Zebrafish brain RNA sequencing reveals that cell adhesion molecules are critical in brain aging

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A B S T R A C T

Brain aging is a complex process, which involves multiple pathways including various components from cellular to molecular. This study aimed to investigate the gene expression changes in zebrafish brains through young-adult to adult, and adult to old age. RNA sequencing was performed on isolated neuronal cells from zebrafish brains. The cells were enriched in progenitor cell markers, which are known to diminish throughout the aging process. We found 176 statistically significant, differentially expressed genes among the groups, and identified a group of genes based on gene ontology descriptions, which were classified as cell adhesion molecules. The relevance of these genes was further tested in another set of zebrafish brains, human healthy, and Alzheimer's disease brain samples, as well as in Allen Brain Atlas data. We observed that the expression change of 2 genes, GJC2 and ALCAM, during the aging process was consistent in all experimental sets. Our findings provide a new set of markers for healthy brain aging and suggest new targets for therapeutic approaches to neurodegenerative diseases.

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1. Introduction

The brain is sexually dimorphic in its function, and it shows transcriptional profile changes during a lifetime. Identifying the genetic pathways that are altered during the aging process would enable us to understand the mechanisms related to brain aging. Determining how similar neurodegenerative processes like Alzheimer's (AD) or Parkinson's disease are to healthy brain aging, and finding a therapeutic or prevention strategy for them are critical. These diseases bring a considerable burden to patients and caregivers. However, to interpret the diseased state, we must first have a comprehensive understanding of the healthy aging brain. To date, the human lifespan has elongated, and the same should be possible for the human health span. One of the methods that provide a thorough examination of the molecular and cellular aspects of the brain is RNA sequencing. It is widely used for humans and animals, with numerous different approaches. RNA sequencing data from the aging human brain is accessible to everyone through The Aging, Dementia, and Traumatic Brain Injury (TBI) Study (http://aging.brain-map.org). In this study, human brain regions were sequenced, from donors of 78–100+ ages, with defined dementia or TBI status.

Various factors have been emphasized in the literature having impacts on the aging process, from molecules to cellular components and pathways. To begin with, complement system genes were discovered as biomarkers for aging (Bae et al., 2018). Likewise, immune system genes were found to be upregulated in the hippocampus during aging (Ianov et al., 2017). In another study, micro RNA (miRNA)s were examined for their roles in aging. A specific miRNA, miR-34, for example, was observed to be effective in preventing neurodegeneration (Kennerdell et al., 2018). Another factor affecting aging is post-transcriptional regulation. Increases in RNA editing throughout aging is a conserved feature in primates (Li et al., 2013). It was shown that intron retention plays a differential role in regulating biological functions at different stages of aging (Adusumalli et al., 2019). Analogously, splicing patterns were also investigated during aging. Although most of the splicing is confined to developmental stages, some still persist during aging (Mazin
et al., 2013). Last but not the least, the autophagy-lysosomal-related pathways stood out in AD in the dorsolateral prefrontal cortex (Raj et al., 2018). To overcome the above-mentioned deteriorations occurring with advanced age, it is therefore critical to keep track of identifying new potential therapeutical targets for intervention of AD. Such another advancement is the restoration of the neural stem cell (NSC) proliferative ability via interleukin-4 recovering hampered plasticity due to AD (Papadimitriou et al., 2018).

There is also an inter-individual variability in gene expression, and the heterogeneity increases as the age increases. A recent study found that middle-aged women and Alzheimer’s patients had more similar patterns in the prefrontal cortex than that with the younger women (Sanfilippo et al., 2019). As suggested by Kedl et al. (2019), this gene expression variation occurring with advanced age is not solely restricted to a specific pathway, but instead it is distributed throughout diverse biological pathways.

An earlier study examining the existing databases to find a common pathway contributing to AD pointed out the cell adhesion molecule pathway to be the only consistent one (Liu et al., 2012). In addition, cell adhesion molecules, as a gene ontology (GO) biological process category, were found among the categories that are significantly enriched in genes showing consistent upregulation during postnatal development and downregulation in aging (Dönertas et al., 2017). In like manner, cell-adhesion-related genes along with the genes having roles in cell membrane, regulation of neurogenesis, and long-term potentiation were demonstrated to be downregulating with age in the rat hippocampus (Pereira et al., 2017). Similarly, Frahm et al. (2017) showed that focal adhesion genes are downregulated during aging. Another study analyzing 2 whole-genome AD transcriptome data and 4 brain expression genome-wide association study datasets revealed that cis-regulatory single-nucleotide polymorphisms are enriched in cell adhesion molecules pathway (Bao et al., 2015).

The wet laboratory experiments in this study were performed using zebrafish as model organism. Zebrafish has human orthologs for more than 10,000 genes (Howe et al., 2013) and ages gradually (Kishi 2014). In the zebrafish brain, neurogenesis continues throughout the lifespan (Kizil, 2018). Furthermore, they can repair a damage to the optic tectum and retain visual capacity, even at older ages. However, in old fish, the original number of regenerating axons and the outgrowth is diminished, compared to young animals (Van Houcke et al., 2017). Although it has been used widely in biological research, most of the experiments are performed in larval or young zebrafish (younger than 6 months). However, it is possible to study zebrafish at older ages (>24 months). For example, our group analyzed the brains of fish that were 34–43 months old (Karoglu et al., 2017). Also, we previously performed a microarray analysis using individual zebrafish whole brains (Arslan-Ergul and Adams, 2014). There we examined young (7.5–8.5 months) and old (31–36 months) fish brains comparing the gene expression differences in young versus old, and male versus female brains. By using GO analysis, we found that neurogenesis, neuron formation, and brain development related gene expressions differed between young and old, as well as male and female brains.

The purpose of this study is to reveal the gene expression differences in the aging zebrafish brain by using neural cell populations and comparing these to human brain. For this, we extracted cells from brains of young-adult (3 months old), adult (5.5–8.5 months old), and old (26–31 months old) zebrafish and performed RNA sequencing. Following data analysis, we obtained a list of genes whose expressions were significantly changing with age. After testing our gene list in whole zebrafish brains, human brain samples, and Allen Brain Atlas data, we concluded that cell adhesion molecules, more specifically ALCAM (ALCAM; human, Alcamb; zebrafish) and GJC2 (GJC2; human, Cx47; zebrafish), are critical during brain aging. To our knowledge and according to Gene Expression Omnibus datasets, this is the first RNA sequencing data in old zebrafish brains and comparison of this data to human brain samples.

2. Methods

2.1. Cell extraction from the whole brain

The cell extraction and tissue culturing procedure from zebrafish brains were adapted from Kizil laboratory special thanks to Christos Papadimitriou and Dr Çağhan Kizil. In order for cell extraction from zebrafish brains, Miltenyi Biotec Neural Tissue Dissociation Kit (130-092-628) was used. Briefly, fish were washed in 70% ethanol, and brain dissection was performed as described previously (Arslan-Ergul et al., 2016), except for the fact that it was performed in Dulbecco’s phosphate buffered saline (DPBS) + 1% penicillin-streptomycin + 1% anti-anti solution (all from Gibco). Up to 5 brains per reaction, individual brains were pooled in isolation medium containing 4 mL Leibowitz’s (L1) 15 medium and 40 μL pen-strep 1%. Enzyme mixture 1 was prepared as 25 μL papain enzyme in 950 μL enzyme buffer, and added onto the brains in isolation medium and incubated in this solution at 28 °C for 25 minutes. Then, enzyme mixture 2 (5 μL enzyme A and 10 μL buffer Y) was added onto the brain tissue mixture and similarly incubated in this solution at 28 °C for 25 minutes. After triturated glass pipettes were washed one after the other with first DPBS, then 70% ethanol, then again with DPBS, the brain tissue mixture was passed through these pipettes 20 times for homogenization. Lysates were then passed through 70 μm strainers for single cell dissociation and washed with washing buffer (L15 medium supplemented with 5% fetal bovine serum (FBS), 1% pen-strep, and 1% anti-anti solution). Finally, the lysates were centrifuged, the pellet was dissolved in washing buffer, and then centrifuged again. The final pellet was dissolved in culture medium (L15 supplemented with 5% FBS, 1% pen-strep, 20 ng/mL epidermal growth factor, and 20 ng/mL basic fibroblast growth factor). At the end, cells were incubated in 28 °C incubator without CO2. The medium was renewed every other day by half. Cells were plated in CellBIND (Corning) flasks.

2.2. Animals

Wild-type AB strain, zebrafish were used in the experiments. Animals were fed twice a day with dry food and 3 times a week with artesia and kept in stable conditions. The methods used in this study were approved by the Bilkent University Animal Ethics Committee with the decisions 2014/18 and 2016/1.

2.3. RNA isolation

Total RNAs were isolated from the scraped cell cultures, using Qiagen’s RNeasy Mini Kit according to the manufacturer’s instructions. The isolation protocol was followed by DNase treatment with Ambion TURBO DNA-free Kit from Thermo Fisher Scientific. Human RNA samples were obtained from Biochain (USA). Healthy RNA samples (code: R1234035-P) were pooled from 5 donors of ages 21, 24, 26, 26, and 66. Alzheimer RNA sample (code: R1236035Alz-50) was from an Alzheimer patient of age 87. As commercially indicated, all human samples were collected from males.

2.4. RNA sequencing

For these experiments, RNA samples were sent to TUBITAK-MAM Gebze Laboratories. There, Illumina HiSeq 2500 was used according to its routine operating steps. Briefly, RNA libraries were
prepared using TruSeq Stranded Total RNA Library Prep Human/Mouse/Rat (ILLUMINA), which also supports zebrafish, according to the manufacturer’s protocol. The first step of the library preparation was the removal of cytoplasmic ribosomal RNA using biotinylated, target-specific oligos combined with Ribo-Zero ribosomal RNA removal beads. Following this depletion, the RNA was fragmented into small pieces using divalent cations under elevated temperature. The cleaved RNA fragments were copied into first strand complementary (c) DNA using reverse transcriptase and random primers, followed by second strand cDNA synthesis using DNA polymerase I and RNase H. Double-stranded cDNA was subjected to the A-tailing and barcoding indices were added via ligation enzyme. All the samples were single indexed. The products were then purified and enriched with polymerase chain reaction (PCR) to create the final cDNA library.

Data analysis was performed by our group members and Novogene Ltd (China). In brief, raw FASTQ files were cleaned off from the adapters, Q20, Q30, and GC content were calculated. In the downstream analyses, clean data with high quality were used. Index of the reference genome was built using Bowtie v2.2.3, and paired-end clean reads were aligned using TopHat v2.2.0.12. Fragments per kilobase of transcript was calculated after read numbers mapped to each gene via HTSeq v0.6.1. Differential expression analysis was performed with DESeq R package (1.18.0). Genes with an adjusted p-value < 0.05 (Benjamini and Hochberg approach) assigned as differentially expressed. GO enrichment analysis was implemented by the GOseq R package (Ashburner et al., 2000; Young et al., 2010), with corrected p-value < 0.05. Pathway and process enrichment analysis was done in Metascape (Zhou et al., 2019). All the raw and processed data were uploaded to Gene Expression Omnibus and can be accessed through the record GSE133436.

2.5. Real-time PCR

We used Roche Transcriptor HiFi cDNA Synthesis Kit to synthesize cDNA. For quantitative (q) PCR experiments, we used Roche FastStart Essential DNA Green Master in a LightCycler 480 machine and SensiFAST SYBR Mix in a Mic PCR from Biomolecular Systems. The primers used are given in Supplementary Table 1.

2.6. Statistics

All statistics and graphics were performed in Prism 8 for macOS, version 8.1.1. For the experiments performed on zebrafish brain cells, there were 3 samples: old, adult, and young-adult. Hence, we applied ordinary 2-way analysis of variance, followed by Tukey’s multiple comparison test and p < 0.05 was considered significant. For the data whole zebra fish (März et al., 2010), tmem119 for microglia (Satoh et al., 2016), and neuron (rbfox3b) for marking post-mitotic neurons (Won et al., 2016). In terms of nestin and neuron, the young-adult group had a significantly higher expression (p = 0.0160 and p = 0.0103, respectively) when compared to the old group. For sox2 and tmem119, the young-adult group had higher expression when compared to the adult and old groups (sox2: p = 0.085, p = 0.0003; tmem119: p = 0.0231, p = 0.0005). For s100beta, the expression levels increased numerically during aging, though slightly. Thus, 2 markers of progenitor cells, nestin and sox2, were highly expressed in young-adult, while showing lower expression in older groups despite retaining the expression of these markers. Neurons, glia, and microglia were also represented in our cell culture. Young-adult group had a higher ratio of neurons and microglia, whereas old group was slightly enriched in glia. For all markers, we observed a gradual change during aging. Thus, our cell culture represents brain cells and progenitors for all age groups.

After the quality controls, bioinformatic analysis was performed to identify differentially expressed genes, which are shown in a Venn diagram (Fig. 2A). Here we compared the young-adult to adult and adult to old groups. We did not include a young-adult to old comparison because we wanted to focus on the genes that gradually and consistently change during aging. So we analyzed only the genes that were differentially expressed between old and adult groups, as well as adult and young-adult groups. That being said, we performed a young-adult to old comparison and drew a 3-component Venn diagram (Supplementary Fig. 1), besides drawing the volcano plots to discriminate the genes that were differentially expressed between groups (Supplementary Fig. 2). Between old and adult, 37 genes were differentially expressed, and between adult and young-adult, 153 genes differed in expression. Out of these, 14 genes were common in both lists, and the full list of these genes are given in Supplementary Table 2. In total, 176 genes were found to be significantly different in either comparison. We drew a cluster map out of these 176 genes (Fig. 2B). Looking at the hierarchical clustering of the genes and samples, adult and old were found closer to each other, whereas young-adult profile formed a separate group. Furthermore, we tried to identify whether groups of genes are expressed distinctly in developmental stages. Principal component analysis was performed for the whole data and the groups were separated distinctly as expected (Supplementary Fig. 3). This confirms that the stages of adulthood are distinct from each other in terms of gene expression.

The GO terms were used to describe the cellular component, molecular function, and biological processes of the genes. The GO enrichment chart shows the number of differentially expressed genes associated with specified GO terms. Running our 176 genes in GO analysis, we observed that several GO terms were associated with our genes (Fig. 2C). Likewise, we performed Metascape

### Table 1

<table>
<thead>
<tr>
<th>Sample</th>
<th>Age</th>
<th>Pool</th>
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<tbody>
<tr>
<td>Old1</td>
<td>26–28 months old</td>
<td>5 brains</td>
</tr>
<tr>
<td>Old2</td>
<td>30–31 months old</td>
<td>5 brains</td>
</tr>
<tr>
<td>Old3</td>
<td>30–31 months old</td>
<td>5 brains</td>
</tr>
<tr>
<td>Adult1</td>
<td>8.5 months old</td>
<td>5 brains</td>
</tr>
<tr>
<td>Adult2</td>
<td>5.5 months old</td>
<td>5 brains</td>
</tr>
<tr>
<td>Adult3</td>
<td>5.5 months old</td>
<td>5 brains</td>
</tr>
<tr>
<td>Young-adult1</td>
<td>3 months old</td>
<td>10 brains</td>
</tr>
<tr>
<td>Young-adult2</td>
<td>3 months old</td>
<td>10 brains</td>
</tr>
<tr>
<td>Young-adult3</td>
<td>3 months old</td>
<td>10 brains</td>
</tr>
</tbody>
</table>

3. Results

3.1. Gene expression profiling reveals distinct grouping

In this study, we aimed to obtain cell suspensions from zebrafish brain of different ages. Assembling cohorts of young-adult (3 months), adult (5.5–8.5 months), and old (28–31 months) zebrafish (Table 1), we obtained a progenitor-enriched cell culture from the brains of these fish and did RNA sequencing from the total RNA (Fig. 1A). Progenitor state of these samples was confirmed via qPCR by measuring nestin, sox2, neun, s100b, and tmem119 (Fig. 1B). Nestin was used as a marker for proliferating precursors (Mahler and Driever, 2007), sox2 for progenitor cells, s100beta for glial cells (März et al., 2010), tmem119 for microglia (Satoh et al., 2016), and neuron (rbfox3b) for marking post-mitotic neurons (Won et al., 2016).
analysis to the same gene set (Fig. 2D). Among the profoundly enriched terms there were cell adhesion, biological adhesion, locomotion, axon guidance, and neuron development from biological processes; membrane from the cellular compartment; and receptor binding, cell adhesion molecule binding, and protein binding from molecular function. Cell adhesion and biological adhesion terms were found to be statistically significant. Hence, from the GO analysis, cell adhesion description stands out as an essential pathway altered throughout aging.

3.2. Functional classification reveals 2 classes of cell adhesion molecules

Of the 176 significant genes, we aimed at finding the human orthologs. For this purpose, we first ran The database for annotation, visualization, and integrated discovery (DAVID) (Huang et al., 2009a) gene conversion tool to convert the gene IDs; 134 genes were assigned with an Entrez Gene ID (Supplementary Table 3). From this list, 2 comparison lists were created with the fold changes from old to adult, and from adult to young-adult. Here we aimed to find the significant changes through aging. Out of the statistically significant differentially expressed genes (DEG) list, we manually chose genes that are meeting 3 criteria decided as follows: (1) either had a fold change of <0.5 or >2, (2) changed in the same direction in both comparison lists, and (3) had a gene name, meaning that the transcript is annotated. The final list had 39 genes. Of the selected genes, we ran a functional classification in DAVID. This tool is useful in reducing large lists into functional groups so that a related subgroup of genes can be investigated further (Huang et al., 2009b). There, we identified 2 gene groups, which are given in Table 2.

All these genes had human orthologs, except cldnk. Although there were 2 groups, all had a standard description: cell adhesion. Thus, both in this analysis and previously in GO analysis, we found cell adhesion to be a significant class of genes, and therefore decided to focus on these genes.

3.3. Relative expression levels of cell adhesion molecules in progenitor cell culture and whole zebrafish brains

In order for verification, different RNA samples were obtained from different sets of fish cohorts from same age groups. Following RNA isolation, qPCR experiments were conducted. For the genes given in Table 2, the qPCR results were as expected. The fold changes were in the same direction for alcam, cldn19, nadl1,2, and mag. For the nfasca, timd4 (p = 0.04, between old and young-adult), and cx47 genes, fold changes were in different direction than the RNA sequencing results (Fig. 3).

The results up to now were from progenitor cell culture obtained from pooled zebrafish brains. We also wanted to test the expression of these genes in the individual whole brains via qPCR in order to detect the expression difference or similarity, if any, between the progenitor cells and the whole brains. For this purpose, 7-month and 17-month-old fish were used. The gene expression was significantly changed between 7-month and 17-month old samples, for all genes except nfasca (Fig. 4). Expression of alcam (p = 0.0011), cldn19 (p < 0.0001), timd4 (p = 0.0003), cx47 (p < 0.0001), nadl1 (p < 0.0001), and mag (p < 0.0001) decreased, whereas nfasca increased with age.

Up to this point, we had a differential expression of the cell adhesion genes in different samples. Because it would be possible to have a different expression pattern in whole brain and isolated...
cells for the fact that the cells are a subset of the brain and progenitor cells have a distinct expression, we thought that at this stage it is necessary to test the validity of the expression pattern difference of these genes in the human brain as well.

### 3.4. Analysis of the selected genes in healthy and Alzheimer human brain samples

Therefore as a next step, the expression levels of the selected genes were analyzed in human samples. For that, we ordered human brain RNA samples from Biochain. One RNA sample was indicated by the company to be extracted and pooled from 5 young human male brains, and the other RNA was indicated to be collected from the brain of an Alzheimer’s patient of age 87. We first wanted to test the samples for known markers of aging and development. Human telomerase complex has an RNA component called TR, and an enzyme component called TERT. Expression of both TR and TERT genes decline greatly during adulthood; however, they still remain at detectable levels (Spilsbury et al., 2015; Ishaq et al., 2016). In accordance, we observed a deficient but detectable TERT level in our Alzheimer sample RNA ($p = 0.0297$). In the same manner, TR expression was also deficient in the Alzheimer sample ($p = 0.0013$). For the undifferentiated stem cells, POU Class 5 Homeobox 1 (POU5F1/OCT4) and Nanog homeobox (NANOG) gene expressions are known to be critical for cell proliferation, while Tubulin beta 3 class III (TUBB3, TUJ1) gene is known to be primarily expressed in

<table>
<thead>
<tr>
<th>Table 2</th>
<th>DAVID gene functional classification result</th>
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<tr>
<td><strong>Gene group 1</strong></td>
<td></td>
</tr>
<tr>
<td>30634</td>
<td>Neural adhesion molecule L1.2 (nadl1.2)</td>
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<tr>
<td>100141490</td>
<td>Neurofascin homolog (chicken) a (nfasca)</td>
</tr>
<tr>
<td>474346</td>
<td>Myelin associated glycoprotein (mag)</td>
</tr>
<tr>
<td>100142639</td>
<td>T-cell immunoglobulin and mucin domain containing 4 (timd4)</td>
</tr>
<tr>
<td>323919</td>
<td>Activated leukocyte cell adhesion molecule b (alcamb)</td>
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</table>

Fig. 2. Description of the differentially expressed genes. (A) Venn diagram for the differentially expressed genes in young-adult versus adult and adult versus old groups. (B) Heat map of the differentially expressed genes. Blue shows low expression, red shows high, and white shows middle. Experiment groups are in columns and genes in rows. It is an unsupervised clustering and subclusters are shown by diagrams on the left and top of the map. (C) Directed acyclic graph diagram of the most significant GO term descriptions of the differentially expressed genes. GO terms are given in rows. In the biological process description, there are 89 genes from our list. Within this group, 13 genes are present in the biological adhesion and cell adhesion descriptions, with a $p = 0.0348$. (D) Metascape pathway and process analysis of the 176 genes from (A). Abbreviation: GO, gene ontology. (For interpretation of the references to color in this figure legend, the reader is referred to the Web version of this article.)
neurons. For these 3 genes, we ran qPCR for healthy and Alzheimer samples and saw that all 3 reduced in Alzheimer samples in sub-neurons. For these 3 genes, we ran qPCR for healthy and Alzheimer gene expression levels. qPCR results for each gene: old is represented by black bars, adult by squared bars, and young-adult by horizontal lines. *p < 0.05.

In order to examine the interaction of these proteins with each other and with some aging-related proteins, we used Cytoscape software (Supplementary Fig. 4), in which L1CAM and ALCAM were identified to interact with each other, while MAG and GJC2 formed another separate interaction together. Furthermore, ALCAM was also found to be interacting with NANOG, OCT4, and SOX2 directly, and with TERT and DNMT1 indirectly.

3.5. Expression data of the selected genes in Allen Brain Atlas dataset

Finally, we wanted to see the expression profile of the selected genes in existing RNA sequencing data. From the Allen Brain Atlas study, “Aging, Dementia, and TBI” database, we analyzed the distribution of our selected genes. First, we did a cluster analysis according to increasing age. Here we could not see any grouping (data not shown). We assumed that the reason for obtaining no grouping in the data was that the youngest person included in this database was 78 years old. So our analysis was actually a clustering analysis of old versus very old. However, very interestingly, when we clustered our selected genes according to brain regions, we noticed a remarkable grouping. Our selected genes were highly expressed in the frontal white matter, which is a brain region of particular importance in aging. So as seen in Fig. 6, in the old human brain, NFASC, ALCAM, DLG1, and MAG genes were highly expressed in the frontal white matter when compared to other brain regions that were studied here. Moreover, L1CAM gene was highly expressed in the parietal and temporal cortex when compared to the hippocampus and frontal white matter. Although we could not infer the age differences in this data, in the old brain, the gene expression patterns seem to vary greatly among the brain regions.

For a better comprehension, our results are gathered in a figure complemented with a visual presentation (Fig. 7). Fig. 7A demonstrates the positions of the cell adhesion proteins and possible interactions, while Fig. 7B summarizes our findings in this study in a table view. In the figure, ALCAM proteins are presented in the membranes of antigen presenting cells pairing with CD6 proteins residing on T cells (Fig. 7A, label 1), L1CAM is shown forming homodimers through pre- and post-synaptic neurons (Fig. 7A, label 2), CLDN (CLDN19 and CLDN16) is demonstrated forming junctions in endothelial cells (Fig. 7A, label 3), MAG is represented in myelin-producing Schwann cells, and NFASC is in myelin-producing cells interacting with CNTN1 protein in axons (Fig. 7A, label 4), and finally GJC2 (also known as CX47) is drawn in oligodendrocytes either forming homodimers in oligodendrocytes or heterodimers with a CX43 protein residing in an astrocyte (Fig. 7A, label 5).
results showed that ALCAM and GJC2 are consistently decreasing with age in all of our analysis, while CLDN19, L1CAM, MAG, and NFASC establish a decreasing expression pattern in 2 out of 4 experiments, and increasing in others.

4. Discussion

In this study, we investigated the age-related gene expression level differences of a group of cell adhesion molecules. We tested the genes in zebrafish brain derived progenitor-enriched cells via RNA sequencing and qPCR, as well as in whole zebrafish brain and commercially available human brain samples via qPCR. Our results showed that the cell adhesion molecules, especially the ones we included in this study, play critical roles in aging. Among all the genes, the decreasing pattern of expression of alcam/ALCAM and cx47/GJC2 with age remarkably showed consistency in all our data, which highlights a prospective target for age-related neurodegenerative diseases opening the doors for new therapeutic strategies and drug targets for the age-related deterioration. Although some of these genes were well-studied and some are not, there are not enough studies on literature regarding the most profound ones: ALCAM, GJC2, L1CAM, CLDN19, MAG, NFASC, and TIMD4.

ALCAM (also known as CD166) is a member of the type I transmembrane immunoglobulin superfamily of cell adhesion molecules, sharing 93% similarity between human and mouse orthologs (Bowen et al., 1997). The name stands for activated leukocyte cell adhesion molecule (ALCAM), mainly because it is expressed in the surface of activated leukocytes (Bowen et al., 1995), being involved in leukocyte migration through blood-brain barrier (BBB) tight junctions and having roles in synaptic transmission modulation (Cayrol et al., 2008; Curis et al., 2016; Lécuyer et al., 2017; Park et al., 2017). On leukocytes it acts by attracting and recruiting leukocytes into the inflammation site (Shahaduzzaman et al., 2015). Cayrol et al. (2008) stated that blocking ALCAM reduced neuroinflammation via promoting leukocyte recruitment to the central nervous system. In addition to leukocytes, ALCAM is widely expressed in various cells, including neurons, epithelial cells, fibroblasts, lymphoid cells, myeloid cells, as well as being an accepted cell surface marker of mesenchymal and hematopoietic progenitor cells (Yasen, 2013; Shahaduzzaman et al., 2015). With the upregulation of ALCAM, migration and adhesion were observed to be increased in synovial fluid-derived mesenchymal stem cells (Kim et al., 2017). And in like manner, Alcam-null (ALCAM−/−) mouse model hematopoietic stem cells were demonstrated to be associated with reduced long-term engraftment potential following transplantation, having no effect on cell cycle distribution regardless of age (Jeannet et al., 2014). In the study, Alcam expression was also claimed to be increased by several fold in aged murine hematopoietic stem cells compared to young ones. In relation to cancer, being in accordance with the fact that natural ligands to cell adhesion molecules were shown to be highly expressed in metastatic tumor cells, antibody neutralization of ALCAM (antibody blocking specific tumor-expressed ligands) reduced the tumor seeding in the brain (Soto et al., 2014). From these aspects, increased ALCAM expression levels seem to be correlated with cancerous and inflammatory phenotypes in some cell types while...
being associated somehow with stem cell ability, which makes its fine-tuning important for the proper functioning of the cells. Considering the research literature gap for ALCAM age-related alterations in the brain, testing the gene expression change with age is thought to have crucial contributions to our understanding of the age-related increased susceptibility for neurodegenerative diseases, as an addition to its known roles in metastasis and trafficking through BBB. Moreover, since it has known effects on
neuroinflammation, it is possible to think of a relation between ALCAM gene expression and AD.

GJC2 gene is expressed in oligodendrocytes, and the protein is crucial for gap junctional communication (Georgiou et al., 2017). A mutation in GJC2, leading to the loss of connexin 47 protein, was implicated in hypomyelinating leukodystrophy-2 disease (Pelizaeus-Merzbacher-like disease), which in its severe forms results in hypomyelination (Al-Yahyaee et al., 2013), and in hereditary spastic paraplegia (Abrams, 2019). Therefore mutations occurring in GJC2 gene have severe consequences. Though, to our knowledge there is no study investigating how GJC2 is affected in the brain through aging and AD progression. Having a function in myelination, we think that the gene could have a vital role in both aging and AD.

Immunoglobulin-like cell adhesion molecule (L1CAM), also known as CD171, has a dual function in cells: one being as a static cell adhesion molecule and the other as a motility promoting molecule, which has functions during neural development and metastasis (Kiefel et al., 2012). Similarly, the integrin family of receptor and cell adhesion proteins provide communication between cell to cell as well as cell to extracellular matrix, besides being known to correlate with cancer incidence and metastasis. The ability of LICAM binding to integrins was suggested to be associated with its promigratory and proinvasive effects in cancer cells (Burgett et al., 2016; Nieberler, 2017). In another study, a zebrafish knockdown model of nad1.2 (LICAM zebrafish ortholog; Table 2) was indicated having problems in axonal growth, and also myelination abnormalities (Linneberg et al., 2019). Furthermore, as suggested by Shi et al. (2016), tau proteins reside explicitly within L1CAM expressing exosomes in human plasma, via which toxic tau aggregates are transported from cell to cell contributing to the formation of tauopathies according to a hypothesis. To mention its relation to senescence, L1CAM is stated to be enriched on the surface of senescent fibroblasts and aid their migration (Mrazkova et al., 2018).

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**Table 2**

<table>
<thead>
<tr>
<th>Brain region</th>
<th>Dementia</th>
<th>Frontal White Matter</th>
<th>Hippocampus</th>
<th>Parietal Cortex</th>
<th>Temporal Cortex</th>
</tr>
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<td>No</td>
<td>Yes</td>
<td>No</td>
<td>Yes</td>
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<tr>
<td>CLDN19</td>
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<td>No</td>
<td>Yes</td>
<td>No</td>
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<td>Yes</td>
<td>No</td>
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<td>No</td>
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<tr>
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<td>No</td>
<td>Yes</td>
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<tr>
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**Fig. 6.** Clustering of the selected genes in Aging, Dementia, and TBI data. Brain regions; frontal white matter, hippocampus, parietal cortex, and temporal cortex were included in the study. Each brain region is grouped into patients with or without dementia, and then further organized in increasing age from 78 to 100+. Red shows the highest expression and blue shows the lowest expression. Genes are in rows. Aging, Dementia and TBI Study. 2016 Allen Institute for Brain Science. Allen Human Brain Atlas is available from human.brain-map.org. Abbreviation: TBI, Traumatic Brain Injury. (For interpretation of the references to color in this figure legend, the reader is referred to the Web version of this article.)

**Fig. 7.** Figure depicting the interactions of cell adhesion molecules in the junctions and a summary of gene expression levels in different experimental set-ups. (A) Cell adhesion molecules, the cells on which they are presented, and their interaction partners are shown in diagrams. (B) In the table, gene names of zebrafish and human are given in the first and last column. A blue circle means that gene expression is decreased during aging, and the red circle means that it is increased in aging. One yellow circle means the expression remains the same. Stars in the middle of circles indicate significance. Adapted from Kegg pathway no: hsa04514, except cx47 interactions, which is adapted from Kim et al. (2013). Abbreviations: APC, antigen presenting cell; Astro, astrocyte; Axon, axon of a neuron; endothelial, endothelial cell; Myelin, myelin-producing oligodendrocyte; Olig, oligodendrocyte; Post, postsynaptic neuron; Pre, presynaptic neuron; Schwann, Schwann cell; T cell, T lymphocytes. (For interpretation of the references to color in this figure legend, the reader is referred to the Web version of this article.)
Another cell adhesion molecule, CLDN19, encodes a tight junction protein (Perdome-Ramirez et al., 2019). Together with CLDN16, it forms a permeable channel for magnesium and calcium ions (Giménez-Mascarell et al., 2018). Mutations of CLDN19 were shown to disrupt retinal neurogenesis (Wang et al., 2019), leading to hypomagnesemia, hypercalciuria, and ocular abnormalities (Perdome-Ramirez et al., 2019; Vall-Palmar et al., 2018). Thus, whether this gene has a role in neurogenesis through aging is intriguing to follow. The fact that CLDN19 is expressed in endothelial cells forming a magnesium channel, suggests a role in aging by again regulating the BBB traffic.

MAG is a type I transmembrane glycoprotein being expressed in myelinating oligodendrocytes and Schwann cells (Pronker et al., 2016). It takes part in myelin maintenance, as in GJC2, and survival of oligodendrocytes (Quaries, 2007). Mutations in this gene cause the Pelizaeus-Merzbacher-like disease as in the case of GJC2 gene mutations with ill-formed onion bulb structures and thin myelin sheath of the affected axons (Lossos et al., 2015). In like manner, mice deficient in MAG have defects in the formation and maintenance of myelin and possibly in axonal function (Bartsch et al., 1996; Kamil et al., 2019).

Besides above-mentioned ones, neurofascin cell adhesion proteins take part in the node of Ranvier assembling the nodal macromolecular complex which includes the voltage-gated sodium channels (Zhang et al., 2015). They also have roles in the maintenance of the node and myelinated axon function (Ebel et al., 2014; Taylor et al., 2017), in addition to providing a permissive environment for neurite extension together with N1CAM ( Volkmer et al., 1996). Therefore consistently, antibodies to neurofascin isoforms result in inflammatory demyelinating neuropathy and multifocal motor neuropathy (Delmont et al., 2017; Devaux et al., 2016; Notturno et al., 2014). Last but not the least, being another cell adhesion protein, TIMD4 is a type I membrane protein known to regulate TH1 and TH2 cell response (Yano et al., 2017). It contains an arginine-glycine-aspartate tripeptide (RGD domain), which indicates a role in cell adhesion.

Our study had several weaknesses though. First, we could not find commercially available healthy brain sample collected from an old age human individual. Therefore, although we wanted to compare the healthy old aged human sample with the old aged Alzheimer sample, and healthy old aged human sample with young aged one, due to ethical constraints we could not be able to perform such tests. Second, in Allen Brain Atlas, the ages were again advanced, ranging from 78 to 100+. So again in this data we could only be able to work with the old age group. Nevertheless, by combining data from different sources, we tried to deduce meaningful results.

This study provides evidence that the gene expression levels of a group of cell adhesion molecules change during aging and with AD. Therefore, the findings suggest a prominent role for these molecules during brain aging, besides having a potential for them to take part in future studies on aging and neurodegenerative diseases.

Disclosure statement

The authors report no conflicts of interest.

CRedit authorship contribution statement


Ayça Arslan-Ergül: Conceptualization, Methodology, Resources, Data curation, Writing - original draft, Writing - review & editing, Supervision, Project administration, Funding acquisition.

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Appendix A. Supplementary data

Supplementary data associated with this article can be found, in the online version, at https://doi.org/10.1016/j.neurobiolaging.2020.04.017.

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