

1 *Review*

2 **High-throughput direct mass spectrometry-based** 3 **metabolomics to characterize metabolite fingerprints** 4 **associated with Alzheimer's disease pathogenesis**

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16 **Abstract:** Direct mass spectrometry-based metabolomics has been widely employed in the last
17 years to characterize metabolic alterations underlying to Alzheimer's disease development and
18 progression. This high-throughput approach presents a great potential for fast and simultaneous
19 fingerprinting of a vast number of metabolites, which can be applied to multiple biological samples
20 such as serum/plasma, urine, cerebrospinal fluid and tissues. In this review article we present the
21 main advantages and drawbacks of metabolomics based on direct mass spectrometry compared
22 with conventional analytical techniques, and provide a comprehensive revision of the literature on
23 the application of these tools in Alzheimer's disease research.

24 **Keywords:** Metabolomics, direct mass spectrometry, Alzheimer's disease, pathogenesis,
25 biomarkers
26

27 **1. The potential of direct mass spectrometry-based metabolomics**

28 Metabolomics requires the use of powerful and versatile analytical techniques with the aim of
29 covering the largest number of compounds comprising the great complexity of the metabolome,
30 composed of metabolites with diverse molecular weights, polarities, acid-base properties, and other
31 physicochemical characteristics. To this end, multiple metabolomic platforms have been proposed in
32 the literature, including nuclear magnetic resonance (NMR), or mass spectrometry (MS) coupled to
33 liquid chromatography (LC), to gas chromatography (GC), or to capillary electrophoresis (CE), each
34 one of them having their own strengths and weakness. For this reason, the combination of several of
35 these complementary techniques is emerging in the last years as the most suitable strategy to
36 accomplish a comprehensive characterization of the metabolome [1–3]. Among these analytical
37 tools, direct mass spectrometry (DMS)-based metabolomics has been usually relegated to the
38 background due to its inherent drawbacks, such as the impossibility of resolving chemical isomers
39 and problems associated with ion suppression due to the direct introduction of the whole sample
40 into the mass spectrometer without a previous chromatographic or electrophoretic separation.
41 However, some recently published review articles have also highlighted the great potentials of this
42 metabolomic approach [4,5]. The most notable advantage of this tool is its high-throughput
43 screening capability due to the absence of a previous time-consuming separation step, which
44 considerably reduces the total analysis time, thus allowing the analysis of hundreds of samples per
45 day. The elimination of this chromatographic/electrophoretic separation also prevents the

46 introduction of biased and selective retention mechanisms, so that DMS can be used for the
47 simultaneous determination of a huge number of metabolites, covering a wide physicochemical
48 space. In this sense, it should be also noted that multiple instrumental configurations are available
49 for performing DMS-based metabolomics, which can be combined to increase the metabolome
50 coverage. For non-targeted metabolomics, direct infusion mass spectrometry (DIMS) is the simplest
51 approach since only needs a syringe pump to constantly introduce the sample extract into the mass
52 spectrometer. Complementarily, the sample can also be delivered by flow injection (FIMS) as a plug
53 into a stream of solvent delivered by a LC pump. On the other hand, the multi-dimensional mass
54 spectrometry-based shotgun lipidomic (MDMS-SL) approach developed by Han *et al.* allows the
55 direct quantitation of hundreds of individual lipid species by means of a selective ionization of
56 certain category of lipid classes at certain MS conditions [6]. In this context, simpler targeted
57 metabolomic platforms are the AbsoluteIDQ™ kits developed by Biocrates Life Sciences AG
58 (Innsbruck, Austria), focused on the FI-MS/MS-based quantification of multiple metabolite classes,
59 including lipids (phospholipids, sphingolipids, acyl-carnitines, glycerolipids), amino acids, hexoses
60 and biogenic amines [7]. In turn, most of these DMS-based configurations can be coupled with
61 various complementary atmospheric pressure ionization sources. Electrospray ionization (ESI) is the
62 most commonly employed source in non-targeted metabolomics, applicable for the simultaneous
63 detection of compounds with very diverse polarities and molecular weights due to its sensitivity and
64 versatility. Complementarily, atmospheric pressure chemical ionization (APCI) and atmospheric
65 pressure photoionization (APPI) sources can also be employed for the ionization of less polar
66 compounds. Thus, the combination of complementary ion sources and ionization modes (i.e.
67 positive and negative polarities), is recommended in order to maximize the metabolome coverage.
68 To conclude, it is also worth noting that the lack of a separation step prior to MS detection facilitates
69 the experimental design by avoiding common troubles associated with chromatography and
70 electrophoresis, such as column/capillary clogging and deterioration, the need of complex data
71 processing packages to align retention/migration times, as well as the minimization of the
72 instrumental drift along batch analysis thanks to the reduced acquisition times usually employed in
73 these approaches.

74 2. Alzheimer's disease, mild cognitive impairment and animal models

75 Alzheimer's disease (AD) is the most common neurodegenerative disorder worldwide in the
76 elderly, which is primarily characterized by neuropathological alterations associated with the
77 deposition of amyloid β plaques and the formation of intra-neuronal neurofibrillary tangles of
78 hyperphosphorylated tau protein. Anyway, numerous authors have proposed that other multiple
79 pathological processes can also play a pivotal role in the development of this disease, such as
80 oxidative stress, mitochondrial dysfunction, neuroinflammatory mechanisms, abnormal metal
81 homeostasis and many others [8–10]. The investigation of AD etiology involves a great challenge to
82 the scientific community due to its great complexity and variability of clinical symptoms, its long
83 pre-symptomatic period, and the impossibility of studying brain microscopic changes until the final
84 stages of the disease. For these reasons, diagnosis of AD nowadays relies on the combination of
85 various physical, neuropsychological and laboratory tests according to the clinical criteria of the
86 National Institute of Neurological and Communicative Disorders and the Alzheimer's Disease and
87 Related Disorders Association (NINCDS-ADRDA) [11]. However, this diagnostic method is only
88 effective at advanced stages of disease, which hinders the application of pharmacological
89 interventions, and in addition suffers of low specificity against other dementias as demonstrated
90 after *post mortem* histopathological verification [12]. Thus, the discovery of novel biomarkers for
91 accurate diagnosis of AD is mandatory, especially for predicting the development of disease from
92 pre-dementia phases, also known as mild cognitive impairment (MCI). MCI is a heterogeneous
93 syndrome characterized by very mild symptoms of cognitive dysfunction, which is usually
94 considered an intermediate stage in the development of Alzheimer's disease from normal aging.
95 Although MCI shares many features with early AD, current data suggest that some patients may
96 have a benign form of MCI as part of the normal aging process [13]. Therefore, there is a great need

97 to discover potential biomarkers for diagnosis and to investigate the pathological mechanisms
98 associated with AD and MCI development and progression.
99

100 On the other hand, animal models are very useful tools for investigating the pathogenesis of
101 AD and associated alterations in the central nervous system at different stages along the progression
102 of disease [14], while studies in human cohorts are limited to *post-mortem* brain tissue, when disease
103 is in its final stage. Transgenic mice, obtained by the over-expression of mutated forms of human
104 genes associated with AD such as the amyloid precursor protein (APP), presenilin 1 (PS1), presenilin
105 2 (PS2) or apolipoprotein E (ApoE), are the most useful models since the neuropathology elicited by
106 these animals is analogous to that observed in human AD, and furthermore biochemical routes in
107 humans and rodents are very similar [15]. The most commonly transgenic mice employed in AD
108 research are based on the over-expression of mutated forms of the APP, including the APP_{Tg2576},
109 APP_{V717F} and CRND8 transgenic lines, which usually show amyloid deposition in hippocampus and
110 cortex and memory deficits, but not neuronal loss. In this vein, it has been demonstrated that the
111 co-expression of mutated PS1, and to a lesser extent PS2, accelerates amyloid deposition, thus
112 facilitating the appearance of the characteristic AD phenotype (APP×PS1, TASTPM). Taking into
113 account that ε4 allele of the ApoE is the strongest risk factor for AD, several knock-in mice in which
114 this protein is expressed have been developed, which show significant cognitive and synaptic
115 plasticity impairments. On the contrary, only a few transgenic models expressing tauopathy have
116 been developed to date due to the ignorance of genes involved in this process in AD (TAPP, 3×Tg).

117 3. Application of direct mass spectrometry-based metabolomics to AD research

118 Considering the multifactorial nature of AD etiology, the application of holistic metabolomic
119 approaches has emerged in recent years for the investigation of pathological mechanisms
120 underlying to this neurodegenerative disorder and for the identification of novel diagnostic
121 biomarkers. In particular, DMS-based metabolomics has demonstrated a great potential to
122 characterize the AD metabolome in a comprehensive manner, as discussed in this section.
123

124 Numerous non-targeted DMS-based metabolomic studies have been conducted in serum
125 samples, which is a very useful biofluid in the clinical practice for the identification of diagnostic
126 biomarkers in a non-invasive manner. González-Domínguez *et al.* employed a DIMS platform based
127 on a two-step treatment of serum samples from AD patients to obtain a holistic snapshot of
128 metabolite alterations associated to the early development of this neurodegenerative disorder
129 [16,17]. The most notable findings could be associated with an abnormal homeostasis of neural
130 membrane lipids, evidenced by reduced levels of circulating phospholipids containing
131 polyunsaturated fatty acids (PUFAs) and increased content of lipid species composed of saturated
132 fatty acids (SFAs) and some breakdown products (e.g. choline, glycerophosphocholine).
133 Furthermore, significant impairments were also observed in biological pathways related to energy
134 metabolism, neurotransmitter levels and fatty acid homeostasis. To complement this study, a
135 FI-APPI-MS approach was subsequently applied to focus on the less polar metabolome, non-readily
136 detectable by ESI-based metabolomics [18]. Increased serum levels of diacylglycerols and ceramides
137 were detected in AD patients, indicative of up-regulated degradation of membrane phospholipids
138 and sphingolipids by the action of phospholipases and sphingomyelinases, in line with results from
139 DIMS analyses. Due to the central role that lipid dyshomeostasis seems to play in AD pathogenesis,
140 serum samples from the same cohort of AD patients were subjected to DIMS-based lipidomics using
141 a modification of the Bligh-Dyer extraction method [19]. Again, it was observed a reduced content of
142 PUFA-containing phospholipids and increased levels of diacylglycerols, corroborating previous
143 hypotheses. Furthermore, changes in other low molecular weight metabolites also evidenced severe
144 impairments in the homeostasis of various neurotransmitter systems, nitrogen metabolism and
145 oxidative stress. Taking into account this evidence about the major role that phospholipids play in
146 AD etiology, a metabolomic multiplatform based on the combination of DIMS and LC-MS, this later
147 coupled to both molecular (ESI) and elemental (inductively coupled plasma, ICP) mass spectrometry

148 was employed to get a deeper understanding about the AD-associated phospholipidome [20]. Thus,
149 results evidenced that multiple factors are involved in this abnormal phospholipid homeostasis,
150 including the imbalance of PUFA/SFA contained in their structure, the over-activation of
151 phospholipases, the implication of oxidative stress and peroxysomal dysfunction, among others.
152 Complementarily, González-Domínguez *et al.* also employed the DIMS and FI-APPI-MS approaches
153 previously described to investigate the AD-like pathology in various transgenic mice models. The
154 analysis of serum samples from APP×PS1 mice revealed analogous metabolomic disturbances to
155 those detected in previous studies with human cohorts, demonstrating the potential of these
156 transgenic animals to model AD [21]. Additionally, DIMS-based fingerprinting has been also
157 applied to the APP×PS1×IL4-KO transgenic model with the aim of investigating the role of
158 inflammation induced by means of interleukin-4 depletion in AD pathology [22]. Alterations in
159 serum levels of eicosanoids, amino acids and related compounds, and metabolites involved in the
160 urea cycle demonstrated that depletion of interleukin-4 might potentiate AD pathology in the
161 APP×PS1 model. It should be noted that all these results obtained by DMS analysis were
162 subsequently validated by applying various orthogonal metabolomic techniques, including LC-MS,
163 GC-MS and CE-MS [23–26], thus demonstrating the potential of MS-fingerprinting approaches to
164 carry out fast and accurate screening of complex metabolic networks.
165

166 Other published studies on DMS-based metabolomics have focused on the characterization of
167 metabolic impairments observed in brain from various transgenic mice models, a tissue of great
168 interest in AD research since allows the *in situ* investigation of neuropathological processes
169 associated with this neurodegenerative disorder. Lin *et al.* applied an optimized DIMS platform to
170 look for characteristic metabolic impairments in hippocampus [27] and cerebellum [28] from the
171 CRND8 mouse model. Major findings were observed with regards to an abnormal metabolism of
172 amino acids and nucleotides, as well as the over-production of eicosanoids. In this line, DIMS-based
173 analysis of various brain regions from the APP×PS1 mouse model (i.e. hippocampus, cortex,
174 cerebellum, striatum, and olfactory bulbs) evidenced that hippocampus and cortex are the most
175 affected areas by AD pathology [29]. Similarly to previous studies, significant differences were
176 observed in levels of phospholipids, acyl-carnitines, fatty acids, nucleotides, amino acids and many
177 other metabolites, results which were then confirmed by LC/GC-MS metabolomic analysis [30].
178 Recently, Wood *et al.* also employed a lipidomic approach based on DIMS to define potential
179 biomarkers with the aim of distinguishing healthy controls from MCI and AD patients [31]. They
180 analyzed frontal cortex grey, white matter and cerebrospinal fluid (CSF), an interesting biofluid that
181 directly reflects the brain metabolic production, and detected abnormal levels of various lipid classes
182 (e.g. plasmalogens, phosphatidylethanolamines, diglycerides), in agreement with previous studies.
183 Alternatively, other peripheral organs from the APP×PS1 model have also been investigated to
184 assess the possible systemic nature of AD, including the liver, kidneys, spleen and thymus [32]. In
185 this work, authors found significant impairments associated with oxidative stress, lipid
186 dyshomeostasis and imbalances in energy metabolism, among other processes, results which were
187 subsequently validated by using a metabolomic multiplatform based on the combination of LC and
188 GC coupled to MS [33,34]. Moreover, urine can also serve as a valid biological sample to study
189 metabolomic perturbations associated with AD by using DIMS-based approaches, as demonstrated
190 by González-Domínguez *et al.* [35]. For this purpose, various sample preparation methods and
191 normalization strategies were tested, evidencing that ten-fold dilution of urine prior to
192 MS-fingerprinting and subsequent statistical data normalization is enough to minimize ion
193 suppression and to correct the inherent inter-individual variability of this matrix, respectively.
194

195 From a targeted perspective, the MDMS-SL platform optimized by Han *et al.* is a very
196 interesting alternative for the comprehensive investigation of lipidomic alterations associated with
197 AD, in samples coming from both human and animal models. The application of this tool to plasma
198 and brain samples showed significant changes in the levels of plasmalogens [36], sulfatides [37–39],
199 ceramides [37,40] and sphingomyelins [40], thus corroborating the pivotal role of lipid metabolism

200 in pathogenesis of AD. On the other hand, other authors proposed the use of AbsoluteIDQ™ kits to
201 analyze blood, brain and CSF samples from AD and MCI patients, observing major changes in the
202 content of phospholipids and acyl-carnitines [41–46]. However, it should be noted that this tool
203 present a great drawback regarding its low metabolome coverage.

204 4. Conclusions

205 Metabolomic approaches based on DMS analysis are gaining great importance in the last years
206 because of their high-throughput screening potential, reduced analysis time and wide metabolome
207 coverage. Particularly, these platforms have been widely applied for studying complex and
208 multifactorial disorders such as Alzheimer's disease, with the aim of elucidating pathological
209 mechanisms underlying to disease development and progression and discovering potential
210 diagnostic biomarkers. The analysis of multiple biological samples, including serum/plasma, urine,
211 brain (hippocampus, cortex, cerebellum, etc.), cerebrospinal fluid and other organs (liver, kidney,
212 spleen, thymus), has enabled obtaining a comprehensive snapshot of the major metabolic hallmarks
213 associated with this neurodegenerative disorder, such as impairments in the homeostasis of
214 membrane lipids, oxidative stress, inflammatory processes, imbalance in energy metabolism and
215 neurotransmitter metabolism, among many others.

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