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5	Corresponding Author	Family Name	Lupo
6		Particle	
7		Given Name	Giuseppe
8		Suffix	
9		Organization	Sapienza University of Rome
10		Division	Department of Chemistry
11		Address	Piazzale A. Moro, Rome 00185, Italy
12		e-mail	giuseppe.lupo@uniroma1.it
13	Author	Family Name	Gaetani
14		Particle	
15		Given Name	Silvana
16		Suffix	
17		Organization	Sapienza University of Rome
18		Division	Department of Physiology and Farmacology “V. Erspamer”
19		Address	Piazzale A. Moro, Rome 00185, Italy
20		e-mail	
21	Author	Family Name	Cacci
22		Particle	
23		Given Name	Emanuele
24		Suffix	
25		Organization	Sapienza University of Rome
26		Division	Department of Biology and Biotechnology “C. Darwin”
27		Address	Piazzale A. Moro, Rome 00185, Italy
28		e-mail	
29	Author	Family Name	Biagioni
30		Particle	

31		Given Name	Stefano
32		Suffix	
33		Organization	Sapienza University of Rome
34		Division	Department of Biology and Biotechnology “C. Darwin”
35		Address	Piazzale A. Moro, Rome 00185, Italy
36		e-mail	
<hr/>			
37		Family Name	Negri
38		Particle	
39		Given Name	Rodolfo
40	Author	Suffix	
41		Organization	Sapienza University of Rome
42		Division	Department of Biology and Biotechnology “C. Darwin”
43		Address	Piazzale A. Moro, Rome 00185, Italy
44		e-mail	
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Molecular Signatures of the Aging Brain: Finding the Links Between Genes and Phenotypes

Giuseppe Lupo¹ · Silvana Gaetani² · Emanuele Cacci³ · Stefano Biagioni³ · Rodolfo Negri³

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Abstract

Aging is associated with cognitive decline and increased vulnerability to neurodegenerative diseases. The progressive extension of the average human lifespan is bound to lead to a corresponding increase in the fraction of cognitively impaired elderly individuals among the human population, with an enormous societal and economic burden. At the cellular and tissue levels, cognitive decline is linked to a reduction in specific neuronal subpopulations, a widespread decrease in synaptic plasticity and an increase in neuroinflammation due to an enhanced activation of astrocytes and microglia, but the molecular mechanisms underlying these functional changes during normal aging and in neuropathological conditions remain poorly understood. In this review, we summarize very recent and outstanding progress in elucidating the molecular changes associated with cognitive decline through the genome-wide profiling of aging brain cells at different molecular levels (genomic, epigenomic, transcriptomic, proteomic). We discuss how the correlation of different molecular and phenotypic traits driven by mathematical and computational analyses of large datasets has led to the prediction of key molecular nodes of neurodegenerative pathways, and provide a few examples of candidate regulators of cognitive decline identified with these approaches. Furthermore, we highlight the dysregulation of the synaptic transcriptome in neuronal cells and of the inflammatory transcriptome in glial cells as some of the key events during normal and neuropathological human brain aging.

Key Words Brain · Aging · Cognitive decline · Gene expression · DNA methylation · Histone acetylation

Introduction

In the mammalian brain, aging is associated with progressive cognitive decline and with enhanced susceptibility to late onset neurodegenerative diseases. The cognitive deficits linked to normal and pathological aging are becoming widespread in the human population, mainly due to its increasing average life expectancy. Consequently, the economic and societal

burden of age-related neurological conditions is increasingly growing and difficult to sustain. This has prompted extensive efforts to investigate the mechanisms underlying the functional impairment of the aged brain, to develop novel therapeutic approaches able to slow down the cognitive decline of the elderly population and to prevent/treat age-related neuropathologies.

At the cellular level, aging is not linked to an overall reduction in the amount of pre-existing neurons in the central nervous system, although certain brain structures and neuronal subtypes may be more vulnerable than others and selectively prone to cell loss [1, 2]. For example, age-related neuronal death has been described in specific subpopulations of the cerebral cortex and the cerebellum [3, 4]. The strongest effects of aging, however, appear to be exerted at the level of synaptic activity, as shown by the age-associated decline in the number of synapses and in synaptic plasticity in several brain regions [5–8]. Moreover, a landmark of the aged brain is an increased level of neuroinflammation, which can contribute to the alterations in neuronal/synaptic function [9]. All these deficits worsen in age-related neurodegenerative conditions, such as

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✉ Giuseppe Lupo
giuseppe.lupo@uniroma1.it

¹ Department of Chemistry, Sapienza University of Rome, Piazzale A. Moro, 00185 Rome, Italy

² Department of Physiology and Pharmacology “V. Erspamer”, Sapienza University of Rome, Piazzale A. Moro, 00185 Rome, Italy

³ Department of Biology and Biotechnology “C. Darwin”, Sapienza University of Rome, Piazzale A. Moro, 00185 Rome, Italy

55 late onset Alzheimer's disease (AD), although there are also
56 peculiar differences between normal and neuropathological
57 aging [2, 9, 10]. Finally, aging has a major impact on the
58 generation of new neurons in the adult brain. At least in ro-
59 dents, in which adult neurogenesis has been best character-
60 ized, a sharp age-dependent drop in the production of new
61 neurons is detectable at the level of both of the main adult
62 neurogenic niches, the subventricular zone (SVZ), flanking
63 the lateral ventricles, and the hippocampal subgranular zone
64 (SGZ) [11, 12]. Thus, aging can hinder the remodelling of
65 neural circuits by affecting synaptic connectivity and the inte-
66 gration of new neuronal elements, leading to a progressive
67 decline in the functional plasticity of the brain and in cognitive
68 abilities.

69 Very recently, several studies have investigated the molec-
70 ular mechanisms underlying the functional changes taking
71 place in the aging brain, by exploiting genome-wide next-
72 generation sequencing (NGS) technologies to dissect the
73 age-related molecular signatures of different brain cell types.
74 In this review, we summarize the main experimental ap-
75 proaches that have been employed so far and the common
76 themes emerging from the complexity of these high-
77 throughput analyses. We focus on cell type-specific, region-
78 specific, and age-specific misregulation of genes related to
79 neuronal/synaptic activity, immune response/inflammation,
80 autophagy/lysosome function, cell death/stress regulation,
81 and cell signalling as some of the molecular events implicated
82 in normal and pathological brain cell aging. Furthermore, we
83 discuss how the integrated analyses of several NGS datasets is
84 starting to shed light on the genetic/epigenetic mechanisms
85 underpinning these molecular alterations and the causative
86 links between age-related changes in gene activity and cellular
87 phenotypes.

88 **Molecular Profiling of Aged Brain Cells: 89 **Insights from Genome-Wide Transcriptomic 90 **and Epigenomic Studies******

91 Over the last 15 years, several transcriptomic datasets from
92 human and rodent samples of different ages and brain regions
93 have been reported. In the cases of human studies, the cohorts
94 of individuals used for these analyses have often included
95 subjects with clinical diagnosis of AD at the time of sampling,
96 thus providing insights into the mRNA changes associated
97 with both normal and neuropathological aging. Several
98 genome-wide analyses of age-related epigenetic modifica-
99 tions, especially DNA methylation and histone acetylation,
100 have also been described, although the available epigenomic
101 datasets are still limited, when compared with transcriptomic
102 resources. Recently, the shift from microarray-based to NGS-
103 based approaches and the processing of larger numbers of
104 samples have enhanced the power and resolution of these

105 studies. This has shifted the focus from detecting
106 transcriptional/epigenetic alterations in a few genes or gene
107 categories to deciphering coordinated changes in broad geno-
108 mic regions influencing multiple biological processes, al-
109 though also bringing alternative splicing and moderately
110 expressed non-coding RNAs into the picture. Another critical
111 breakthrough made possible by the improved technological
112 platforms is the gradual transition from bulk tissue to cell
113 type-specific datasets, unveiling more subtle, but functionally
114 relevant, changes in gene activities. Crucially, these investiga-
115 tions now routinely include a detailed phenotypic character-
116 ization of the tissue samples used for the molecular analyses,
117 and a long-term cognitive profile of the tissue donors, thus
118 allowing to correlate transcriptomic and epigenomic changes
119 with cellular and cognitive alterations. The frontier of this
120 approach is represented by single-cell NGS analyses. On the
121 one hand, single-cell transcriptomics can reveal gene signa-
122 tures associated with aging in distinct neural stem/progenitor
123 cell (NSPC) subpopulations [13]. On the other hand, single-
124 neuron whole genome sequencing has been recently used to
125 analyze the accumulation of somatic mutations with aging in
126 humans, a phenomenon named *genosenium*, which shows
127 age-related, region-related, and disease-related molecular sig-
128 natures [14].

129 Altogether, this work has already provided extensive evi-
130 dence that both normal and neuropathological aging, as well
131 as the associated functional modifications of brain cells, cor-
132 relate with broad changes at the transcriptomic and
133 epigenomic levels. In spite of the large variability among dif-
134 ferent studies due to significant differences in the sourced
135 samples and in the experimental approaches used, consistent
136 patterns of age-related gene regulation are starting to emerge,
137 raising the hopes of eventually explaining the age-dependent
138 cognitive decline at the molecular level. Yet, a clear associa-
139 tion between a transcriptional/epigenetic trait and a phenotyp-
140 ic trait does not provide a causal relationship between them,
141 because the molecular change could lie upstream or down-
142 stream of the associated phenotype. Another critical and par-
143 tially overlapping issue is how many of the observed gene
144 expression and epigenetic changes associated with age are
145 actually related to the intrinsic genotypic variation and how
146 many can be rather ascribed to extrinsic non-genetic influ-
147 ences. Clearly, functional studies in appropriate experimental
148 settings (e.g., gain- and loss-of-function approaches *in vitro*
149 and *in vivo*) are mandatory to prove the causative roles of key
150 genes and molecular pathways in cognitive decline. Given the
151 huge number of molecular changes correlating with age, how-
152 ever, a systematic functional characterization of genes/
153 pathways needs to be preceded by strategies to select the can-
154 didates most likely to act upstream of cognitive impairment. In
155 the next section, we will highlight recent high-throughput
156 studies in human subjects showing how it is possible to inte-
157 grate several large-scale datasets (genomic, transcriptomic,

158 epigenomic, proteomic, phenotypic) to make predictions
159 about the molecular nodes driving cognitive decline. In the
160 following sections, we will focus on the key age-related mo-
161 lecular changes revealed in specific cell types and regions of
162 the human brain. Recent studies using mouse models to ana-
163 lyze the genome-wide transcriptomic and epigenomic modifi-
164 cations occurring in NSPCs of the aging neurogenic niches
165 have been reviewed elsewhere [12].

166 **Harnessing High-Throughput Human Brain Datasets** 167 **to Identify Candidate Regulatory Pathways** 168 **of Cognitive Decline**

169 In 4 related studies, Mostafavi, Tasaki, Klein, De Jager,
170 Bennett, and co-authors [15–18] have recently used different
171 bioinformatic pipelines to analyze genome-wide
172 transcriptomic [15], DNA methylation [19], and histone H3
173 lysine 9 acetylation (H3K9ac) [18] datasets from dorsolateral
174 prefrontal cortex (DLPFC) postmortem samples of longitudi-
175 nal cohorts of human subjects of different ages. Donors had
176 different degrees of cognitive decline and neuropathological
177 traits, such as amyloid beta (A β) levels, hyperphosphorylated
178 Tau (pTau) levels, and clinical diagnosis of AD, which had
179 been characterized at the time of tissue sampling. These mo-
180 lecular and phenotypic datasets were also superimposed on
181 known genomic maps of single nucleotide polymorphisms
182 (SNPs) associated with cognitive decline and/or AD in avail-
183 able genome-wide association studies (GWAS).

184 In one study [15], the above described datasets were ana-
185 lyzed using the module-trait network (MTN) method, mining
186 them for modules of co-expressed genes that were then ranked
187 based on their association with cognitive decline and neuro-
188 pathological AD-related traits (e.g., A β and/or pTau burden).
189 This analysis pointed to a module of 390 co-expressed genes
190 (named module 109) as the one most strongly associated with
191 cognitive impairment. Genes within this module were then
192 prioritized to identify the most likely candidates upstream of
193 cognitive-related phenotypes on the basis of several param-
194 eters, including gene network connectivity, gene-phenotype as-
195 sociation, and expression levels in astrocyte and neuron
196 *in vitro* cultures. Twelve of the top hits in this module
197 (Table 1) were selected to analyze the levels of their protein
198 products in DLPFC samples by means of selection reaction
199 motif (SRM) quantitative proteomics, revealing associations
200 between protein levels and cognitive decline for several of
201 them [20]. Two of the top hits, the semaphorin receptor-
202 encoding gene *PLXNB1* and *INPPL1*, a gene involved in the
203 modulation of insulin/insulin growth factor (IGF) signalling
204 pathways, were further validated by functional assays. Both
205 genes were found to be expressed at the protein level in astro-
206 cytes of the frontal cortex in AD patients, and their knock-
207 down in primary astrocyte *in vitro* cultures reduced the extra-
208 cellular levels of A β 42, confirming a role upstream of

neuropathological phenotypic traits [15]. Further functional
work will be needed to assess the false positive rate of the
MTN approach and to reveal roles of the candidate genes in
aspects of cognitive decline other than A β 42 accumulation.

Another study [16] employed the above described multi-
omic (mRNA, DNA methylation, H3K9ac, SNPs) datasets to
perform a global correlation analysis among transcription, epi-
genetic marks, and genomic variants aimed at mining genes,
whose expression levels were significantly associated with
SNPs and with changes in either DNA methylation or
H3K9ac patterns. The molecular data were then superimposed
on phenotypic datasets to define, for each of the genes meeting
the above criteria, a local regulatory network (LRN),
predicting causal relationships (i.e., an upstream, downstream,
or independent hierarchical level in the network) among
mRNA levels, epigenetic modifications, and phenotypes.
This pipeline allowed the identification of a subset of genes
predicted to act upstream of cognitive decline, which were
further prioritized based on additional parameters, including
the association with cognitive deficits/AD according to inde-
pendent GWAS analyses. Notably, the top hits resulting from
this work (Table 1) included *SEMA3F*, encoding for a
semaphorin family member that has been implicated in the
regulation of PI3K and AKT kinases, key intracellular trans-
ducers of the insulin/IGF signalling pathway [16, 21]. No
functional experiments, however, were performed to validate
the identified candidate genes. In future work, it would be
important to implement this approach by including additional
omic datasets (e.g., genome-wide profiling of miRNAs and
additional epigenetic modifications) and by extending single
gene LRNs to multi-gene networks featuring the interactions
between different genes. Furthermore, for both MTN and
LRN approaches, follow-up studies will benefit from the anal-
ysis of other brain regions and of additional cohorts of subjects
with different ancestry composition and education levels.

A third study [17] took advantage of available GWAS
datasets providing 457 SNPs correlated to AD, to define a
polygenic risk score associated with AD (AD-PRS) for
each individual in longitudinal cohorts of aging human
subjects genotyped for the AD-related SNPs. To evaluate
AD-PRS correlations with specific phenotypic and molecu-
lar traits, this analysis was then integrated with pheno-
typic profiles, the abovementioned multi-omic datasets and
SRM proteomic quantification in DLPFC samples for a
group of proteins potentially linked to AD. As expected,
a high AD-PRS correlated with cognitive decline and neu-
ropathological phenotypes (e.g., A β and pTau burden).
Remarkably, this pipeline also revealed the association of
AD-PRS with distinct molecular features, including the
expression levels of the IGFBP5 protein, a key regulator
of insulin/IGF signalling encoded by one of the candidate
genes in the above described, cognitive-related, gene mod-
ule (module 109) [15, 17].

Table 1 Candidate regulators of cognitive decline identified through genome-wide molecular profiling of human brain cells. Representative candidate genes associated with cognitive decline identified through module-trait network, local regulatory network, or key driver analyses. For the genes presented in bold, the association with neuropathological traits (e.g., increased A β and/or pTau levels) was further validated by means of functional assays in cell cultures (see text for further details and references)

Module-trait network analysis	Local regulatory network analysis	Key driver analysis
<i>AK4 (adenylate kinase 4)</i>	<i>CIQTNF4 (C1q and TNF related 4)</i>	<i>CCT5 (chaperonin containing TCPI subunit 5)</i>
<i>ANKRD40 (ankyrin repeat domain 40)</i>	<i>DHRS11 (dehydrogenase/reductase 11)</i>	<i>COMT (catechol-O-methyltransferase)</i>
<i>BCL2L1 (BCL2 like 1)</i>	<i>NPM3 (nucleophosmin/nucleoplasmin 3)</i>	<i>GNA12 (G protein subunit alpha 12)</i>
<i>FBXO2 (F-box protein 2)</i>	<i>NUPR1 (nuclear protein 1, transcriptional regulator)</i>	<i>HSPA2 (heat shock protein family A member 2)</i>
<i>HSPB2 (heat shock protein family B member 2)</i>	<i>RABEP2 (rabaptin, RAB GTPase binding effector protein 2)</i>	<i>PDHB (pyruvate dehydrogenase E1 beta subunit)</i>
<i>IGFBP5 (insulin like growth factor binding protein 5)</i>	<i>SCG3 (secretogranin III)</i>	<i>RGS4 (regulator of G protein signaling 4)</i>
<i>INPPL1 (inositol polyphosphate phosphatase like 1)</i>	<i>SEMA3F (semaphorin 3F)</i>	<i>ST18 (C2H2C-type zinc finger transcription factor)</i>
<i>ITPK1 (inositol-tetrakisphosphate 1-kinase)</i>	<i>SLC22A23 (solute carrier family 22 member 23)</i>	
<i>KIF5B (kinesin family member 5B)</i>	<i>STAU1 (staufen double-stranded RNA binding protein 1)</i>	
<i>PLXNB1 (plexin B1)</i>	<i>TRIOBP (TRIO and F-actin binding protein)</i>	
<i>SASH1 (SAM and SH3 domain containing 1)</i>		
<i>SLC6A12 (solute carrier family 6 member 12)</i>		
<i>VAT1 (vesicle amine transport 1)</i>		

Building up on the same multi-omic datasets and especially focusing on H3K9ac profiling, a fourth study [18] mapped the age-associated genome-wide changes in H3K9ac deposition in the human DLPFC and compared them with transcriptional and phenotypic patterns, identifying genomic segments in which H3K9ac distribution was linked with transcriptional activity and with Tau-related phenotypic traits. The authors then screened drug repurposing databases to mine for compounds associated with gene expression signatures that were predicted to negatively correlate with the transcriptional profiles of Tau-related H3K9ac domains. This led to the identification and functional validation of a small molecule inhibitor of the heat shock protein HSP90, which could attenuate some of the effects of Tau overexpression in neuronal *in vitro* cultures [18]. Notably, the gene encoding for another heat shock protein, HSPB2, was part of the cognitive-related gene module (module 109) identified by MTN analysis [15]. Moreover, increased levels of HSPB2 were shown to be associated with cognitive impairment by proteomic assays in DLPFC samples [20]. Remarkably, a different study also disclosed a heat shock protein, HSPA2, as a candidate key driver of AD-related neuropathology [22]. This conclusion was reached by means of a pipeline that included the identification of modules of differentially co-expressed mRNAs and proteins by transcriptomic and proteomic comparison of frontal/temporal cortex samples from control and AD-affected subjects, followed by network

analysis to predict causal interactions between different targets (mRNAs and proteins) within modules. Key driver analysis was then used to mine for the targets with the higher number of connections, which were further prioritized based on the association of mRNA expression changes with SNPs. HSPA2 emerged as one of the top hits from this approach (Table 1) and was functionally validated by showing that its overexpression could significantly increase both A β and Tau/pTau levels using Tau-expressing H4 *in vitro* cell cultures, an observation consistent with its elevated levels in AD subjects [22].

Altogether, the studies discussed above provide an important proof of principle that different high-throughput datasets, by means of carefully crafted pipelines integrating several layers of information, may be successfully employed to predict genuine molecular drivers of cognitive impairment, even though an association between certain molecular and phenotypic traits does not formally imply a causal relationship between the associated molecules and phenotypes. Importantly, these studies also suggest that different bioinformatic pipelines can single out candidate regulators of cognitive decline that converge on shared molecular pathways (e.g., INPPL1, IGFBP5, and SEMA3F, all implicated in insulin/IGF signaling). This can significantly increase the chance of pinpointing the key pathways upstream of neuropathological events, thus facilitating the identification of optimal targets for therapeutic approaches. Although the above described studies represent

314 an initial step in this direction, with several limitations related
 315 to sample type and size, to the depth of omic datasets and to
 316 downstream functional validation, they suggest potentially
 317 important roles for the molecular pathways mediated by
 318 semaphorins/plexins, insulin/IGFs, and HSPs in the etiology
 319 of cognitive decline during normal and/or pathological aging
 320 (Fig. 1).

321 It is worth noting that AKT, a critical node in the insulin/
 322 IGF pathway, can regulate the activity of the GSK3 β kinase,
 323 which plays a key role in Wnt/ β -catenin-dependent signal
 324 transduction. Previous work has shown that Wnt/ β -catenin-
 325 dependent signalling can promote the nuclear accumulation
 326 of REST, a master transcriptional regulator of neuronal differ-
 327 entiation genes that also plays a protective function in cogni-
 328 tive decline. Nuclear REST levels in neurons of the human
 329 PFC and hippocampus increase during aging, but this effect is
 330 largely lost in cognitively impaired subjects. Functional
 331 *in vitro* and *in vivo* experiments showed that REST acts as a
 332 repressor of genes involved in cell death, A β production, and
 333 Tau phosphorylation and that lack of REST function increases
 334 neuronal sensitivity to oxidative stress and A β and accelerates
 335 age-dependent neurodegeneration in the brain of knockout
 336 mice [23]. Based on these results, it is tempting to speculate
 337 that changes in the levels of proteins implicated in the regula-
 338 tion of AKT-dependent signalling and associated with cogni-
 339 tive decline (such as INPPL1, IGFBP5, SEMA3F) may be

functionally linked, at least in part, to neuropathological phe- 340
 341 notypes through alterations in REST activity. 341

342 The observations described above also suggest that neuro- 342
 343 pathological aging is not simply an accelerated/enhanced ver- 343
 344 sion of normal aging but involves the dysregulation of molec- 344
 345 ular pathways that play a protective role in healthy aging cells. 345
 346 This hypothesis has received strong support by a recent study 346
 347 that employed lateral temporal lobe samples from younger 347
 348 non-AD, elder non-AD, and elder AD subjects for genome- 348
 349 wide profiling of histone H4 lysine 16 acetylation (H4K16ac) 349
 350 [24]. Three-way comparison of H4K16ac peaks in these 350
 351 groups of samples revealed three classes of changes: i) age- 351
 352 regulated changes (present both in elder non-AD and in elder 352
 353 AD samples, as compared to younger samples); ii) age- 353
 354 dysregulated changes (present only in elder non-AD samples, 354
 355 as compared to younger samples, but not in the elder AD 355
 356 group); iii) disease-specific changes (distinguishing AD sam- 356
 357 ples from both younger and elder non-AD samples). 357
 358 Furthermore, H4K16ac changes between younger and elder 358
 359 non-AD samples negatively correlated with the changes ob- 359
 360 served between elder non-AD and AD samples. Parallel 360
 361 transcriptomic analyses showed a positive correlation between 361
 362 H4K16ac peak levels and the mRNA levels of nearby genes, 362
 363 indicating that the observed differences in H4K16ac among 363
 364 the three groups of samples are likely to be associated with 364
 365 gene expression changes. Genes with decreased H4K16ac 365
 366 peaks in elder non-AD samples, as compared with younger 366
 367 samples, were enriched for REST binding motifs, thus 367
 368 supporting a role of REST in the regulation of aging-related 368
 369 transcriptional networks. Moreover, genes related to cell death 369
 370 regulation were the most significantly enriched category 370
 371 among age-dysregulated genes, in agreement with the pro- 371
 372 posed misregulation of cell death-related REST targets in neu- 372
 373 ropathological aging [24]. All together, these data indicate that 373
 374 some of the epigenetic and gene expression changes associ- 374
 375 ated with normal brain aging are reversed during pathological 375
 376 aging, and that neuropathological conditions are also associ- 376
 377 ated with specific changes that distinguish them from non- 377
 378 pathological aging. Pipelines designed to prioritize candidate 378
 379 molecular pathways for functional assays among the genes 379
 380 associated with age-dysregulated and disease-specific changes 380
 381 are expected to provide key insights into the molecular mech- 381
 382 anisms of cognitive decline and unveil new potential targets 382
 383 for better therapeutic approaches. 383

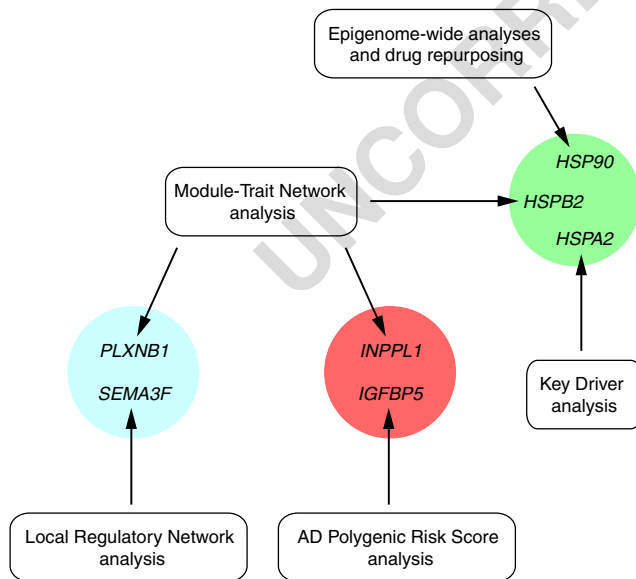


Fig. 1 Identification of candidate regulators of cognitive decline through genome-wide molecular profiling of human brain cells. Schematic description of recent bioinformatic pipelines used to prioritize candidate genes with potential causative roles in cognitive decline, and of some of the top hits resulting from these approaches. Colored circles highlight genes identified with different strategies that are implicated in similar molecular pathways (pale blue, semaphorin/plexin signalling; red, insulin/IGF signalling; green, heat shock protein networks). See text for further details

Age-Related Transcriptomic and Epigenomic Changes in Human Brain Neuronal Cells Include the Dysregulation of the Gene Networks Implicated in Synaptic Function 384
 385
 386
 387

388 The studies described in the previous section reported an as- 388
 389 sociation between the downregulation of gene categories re- 389
 390 lated to neuronal/synaptic function and cognitive decline. For 390

391 example, these categories were enriched in the modules of co-
392 expressed genes showing a negative correlation between ex-
393 pression levels and cognitive impairment according to MTN
394 analysis [15]. Moreover, according to LRN analysis, the genes
395 predicted to be upstream of cognitive processes showed en-
396 richment for categories linked to neurons and myelin [16].
397 Similar enrichment of gene categories related to neuronal
398 function among the genes downregulated with age has been
399 reported in other transcriptomic analyses of bulk tissue from
400 human brain samples [25]. These observations may reflect the
401 association of cognitive decline with the depletion of neuronal
402 cell populations, with the dysregulation of gene expression
403 programmes in neuronal cells, or with both these effects. To
404 address this issue, several studies have specifically investigat-
405 ed the neuronal transcriptome in brain samples from aged or
406 cognitively impaired individuals using different technical
407 approaches.

408 To gain insight into the age-related gene expression chang-
409 es that specifically take place in neurons, recent investigations
410 have superimposed transcriptomic datasets of bulk brain sam-
411 ples from human subjects of different ages on available tran-
412 scriptional signatures of seven purified murine cell types [3].
413 Notably, human samples were collected from ten different
414 brain regions to characterize both region-dependent and age-
415 associated transcriptional changes. This analysis showed that
416 the genes differentially expressed between younger and elder
417 samples could be classified as region-selective (i.e., the chang-
418 es were specific for one or a few regions) or multi-regional
419 (changes detectable in most or all of the tested regions). When
420 the differentially expressed genes were also categorized based
421 on their association with a specific cell type, neuronal-specific
422 genes showing age-related changes resulted enriched for
423 region-selective genes and were often downregulated with
424 age. Nonetheless, when the analyzed samples were clustered
425 on the basis of the expression levels of the neuronal genes,
426 they could be mainly discriminated based on regional charac-
427 ter rather than age, indicating that region-dependent mecha-
428 nisms have a stronger influence than age-related factors on the
429 overall transcriptional profiles of neuronal cells. These data
430 extend to human brain cells, and on a genome-wide scale,
431 previous results obtained in mouse neural progenitors suggest-
432 ing that genes controlling positional identity, such as *Hox*
433 genes, tend to maintain regional transcriptional/epigenetic pat-
434 terns throughout life [26]. The same study quantified the num-
435 ber of NeuN-positive neurons in sections of the frontal cere-
436 bral cortex from a subset of the brain samples used for the
437 transcriptomic assays. This analysis showed a selective de-
438 crease in neuronal cells with large cell bodies, which account
439 for roughly 20% of the analyzed neurons, although the use of
440 a low sample number and of a single neuronal marker repre-
441 sent possible caveats in these assays. Furthermore, the neuro-
442 nal genes showing age-related downregulation were enriched
443 for specific categories related to synaptic function, ion

transport, and cell death regulation [3]. These results suggest
that brain aging is associated with a decrease in the expression
levels of a subset of the neuronal transcriptome. This may be
due to the combination of a loss of neuronal cells among
selectively vulnerable neuronal subpopulations with the dys-
regulation of the molecular networks implicated in neuronal/
synaptic activity.

Another study has superimposed transcriptomic analyses
of frontal cortex samples from human subjects of different
ages on a single-cell transcriptome of the mouse visual cortex
to extrapolate neuronal cell type-specific transcriptional sig-
natures in the human brain datasets and evaluate the extent of
their age-related changes [27]. Although neuronal genes were
generally downregulated with age, the changes were consid-
ered to be significant only for three subtype-specific gene
signatures, namely those associated with *Ptgs2*-expressing,
Arf5-expressing, and *Batf3*-expressing glutamatergic neurons
of cortical layers 2–5. Additional subtype-specific signatures
showed significant changes that were limited to selected gene
categories. In particular, age-dependent downregulation was
detected for genes related to synaptic transmission and cell-
cell signalling that were also associated with *Sst/Cdk6*-ex-
pressing and/or *Vip/Mybpcc*-expressing inhibitory neurons.
Finally, the authors found many genes that showed an age-
related decrease in their expression levels without being asso-
ciated with any particular cell type. Notably, these genes were
enriched in categories related to neuronal function, such as
synaptic transmission and ion membrane transport [27].
Taken together, these results support the idea that certain neu-
ronal subtypes may be especially vulnerable to age-dependent
cell loss, although the global neuronal population is largely
preserved in the aged brain. Furthermore, they suggest that
aging is associated with a dysregulation of the gene expression
programmes regulating synaptic activity, which may have a
stronger impact on specific subtypes, but is likely to affect the
neuronal lineage in general. Nonetheless, the use of mouse
transcriptomic datasets to identify human neuron-specific ex-
pression signatures represents a possible limitation of the
above described studies.

Further evidence for the association of cognitive decline
with the altered regulation of the synapse-related tran-
scriptome is provided by a previous analysis of the expression
levels of 340 synaptic genes in 4 different brain regions of
younger non-AD, elder non-AD, and elder AD human sub-
jects [28]. Many of the analyzed genes were found to be dif-
ferentially expressed in elder non-AD samples in comparison
with younger samples, but also when comparing age-matched
AD and non-AD samples, with most of the changes consisting
in an age-associated or AD-associated downregulation.
Notably, aging and AD mainly targeted different brain re-
gions, as the majority of the expression changes in the non-
AD younger-elder comparison were observed in cortical re-
gions, whereas most of the changes between AD and non-AD

497 samples were detected in the hippocampus. A linear regres- 550
498 sion analysis across all sample groups (younger, elder, AD), 551
499 however, revealed several genes showing progressive age- 552
500 related and AD-related changes (usually a downregulation) 553
501 in both cortical and hippocampal samples (i.e., the change 554
502 detected in elder samples was further enhanced in AD sam- 555
503 ples). The genes affected by aging and AD were enriched for 556
504 key categories related to synaptic function, such as vesicle 557
505 trafficking, neurotransmitter receptors, synaptic stability/scaf- 558
506 folding, and neuromodulatory systems [28]. These results 559
507 present some limitations, because the analyzed genes may 560
508 not be entirely neuron-specific and/or synapse-specific, al- 561
509 though transcriptional changes may not necessarily result in 562
510 altered protein levels. Nevertheless, they suggest that both 563
511 normal aging and neuropathological conditions may impinge 564
512 on the synaptic transcriptome causing an extensive downreg- 565
513 ulation of genes implicated in key synaptic processes. The 566
514 effects of aging and AD on the synaptic transcriptome display 567
515 regional specificity, although a subset of the synaptic genes is 568
516 vulnerable to the converging action of age-related and AD- 569
517 dependent mechanisms. It is worth noting that, according to a 570
518 follow-up study comparing the transcriptome of the same 4 571
519 brain regions in cognitively normal, mildly cognitively im- 572
520 paired (MCI) and AD elderly subjects, genes enriched for 573
521 categories related to synaptic function, proteostasis, and mito- 574
522 chondrial metabolism were differentially expressed in MCI 575
523 samples compared to both normal and AD samples. In partic- 576
524 ular, synaptic genes were often downregulated in AD com- 577
525 pared to normal samples, but they were frequently upregulated 578
526 in MCI samples [29]. Furthermore, a recent transcriptomic 579
527 and phosphoproteomic profiling of temporal cortex samples 580
528 from patients at different AD-related pathological stages has 581
529 revealed fluctuating mRNA/protein levels (down-up-down 582
530 patterns) of genes linked to synaptic and mitochondrial func- 583
531 tion during early phases of disease [30]. These observations 584
532 suggest that the gene expression changes associated with cog- 585
533 nitive decline, such as those in the synaptic transcriptome, 586
534 follow, at least in part, non-monotonic trajectories, with trend 587
535 reversals during early neuropathological stages. These obser- 588
536 vations may be critical to develop therapies intercepting cog- 589
537 nitive impairment at its onset.

538 Further insight into the transcriptional patterns associated 588
539 with brain aging has been gained through studies that com- 589
540 pared the transcriptomic changes taking place in human brain 590
541 samples during different temporal windows of the lifespan, 591
542 such as fetal development, postnatal development, and aging 592
543 [31, 32]. These analyses showed that synapse-related genes 593
544 were enriched among the genes showing increases in their 594
545 expression levels during both fetal and early postnatal stages, 595
546 as expected from the extensive development of neural circuit- 596
547 ries during both stages [32]. In contrast, a reversal in the tra- 597
548 jectory of gene expression changes was detected during brain 598
549 aging, because gene categories related to neuronal/synaptic 599
600

601 function were enriched for genes upregulated during postnatal 602
602 development and downregulated during aging [31]. Genes 603
603 related to AD also tended to show up-down trends when their 604
604 expression changes during fetal/postnatal development and 605
605 during aging were compared [31, 32]. Although these data 606
606 are limited to the PFC [32], or the analysis did not investigate 607
607 region-specific changes when samples from several brain re- 608
608 gions were included [31], they suggest the intriguing possibil- 609
609 ity that the alterations in the neuronal/synaptic transcriptome 610
610 observed in the aged brain result, at least in part, from the 611
611 dysregulation of the molecular networks regulating neural 612
612 development.

613 Epigenomic analyses are starting to reveal the epigenetic 614
614 mechanisms that underlie these age-related and disease- 615
615 dependant transcriptional changes, pointing to a prominent 616
616 role of histone acetylation. For example, in the above de- 617
617 scribed genome-wide profiling of H4K16ac deposition in the 618
618 temporal cortex from younger non-AD, elder non-AD and 619
619 elder AD subjects, the groups of genes showing age- 620
620 regulated changes and disease-specific changes in the levels 621
621 of H4K16ac peaks were enriched for categories related to 622
622 neurons and synapses [24]. These data are consistent with 623
623 the aforementioned investigation of the synaptic tran- 624
624 scriptome, showing that both age and AD affect the expres- 625
625 sion of synaptic genes, with partially collaborative, but also 626
626 distinct effects [28]. Furthermore, transcriptomic and genome- 627
627 wide profiling of histone H3 lysine 27 acetylation (H3K27ac) 628
628 found that, in PFC samples from human subjects of different 629
629 ages, the genes downregulated in aged samples also showed 630
630 reduced levels of H3K27ac at their promoters and were 631
631 enriched for categories related to neuronal and synaptic func- 632
632 tion. Notably, these patterns were observed also in the aged 633
633 mouse brain and could be reversed by *in vivo* treatments with 634
634 histone deacetylase inhibitors, such as suberanilohydroxamic 635
635 acid (SAHA), providing an initial evidence that the changes in 636
636 histone acetylation may play a causative role in the age- 637
637 associated dysregulation of the neuronal/synaptic tran- 638
638 scriptome [25].

639 Alterations in DNA methylation may also be implicated in 639
639 these transcriptional changes. By profiling DNA methylation 640
640 in sorted NeuN-positive (neuronal) and NeuN-negative (non- 641
641 neuronal) nuclei from the frontal/temporal cortex of non-AD 642
642 and AD human subjects of different ages, a recent study de- 643
643 scribed several hundred CpGs showing age-related or AD- 644
644 related methylation changes specifically in neurons, with near- 645
645 by genes including categories linked to neuronal processes 646
646 [33]. Although this work did not provide transcriptomic data 647
647 to directly correlate DNA methylation changes with gene ex- 648
648 pression changes, studies comparing PFC samples from 649
649 young and aged rats, or from age-matched cognitively un- 650
650 impaired and impaired rats, demonstrated a correlation be- 651
651 tween changes in DNA methylation and mRNA levels in 652
652 genes implicated in neuronal/synaptic function [34, 35]. 653
653

603 The epigenomic changes in brain cells related to cognitive
604 decline may also affect mRNA splicing besides expression
605 levels, as shown by a recent genome-wide mapping of splic-
606 ing variants in DLPFC samples from AD and non-AD human
607 subjects [36], which revealed a large number of genes under-
608 going alternative splicing, and several genes in which splicing
609 differences were associated with AD neuropathological traits.
610 This study also found a correlation between SNPs associated
611 with alternatively spliced genes and SNPs linked to DNA
612 methylation/H3K9ac changes, and between splicing-related
613 SNPs and known AD-associated SNP susceptibility variants,
614 suggesting that some of the AD-related alternative splicing
615 may be regulated at the epigenetic level. To identify more
616 precisely genes showing differential splicing or transcription
617 levels that are associated with AD and mediated by genetic
618 variation, the authors performed transcriptome-wide associa-
619 tion studies (TWAS). Notably, the genes emerging from
620 TWAS analysis were predicted to be interconnected, at the
621 protein level, with other known AD susceptibility genes,
622 forming a AD-related network that is enriched for categories
623 linked to endocytosis, autophagy, and lysosome function [36].
624 This work, however, does not distinguish RNA changes oc-
625 ccurring upstream or downstream of neuropathological traits or
626 representing indirect effects due to pathological alterations in
627 other brain regions. A previous transcriptomic analysis of
628 frontal cortex and cerebellum tissue samples and of laser-
629 capture microdissected Purkinje neurons from human subjects
630 of different ages found a few age-related changes that were
631 shared between bulk tissues and Purkinje neurons, including
632 genes implicated in endosomal/lysosomal functions [37]. This
633 suggests that at least some of the age-associated or disease-
634 dependent mRNA changes related to autophagic/lysosomal
635 processes can take place in neuronal cell types.

636 Given the complexity of the epigenetic regulatory mecha-
637 nisms acting in neural cells [38], the studies described above
638 are only starting to shed light on the interactions between
639 normal and neuropathological aging processes and the neuro-
640 nal epigenome, but we can expect that this will be a major
641 avenue for future investigations.

642 **Age-Related Transcriptomic and Epigenomic Changes** 643 **in Human Brain Glial Cells Include the Dysregulation** 644 **of the Gene Networks Implicated in Inflammatory** 645 **Responses**

646 Transcriptomic datasets of bulk tissue samples from the hu-
647 man PFC show that, opposite to the age-related downregula-
648 tion of genes implicated in neuronal/synaptic function, aging
649 and cognitive decline are associated with an increase in the
650 expression levels of genes enriched for categories linked to
651 immune functions and inflammation [15, 25]. Remarkably,
652 the genes upregulated during PFC aging, including immune/
653 inflammation-related categories, were recently shown to be

654 associated with a specific epigenomic signature, whereby both
655 their promoters and gene bodies were marked by H3K27ac
656 deposition, unlike age-downregulated genes, in which
657 H3K27ac peaks were found only at promoter regions.
658 Furthermore, ageing was associated with a loss of H3K27ac
659 at gene bodies, which strongly correlated with an age-
660 dependent increase in transcription levels. *In vivo* treatments
661 of adult mice with SAHA prevented the age-related decrease
662 of H3K27ac at gene bodies and the transcriptional upregula-
663 tion associated with this decrease, although also improving the
664 cognitive performance of treated mice in specific behavioral
665 tasks [25].

666 It would be crucial to extend these studies using a wider
667 range of epigenomic analyses focused not only on fast-
668 turnover histone acetylations like H3K9ac, H4K16ac and
669 H3K27ac [39], but also on more stable histone acetylations
670 such as H3K4ac or H3K56ac and histone methylations at
671 H3K4, H3K9, H3K27, and H4K20, in order to evaluate the
672 persistence and potential reversibility of age-related signa-
673 tures. Notwithstanding these caveats, the above described re-
674 sults suggest the existence of an epigenomic mechanism con-
675 trolling the activation levels of the brain inflammatory tran-
676 scriptome, which may be prone to an age-related dysregula-
677 tion predisposing the brain to increased inflammatory re-
678 sponses and cognitive impairment.

679 Inflammatory responses in the brain are mainly mediated
680 by resident glial populations, especially astrocytes and mi-
681 croglia. In agreement with this notion, genome-wide
682 transcriptomic analyses of the cell type-specific and the
683 region-specific gene expression changes associated with hu-
684 man brain aging found that the genes showing multi-regional
685 changes (i.e., detected across most of the tested brain areas)
686 were often upregulated, microglial-specific and enriched for
687 categories related to immune response. Consequently, in con-
688 trast to neuronal gene expression, microglial transcriptional
689 signatures mainly clustered the analyzed samples according
690 to age, rather than regional identity, indicating that the age-
691 related transcriptional response of microglia generally lacks a
692 specific regional character, with the exception of the cerebel-
693 lum [3]. Therefore, aging is associated with a widespread in-
694 crease of immune-related gene expression in microglial cells,
695 which is likely to be functionally linked to the inflammatory
696 environment of the aged brain.

697 Two recent studies have profiled the transcriptome of hu-
698 man microglial cells that were sorted either from the DLPFC
699 [40] or from the parietal cortex [41]. The DLPC dataset was
700 derived from elderly subjects, some of them with a diagnosis
701 of AD, whereas the parietal cortex dataset was obtained from
702 cognitively normal subjects of various ages. Both studies con-
703 firmed that the human microglia transcriptome is differentially
704 regulated with age. In particular, the analysis of gene expres-
705 sion levels in relation to sample age in the parietal cortex
706 dataset revealed several hundred genes that were differentially

707 expressed in younger and elder microglia and that were
708 enriched for categories related to immune response, cytoskel-
709 eton, and cell adhesion [41]. A large number of age-related
710 transcriptional changes were also found by comparing the
711 DLPFC dataset with previous microglia signatures obtained
712 from middle-aged subjects. The genes upregulated in the aged
713 DLPFC microglia were enriched for categories related to amy-
714 loid fiber formation, immune functions, and endosomal/
715 vacuolar pathways [40].

716 A substantial group of genes preferentially expressed in the
717 aged human microglia, named HuMi_Aged gene set, was iden-
718 tified by superimposing the microglial-specific DLPFC profiles
719 with the transcriptome of bulk DLPFC tissue and selecting the
720 genes robustly enriched in sorted cells compared to tissue sam-
721 ples. This set contained several known susceptibility genes for
722 late onset AD, although the sample number used for microglia
723 purification was too small to find significant associations be-
724 tween AD-related neuropathological traits in these samples and
725 HuMi_Aged genes [40]. A recent study combined a morpho-
726 logical evaluation of the proportion of active microglia (PAM)
727 in postmortem cortical samples from elderly subjects with
728 GWAS to correlate genomic variation with microglia activation
729 and AD. These analyses revealed an association of PAM-
730 related and AD-related susceptibility variants, suggesting that
731 the mechanisms leading to increased microglia activation in the
732 aged brain have a genotypic component, which also seems to
733 predispose to cognitive impairment [42]. All together, the ob-
734 servations described above suggest that the inflammatory tran-
735 scriptome of microglia cells is broadly hyperactivated in the
736 aging brain, that both genetic variation and epigenetic regula-
737 tion contribute to this age-related upregulation, and that the
738 genetic and epigenetic factors linked to increased microglia
739 activation may contribute to cognitive decline.

740 Unlike the microglia-specific transcriptome, which appears
741 to be mainly affected by age with limited regional differences,
742 astrocyte-specific genes in the human brain show complex
743 patterns that are dependent on both regional identity and
744 age, as shown by superimposing transcriptomic profiling of
745 bulk human brain tissues to mouse brain single-cell datasets
746 [3]. In particular, these genes reveal a significant degree of
747 regional diversity in the younger brain, clearly separating the
748 transcriptional profiles of different brain areas, such as the
749 cerebral cortex, the hippocampus, the cerebellum, and the
750 brainstem. Aging brings about marked shifts in the regional
751 character of astrocytic gene expression, causing loss of region-
752 al differences in the transcriptional profiles of the aged brain
753 [3]. These age-related changes may be causally relevant to the
754 alterations in the neural circuitry underlying cognitive decline,
755 given the extensive functional interactions between astroglial
756 and neuronal cell populations. In support of this hypothesis,
757 recent transcriptomic profiling of purified astrocytes from
758 young and aged mice showed that aged astrocytes upregulated
759 genes involved in synaptic elimination and neuronal damage,

such as those encoding for key components of the comple- 760
ment cascade [43]. Given the significant differences found in 761
the transcriptome of purified human and mouse microglia [41, 762
44], however, the age-related changes identified in aged mice 763
might not faithfully recapitulate the alterations in the aged 764
human astroglia. Therefore, analyses performed with purified 765
astrocyte cell populations from human brain samples are nec- 766
essary to fill this knowledge gap. 767

768 Conclusions and Future Perspectives

769 The last couple of years have witnessed a major leap forward 770
in employing human postmortem tissue samples as a tractable 771
experimental system to elucidate the molecular changes asso- 772
ciated with human brain aging and to gain insight into the 773
possible mechanisms causing cognitive decline during normal 774
and neuropathological aging. This breakthrough has been 775
made possible by improved and affordable NGS platforms 776
allowing the generation of large molecular datasets, which 777
are amenable to mathematical and computational approaches 778
to predict the likelihood of causal relationships based on sta- 779
tistical associations among diverse molecular and phenotypic 780
traits. Thus, although functional studies are extremely difficult 781
to do in human subjects, mechanistic predictions can be made 782
based on human datasets and subsequently validated in other 783
experimental paradigms.

784 The experimental use of human brain samples is inherently 785
exposed to potentially confounding factors such as the extent 786
of postmortem tissue preservation and the degree to which the 787
general population is represented in the cohort used for sam- 788
pling. Besides these obvious caveats, several challenges will 789
have to be faced to improve the power of molecular investi- 790
gations of human brain aging. First, a better temporal resolu- 791
tion of the age-dependent changes will need to be achieved, 792
with a stronger focus on the early onset of brain aging pro- 793
cesses and cognitive decline. Second, it will be essential to 794
perform molecular profiling of purified neurons and astrocytes 795
from human brain samples, as already done with microglia, 796
and to implement single-cell approaches to dissect the age- 797
related changes taking place in specific cell subpopulations. 798
Third, it will be important to extend epigenomic analyses by 799
the profiling of additional chromatin modifications, and to 800
focus on the age-associated changes in the non-coding tran- 801
scriptome. The modifications in gene expression levels detect- 802
ed during normal or pathological brain aging may also reflect 803
alterations in transcript stability, which may be regulated by 804
microRNAs [45]. In agreement with this hypothesis, age- 805
related modulation of microRNA levels in the rodent brain 806
has been reported [46, 47]. Notably, circular RNAs have been 807
shown to accumulate in the aging brain [48], and long non- 808
coding RNAs are also implicated in various aging processes 809
[49]. Finally, although pluripotent cells and organoid culture

810 methods have allowed to perform functional assays in human
 811 *in vitro* neural cell systems [50], mouse genetic models will
 812 remain essential for functional studies *in vivo* in the foresee-
 813 able future. Therefore, more work is crucially needed to clar-
 814 ify the conserved and the divergent aspects in the genome-
 815 wide mechanisms underlying brain aging in humans and in
 816 animal model systems.

817 **Required Author Forms** Disclosure forms provided by the au-
 818 thors are available with the online version of this article.

819

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