

Neuroimaging and fluid biomarkers in Parkinson's disease in an era of targeted interventions

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A major challenge in Parkinson's disease is the variability in symptoms and rates of progression, underpinned by heterogeneity of pathological processes. Biomarkers are urgently needed for accurate diagnosis, patient stratification, monitoring disease progression and precise treatment. These were previously lacking, but recently, novel imaging and fluid biomarkers have been developed. Here, we consider new imaging approaches showing sensitivity to brain tissue composition, and examine novel fluid biomarkers showing specificity for pathological processes, including seed amplification assays and extra-cellular vesicles. We reflect on these biomarkers in the context of new biological staging systems, and on emerging techniques currently in development.

Parkinson's disease (PD) is the second most common neurodegenerative disorder, affecting over 6 million people worldwide¹, with a marked increase in prevalence in the past 30 years¹. It is characterised by rigidity, slowness and tremor, and causes disabling symptoms including dementia, depression and hallucinations. Although the pathological hallmark of PD is the accumulation of alpha-synuclein-containing Lewy bodies, most patients also have varying degrees of other pathological accumulations, including beta-amyloid plaques and tau-tangles, which are more strongly linked with dementia onset and poor outcomes². A major challenge is the wide heterogeneity in initial presentation and rates of progression, which is partly underpinned by variation in underlying pathological accumulations². This, coupled with a lack of sensitive and specific diagnostic markers, and quantitative biomarkers of progression, has hampered clinical trials for disease modification in PD³.

Quantitative biomarkers are important in PD for three key reasons: i) for early and accurate diagnostic certainty in the clinic and for inclusion in clinical trials; ii) to monitor disease progression and predict outcomes; and iii) to stratify patients with differing underlying pathologies to be targeted for distinct treatments (see Fig. 1). They can

also improve trial efficiency by showing evidence of target engagement as well as aid our understanding of underlying pathological mechanisms of PD spurring further treatment into potential therapeutic targets. The need for disease-specific biomarkers has been heightened further by calls for biological staging systems in PD^{4,5}.

Until recently, neuroimaging measures have been limited as markers for diagnosis and progression in PD. DAT SPECT imaging to detect integrity of striatal dopaminergic terminals can support diagnosis but shows minimal progression over time⁶. MRI was previously unhelpful in diagnosis and monitoring of PD, with conventional measures such as brain atrophy showing inconsistent patterns in PD, and often appearing normal until late stages⁷. Fluid measures of alpha-synuclein levels have, until recently, shown inconsistent relationships with Lewy body pathology and disease presence or progression^{8,9}. Instead, novel imaging markers sensitive to brain tissue composition¹⁰ are emerging, showing stronger relationships with disease severity. Novel measures of alpha-synuclein including its aggregation¹¹, and CSF and plasma measures of other proteins^{12–14} also show promise as diagnostic and progression markers in PD and in related conditions such as dementia with Lewy bodies (DLB)¹⁵.

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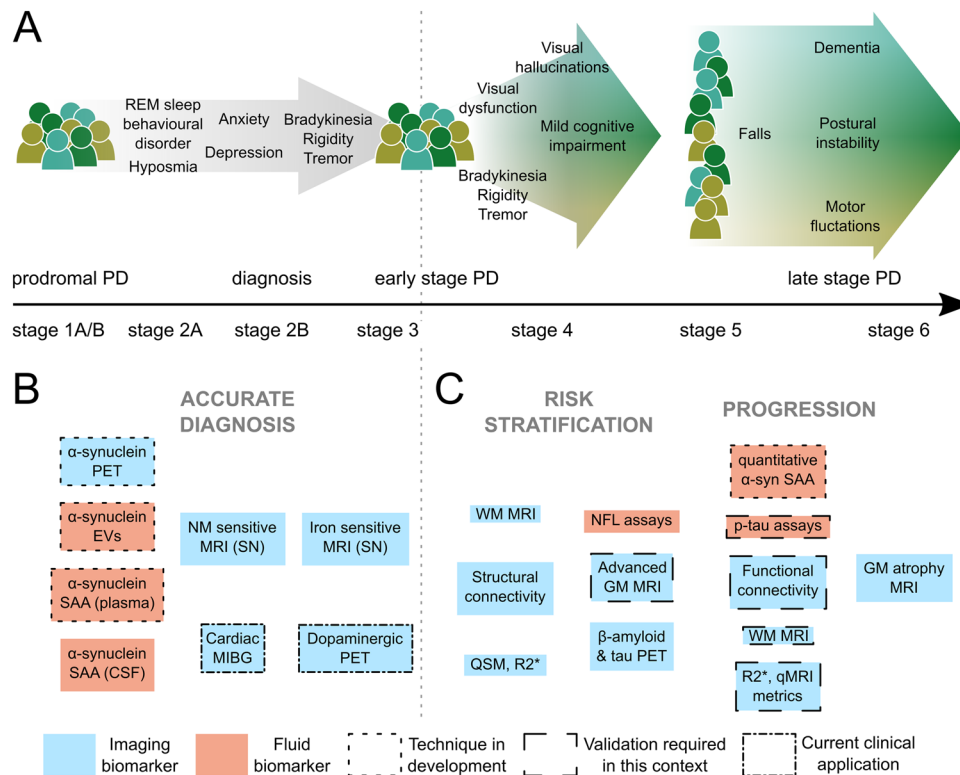


Fig. 1 | Fluid and imaging biomarkers over the course of Parkinson's disease (PD). **A** Key clinical features from prodromal to late-stage PD, emphasising the high degree of heterogeneity. Below the black arrow are the biological stages of alpha-synuclein disease, as proposed by Simuni et al.⁴. Stages 1A/B are characterised by biomarkers of neuronal alpha-synuclein and dopaminergic dysfunction, with clinical signs and symptoms emerging in stages 2A/B and functional impairment emerging from stage 3. **B, C** Key fluid and imaging biomarkers in PD, in orange and blue boxes, respectively. **B** Fluid and imaging biomarkers for accurate diagnosis. Techniques to the left have increased sensitivity at earlier as well as later stages. **C** Fluid and imaging biomarkers for risk stratification and tracking disease

progression in PD. Techniques to the left have greater utility in risk stratification for poor outcomes or faster progression. Techniques to the right show utility or potential for tracking disease progression. Dotted borders indicate techniques still in development. Dashed borders indicate techniques that may be well-established in other contexts but require further validation for these uses. α -syn alpha-synuclein; CSF cerebrospinal fluid; EVs Extracellular vesicles; GM grey matter; MIBG meta-iodoethylguanidine scintigraphy; NFL neurofilament light chain; NM neuromelanin; PET positron emission tomography; qMRI quantitative MRI; QSM quantitative susceptibility mapping; REM rapid eye movement; SAA seed amplification assay; SN substantia nigra; WM white matter.

New treatments for neurodegenerative diseases are emerging, with some showing potential application in PD. These include anti-amyloid therapies^{16,17}, and the p38 alpha-kinase inhibitor, neflamapimod¹⁸. Such interventions will require robust biomarkers to stratify patients for inclusion and as outcome measures to track responses to treatment. Here we provide a perspective of new and emerging neuroimaging and fluid biomarkers in PD. We consider their application for early and accurate diagnosis, in the context of recent proposals of staging systems for PD; in monitoring disease progression in PD; in stratifying patients for more precise treatment; and into quantifying target engagement, both in the clinic and in research. Finally, we consider novel approaches such as high-field MRI and multimodal information from wider body systems that show promise for identifying high-risk groups for more targeted therapeutic interventions.

Biomarkers for early and accurate diagnosis

Imaging biomarkers

Dopaminergic imaging. PD and other parkinsonian disorders are characterised by loss of striatal dopaminergic neurons and synapses. Resultant reduction in dopaminergic activity can be assessed using Positron emission tomography (PET) and single-photon emission computed tomography (SPECT) (Fig. 2a). A 2017 meta-analysis of 142 studies of presynaptic DAT and SPECT imaging in PD found that across all ligands, the greatest reduction in uptake was observed in the putamen, particularly in posterior regions. Across all regions, ligands

for vesicular monoamine transporter 2 (VMAT2, reflecting vesicular dopamine storage) were most sensitive at separating PD from controls, while those for aromatic amino acid decarboxylase (AADC, relating to dopamine synthesis) were least sensitive¹⁹. This likely reflects compensatory mechanisms to increase synaptic availability of dopamine²⁰. Ligands measuring presynaptic processes may be more sensitive in PD due to relative preservation, or even upregulation of postsynaptic dopamine receptors early in the disease. Dopaminergic imaging may also have potential to detect neurodegeneration at premotor stages of PD, although with less sensitivity^{21,22} (Table 1), with reduced striatal uptake observed in idiopathic REM sleep behaviour disorder (RBD)²³ and hyposmia²⁴; and in differentiating PD from atypical²¹ and vascular parkinsonism²⁵, with machine-learning-based approaches potentially able to distinguish these at the individual level²⁶. However, whilst dopaminergic imaging techniques show some decreases in PET signal soon after diagnosis, these begin to plateau within five years' of diagnosis, limiting their ability as markers of disease progression⁶.

Imaging of cerebral glucose metabolism. Another PET-based approach is to quantitatively assess cerebral glucose metabolism as a marker for neuronal function, using the radiolabelled glucose analogue 2-deoxy-2-[18 F]fluoro-D-glucose (18F-FDG). The sensitivity of 18F-FDG PET to functional changes outside the striatum allows it to detect distinctive patterns of functional change in PD²⁷, with greater utility for differential diagnosis from atypical Parkinsonism than dopaminergic PET²⁸. Recent work has also described a pattern of cerebral glucose

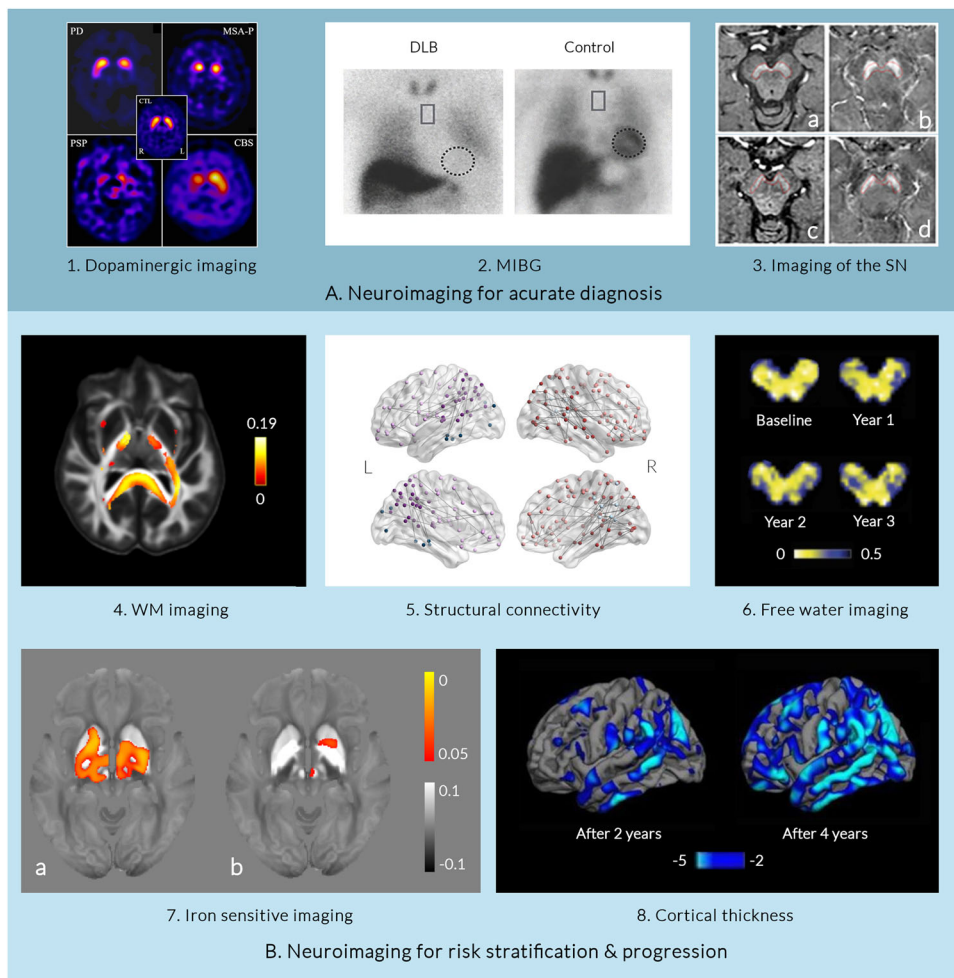


Fig. 2 | Emerging imaging measures in Parkinson's disease. A Imaging measures for accurate diagnosis of Parkinson's disease (PD) include: 1. *Dopaminergic imaging*: axial 123I-FP-CIT single-photon emission computed tomography (SPECT) scans for neuropathologically-confirmed degenerative parkinsonism; adapted from Nicastro et al.²¹ (Released under a Creative Commons Attribution-NonCommercial-NoDerivs 4.0 International license). 2. *Cardiac meta-iodobenzylguanidine (MIBG) scintigraphy*: Absent tracer uptake in the heart (dotted circle) in a patient with dementia with Lewy Bodies (DLB) compared to a control. Heart-to-mediastinum ratios (heart: dotted circles, upper mediastinum: rectangle) can be calculated. Adapted from ref. 36; 3. *Imaging of the Substantia Nigra (SN)*: Neuromelanin-sensitive MRI (NM-MRI) to visualise SN atrophy, with earliest changes involving signal loss in nigrosome-1 territory, caudal and medial-lateral portions of the SN. Loss of nigrosome-1 shown here in a PD patient in an NM-MRI image (c) and QSM image (d) compared to a control (a: NM-MRI, b: QSM). Adapted with permission from ref. 42 (Released under a Creative Commons Attribution-NonCommercial-NoDerivs 4.0 International license). **B** Neuroimaging measures for risk stratification and tracking disease progression in PD: 4. *Imaging of White Matter*: Fixel-based analysis shows reduced white matter fibre cross-section in PD patients who later developed poor outcomes over 3-year follow-up. Colour scale represents percentage reduction in fibre cross-section in PD patients with poor vs PD patients with good outcomes.

Adapted from ref. 142; 5 *Structural connectivity*: Widespread reductions in structural connectivity, mostly in frontal and parietal-occipital connections, seen at baseline in PD patients who progress to poor outcomes within 3 years, from ref. 142; 6. *Free water imaging*: Longitudinal changes in free water within the SN seen in PD patients compared to controls. Loss of signal is seen in the midbrain of a PD patient over 4 years. Adapted with permissions¹⁷⁹, (released under a Creative Commons Attribution-NonCommercial 4.0 International license) 7. *Iron-sensitive imaging*: QSM signal at baseline correlates with longitudinal worsening of (a) motor and (b)cognitive scores in PD patients over 3 years. Colour bar represents p-value family-wise error-corrected for multiple comparisons, grey scale represents susceptibility values. Adapted from ref. 173 8. *Cortical thickness*: Cortical thinning patterns in patients with mild PD compared to controls after 2 and 4-year follow-up. Statistically-significant regions shown in blue, corrected for multiple comparisons using false discovery rate. Colour bar represents *t*-values. Adapted with permission from Fillippi et al.¹²³(Released under a Creative Commons Attribution-NonCommercial-NoDerivs 4.0 International license). CBS Corticobasal syndrome, CTL Control, DLB Dementia with Lewy bodies, L Left; MCI Mild cognitive impairment, MIBG meta-iodobenzylguanidine MSA-P Multiple system atrophy parkinsonian type, NM-MRI Neuromelanin-sensitive MRI, PD Parkinson's disease, PSP Progressive supranuclear palsy; QSM Quantitative susceptibility mapping, R Right; SN Substantia nigra.

hypometabolism in iRBD associated with conversion to PD^{29,30}, with occipital hypometabolism a key feature, although this application of FDG-PET requires further validation.

Cardiac meta-iodobenzylguanidine (MIBG) scintigraphy. MIBG scintigraphy provides a measure of cardiac sympathetic denervation. MIBG is an analogue of an adrenergic blocking agent that is actively taken-up into post-ganglionic presynaptic nerve endings of the

adrenergic nervous system. When labelled with 123Iodine it can be visualised using cardiac scintigraphy, and quantified by comparing uptake in a region of interest over the heart with a control region over the mediastinum, providing the heart-mediastinum ratio (Fig. 2a).

Reduced cardiac MIBG uptake in PD was first shown in the 1990s³¹, and can discriminate PD from atypical parkinsonian disorders and mimics³² with over 85% sensitivity and specificity

Table 1 | Sensitivity and specificity of biomarkers of dopamine dysfunction for early and accurate diagnosis of Parkinson’s disease (PD)

Biomarker	Application	Sensitivity [95% confidence intervals]	Specificity [95% confidence intervals]	Utility & Evidence
Dopaminergic imaging	Overall accuracy of clinical diagnosis of a parkinsonian syndrome ²²	Early: 79% [68–87%] Clinically established diagnosis: 97% [94–99%]	Early: 97% [84–99%] Clinically established diagnosis: 98% [91–100%]	Clinical & Research: Validated biomarker
	Distinguishing PD from vascular or drug induced parkinsonism ²⁵	86% [81–90%]	83% [68–93%]	Clinical: Validated biomarker
Neuromelanin-sensitive MRI (NM-MRI)	Distinguishing PD from controls ¹⁰	89% [83–100%] ^a	83% [53–100%] ^a	Research: Validated biomarker
	Distinguishing PD from atypical Parkinsonian syndromes ⁵⁴	Locus coeruleus: MSA-P: 60% ^b PSP: 63.3% ^b Lateral substantia nigra: PSP: 86.7% ^b	Locus coeruleus: MSA-P: 90% ^b PSP: 92.2% ^b Lateral substantia nigra: PSP: 76.9% ^b	Research: Further validation is needed
Dorsolateral nigral hyperintensity/loss of swallow-tail sign	Distinguishing PD from controls ¹⁰	96% [76–100%] ^a	95% [79–100%] ^a	Clinical: Validated biomarker
Free water imaging for the SN	Distinguishing PD from controls ⁵⁰	Prodromal disease: 74.4% ^b Newly diagnosed PD: 98% ^b	Prodromal disease: 70.8% ^b Newly diagnosed PD: 75% ^b	Research: Validated biomarker
Iron sensitive imaging of the substantia nigra	Distinguishing PD from controls ^{63–79c}	QSM ^{63–77} : 86.9% [58.1–100%] ^a R2 ^{a63,68,77–79} : 65.1% [56%–81%] ^a	QSM ^{63–77} : 77.1% [45–91.3%] ^a R2 ^{a63,68,77–79} : 74.5% [52–85.7%] ^a	Research: Validated biomarker

MSA-P Multiple system atrophy with parkinsonism, PD Parkinson’s disease, PSP Progressive supranuclear palsy, QSM quantitative susceptibility mapping.

^aPooled sensitivities and specificities from meta-analyses. Range of results from individual studies in brackets.

^bConfidence intervals not reported.

^cPooled sensitivities and specificities calculated from the cited studies using a bivariate hierarchical summary receiver operating characteristic model.

Table 2 | Sensitivity and specificity of biomarkers of alpha-synuclein, and other biomarkers, for early and accurate diagnosis of Parkinson’s disease (PD)

Biomarker	Application	Sensitivity [95% confidence intervals]	Specificity [95% confidence intervals]	Utility & Evidence
Alpha-synuclein seed amplification assay (SAA)	Distinguishing PD from controls ^{11,101,103–105}	CSF ¹¹ : 88% [82–93] ^a	CSF ¹¹ : 95% [92–97%] ^a	Research and future clinical applications: Validated in CSF, further validation is needed for other samples
		Skin biopsy ¹⁰¹ : 75% ^b	Skin biopsy ¹⁰¹ : 83% ^b	
		Saliva ¹⁰³ : 76% [66.1–85.9%]	Saliva ¹⁰³ : 94.4% [86.6–100%]	
		Submandibular gland ¹⁰⁴ : 56.1% ^b	Submandibular gland ¹⁰⁴ : 92.9% ^b	
		Plasma ¹⁰⁵ : 94/6%	Plasma ¹⁰⁵ : 92.2%	
		Neuron-derived exosomal alpha-synuclein	Distinguishing PD from controls ¹⁰⁶	
Other biomarkers				
Cardiac meta-iodobenzylguanidine (MIBG) scintigraphy	Distinguishing PD from controls ³³	88% [86–90%] ^a	85% [81–88%] ^a	Clinical & research: Validated biomarker
Imaging of cerebral glucose metabolism	Distinguishing PD from controls ²⁷	93% [84–100%]	93% [85–100%]	Research: Further validation needed
	Distinguishing PD from atypical Parkinsonian syndrome ²⁸	91.4% [76.5–100%] ^a	94.7% [75–98%] ^a	Research: Further validation needed

^aPooled sensitivities and specificities from meta-analyses. Range of results from individual studies in brackets.

^bConfidence intervals not reported.

(Table 2)³³. A recent prospective study of 288 people with de novo PD, showed that cardiac MIBG had higher diagnostic accuracy than DAT-SPECT imaging, with 92% sensitivity and 94% specificity³⁴. In 2015, cardiac MIBG was included as a supportive element for diagnostic criteria for PD³⁵ and in 2017, added to diagnostic criteria for DLB (dementia with Lewy bodies)³⁶. There may also be a role for cardiac MIBG in defining distinct PD phenotypes: early studies suggested greatest loss of uptake in akinetic-rigid than tremor-dominant disease³⁷. More recently, loss of cardiac MIBG uptake was linked with

a “body first” as opposed to “brain first” phenotype, with associated loss of 11C-donepezil PET-CT signal, reflecting reduced cholinergic parasympathetic innervation³⁸.

MRI techniques sensitive to substantia nigra changes. By the time motor symptoms present in PD, extensive loss of dopaminergic neurons in the substantia nigra (SN) has already occurred³⁹. Therefore, MRI techniques sensitive to related microstructural changes have potential as diagnostic biomarkers of PD.

The pigment neuromelanin (NM) is found at high concentrations in the SN and locus coeruleus (LC), where it is synthesised through polymerisation of dopamine-protein adducts⁴⁰. Degeneration of SN neurons results in loss of NM and is detectable using neuromelanin-sensitive MRI (NM-MRI), which normally comprises a T1-weighted turbo spin echo sequence or magnetisation transfer-weighted gradient echo sequence (Fig. 2a). The NM signal is primarily due to the interaction between NM and the iron it chelates, and the influence this has on both T1-relaxation and magnetisation transfer effects⁴¹. NM-MRI signal loss in PD has been consistently demonstrated in the SN and LC^{42–45}. In a 2021 meta-analysis, NM-MRI of the SN and LC could distinguish between PD and controls with a pooled sensitivity of 89% and specificity 83%¹⁰. NM-MRI may also have utility in detecting preclinical PD, as NM alterations are seen in iRBD^{45,46}, although others have reported negative findings⁴⁷. One study combining NM-MRI with fractional anisotropy measures was able to distinguish between iRBD and control with high diagnostic accuracy⁴⁸. However, NM-MRI cannot be used to distinguish between PD and atypical Parkinsonism^{49,50}, and is so far only sensitive at group level.

The loss of NM also leads to a reduction in dorsolateral nigral hyperintensity in nigrosome 1 of the substantia nigra on susceptibility-weighted images, often referred to as loss of the swallow-tail sign. This is a robust biomarker, with a 2021 meta-analysis showing this can accurately distinguish between PD and controls¹⁰ (Table 1).

Free water imaging, calculated from diffusion weighted MRI, separating the diffusion properties within tissue from those in the extracellular space, can also be used as a measure of SN dopaminergic neuronal degeneration in PD, where increases in extracellular space correspond to loss of neurons⁵¹. Free water content in the posterior SN can distinguish between PD and controls^{52,53}, and incorporating information on free water content in the basal ganglia may help distinguish PD from atypical parkinsonism⁵⁴. Posterior SN free water content also differs between PD and iRBD⁵⁵.

Increased tissue iron content is seen in association with neurodegeneration in the SN in PD at postmortem⁵⁶. Iron can cause cell stress and atrophy through toxic free reactive oxygen species, that in turn cause damage to DNA; affect mitochondrial function; and modify proteins through reactive aldehydes, leading to iron-mediated cell death, or ferroptosis⁵⁷. In addition, excess iron promotes aggregation of alpha-synuclein fibrils⁵⁸. Iron stored in ferritin is the primary contributor to paramagnetic signal in deep brain nuclei when using quantitative susceptibility mapping (QSM) and R2*. Accordingly, robust and consistent increases in iron-related signal in the PD SN are seen using QSM and R2*^{56,59}. While NM loss is localised only to the SN and LC in PD, increases in brain tissue iron concentration in PD relative to controls are seen throughout the brain using QSM^{60–62}, most commonly in the basal ganglia. A number of studies have used QSM^{63–77} or R2* values^{63,68,77–79} in the substantia nigra to discriminate between PD and healthy controls (Table 1) with QSM generally outperforming R2* in this context⁸⁰. Approaches additionally considering QSM values outside of the SN have shown improved diagnostic accuracy between controls and PD⁸¹, as have approaches using QSM in combination with NM-MRI^{70,82} or to augment interpretation of the NI sign⁸³. As with NM-MRI and free water content, increased iron measured using QSM is also found in iRBD^{74,84}. Crucially, increases in both iron⁸⁵ and NM⁸⁶ in the SN are observed across the lifespan during normal aging, including the period where PD typically manifests. As such, age will need to be accounted for when using these tools for diagnostic use.

Diagnostic fluid biomarkers

The pathological hallmark of PD is the build-up of intracellular aggregates of alpha-synuclein, formed due to protein misfolding. Although total alpha-synuclein can be measured in body fluids such as CSF, values are heterogeneous, with overlap between patients and controls. This arises due to extremely low concentrations in CSF (with

high levels in blood causing contamination), and vulnerability to factors such as time before storage and number of freeze-thaw cycles⁸⁷. Measuring total alpha-synuclein is also unlikely to reflect disease status, as the smaller oligomeric species are thought to be the most damaging forms.

Seed amplification assays. A recent approach for detecting alpha-synuclein is based on the ability of alpha-synuclein seeds to cause monomeric alpha-synuclein to form oligomeric and then fibrillary forms. This technique was originally adapted by Green and colleagues from real time quaking induced conversion (RT-QuIC) assays used to detect prion protein⁸⁸. For this assay, biological sample is added to a well containing monomeric alpha-synuclein and thioflavin T (ThT). If the sample contains pathological alpha-synuclein, it induces templating, misfolding and aggregation of the monomers, resulting in a high β -pleated sheet content of the protein, which binds ThT, resulting in increased fluorescence after several hours. A similar approach known as protein misfolding cyclic amplification was used by Soto and colleagues⁸⁹. The common term alpha-synuclein seed amplification assay (SAA) is now used to apply to both approaches, which have since been replicated by other groups^{90–92}. The assay has been made more streamlined and faster⁹¹, and is reproducible across laboratories⁹³ with pooled sensitivity and specificity for CSF alpha-synuclein SAA across 22 studies, 88% and 95% respectively¹¹ (Table 2). Alpha-synuclein SAA can also identify prodromal stages of synucleinopathy in iRBD^{94,95}, with pooled sensitivity of 74% and specificity 93%¹¹. Currently, most SAAs report a dichotomous positive or negative readout, but additional metrics such as lag time, maximum fluorescence, and time to 50% fluorescence can also be calculated, providing potential quantification. Faster kinetics and higher maximum fluorescence values are seen in DLB compared to PD^{91,93}; and more rapid kinetics are seen for PD patients with poorer cognitive scores⁹⁶. However, relationships with motor severity^{92,97} are not seen. A recent study of 1100 patients from the Parkinson's Progression Marker Initiative (PPMI), including a large number with genetic variants, showed that alpha-synuclein SAA positivity varied, depending on genetic and clinical features⁹⁸. Alpha-synuclein SAA positivity in PD was highest for patients with *GBA* mutations (95.9%), and lowest with *LRRK2* mutations (67.5%), consistent with previous studies^{99,100}. Notably, RBD patients with hyposmia had higher levels of SAA positivity, which was also higher than DAT positivity, suggesting that alpha-synuclein SAA becomes abnormal before DAT imaging⁹⁸.

However, obtaining CSF is relatively invasive, and not amenable to repeat sampling, prompting attempts for SAAs in other biological samples. Alpha-synuclein SAAs performed on skin biopsies have shown similar sensitivities and specificities as CSF¹⁰¹; whereas nasal brushing to access the olfactory mucosa has lower accuracy, and the sampling method requires a rigid scope and otolaryngologist to access the deep structures¹⁰². Other tissue suitable for SAA include submandibular glands¹⁰¹; saliva (albeit with lower sensitivity than CSF)¹⁰³; and colonic biopsies¹⁰⁴. More recently, immunoprecipitation-based RT-QuIC, detected pathogenic alpha-synuclein in serum, with high sensitivity and specificity in discriminating PD as well as DLB from controls¹⁰⁵.

Extracellular vesicles. Another approach showing strong potential to provide useful diagnostic blood tests for PD is the quantification of proteins released from brain-derived extracellular vesicles (EVs)¹⁰⁶. These can be released from any cell type and carry proteins, DNA, RNA and glycoconjugates. Critically, they can cross the blood-brain barrier, resulting in contents from neuronal cells being carried into the periphery where they can be more easily measured. A range of techniques can be used to isolate EVs, with L1CAM immunoprecipitation particularly effective for identifying neuronal EVs¹⁰⁷. In a recent study of 275 PD patients and 144 controls, as well as atypical PD cohorts, neuronal

EVs were isolated using LICAM immunocapture, followed by mass spectrometry or electrochemiluminescence to measure proteins including alpha-synuclein and clusterin¹⁰⁶. Compared with controls and other neurodegenerative conditions including MSA, levels of mean neuron-derived exosomal alpha-synuclein were two-fold higher in clinical PD and prodromal PD, with an AUC of 0.86 for separating patients with PD from controls¹⁰⁶. An even higher AUC of 0.98 was found when combined with measures of clusterin, an exosome-associated protein¹⁰⁶. Stuendl et al similarly purified EVs (although without enrichment for brain-derived EVs); quantified alpha-synuclein content using electrochemiluminescence; and showed that EV-derived alpha-synuclein was able to discriminate PD from controls (AUC 0.77) and had an inverse relationship with motor severity¹⁰⁸. More recently Kluge et al isolated EVs from plasma using immunoblotting, dynamic light scattering and transmission electron microscopy, with LICAM signals to confirm neuronal origins of the EVs. They then extracted and amplified pathological alpha-synuclein using an SAA, which was able to correctly identify all 30 PD samples¹⁰⁹, although this has not yet been independently validated.

Omics. In contrast to hypothesis-driven searches for biomarkers, proteomics provide an alternative, data-driven approach. Several groups have recently applied proximity extension assays (PEA) to detect and quantify very large numbers of proteins simultaneously in biological samples, especially plasma. These are immunoassays, leveraging antibody-oligonucleotide protein binding for quantitative real-time polymerase chain reaction measurements. Using this approach, a number of proteins have emerged showing differences between PD patients and controls. CCL28 a protein bridging the innate and adaptive immune system, was increased in the CSF of patients with PD¹¹⁰. In a study using large-scale PEA of 276 proteins in oncology, cardiovascular and metabolism panels, dopa-decarboxylase (DDC), which catalyses the decarboxylation of dopa to dopamine, was also increased in the CSF of PD patients compared to controls and was the most significant protein in separating those groups¹³. Differences in DDC correlated with dopaminergic treatment, but were also seen in drug-naïve patients, showing that this was not a direct effect of treatment¹³.

DDC was also the top biomarker found in a PEA study in 48 PD and 347 controls using an extensive panel of 2493 proteins in CSF and 92 proteins in plasma¹⁴. Higher DDC levels correlated with worse cognitive scores and a higher risk of progressing to clinical LBD during follow-up. A suggested mechanism for this finding was a form of compensatory upregulation in patients with low dopamine levels. Other markers identified using PEA approaches in plasma include up- and down-regulation of inflammation markers in PD compared with controls; and lower levels of markers commonly linked with cardiovascular risk¹¹.

An even more powerful approach in terms of potential scale of proteins that can be tested is aptamer-based technology¹¹². This uses short single-stranded oligonucleotides that can bind with high specificity to proteins and small molecules, enabling highly multiplexed proteomics assays. In plasma samples from two separate cohorts, including a total of 311 patients with PD, 1129 proteins were measured with this type of aptamer-based platform¹¹³. Four proteins differentiated PD from controls: bone sialoprotein, osteomodulin, aminocyclase and growth hormone receptor. This aptamer-based approach can be used for large-scale biomarker discovery. A recent programme examining 4006 proteins from serum, CSF and brain tissue¹¹⁴, identified a signature of proteins more strongly associated with PD (with AUC ranging from 0.66 for serum, to 0.75 for brain tissue), including proteins from cytokine receptor interaction and complement pathways. Similarly, Abdi et al.¹¹⁵ examined 1300 proteins in serum in a PD cohort followed-up for up to 8 years, and revealed 4 proteins potentially discriminating between PD and controls with a relationship to

longitudinal clinical progression. These were NOTCH1, which has a role in neuronal and glial proliferation; ALCAM, a cell adhesion molecule involved in axonal guidance and growth; DUS3, which regulates cell differentiation; and CD36, which is involved in mediating microglial pro-inflammatory responses.

The field is changing rapidly, with multiple techniques emerging in the last 5 years alone, and a number of fluid-based biomarkers already show promise both for clinical application and use in PD research. SAAs have good accuracy in discriminating controls from PD, with AUCs of up to 0.93^{92,116}, and potentially even discriminating DLB from PD¹⁰⁵. This is still below the diagnostic accuracy displayed by fluid biomarkers in Alzheimer's disease. For example p-tau217 has been shown to distinguish between Alzheimer's and controls with an AUC of 0.98¹¹⁷, although SAA accuracy may increase in the coming years. However, the gold-standard fluid required for SAA remains CSF, which means relatively invasive collection compared with plasma. This, combined with the lack of meaningful quantification may limit SAA usage to diagnosis, rather than as a marker of progression in PD. Omics show significant promise in elucidating information about PD disease mechanisms, and the data-driven approach could reveal new targets for future clinical trials or new assay development. For example DDC could distinguish between controls and synucleinopathies with an AUC of 0.89. However, such approaches are less likely to be useful clinically in of themselves due to lack of precisely-defined outcome measures and considerable overlap between disease groups. EVs show the greatest promise for future clinical application. They are accessible from blood, and early results indicate that EV-based assays display extremely high accuracy in distinguishing between PD and controls¹⁰⁹. Approaches that combine several EV protein measures may be the most fruitful for the clinic and for trials, as these significantly increase AUCs for separating PD from other proteinopathies (to 0.98)¹⁰⁶, and some studies suggest potential for meaningful quantification¹⁰⁸.

Staging systems leveraging diagnostic measures

Two biomarker-based staging systems have recently been proposed that leverage some of these methods. They are motivated by a need to consider PD biologically, rather than as a clinical syndrome, and follow a similar framework to the A/T/N approach established for Alzheimer's¹¹⁸. This is a system that categorizes individuals based on the presence of specific disease-relevant biomarkers, with beta-amyloid (A, measured using CSF or amyloid PET), hyperphosphorylated tau (T, measured using CSF p-tau or tau PET) and neurodegeneration (N, measured by atrophy on structural MRI, FDG PET or CSF total tau).

Both of the proposed systems for staging in PD are based on the concept that pathological alpha-synuclein aggregation is the fundamental pathological process defining PD; that dopaminergic dysfunction in the midbrain is eventually universally present; and move away from purely clinically-defined syndromes.

Simuni et al.⁴ propose a biological definition for neuronal alpha-synuclein disease, a unifying term that encompasses alpha-synucleinopathies such as PD and DLB. They define two key anchors: S, representing neuronal alpha-synuclein, measured using CSF SAA, with two levels S- or S+; and D, representing midbrain dopaminergic neuronal degeneration, detected using DAT-SPECT imaging, defined as D- or D+. They also consider genetic status, particularly for fully penetrant genes such as *SNCA*. They propose a staging system that takes into account these anchors, and starts from Stage 0, preclinical disease, defined by the presence of fully penetrant *SNCA* variants. Stage 1 and beyond all require SAA positivity, and from stage 2B, evidence of dopaminergic dysfunction is also required, although this may be present at Stage 1B. Stage 1 is defined as no clinical signs or symptoms, with subtle signs being present at Stage 2 that are mainly non-motor in nature. Patients in Stages 3-6 show functional impairment "that can be driven by motor, cognitive or other non-motor clinical

signs or symptoms". Each stage from 3 onwards is defined respectively as slight, mild, moderate and severe. Patients currently undergoing clinical trials for neuroprotection at early stages are already in Stage 3, and motor functional impairment is not present before this stage. In effect, this proposed staging system will mean that patients can be identified and staged for recruitment into neuroprotective trials at even earlier stages of PD, including prodromal stages, at Stages 1 and 2 of the system.

Höglinger et al.⁵ similarly consider alpha-synuclein positivity using the alpha-synuclein SAA from CSF or skin biopsies, as well as neurodegeneration, measured with imaging including dopaminergic PET/SPECT. In addition, they consider genetic variants strongly linked with PD, including *SNCA*, *GBA*, and *LRK2*.

Both approaches acknowledge that future versions will likely incorporate novel alpha-synuclein measures, and additional biomarkers reflecting co-pathology. Whilst designed primarily for PD, they allow for prodromal forms and other synucleinopathies such as DLB to be included in their framework. They have the potential to enable better biological characterisation and matching of clinical cohorts across studies, as well as stratification for clinical trials, and more targeted treatments; although they both currently lack a system to track mid and late stage progression.

Biomarkers of disease progression

In addition to aiding precise and early diagnosis of PD, biomarkers of disease progression are needed for monitoring outcomes in the clinic and in clinical trials. Quantitative and sensitive markers of progression have the additional benefit of risk stratification by identifying individuals at risk of faster progression and poorer outcomes. Accurate risk stratification at the time of diagnosis will allow earlier and more targeted selection of patients for treatments and clinical trials.

Imaging markers

Imaging grey matter integrity. Markers of grey matter atrophy, using voxel-based morphometry or cortical thickness analyses have been extensively investigated as potential biomarkers of disease progression in PD, but with conflicting results. Individual PD patients do not often show easily detectable grey matter loss at diagnosis, although at group level, PD patients show cortical thinning and reduced sub-cortical and cerebellar volume compared to controls^{119,120}. They also accumulate more cortical and basal ganglia atrophy than age-matched controls at longitudinal follow-up^{7,119,121}. However, measures of grey matter atrophy are less useful in differentiating between patients with PD with different disease stages or outcomes: some studies have shown greater volume loss within right thalamus, hippocampus and fronto-temporal regions in PD with mild cognitive impairment (MCI) compared to those with intact cognition^{122,123} but this was not replicated in a recent multicentre study that only found differences with behavioural symptoms but not cognition at 1 year follow-up¹²⁴.

High inter-individual variability in cortical atrophy patterns in PD may account for the lack of group differences in case-control statistical approaches. Normative modelling, which maps individual patterns of variation from the expected norm, takes advantage of this heterogeneity and could be a more sensitive approach. In Alzheimer's disease, the number of regions that are outliers in cortical atrophy correlate with poorer cognition and predict future conversion to dementia¹²⁵. In PD and DLB, outlier counts are also higher in patients with dementia, and correlate with worse cognition¹²⁶.

Another potentially more sensitive technique for assessing grey matter is to examine microstructure using diffusion weighted imaging. Higher mean diffusivity within the grey matter is thought to reflect damage to dendrites or cellular membranes causing less restricted water diffusion within the grey matter. In a region-of-interest analysis of the basal ganglia in the PPMI cohort, baseline mean diffusivity of the globus pallidus was associated with worsening motor severity,

cognition, and total symptom burden after 4.5 years' follow-up¹²⁷. In the same cohort, using a different subset of patients and whole brain data with 2 year follow-up, increased mean diffusivity within frontal and temporal cortical regions were associated with faster decline in cognition and faster increases in non-motor, but not motor disease burden¹²⁸.

Imaging white matter integrity, micro- and macro-structure. Cell and animal models of PD suggest that axonal changes precede neuronal loss in PD¹²⁹. Therefore, measures sensitive to changes in white matter tracts are likely to be more sensitive in PD. Indeed, in a recent cross-sectional simulation of progression of structural alterations in 130 patients with PD, Park et al showed earlier white matter changes than cortical thickness loss¹³⁰. Additionally, several studies using diffusion tensor imaging have shown changes in fractional anisotropy or mean diffusivity within the white matter, in the absence of grey matter changes in PD¹³¹⁻¹³³. These white matter alterations increase with worsening cognitive decline, particularly within the corpus callosum¹³⁴, and with motor severity¹³⁵.

However, diffusion tensor imaging results have not been fully consistent: a longitudinal multicentre study of 423 newly diagnosed PD patients followed-up over 3 years showed no correlation of baseline, or longitudinal diffusion tensor-derived measures with subsequent cognitive impairment¹³⁶, with negative results seen in another independent cohort¹³⁷. A potential reason for inconsistency in diffusion tensor-derived metrics is that they cannot accurately model crossing white matter fibres, which represent the majority of white matter tracts in the human brain¹³⁸.

Newer higher-order diffusion models have emerged with more robust estimations for crossing fibres, with promising results in PD. Fixel-based analysis, one such higher-order model, has revealed reduced white matter fibre cross-section in the corpus callosum with disease progression in PD¹³⁹; and widespread reductions are seen at baseline and longitudinally in patients with PD and poor visuo-perceptual function, who are at higher risk of dementia¹⁴⁰. We have recently shown that fibre cross-section is reduced at baseline with up to 19% reduction amongst several white matter tracts in cognitively intact patients with PD who progress to poor outcomes within 3-year follow-up¹⁴¹ (Fig. 2b). Fibre-specific measures of white matter integrity may be useful biomarkers predicting outcomes in PD, although further validation of their ability to track disease progression is needed.

Structural and functional connectivity

Approaches that consider structural and functional connections at network level provide additional metrics and even mechanistic insights, with structural connectivity of particular interest given early white matter involvement. Reductions in structural connectivity are seen in medication-naïve PD patients increasing with disease duration¹⁴² and preserved structural connectivity within frontal and cerebellar regions are associated with slower motor symptom progression¹⁴³. Baseline reductions in structural connectivity also predict progression to poor cognitive outcomes at 3-year follow-up¹⁴¹ and in PD patients with MCI they predict progression to dementia within 5 years¹⁴⁴, with interhemispheric frontal and parietal connections being most implicated in cognitive impairment^{141,144,145}.

Changes in functional connectivity have also been shown using functional MRI, where functional connectivity between two regions is defined as the similarity (commonly Pearson correlation) between the BOLD signal from those regions. Both increases and reductions in resting state functional connectivity have been reported in relation to disease severity in PD, with the most consistent findings including reduced connectivity within the posterior putamen; however there is still variability in results between studies, likely due to divergent design, cohorts, and analytical methods¹⁴⁶. Task-related functional connectivity is also altered in PD. In a recent meta-analysis of 39 studies of functional connectivity during motor tasks, patients with PD

consistently showed reduction in activation of the posterior putamen, cerebellum, primary motor cortex and supplementary motor area, but also increased activation to cortical regions connected to the anterior putamen¹⁴⁷. However, these changes were not correlated with disease motor severity¹⁴⁷.

Functional connectivity of specific brain regions may be more relevant to disease progression: a recent study assessing resting state functional connectivity of the striatum showed differences in spatial connectivity patterns in patients with PD which related to motor severity and levodopa administration¹⁴⁸. In addition, key changes in resting state functional connectivity have emerged across studies relating to cognitive decline in PD. These are an overall increase in local connectivity, with reduction in long-range functional connectivity¹⁴⁹; a temporal shift towards a more segregated functional connectivity^{150,151}; and an uncoupling of functional from underlying structural connectivity in transmodal (higher association) regions¹⁵².

These changes in resting-state and task-induced functional connectivity provide useful mechanistic insights, but may not be as sensitive as white matter imaging in predicting longitudinal progression in PD¹⁴¹. As in the case of grey matter imaging, individual variability may partly underlie this, and recent approaches that leverage individual heterogeneity can predict motor outcomes in PD with potential for individual-level predictions¹⁵³. An alternative explanation for the somewhat heterogeneous functional connectivity changes would be variable compensatory changes in functional connectivity within motor and cognitive networks¹⁵⁴. This is in keeping with the upregulation of cortical areas seen during motor tasks in PD¹⁴⁷. Interindividual variability in compensatory mechanisms which in itself, varies during the course of the disease, could be contributing to the heterogeneity of functional changes in PD as well as to heterogeneity in disease severity. Experiments could be designed to specifically assess compensation: a recent study using an action selection task, showed increased selection-related activation of parietal-premotor cortex in PD with milder clinical phenotype compared to both controls (suggesting upregulation of normal activity) and to PD patients with worse motor disease (suggesting that increased activity in these regions may be related to milder motor symptoms)¹⁵⁵. Quantifying compensatory mechanisms could be an important biomarker to predict clinical progression in PD.

Imaging cerebrovascular effects. Coexistent cerebrovascular disease is increasingly recognised as having a synergistic effect on synucleopathies¹⁵⁶ and multiple studies have assessed imaging markers of cerebrovascular disease in relation to outcomes in PD. White matter hyperintensity burden is higher in patients with PD dementia and MCI compared to patients with intact cognition¹⁵⁷. Total white matter hyperintensity volume at baseline predicts cognitive decline, but not motor progression, after follow-up^{158,159}. Higher burden of cerebral microbleeds¹⁶⁰ and lacunes¹⁶¹ is also associated with poorer cognitive scores. But other markers of cerebrovascular disease such as enlarged perivascular spaces are less explored to-date. Concurrent multimodal assessment of different markers of cerebrovascular disease and how these relate to white matter imaging changes are needed to disentangle the contribution of neurodegenerative and cerebrovascular causes on white matter degeneration in PD.

Cardiac MIBG scintigraphy. Although useful for PD diagnosis, it is less clear whether cardiac MIBG has a role in measuring disease progression. Early studies suggested a relationship with motor severity, especially for the early heart-mediastinum ratio^{32,37}, but other studies have not found this^{162,163}. In part, this may relate to a ceiling effect reached early in the disease course¹⁶⁴. More recently, a relationship with other biomarkers has emerged, with correlation between cardiac MIBG uptake and plasma alpha-synuclein measured using ELISA¹⁶⁵; and phosphorylated alpha-synuclein in autonomic nerve bundles on

skin biopsy¹⁶⁶. Finally, cluster analyses have shown that early MIBG changes and cortical atrophy are linked to a more malignant PD phenotype with more rapid decline¹⁶⁷.

Iron-sensitive imaging. NM-related signal, detected using NM-MRI, has consistently shown decreases over the disease course in PD^{45,168,169}. While this is useful in detecting changes, a key limitation of NM-MRI is its specificity to NM-containing regions, namely the SN and LC, making it relatively insensitive to markers of progression that manifest in other anatomical locations. This is particularly an issue in later disease stages, when the vast majority of NM-containing neurons have degenerated and a ceiling effect is observed.

Tissue iron content in the basal ganglia outside of the SN measured using QSM and R2* in PD is associated with disease duration and severity of motor symptoms⁶⁰ (Fig. 2b), although not as consistently as within the SN¹⁷⁰. Increased magnetic susceptibility in hippocampal and cortical regions is associated with cognitive severity at a single timepoint^{62,171}, and increased susceptibility in the nucleus basalis of Meynert is predictive of poorer cognitive performance after follow-up¹⁷². However, longitudinal studies examining changes in deep brain nuclei QSM over the disease course in PD demonstrate inconsistent results, with increases¹⁷³, decreases^{174,175} and no change¹⁷² all reported. R2*, which is additively sensitive to a wider range of causes of tissue changes, including cell-count, shows more consistent longitudinal increases during PD than QSM^{176,177}, as does free-water imaging of the SN¹⁷⁸. This suggests that tissue microstructure changes over the disease course are likely to include additional factors rather than simply a gross increase in ferritin-bound tissue iron. Such tissue changes may involve alterations in myelination¹⁷⁹, pathological protein accumulation¹⁸⁰, or changes in concentrations of other magnetically-reactive metals such as calcium, copper, and magnesium¹⁷⁹.

CSF and plasma biomarkers

Plasma and CSF biomarkers that can predict or track disease progression are increasingly being developed. Given the core role of alpha-synuclein aggregation in PD, great attention has been paid to quantify alpha-synuclein using varied immunoassays including ELISA, xMAP or electrochemiluminescence. However, total alpha-synuclein in the CSF, does not show a consistent relationship with Lewy body pathology, or change over time in relation to motor or cognitive measures^{8,9}, and quantification metrics of the alpha-synuclein SAA are not yet validated or robust across different laboratories.

In contrast, CSF biomarkers of Alzheimer's pathology at baseline, specifically higher CSF phosphorylated tau (p-tau), lower CSF beta-42 and lower ratio of beta-42 to total tau are all associated with worsening motor and cognitive performance at 3-year follow-up^{181,182}. This is consistent with ex-vivo evidence for coexistent beta-amyloid and tau pathology in PD². Plasma p-tau levels have not been as successful in predicting outcomes in PD, although they correlate well with CSF p-tau levels and are established in Alzheimer's as a biomarker of tau and beta-amyloid pathology: levels of phosphorylated tau at threonine 181 (p-tau181) are not increased in PD patients who progress to dementia^{141,183}. Alternative targets such as p-tau at threonine 217 (p-tau217) have been recently developed, showing better accuracy in Alzheimer's¹⁸⁴ and may be more relevant to poor cognitive outcomes in PD. Importantly, these biomarkers can now be reliably measured in blood.

Neurofilament light chain (NFL), a marker of underlying axonal damage, can be reliably quantified using immunoassay in both CSF and plasma, even at low concentrations, and has been shown across several cohorts to correlate with poorer outcomes: CSF NFL is increased in PD patients with cognitive impairment¹⁸⁵ and higher plasma NFL values predict subsequent MCI and dementia^{141,183,186}. However, NFL is a non-specific marker and rises in relation to white matter lesions across inflammatory and other degenerative conditions^{187,188}, and with age,

which may limit its use in clinical settings where comorbidities cannot be controlled. Plasma NFL has recently shown clinical utility in differentiating PD from atypical Parkinsonian disorders¹⁸⁹.

Some of these markers can be combined in multiplex assays, improving the efficiency of testing, with reduced time and cost, and avoiding the same sample being handled multiple times. For example, the Simoa tetraplex panel provides information on NFL, GFAP, tau and UCHL-1. In a study comparing 29 PD patients and 30 controls, no difference was seen between groups, but a relationship between NFL and GFAP was seen for motor severity¹⁹⁰.

Biomarkers of lysosomal and inflammatory pathology are also starting to be explored in PD as markers of poor outcomes. In the PPMI cohort, amongst patients with sporadic PD (negative for *GBA* mutations), patients with higher CSF glucocerebrosidase-to-sphingomyelin ratios showed faster cognitive decline over 3 years, although the ratio did not show significant longitudinal changes¹⁹¹. In another cross-sectional cohort, reduced glucocerebrosidase activity was seen in patients with more advanced motor disease (together with lower cathepsin D activity) and correlated with worse cognition¹⁹².

Increased inflammatory cytokines including interleukins 6, 4 and 8, C-reactive protein, tumour necrosis factor α , and transforming growth factor levels are correlated with worse motor function, whilst worse cognition is reported with higher levels of interleukins 6, 2, 8, 17, tumour necrosis factor α , and C-reactive protein in several studies¹⁹³. However, significant heterogeneity in study design, measures of inflammatory biomarkers and analysis techniques¹⁹³, makes evaluation of inflammatory markers compared to other markers of progression in PD difficult, and their clinical usefulness remains uncertain.

Clinical markers of progression

Older age, male sex, higher baseline motor scores and worse baseline cognition have all been linked to worse motor and cognitive outcomes at longitudinal follow-up^{141,182,194–196} and have been integrated with sleep disorder, presence of depression and various genetic, imaging or CSF biomarkers into clinical predictive scores^{182,196}. Increasing evidence also implicates visual system dysfunction as an early and predictive biomarker of poor outcomes in PD. The visual system is affected in PD, with worse performance of affected individuals across a range of visual tasks: spatial navigation, line orientation, visual rotation, facial recognition, contrast sensitivity, colour vision¹⁹⁷. In a large-scale population study, patients with PD showed worse visual acuity than age-matched unaffected individuals¹⁹⁸. Changes within the visual system in PD are even seen in the retina, with loss of dopamine in post-mortem and animal models¹⁹⁹ and several studies showing thinning of ganglion cell and inner plexiform layers in PD patients²⁰⁰; with thinning correlating with poorer cognitive scores^{201,202}, and seen years before clinical presentation²⁰³, albeit with small effect sizes. Although the whole of the visual system is affected in PD, higher-order visual tasks, rather than retinal changes, are stronger predictors of poor outcomes²⁰² and, in predictive cross-sectional modelling, these occur before retinal changes in the sequence of progression to PD dementia²⁰⁴.

Biomarkers for precise treatment

With novel treatments emerging for neurodegenerative diseases that target specific protein aggregations, stratification of subgroups enriched for specific targets will be crucial for clinical trials of novel interventions; and biomarkers revealing evidence of target engagement will be needed as outcome measures in these trials. Ultimately, disease modifying treatment in PD is likely to be individualised, with patients receiving the combination of therapies that best-matches their precise pathological phenotype, targeting not only alpha-synuclein but potentially amyloid and tau co-pathology, cerebrovascular disease or neuroinflammation. Similar pathological processes with prominent alpha-synuclein accumulation underlie PD, PD dementia and DLB making them challenging to distinguish using

biomarkers. Instead, they should be considered a spectrum of disorders, separated by the timing of the onset of dementia. They can each involve differing combinations of underlying pathological processes. Robust and accurate measurements of these underlying pathological accumulations will be a critical part of applying new interventions in PD, as well as other synucleinopathies.

Measures of underlying alpha-synuclein pathology

As the characteristic neuropathological hallmark of PD, alpha-synuclein would be expected to be detected in almost all patients. However, its presence alone does not explain progression to dementia, and total measured CSF alpha-synuclein does not correlate with disease progression or severity^{8,9}. Serum and plasma alpha-synuclein immunoassays are even less successful, with reports of both higher, lower and no difference in PD compared to controls²⁰⁵. This overlap between groups may be due to alpha-synuclein occurring as multiple different species, and therefore greater precision in identifying relevant species may provide greater sensitivity. CSF oligomeric alpha-synuclein and oligomeric/total alpha-synuclein ratio, for example is increased in PD and associated with worsening motor signs, in particular in postural-instability PD subtypes²⁰⁶ although specificity remains low²⁰⁵. Higher oligomeric but not total-alpha-synuclein was also seen in the serum of PD patients compared to Alzheimer's and controls in a small study²⁰⁷. Phosphorylated alpha-synuclein is similarly increased in plasma and red blood cells of PD patients compared to controls²⁰⁵. The alpha-synuclein SAAs, using CSF or more accessible fluids such as plasma, are likely to be important in stratification for trials and for targeted treatment, with potential for monitoring disease progression as quantification metrics become more reproducible.

An important gap in current efforts to image pathological protein accumulation in PD is the lack of a validated PET radiotracer for alpha-synuclein, restricting both diagnosis and assessment of disease progression. Development of potential tracers is hindered by several factors. The absolute concentration of alpha-synuclein aggregates within the brain is far lower than that of tau or beta-amyloid, and alpha-synuclein aggregates are often co-localised with tau and beta-amyloid aggregates, which are structurally similar due to the high number of beta-pleated sheets. This means any alpha-synuclein tracer would need to have extremely high affinity and selectivity for its target. Additionally, alpha-synuclein inclusions are intracellular, meaning tracers would need to cross both the blood brain barrier and the cell membrane. Finally, a lack of reliable assays restricts evaluation of candidate compounds²⁰⁸. Despite this, there have been promising recent reports of the use of a PET tracer, 18F-F0502B, with high selectivity for alpha-synuclein over tau and beta-amyloid, and displaying preferential uptake in non-human primate models of PD over controls²⁰⁹. Another potential PET tracer, 18F-ACI-12589, also with high alpha-synuclein selectivity, was recently tested in-vivo in humans, and was able to successfully discriminate MSA from controls, Alzheimer's and progressive supranuclear palsy, although displayed relatively low uptake in PD²¹⁰. If successful, an alpha-synuclein tracer will be transformative in the field. Firstly, it may reveal differing patterns of alpha-synuclein deposition that could aid in the differential diagnosis of PD, PDD and DLB, as well as atypical Parkinsonian syndromes, potentially in combination with amyloid and tau PET. Secondly, it would allow in-vivo modelling of the spatiotemporal dynamics of alpha-synuclein spread in synucleinopathies in large cohorts, how this varies between individuals, and how it is associated with the development of different symptoms. Finally, it could provide evidence of target engagement in clinical trials, in the way that beta-amyloid tracers showed unequivocally for anti-amyloid therapies¹⁷.

Measures of beta-amyloid and tau co-existent pathology

The development of [¹¹C] Pittsburgh compound-B (11C-PIB) combined with PET imaging allowed visualisation of beta-amyloid in vivo, and

beta-amyloid PET positivity is now a key biomarker for Alzheimer's disease, with other radioligands including [18 F] florbetapen and [18 F] florbetapir since developed. Early studies of 11C-PIB uptake in PD suggested very little difference from controls^{211,212}, and in a meta-analysis, while high beta-amyloid positivity was seen for DLB and PDD compared to controls, the positivity rate in PD was actually lower than in controls²¹³. Interestingly, however, several studies have reported associations between increased beta-amyloid burden and poorer cognitive ability in PD, even at very low levels^{214,215}, with increased beta-amyloid burden in PD predicting poorer future cognition²¹⁶ and more rapid progression²¹⁷.

In-vivo PET imaging of tau binding is also possible using radiotracers such as 18F-AV-1451, 18F-FDDNP and, more recently, 18F-PI-2620 and 18F-MK-6240. A recent meta-analysis found minimal differences between controls and PD²¹⁸, and while some studies have reported decreased tau uptake in the SN in PD, this is due to off-target binding of NM by 18F-AV-1451. Second generation tau tracers do not display this off-target binding²¹⁹, and when combined across studies, increased tau uptake is found in inferior temporal cortex in PD compared with controls, entorhinal cortex in PD with cognitive impairment compared with PD without cognitive impairment, and inferior temporal and occipital cortex in DLB compared to PD²¹⁸. Like beta-amyloid, tau burden is associated with cognitive severity in PD^{220,221}.

Plasma biomarkers of amyloid and tau pathology are becoming established in clinical practice and show excellent correlation with PET markers and underlying Alzheimer's pathology¹². In a small study of PD patients, fluid markers have also been concordant with PET imaging²²², although their sensitivity and specificity in identifying underlying beta-amyloid and tau pathology in PD has not yet been shown. Combined methods that characterise both beta-amyloid and alpha-synuclein aggregates in the same sample may be even more informative. For example, optical single-molecule imaging methods can assess aggregates with intermolecular beta-sheet structure and in a small number of PD patients showed increased large aggregates compared with controls, with higher alpha-synuclein and lower beta-amyloid²²³.

It is perhaps counter-intuitive that the incidence and extent of beta-amyloid and tau pathology on imaging is so low in PD, when a large proportion of PDD patients show both tau and amyloid pathology at postmortem¹⁸¹. However, there is evidence that alpha-synuclein may inhibit beta-amyloid plaque formation²²⁴, while aggregated forms of alpha-synuclein promote the aggregation of tau²²⁵. Taken together, this suggests that beta-amyloid formation is inhibited in PD, but that tau aggregation, and its associated cognitive deficits²²⁶, are present even in PD patients with low beta-amyloid burden^{220,221}. There may be a shorter asymptomatic phase of beta-amyloid deposition in pre-symptomatic PD dementia compared to Alzheimer's disease²²⁷, which could include a relatively accelerated deposition of beta-amyloid and tau.

Markers of mitochondrial sub-types of PD

It is well-established that mitochondrial dysfunction contributes to the pathogenesis of PD: cell models of PD include those using mitochondrial toxins²²⁸; exposure to environmental agents that target mitochondrial respiratory chain complexes is linked to higher risk of PD²²⁹; mitochondrial genetic defects are seen in post-mortem PD brain²³⁰ and variations in pathways controlling mitochondrial DNA increase the risk of PD. Subgroups of *LRRK2* mutations (one of the commonest genetic forms of PD) are thought to relate to mitochondrial dysfunction²³¹. However, previous assays required live cells to detect mitochondrial dysfunction, which are not easily accessible in clinical practice or scalable for large cohorts. Recently, Qi et al described a blood-based marker of mitochondrial dysfunction using a PCR-based assay²³². The authors were able to show increased damage to mitochondrial DNA in both idiopathic PD and in patients carrying the *G2019S LRRK2* mutation, and found that mitochondrial DNA damage correlated with

LRRK2 kinase activity. This approach could allow precision-targeting of PD subtypes with mitochondrial dysfunction, as well as monitoring of *LRRK2* inhibition in trials.

Imaging of underlying neuroinflammation

Alongside plasma and CSF biomarkers of inflammation, which are not specific to central inflammation, PET imaging of neuroinflammatory processes may offer another avenue for stratification for treatment targeting inflammation, or monitoring disease in PD²³³. For example, the radioligand 11 C(R)PK11195 binds to the mitochondrial translocator protein, TSPO, which is preferentially expressed in activated microglia. Increased 11 C(R)PK11195 binding has been reported in temporal, parietal and occipital cortices in PD compared to controls, and even more strongly in PDD; the extent of binding shows an inverse relationship with cognitive ability²³⁴. However, other studies have found increased binding in the SN and putamen, but no differences in cortical regions between PD and controls²³⁵. Increased 11 C(R)PK11195 binding may even relate to preserved cognition and white matter integrity in PD²³⁶. However, the assumption that TSPO PET accurately reflects activated microglia in humans is based on the observation of increased TSPO expression seen in mice²³⁷. A recent meta-analysis of mouse datasets confirmed TSPO upregulation in mice, but increased TSPO expression was not seen in human macrophages²³⁸. The authors then directly examined TSPO expression in a range of mammalian species and showed that while it is increased in rodents, it is not increased in other mammals, including humans. They further compared microglial TSPO expression in neurodegenerative and neuroinflammatory diseases (Alzheimer's, amyotrophic lateral sclerosis and multiple sclerosis), and in mouse models of these conditions. They showed that although TSPO expression increased significantly in the mouse models of each condition, it did not increase in humans. The authors suggest that their findings transform how TSPO PET signal is interpreted, and that it reflects local cell density than microglial activation in humans. As such, further work is needed to validate PET imaging of neuroinflammation as a biomarker in PD.

Future directions

Multimodal approaches within the brain and beyond

Methods that combine information across modalities may be more sensitive in tracking disease progression and predicting outcomes. These include approaches that combine information from both structural and functional organisational brain changes. The relationship between structural and functional connectivity is altered in PD, with less close coupling, particularly in PD patients at higher risk of dementia¹⁵² and those with visual hallucinations²³⁹. Another technique relates the effects of structural brain network alterations to brain function using network control theory. This approach can predict electrophysiological neural responses after stimulation in patients with epilepsy²⁴⁰ and response to non-invasive transcranial magnetic stimulation during language tasks²⁴¹. In PD, recent applications of this framework provided insights into mechanisms of visual hallucinations²⁴². Such approaches have potential to discriminate disease stages, predict progression as well as effects of brain activation on brain function²⁴³.

Multimodal information from wider body systems may even have promise as markers of disease progression in PD. In a large recent longitudinal study of healthy aging, in over 1200 healthy people using seven body and three brain modalities, Tian et al showed that body aging phenotypes, particularly of the cardiovascular system, preceded brain measures²⁴⁴. Attempts to incorporate information from other body systems in PD diagnosis are already underway. For example, for the cardiovascular system, MIBG is now an established supportive biomarker of PD. More recently data from the gastrointestinal system have been used. Work examining changes in the gut microbiome in PD patients revealed a potentially different microbiome structure

compared to controls, with depletion of bacteria important for short-chain fatty-acid production²⁴⁵, and two bacteria species possibly related to disease severity²⁴⁶. As PD is a multisystem disorder with a long prodromal phase affecting autonomic, bowel, bladder and other systems, incorporating multimodal information from the wider body could be particularly informative for earlier diagnosis, and as a measure of progression.

MRI quantification for information about tissue structure

Multiparametric quantitative MRI provides parametric maps of biophysical tissue properties that reflect the microenvironment of the underlying tissue, aspiring to provide “in vivo histology” of the human brain. Of these multiparametric maps, R2* is already relatively well-established^{59,124,176,177}. However, other sequences provide additional information likely to be relevant in PD. The longitudinal relaxation rate (R1), which is sensitive to myelin, water content, and iron concentration is organised in specific spatial gradients in the human putamen and caudate and changes with age²⁴⁷, but has not yet been examined in PD. In addition to assessing individual signals (i.e. R1, R2*, etc), regional changes in the ratio of these signals may also provide useful information. In 99 patients with early-stage PD from the PPMI dataset, decreased T1w/T2w in the posterior putamen correlated with contralateral motor signs and uptake on DAT scans²⁴⁷. More recently the R1/R2* relaxivity ratio has been proposed as a signal sensitive to iron mobilisation capacity within human brain tissue, rather than just iron deposition²⁴⁸. This may have particular application at very early stages of PD when iron homeostasis could be affected prior to significant iron deposition.

Magnetic resonance electrical properties tomography (also called quantitative conductivity mapping) can be used to measure tissue conductivity and permittivity, reflecting changes in the ionic content of tissues. Although still in development and with several algorithms proposed, this approach has been recently successfully applied to brain tumours²⁴⁹ and in Alzheimer’s shows a relationship with cognitive severity²⁵⁰ and may have applications in PD.

Recently, PET ligands binding to synaptic vesicle 2a (SV2a), such as IIC-UCB-J, have been developed, providing a in vivo marker of synaptic density, thought to be a key pathomechanism in PD and PD progression to dementia²⁵¹. Studies of IIC-UCB-J show promising early results, and have been successfully used to demonstrate synaptic reduction in preclinical models of PD²⁵².

Increasing signal-to-noise: Ultra-high field MRI

Technical advancements in neuroimaging leading to ultra-high field MRI, increase signal-to-noise ratios providing improved spatial resolution and tissue contrast. Ultra-high field MRI at 7 Tesla (7T) can identify small subcortical structures such as the SN sub-regions and locus coeruleus^{253,254}. Although study numbers to date are low, 7T-derived measures of the SN show potential as a biomarker in PD: loss of the SN’s normally smooth lateral boundary is seen in PD²⁵⁵, with reduction in SN volume²⁵⁶ and lower fractional anisotropy²⁵⁷ compared to controls. SN volume was also related to longer disease duration and worse motor severity in a small cohort of 32 patients²⁵⁶.

Locus coeruleus integrity is also affected in PD, with increased neuromelanin signal in 78 patients with early PD compared to 36 controls²⁵⁸ whilst reduced T1 signal relates to worse cognition²⁵⁹. 7T MRI of these small subcortical structures and their connectivity patterns could potentially be useful in diagnosis, risk stratifying and predicting response to treatment, particularly response to DBS where 7T MRI has already been applied²⁶⁰. The sub-millimetre spatial resolution of ultra-high field MRI allows it to detect changes in microstructure and functional activity within cortical layers²⁶¹. This may have application in PD, based on animal models that suggest early and preferential involvement of deep layers²⁶². In the coming years, spatial resolution of MRI will likely reach 200µm with 11.7T and 14T MRI

currently under development^{263,264}; this will allow the visualisation of myelination differences within the cerebellum and the true separation of BOLD signals from all cortical layers as well as the quantification of brain metabolites including GABA via MR spectroscopy²⁶⁴. This will open new avenues for research into mechanistic understanding of PD and potential new treatment targets.

Conclusions

We are moving from a conception of PD as a clinical syndrome, and instead adopting frameworks with precise biological definitions. It is important to note that most biomarker studies to date are based on analyses of group information, rather than at the level of the individual patient. How they can be applied at the individual level, as part of proposed staging systems and in clinical practice, will require further validation and investigation. Interpretation of both imaging and fluid biomarkers needs to be done in the context of clinical information; and for these to be useful in wider settings, we will require standardisation of sample collection and analysis pipelines and in many cases consideration of cost. Despite these challenges, by providing information about underlying pathological processes, imaging and fluid biomarkers bring important opportunities to enhance diagnostic frameworks, enable early diagnosis, capture underlying causes of heterogeneity of PD, and quantitatively track disease progression. In this way they ultimately provide hope for new interventions to slow the progression of PD.

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Author contributions

A.Z. designed and drafted the work. G.T. designed and drafted the work. H.Z. drafted the work. R.S.W. Conceived and designed the work; drafted the work; All authors approved the submitted version and agree to be personally accountable for their contributions.

Competing interests

R.W. has received speaking and writing honoraria respectively from GE Healthcare and Britannia, and consultancy fees from Therakind, and is local PI for an EIP neflamapimod trial. H.Z. has served at scientific advisory boards and/or as a consultant for Abbvie, Acumen, Alecor, Alzinova, ALZPath, Annexon, Apellis, Artery Therapeutics, AZTherapies, Cognito Therapeutics, CogRx, Denali, Eisai, Merry Life, Nervgen, Novo Nordisk, Optoceutics, Passage Bio, Pinteon Therapeutics, Prothena, Red Abbey Labs, reMYND, Roche, Samumed, Siemens Healthineers, Triplet Therapeutics, and Wave, has given lectures in symposia sponsored by Alzecure, Biogen, Celectricon, Fujirebio, Lilly, and Roche, and is a co-founder of Brain Biomarker Solutions in Gothenburg AB (BBS), which is a

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Additional information

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