrated that during the early stages of cognitive decline, the BBB exhibits significant leakage within the hippocampus and its subfields, independent of changes in A $\beta$  and Tau pathology. Concurrently, elevated levels of soluble platelet-derived growth factor receptor  $\beta$  (sPDGFR $\beta$ ) in the CSF also showed an association between BBB disruption and injury to cerebral capillary pericytes. Hence, BBB disruption is an independent biomarker for early cognitive impairment, highlighting

the independent role of NVU dysfunction in the pathogenesis of AD and suggesting a new direction for early diagnosis and intervention.

Furthermore, DCE-MRI has been applied to psychiatric disorders. Shang et al. (2024) provided the first in vivo evidence of significantly elevated BBB permeability in emotion-related brain regions (olfactory cortex, caudate nucleus, and thalamus) of patients with major depressive disorder (MDD). Importantly, this increase in permeability was positively correlated with the severity of depressive symptoms, providing direct imaging evidence for BBB disruption in MDD pathophysiology.

In summary, DCE-MRI has become a valuable tool for quantifying BBB permeability in vivo, though the main challenge remains standardizing acquisition and analytical pipelines to enable robust multicenter comparisons. Key frontiers include defining the prognostic value of distinct leakage phenotypes and validating BBB permeability metrics as endpoints for interventional trials. Looking ahead, DCE-MRI holds substantial translational potential for guiding precision medicine, validating BBB integrity as a therapeutic target, and serving as a biomarker in early-phase clinical trials.

### 3.2 CT

CT imaging produces cross-sectional images based on the differential X-ray absorption by various tissues. In BBB permeability assessment, CT relies on iodinated contrast agents: an intact BBB prevents contrast agent extravasation, whereas BBB disruption allows leakage into the interstitium, resulting in localized hyperdensity on CT images, termed “contrast enhancement.” CT offers rapid scanning, widespread availability, and high sensitivity for detecting hemorrhage and calcification, making it suitable for initial screening in acute brain injury. However, its application in BBB research is limited by insufficient quantitative capability, low soft-tissue contrast, radiation exposure, and limited functional information (Pardridge, 2005). CT perfusion (CTP) imaging has advanced CT-based BBB research. This technique involves the rapid intravenous bolus administration of a contrast agent, followed by continuous dynamic scanning of a selected tissue plane to generate time–density curves. BBB permeability parameters are then calculated by applying pharmacokinetic models (Miles and Griffiths, 2003). For instance, Xu et al. (2017) used CTP to

demonstrate a significant increase in BBB permeability in the perihematomal region within 24–72 h following acute spontaneous basal ganglia hemorrhage, which correlated positively with both hematoma and perihematomal edema volumes. Furthermore, Chen XY et al. (2024) found that relative permeability parameters from CTP in hypoperfused regions could predict the risk of parenchymal hematoma after mechanical thrombectomy in ischemic stroke patients, with better predictive value than that of conventional imaging markers.

Spectral CT, another major technological advancement, works on the principle that different materials exhibit distinct X-ray attenuation characteristics at different energy levels. Performing scans with two different X-ray energy spectra simultaneously or in rapid succession enables material decomposition (Johnson et al., 2007; Siegel et al., 2016; Winklhofer et al., 2017). Spectral CT can generate iodine density maps and virtual non-contrast images, improving detection of subtle BBB leakage while reducing artifacts and contrast agent dosage (Grkovski et al., 2023b). Research by Grkovski et al. (2023a) demonstrated that this technique allows for clearer differentiation between ischemic tissue and normal brain parenchyma post-stroke, enabling more precise lesion characterization and quantitative assessment. Consequently, spectral CT provides a powerful imaging tool for the more accurate evaluation of BBB injury in stroke, with the potential to optimize clinical management and prognostic evaluation.

### 3.3 PET

PET, as a highly sensitive molecular imaging technique, enables *in vivo*, non-invasive assessment of BBB permeability by tracking radiolabeled tracers. It detects ligand concentrations in the picomolar to femtomolar levels and quantifies parameters such as the tracer uptake rate or permeability–surface area product (PS) at the BBB. This transcends simple qualitative assessments of “present or absent,” enabling precise comparisons across individuals and time points, thereby providing objective metrics for monitoring disease progression or therapeutic response. Additionally, PET offers targeting specificity; by selecting different radiotracers, it can specifically investigate various BBB functions. PET has been widely used in neurodegenerative diseases, brain tumor imaging, and drug development.

Chen et al. (2017) developed a radiotracer ( $^{18}\text{F}$ -aluminum-NOTA-conjugated truncated Evans Blue (EB), namely [ $^{18}\text{F}$ ] FAI-NEB) that rapidly binds to serum albumin and used dynamic PET to assess tumor vascular leakage via a simplified quantitative parameter ( $P_s$ ), which correlated well with traditional EB extraction and DCE-MRI. In another study, Chung et al. (2025) used a non-invasive method that combined high temporal-resolution total-body PET imaging with kinetic modeling to quantitatively measure the PS of molecules at the BBB. Using multiple tracers (e.g.,  $^{18}\text{F}$ -fluorodeoxyglucose [FDG],  $^{18}\text{F}$ -fluciclovine, and  $^{11}\text{C}$ -butanol), they simultaneously estimated cerebral blood flow and the BBB transport rate constant  $K_1$  in a single scan, from which the molecular PS was calculated. In studies of healthy aging and patients with metabolic dysfunction-associated steatohepatitis (MASH), they found that BBB permeability to FDG decreases with advancing age. Dai et al. (2025) designed a novel fluorescent probe based on a quino-line-malonitrile (QM) scaffold, which has the capabilities of near-infrared fluorescence (NIRF), MRI, and PET trimodal imaging. This probe can cross the BBB and specifically bind to cerebral A $\beta$  plaques. In AD mice, it showed higher brain radioactive accumulation (AD 1.7 %ID/g vs. wild-type controls 0.9 %ID/g, where ID refers to the injected dose) and enabled dynamic metabolic monitoring. This work integrated NIRF, MRI, and PET imaging capabilities into a single molecular probe, providing a powerful multifunctional imaging tool for the early diagnosis of AD.

PET imaging has also been used to evaluate BBB modulation in AD mice. Li WJ et al. (2025) used focused ultrasound (FUS) to induce transient, reversible BBB opening, enhancing the cerebral delivery of the radiotracer [ $^{68}\text{Ga}$ ] STZL4110 in AD mice and improving PET signal-to-noise ratio. These findings contribute to the development of novel strategies for optimizing brain-targeted drug delivery and pathological imaging. Toyohara (2016) has elucidated the fundamental principles of PET imaging for investigating P-gp function at the BBB. As a critical efflux transporter, P-gp directly modulates the permeation of pharmaceuticals and toxins into the CNS, thereby influencing therapeutic efficacy and toxicological profiles in CNS disorders (Juliano and Ling, 1976; Ambudkar et al., 1999). P-gp hyperfunction may contribute to multidrug resistance, while hypofunction potentially elevates neurotoxicity risks. These advancements

highlight the utility of P-gp PET imaging for mechanistic investigations and clinical pharmacokinetic guidance for conditions such as epilepsy and depression.

In summary, PET is evolving from a descriptive research tool into a quantitative platform for precision neurovascular medicine. Standardizing tracer protocols and simplifying complex kinetic models will support the early diagnosis of neurological disorders, drug delivery, and efficacy assessment.

### 3.4 Fluorescence imaging

Fluorescence imaging (FI), with its non-invasive nature, high sensitivity, superior spatiotemporal resolution, and real-time dynamic imaging capabilities, has become a key tool in biomedical research. It enables multidimensional visualization from molecular to in vivo levels. In preclinical investigations of BBB permeability, FI offers rapid, intuitive dynamic monitoring (Hong et al., 2017). Advances in super-resolution optical imaging instruments, in vivo imaging equipment, and imaging probes have broadened its use for dynamic BBB detection in living organisms and high-resolution imaging of in vitro BBB models or excised tissues.

#### 3.4.1 Ex vivo brain tissue BBB FI

Ex vivo FI is a standard method for evaluating BBB permeability. It typically involves intravenous administration of fluorescent tracers such as fluorescein isothiocyanate (FITC)-dextran or EB, followed by transcranial perfusion, brain extraction, and confocal microscopy to visualize tracer distribution and extravasation. This approach clearly visualizes whether the tracer remains confined within the blood vessels or has extravasated into the perivascular parenchyma, enabling precise anatomical localization of leakage sites. Zhang et al. (2020) systematically evaluated electroacupuncture stimulation at specific frequencies (2/100 Hz) for its effect on rat BBB permeability using EB and 20 kDa FITC-dextran combined with confocal microscopy. The results demonstrated that electroacupuncture reversibly enhanced BBB permeability in rats, with restoration within 12 h post-stimulation. Imaging revealed tracer extravasation in the cerebral cortex and perivascular regions, indicating that electroacupuncture induces reversible BBB opening by modulating tight junction proteins (e.g., ZO-1 and occludin). Shumer-Elbaz et al. (2025) combined low-frequency FUS with microbubbles to deliver ionizable

lipid nanoparticles (LNPs) to glioblastoma (GBM) regions. They assessed BBB opening by examining EB and fluorescent dextran extravasation in brain sections, and used cyanine 5 (Cy5)-labeled small interfering RNA (siRNA)-LNP fluorescence to confirm translocation and tumor accumulation. Confocal microscopy revealed the distribution of Cy5-siRNA-LNPs within tumor tissue, with co-localization in green fluorescent protein (GFP)-labeled tumor cells indicating cellular uptake. 4',6-Diamidino-2-phenylindole (DAPI) staining further illustrated the distribution of LNPs among non-tumor cells in the tumor microenvironment. This strategy overcomes the limitations of the BBB for LNP delivery, enabling LNP-based therapies for neurological disorders. Ex vivo imaging provides intuitive spatial localization and semi-quantitative data, but in vivo imaging technologies are needed for dynamic monitoring of BBB permeability.

#### 3.4.2 In vivo FI techniques

In vivo FI techniques include wide-field, near-infrared I/II (NIR-I/II), and multiphoton FI. Conventional wide-field FI requires cranial window implantation or skull thinning (Yang et al., 2010; Wang et al., 2019) and offers limited resolution and depth, but is simple, cost-effective, compatible with diverse fluorescent probes, and suitable for large-scale screening and longitudinal studies. With advances in microscopy, NIR-I/II FI imaging has reduced tissue absorption and scattering of NIR light. NIR-II (1000–1700 nm) achieves approximately 3 cm depth, superior signal-to-noise ratio, and higher spatial resolution than visible or NIR-I (Hong et al., 2012; Cao et al., 2020). Wang et al. (2022) systematically modified aniline groups in dye molecules to create a series of NIR-II fluorophores with emission wavelengths ranging from 800 to 1400 nm. Certain dyes (e.g., BF<sub>1</sub> and BF<sub>6</sub>) with molecular weights below 450 kDa cross the intact BBB for non-invasive, high-fidelity NIR-II FI of the mouse brain parenchyma. Lu et al. (2024) developed a novel NIR-II imaging probe (named DGC) that targets connective tissue growth factor (CTGF) for early AD detection. Comprising gold nanoclusters and cyclic peptides, DGC exhibits high BBB affinity ( $K_D = 21.9$  nmol/L, where  $K_D$  is the dissociation equilibrium constant) and penetration ability, allowing clear visualization of elevated cerebral CTGF levels in 1-month-old amyloid precursor protein/presenilin 1 (APP/PS1) transgenic mice. Additionally, this probe supports

multimodal analysis, including FI, chromogenic reaction, and inductively coupled plasma mass spectrometry (ICP-MS) quantification, and has been validated in human AD brain sections. Notably, Zhang et al. (2023) conducted pioneering research that achieved real-time multiplexed NIR-II imaging enhanced by deep learning—a breakthrough marking a crucial step toward the clinical translation of biomedical optical imaging technologies. However, the practical application of this technology in assessing BBB permeability still faces several translational hurdles. These include reduced signal fidelity due to increased imaging depth, the lack of clinically validated BBB-specific tracers, and challenges related to the *in vivo* stability, biosafety, and scalable production of NIR-II imaging probes.

Multiphoton microscopy (MPM), particularly two-photon microscopy (TPM), uses NIR femtosecond laser pulses to enable real-time, label-free, subcellular resolution imaging of biological tissues with minimal photodamage—achieving depths of hundreds of micrometers, low phototoxicity, and rapid acquisition (Denk et al., 1990). In BBB research, TPM allows dynamic, real-time monitoring of leakage processes. Lee et al. (2018) used real-time *in vivo* two-photon imaging to study the effects of chronic restraint stress on the mouse cerebrovascular system. Their research revealed that chronic stress induced generalized vasoconstriction, reduced total cerebrovascular volume, and concomitant BBB hyperpermeability. These pathological changes were associated with upregulated cerebral vascular endothelial growth factor-A (VEGF-A) expression and downregulated tight junction protein Claudin-5. TPM also serves as a tool for evaluating the impact of pharmacological or physical interventions (e.g., FUS) on BBB permeability. Poon et al. (2018) investigated the effects of FUS combined with microbubbles on A $\beta$  plaques in AD model mice. Using *in vivo* real-time two-photon imaging, they visualized immediate BBB opening post-FUS (evidenced by dextran leakage) and tracked individual A $\beta$  plaques over time. The results indicated that a single FUS session simultaneously enhanced BBB permeability and reduced plaque volume by 62% within two days, persisting for two weeks, suggesting that FUS promotes A $\beta$  clearance via transient BBB opening. Subsequent MRI-guided repeated FUS treatments confirmed the feasibility and cumulative efficacy of this protocol.

Although TPM offers unparalleled advantages in preclinical studies, its extremely shallow penetration depth and the need for craniotomy or the implantation of a transparent cranial window limit its clinical translation potential. Consequently, TPM is primarily used for basic research.

### 3.5 Photoacoustic imaging

Photoacoustic imaging (PAI) is an emerging non-invasive biomedical imaging modality that combines optical contrast with the ultrasonic penetration depth. It generates ultrasonic waves via thermoelastic expansion following light absorption in biological tissues, with signal amplitude reflecting local optical properties (Hui et al., 2016; Chen HY et al., 2024; Wang et al., 2024). Multiwavelength PAI enables high-contrast imaging of molecular targets at centimeter depth, overcoming the depth limits of conventional optical imaging (Liu XG et al., 2021; Lin and Wang, 2022).

The potential of PAI for assessing BBB integrity has been vividly demonstrated by recent studies utilizing novel molecular probes. Qin et al. (2025) developed a novel fluorescence/photoacoustic probe (namely, IVTPO) based on an electron donor- $\pi$ -acceptor (D- $\pi$ -A) system for GBM imaging. This probe exhibits strong optical absorption (572 nm), high molar extinction coefficient ( $12.29 \times 10^4 \text{ M}^{-1} \cdot \text{cm}^{-1}$ , where M represents mol/L), and outstanding BBB penetration. In GBM mouse models, IVTPO enabled real-time PAI through the intact skull, clearly delineating tumor location, abnormal vascular structures, and early probe extravasation in peritumoral regions. Compared to EB, IVTPO showed superior sensitivity and targeting, detecting subtle leakage before full BBB disruption, offering a powerful tool for early GBM diagnosis. Li J et al. (2025) reported a D- $\pi$ -A molecular probe DMA-Te, which was constructed by chalcogenapyrylium modulation with *N,N*-dimethylaniline (DMA) as the donor and the chalcogenapyrylium moiety as the acceptor, for high-contrast PAI of A $\beta$  plaques in AD. By incorporating a Te atom as an electron acceptor, this probe achieves a red-shifted absorption peak at 794 nm while maintaining a low fluorescence quantum yield (0.0006), favoring strong photoacoustic signal generation. The probe exhibits 20.8% BBB penetration rate and a high A $\beta$ -binding affinity. In AD mice, a dual-wavelength PAI system successfully imaged A $\beta$  plaques and cerebral vasculature simultaneously with high resolution.

Despite promising preclinical advances, the clinical translation of PAI faces dual challenges. First, the development of molecular probes is constrained by challenges in biocompatibility, pharmacokinetics, and scalable production. Probes such as IVTPO and DMA-Te, for instance, require extensive safety validation beyond rodent models. Second, the field lacks standardized quantitative methods, currently relying on descriptors such as “signal enhancement” rather than offering universally comparable, dynamic metrics of BBB permeability. Future efforts should focus on developing novel probes, establishing robust quantitative standards, and validating reliability through multicenter clinical studies.

#### 4 Artificial intelligence in enhancing BBB permeability assessment

In vitro BBB models and in vivo imaging techniques remain essential tools for assessing BBB permeability, but they face challenges such as high costs, long time, and limited mechanistic insights. Artificial intelligence (AI) is reshaping the field by integrating multi-omics data, simulating molecule–barrier interactions, optimizing drug design workflows, and analyzing biomedical images (Clark, 2003; Saxena et al., 2019).

Currently, key AI approaches for predicting BBB permeability (Grant et al., 2025) include classical machine learning based on chemical descriptors (Yuan et al., 2018; Shaker et al., 2021), deep learning utilizing graph neural networks (Withnall et al., 2020; Jiang et al., 2021; Dey and Ning, 2024), and natural language processing based on simplified molecular input line entry system (SMILES) sequences (Tevosyan et al., 2022; Wen et al., 2022; Rollins et al., 2024). These models significantly accelerate drug development by predicting candidate permeability and elucidating BBB interactions. In particular, descriptor-based models offer a key advantage in their interpretability, providing mechanistic insights into prediction outcomes. For instance, Liu LL et al. (2021) employed support vector machine (SVM), random forest (RF), and extreme gradient boosting (XGBoost) algorithms with 1757 compounds and nine molecular fingerprints to construct an ensemble of 27 base classifiers, optimizing BBB permeability predictions.

Their study calculated the average reduction in the Gini coefficient for molecular fingerprints using an RF model, enabling clear quantification of the contributions of different molecular substructures to prediction outcomes. It explicitly identified a strong correlation between oxygen-containing polar groups (e.g., carboxylic acids and polyhydroxy structures) and BBB impermeability. However, these models rely on predefined molecular descriptors and may overlook complex interactions. In contrast, deep learning with graph neural networks demonstrates significant potential by directly representing molecular structure and spatial information. For example, Group Graph is a novel substructure-level method that balances computational efficiency and representational capacity by simplifying graphs while preserving chemical information (Cao et al., 2024). However, this process simplifies stereochemical information and omits 3D structural details, which may limit its use in high-value contexts. Language-based BBB permeability prediction methods treat SMILES strings as textual input, enabling end-to-end learning without predefined features, making them suitable for large-scale pretraining. However, their performance typically lags behind that of traditional classical and graph-based models, and they suffer from poor interpretability. Molecular Prediction Model Fine-Tuning (MolPMoFiT) is a transfer learning approach pretrained on millions of compounds and fine-tuned on small datasets. It matches or surpasses state-of-the-art models for BBB permeability prediction, especially in data-sparse scenarios (Li and Fourches, 2020).

AI also aids in the quantitative analysis of biomedical images. Jackson et al. (2021) developed a novel computational tool for CNS PET tracer design using a dataset of 140 tracers. By analyzing seven physicochemical properties—including computed logarithm of the partition coefficient (clogP) and topological polar surface area (tPSA)—they identified differences between successful and non-penetrating tracers in tPSA and clogP. Their screening tool achieved 96.7% specificity and 97.4% positive predictive value, enabling early elimination of failures. This work provides valuable data and efficient computational tools to support the rational design of future CNS PET tracers. In a different application, Zhang et al. (2023) developed a novel method combining optical coherence tomography (OCT) with deep learning (pix2pix cGAN)

for rapid, non-invasive visualization and quantification of microvessels in 3D BBB models. By outperforming traditional morphological processing and RF methods, this deep learning approach achieves superior accuracy in segmenting vessels in low-contrast OCT images. It thus provides an efficient, label-free analytical tool for assessing permeability in BBB models and for drug screening applications.

## 5 Conclusions and future directions

In conclusion, precise assessment of BBB permeability is essential for understanding mechanisms of neurological disease and improving drug development. In vitro models serve as crucial screening platforms in early drug discovery, evolving from traditional static transwell systems to dynamic models that incorporate fluid shear stress (e.g., hollow-fiber reactors and microfluidic chips), and more recently, to highly biomimetic organoid models—progressively enhancing their capacity to replicate the complex microenvironment of NVU. The permeability of in vitro BBB models is evaluated by measuring  $P_{app}$  and TEER. At the in vivo level, imaging technologies complement one another to establish a comprehensive system for real-time, non-invasive assessment of BBB permeability.

To overcome the limitations of existing technologies in physiological relevance, dynamic monitoring capabilities, and quantitative accuracy, future BBB permeability assessment technologies will exhibit a distinct trend toward multimodal integration and intelligence-driven strategies. On the one hand, in vitro model systems will evolve from “simple simulation” to “precision recapitulation”: patient-specific organ-on-a-chip and organoid technologies will achieve highly biomimetic reconstruction of NVU, closely mimicking the human physiological environment. The integration of high-throughput barrier assessment, transporter functional analysis, and multi-omics technologies will enable multidimensional characterization of functional phenotypes. On the other hand, in vivo imaging will use high-resolution MRI, novel PET probes, and NIR-II optical imaging to achieve non-invasive, dynamic, quantitative monitoring of BBB permeability.

AI will become the core driver of this paradigm. By integrating multi-omics data, simulating molecule–barrier interactions, and optimizing drug design, such

as using deep learning to predict compound permeability or intelligently identifying barrier damage features from imaging data, AI will fundamentally reshape BBB permeability research. The deep integration of AI will establish a closed loop of “computational prediction–in vitro validation–in vivo confirmation,” not only dramatically enhancing drug screening efficiency but also enabling the mechanistic interpretation and dynamic prediction of BBB function through multi-omics data integration. Furthermore, the deep application of AI in image processing will enable the automatic and quantitative extraction of critical information—such as leakage characteristics and vascular network morphology—from vast optical imaging datasets. This advancement will bridge the gap between mere observation and comprehensive understanding, as well as predictive capabilities. These developments will significantly enhance our ability to perform dynamic, holistic analyses of the functional state of the BBB in living organisms, ultimately driving innovation in diagnostic approaches and therapeutic strategies for CNS disorders.

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## Author contributions

Yanhong SUN, Shihua LUO, Ying ZHU, and Miaomiao WANG conceived, designed, discussed the work, and revised the manuscript. Lu GAN, Miaomiao WANG, and Yiru FAN edited the manuscript. Miaomiao WANG and Chenxi DUAN prepared figures. All authors have read and approved the final manuscript, and therefore, have full access to all the data in the study and take responsibility for the integrity and security of the data.

## Compliance with ethics guidelines

Miaomiao WANG, Lu GAN, Yiru FAN, Chenxi DUAN, Ying ZHU, Shihua LUO, and Yanhong SUN declare that they have no conflicts of interest.

This review does not contain any studies with human or animal subjects performed by any of the authors.

## Declaration on the use of generative AI tools

In preparing this work, the authors used DeepSeek to refine language and grammar, and Doubao to improve artwork visual quality (e.g., resolution and color correction). The authors reviewed and edited the output and take full responsibility for the final content.

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