TRANSCRIPTOMIC SIGNATURES OF TELOMERASE-DEPENDENT AND -INDEPENDENT AGEING, IN THE ZEBRAFISH GUT AND BRAIN

Running title: Kinetics of ageing in the zebrafish gut and brain

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SUMMARY

Telomerase is best known for its role in the maintenance of telomere length and its implications for ageing and cancer. The mechanisms, kinetics and tissue-specificity underlying the protective or deleterious mechanisms of telomerase, however, remain largely unknown. Here, we sought to determine the telomerase-dependent and -independent transcriptomic changes with ageing, in the gut and brain, as examples of high and low proliferative tissues, respectively. We hypothesised this could shed light on common telomerase-dependent and -independent therapeutic targets aimed at preventing or ameliorating age-associated dysfunction in both tissues. For this, we used the zebrafish model which, similarly to humans, depends on telomerase for health- and lifespan. We performed whole tissue RNA sequencing of gut and brain, in naturally aged zebrafish alongside prematurely aged telomerase null mutants (*tert*$^{-/-}$), throughout their lifespan. Our study highlights stem cell exhaustion as the first main hallmark of ageing to be de-regulated in WT zebrafish gut and brain. Towards the end of life, altered intercellular communication becomes the main hallmark of ageing de-regulated in both gut and brain, and this is accelerated in both tissues, in the absence of telomerase. Finally, we identify 7 key gene changes common between the gut and brain at the early stages of ageing, highlighting

potential early intervention therapeutic targets for preventing age-associated dysfunction in both tissues.

KEYWORDS: Ageing, telomerase, telomeres, gut, brain, zebrafish, transcriptomics, RNA sequencing

1 INTRODUCTION

Ageing is the strongest risk factor for chronic diseases. How and why this is the case remain important questions in the field, especially as key research has shown that targeting common hallmarks of ageing, such as cellular senescence (Baker et al., 2016; Baker et al., 2011), can have a positive impact across multiple tissues and ameliorate several chronic diseases of ageing at the same time. There are well-known key hallmarks of ageing, such as genomic instability, telomere attrition, epigenetic alterations, loss of proteostasis, de-regulated nutrient sensing, mitochondrial dysfunction, cellular senescence, stem cell exhaustion and altered intercellular communication(Lemoine, 2021; López-Otín, Blasco, Partridge, Serrano, & Kroemer, 2013) . However, a major challenge in ageing research is to identify where and when these potentially pathological changes start and when the tipping point between homeostasis and loss of function takes place (Rando & Wyss-Coray, 2021). Additionally, several lines of evidence suggest there may be specific tissues where age-related changes start earlier, potentially influencing others (de Jong, Gonzalez-Navajas, & Jansen, 2016; Rera, Azizi, & Walker, 2013). One example is the gut, which has been suggested to be a trigger for multiple organ failure (Cardoso et al., 2008). Evidence suggests that the kinetics of ageing can vary dramatically between cells, tissues (Shokhirev & Johnson, 2021; M. J. Zhang, Pisco, Darmanis, & Zou, 2021) and individuals, and that this is influenced not only by intrinsic but also extrinsic factors, recently discussed elsewhere (Rando & Wyss-Coray, 2021). These considerations are of particular importance when the aim is to understand how changes in ageing lead to disease and how, when and where to intervene. This is crucial in order to shift towards a more preventive form of medicine, which is a current global ambition (Rudnicka et al., 2020), aiming to match the dramatic increase of lifespan we have experienced in the past century, with an equivalent increase in years of healthy living, i.e, healthspan (England, 2017).

Tissue-specific transcriptomics analysis over the lifecourse can offer important insights into the downstream molecular mechanisms potentially driving the pathology of ageing.

Significant research is being dedicated to these approaches in different animal models, including in mice (Schaum et al., 2020; Tabula Muris, 2020; M. J. Zhang et al., 2021). Different animal models may offer different insights into the mechanisms of ageing, and some models may be better suited to explore the role of specific human hallmarks of ageing. The role of telomere attrition in natural ageing can be considered a hallmark of ageing that may benefit from additional and complementary models, beyond the mouse (Forsyth, Wright, & Shay, 2002; Gomes et al., 2011; Sullivan et al., 2021). Once such model is the zebrafish that, like humans, age and die in a telomerase-dependent manner (Anchelin et al., 2013; Carneiro, de Castro, & Ferreira, 2016; Madalena C Carneiro et al., 2016; Henriques, Carneiro, Tenente, Jacinto, & Ferreira, 2013; Henriques & Ferreira, 2012). Restricted telomerase expression and function are key determinants of natural ageing in humans, underpinning multiple age-related diseases (Blackburn, Epel, & Lin, 2015). However, the role and the dynamics of telomerase-dependent changes that may contribute to tissue-specific ageing are still poorly understood. This is partially due to the fact that telomerase appears to have multiple functions in the cell, that go beyond the maintenance of telomere length, recently reviewed elsewhere (Segal-Bendirdjian & Geli, 2019).

Telomerase is best known for its telomere-dependent function (i.e. canonical functions), acting as a reverse transcriptase, maintaining telomere length through its catalytic domain (TERT protein) and RNA template (TERC) (Greider & Blackburn, 1985). Telomeres are $(TTAGGG)_n$ DNA repeats that together with a complex of proteins (known as Shelterin) create a “cap-like” structure at the end of linear chromosomes (de Lange, 2004), preventing the ends of linear chromosomes from being recognised as deleterious DNA double strand breaks (Ferreira, Miller, & Cooper, 2004). However, in humans, due to time- and cell-specific-limited telomerase expression, telomeres shorten with ageing, leading to proliferative exhaustion and replicative senescence (Bodnar, 1998; d'Adda di Fagagna et al., 2003). Importantly, there is accumulation of cellular senescence with ageing in humans (Dimri et al., 1995) and senescence has been linked to several age-associated diseases (Ovadya & Krizhanovsky, 2014). Additionally, short telomeres themselves can lead to de-regulated gene expression, particularly in genes near the chromosome ends, due to loss of the “telomere positioning effect” (TPE), which is known to regulate gene expression of genes at least up to 10MB away from the chromosome ends (Robin et al., 2014).

Growing evidence now suggests that telomerase also has activity independent of its action at telomeres, known as non-canonical (Goodman & Jain, 2011; Romaniuk et al., 2018; Segal-Bendirdjian & Geli, 2019; Sung, Ali, & Lee, 2014). In the nucleus, these non-canonical functions include transcriptional regulation of genes involved in inflammation, including nuclear factor kappa B (NFkB) and tumour necrosis factor alpha (TNFα) (Deacon & Knox, 2018; Ghosh et al., 2012; Mattiussi, Tilman, Lenglez, & Decottignies, 2012), as well as genes involved in cell proliferation (Choi et al., 2008; Sarin et al., 2005) and cell survival (Cao, Li, Deb, & Liu, 2002; Rahman, Latonen, & Wiman, 2005). Telomerase can also translocate to the mitochondria, where it has been shown to play a protective role against DNA damage and oxidative stress (Ahmed et al., 2008; Haendeler et al., 2009).

As tissues with high cellular turnover present accelerated telomere erosion (Bodnar, 1998; H. W. Lee et al., 1998), it is reasonable to think that telomerase functions are likely to primarily affect highly proliferative tissues. Accordingly, premature accumulation of critically short telomeres has been identified in high proliferative tissues such as the gut, in tert-deficient animal models. Nonetheless, the role of telomerase and telomeres is not restricted to highly proliferative tissues. In the brain, considered a predominately post-mitotic tissue, telomerase has been shown to have a protective role against excitotoxicity (Eitan et al., 2012), oxidative stress (Spilsbury, Miwa, Attems, & Saretzki, 2015), and neuronal death (J. Lee et al., 2010), all involved in neurodegenerative diseases. Studies in late-generation telomerase-deficient mice have suggested that limited telomerase expression is associated with premature accumulation of senescence-associated markers in different cell populations including Purkinje neurons, cortical neurons and microglia (De Felice et al., 2014; Jaskelioff et al., 2011; Jurk et al., 2012; Raj et al., 2015). Telomerase is therefore a promising target to promote healthy ageing in multiple tissues and so the identification of mechanisms driving telomerase-dependent ageing could enable the identification of targeted therapies to improve healthspan.

In this study, we aimed to determine the telomerase-dependent and -independent transcriptomic changes and their kinetics occurring during ageing, in both brain and gut within the same individuals. We hypothesised that this would allow us to identify key age-associated genes and pathways that become prematurely de-regulated in both or either tissue, providing key insights into the early stages of ageing in these tissues and likely

interactions. We further hypothesised that this may highlight potential common telomerase-dependent and -independent therapeutic targets for early intervention aimed at preventing age-associated dysfunction in both tissues. To address these questions, we performed RNA sequencing in whole tissues (gut and brain) of WT fish (2, 9, 22 and 35 months of age) alongside telomerase mutant fish (*tert*$^{-/-}$) (2, 9 and 22 months of age). *tert*$^{-/-}$ zebrafish, extensively characterised elsewhere (Anchelin et al., 2013; Madalena C Carneiro et al., 2016; Henriques et al., 2013), display no telomerase activity and have significantly shorter telomeres from birth, consequently ageing and dying prematurely. Ageing is usually described as a time-dependent change in tissue homeostasis, that increases the probability of disease and death (Hayflick, 2007). However, whether the genes and pathways driving or accompanying these time-dependent changes are also consistently changing in a time-specific manner, remains unresolved (Rando & Wyss-Coray, 2021). We therefore decided to combine a time-series analysis (STEM), which allowed the identification of genes and pathways that are consistently up or down-regulated over-time, with the more traditional differential gene expression (DEGs) analysis between young and old animals. This combined strategy allowed the identification of genes that change in a monotonic, time-dependent manner (STEM), versus genes that change at specific stages of life (DEGs).

We show that although the gut and brain have distinct transcriptomic signatures of ageing, both tissues display hallmarks of ageing as early as 9 months in the WT zebrafish. Importantly, telomerase depletion accelerates the appearance of such hallmarks in both gut and brain. In particular, we identify stem cell exhaustion as the common principal hallmark of ageing at the early stages of ageing, in both tissues. Further, we identify altered intercellular communication, in which immunity and inflammation play a central role, as the main telomerase-dependent hallmark of ageing common between the gut and brain. Finally, we conclude that the gut displays telomerase-dependent hallmarks of ageing at an earlier age than the brain and that these include changes in several key genes that have also been included in the GenAge database, a benchmark curated database that included genes involved in ageing across different organisms, including in humans (Tacutu et al., 2018). Finally, we identify 7 key gene changes common between the gut and brain at the early stages of ageing, namely *ccnb1, kif2c, serpinh1a, temem37*, *si:ch211-5k11.8*, *cfap45* and *eif4ebp3l*.

2 RESULTS

2.1 Identification of monotonic, time-dependent gene signatures and process changes with ageing in the zebrafish gut and brain

In order to identify telomerase-dependent and -independent transcriptional signatures of ageing in the zebrafish gut and brain, we performed RNA-Sequencing of whole tissues, throughout the lifespan of WT and telomerase-deficient (*tert*$^{-/-}$) fish (**Fig. 1 and source data**). While *tert*$^{-/-}$ fish have a lifespan of c. 12-20 months, WT fish typically die between c. 36-42 months of age (Madalena C Carneiro et al., 2016). The data shown here include 4 age-groups of WT (2, 9, 22 and >30 months), corresponding to young, adult, median lifespan and old. As the *tert*$^{-/-}$ fish have a shorter lifespan compared with their WT siblings, the data include 3 age-groups of telomerase-deficient fish (2, 9, and 22 months), which correspond to young, medium lifespan and old. Each group has a sample size of 3 animals and the brains and guts from within the same groups of animals were used (**Fig 1A**). The reads were aligned to the latest zebrafish genome build GRCz11 (Lawson et al., 2020) and resulted in uniquely mapped read percentages ranging from 92.1% to 94.4%, which is a readout of good quality (Lawson et al., 2020). All samples had at least 10 million uniquely mapped reads, except G11 which had around 8 million. Further quality control, using a principal component analysis (PCA) including all the samples, revealed that one of the gut samples clustered with the brain samples, and not with the gut samples (**Fig. 1B**). This was considered to be a technical error and this sample (G7) was therefore excluded from further analysis. To analyse the overall impact of the genotype and age on transcriptomic regulation, we then performed a PCA in the samples from the gut and brain, separately. We further observed that some samples cluster *per* age, but there are some genotypes that separate quite distinctly, despite being of the same age (**Fig 1C and D**). As an example, the WT and *tert*$^{-/-}$ gut samples are quite distinct, and the *tert*$^{-/-}$ 2-month samples cluster closely to the WT at 9 months than the WT at 2 months (**Fig 1C**), providing the first hint of an acceleration of the ageing transcriptomic profile in the *tert*$^{-}$*/-*. We found that there was higher variability in the gut than in the brain, between samples within each age group (**Fig 1C and D**). A summary of the number of significant differentially expressed genes (DEGs) in all samples is represented in **Fig. 1E**. How these DEGs relate to which other, how many

overlap and how many are in common or accelerated in the absence of telomerase (*tert*$^{-/-}$), will be explored later in the manuscript.

To identify genes and pathways that are consistently up- or down-regulated in a time-dependent manner in natural ageing in the zebrafish gut and brain, we grouped the genes into temporal expression profiles, by time-series analysis using Short Time-series Expression Miner (STEM) software (Ernst & Bar-Joseph, 2006). To determine whether the temporal profiles were associated with specific biological processes and pathways, pathway over-representation analysis (ORA) were performed for the genes assigned to the significant STEM profiles. Enrichments of GO Biological Process (GOBP), GO Molecular Function (GOMF), GO Cellular Compartment (GOCC), Kyoto Encyclopedia of Genes and Genomes (KEGG) terms and REACTOME pathway terms were therefore analysed for each profile. Time-series analysis of WT gut identified 9 different profiles, with 2 of them containing up-regulated genes (profiles 7 and 21. Total of 523 genes), and 7 containing down-regulated genes (profiles 32, 31, 23, 9, 12, 22 and 34. Total of 11,594 genes) (**Fig. 2A1 and source data**). Interestingly, in the *tert*$^{-/-}$ gut, all the profiles identified by time-series analysis contain up-regulated genes (profiles 8, 6, 15 13, 12 and 11. Total of 10,317 genes) (**Fig. 2A2 and source data**). Enrichment analysis showed that the profiles containing up-regulated genes are associated with immune response, while profiles containing down-regulated genes are largely associated with proliferation, cellular response to DNA damage, and DNA damage repair, in both WT and *tert*$^{-/-}$ gut (**Fig. 2 A1.1, 1.2 and A2.1**). To help contextualise our analysis, we performed a further classification of enriched processes according to the hallmarks of ageing, which have been previously identified (Lemoine, 2021; López-Otín et al., 2013). This classification further strengthened the observation that the up-regulated profiles include genes mostly involved in altered intercellular communication, in which immunity and inflammation play a key role, whereas the down-regulated profiles identify stem cell exhaustion and genomic instability as the main hallmarks of ageing, to which the genes affecting proliferation, DNA damage and repair are likely to contribute (**Fig. 2 A1.1, 1.2 and A2.1 and source data)**.

In the WT brain, we identified 9 temporal profiles, 6 of them including up-regulated genes (profiles 42, 29, 40, 30, 21 and 48. Total of 7,230 genes), and 3 including down-regulated genes (profiles 1, 12 and 26. Total of 561 genes) (**Fig. 2B1 and source data**). In the

217 *tert*$^{-/-}$ brain, time-series analysis revealed 6 different profiles. Profiles 4, 0 and 3 containing
218 down-regulated genes (total of 5,374 genes) and profiles 15, 12 and 11 containing up-
219 regulated genes (total of 1,155 genes) (**Fig. 2B2 and source data**). As in the gut, up-
220 regulated profiles reveal genes mostly involved in immune regulation and inflammation and
221 down-regulated profiles are mostly involved in cell cycle, genome stability and DNA damage
222 responses, in both genotypes This is further highlighted when placed into context by the
223 analysis based on the hallmarks of ageing, where up-regulated profiles identify altered
224 intercellular communication whereas the down-regulated ones identify stem cell exhaustion
225 as the main hallmarks affected (**Fig 2 B1.1, 1.2; B2.1, 2.22 and source data**).

226 In summary, STEM analysis and enrichment pathways in both gut and brain ageing
227 show a general trend towards up-regulation of genes involved in immune response and
228 down-regulation of genes involved in cell cycle, DNA damage and repair. This general trend
229 is recapitulated in the absence of telomerase. These mechanisms are all known contributors
230 to the well-described altered intercellular communication, genome stability and stem cell
231 exhaustion hallmarks of ageing, respectively (Lemoine, 2021; López-Otín et al., 2013).

232

233 **2.2 Comparing the hallmarks of ageing over time, in WT and *tert*$^{-/-}$ zebrafish gut and**
234 **brain**

235 More than just identifying the signatures of natural WT ageing in the gut and brain
236 and identifying telomerase-dependent and -independent changes, we wanted to
237 understand if there were particular changes occurring before others, and how their kinetics
238 compared between the gut and the brain. In order to contextualise our analysis in the light
239 of the well-described hallmarks of ageing, (Lemoine, 2021; López-Otín et al., 2013), we used
240 the main hallmarks of ageing as headers in which we could group the different changes in
241 processes that were enriched from both the STEM profiles and all DEGs. This allowed us to
242 compare the effects of age, genotype and tissue on the evolution of the key hallmarks of
243 ageing. When we combine all the gene changes (up- and down-regulated) and associated
244 biological processes affected (**Fig. 3 and source data**), we observe distinct tissue-specific
245 signatures of ageing, namely in the gut (**Fig. 3A**) and brain (**Fig. 3B**), particularly when
246 considering the time-series analysis on its own (STEM) (**Fig 3. A1 and B1**). Whereas ageing in
247 the WT gut seems to be predominantly affected by the de-regulated nutrient sensing

initially, the brain displays stem cell exhaustion as the main hallmark affected at the early age of 9 months (**Fig. 3A1**, **B1**, respectively). When we combine the different analysis, though (**Fig. 3A3 and B3**), both gut and brain have stem cell loss as the main hallmark of ageing associated with the enriched processes found at 9 months of age, highlighting this as a potential contributor to the early stages of ageing in both tissues. Towards the end of life, both gut and brain have altered intercellular communication as the main hallmark of ageing identified. Importantly, independently of the analysis, the *tert*$^{-/-}$ zebrafish show accelerated hallmarks of ageing. In specific, a *tert*$^{-/-}$ 2-month-old gut profile is very similar to a WT 35-month-old gut profile (**Fig. 3A**). In the brain, at 2 months of age the *tert*$^{-/-}$ mutant also displays some of the hallmarks of ageing that also become altered in WT ageing at later ages, particularly altered intercellular communication, but the *tert*$^{-/-}$ brain becomes much more similar to the aged WT from the age of 9 months onwards (**Fig. 3B**). This suggests that hallmarks of ageing accelerated in the absence of telomerase are developing earlier in the gut than in the brain.

Finally, when we look at the overall number of gene expression changes, i.e, not just the ones associated with the hallmarks of ageing, we observe that there is a general increase in the number of changes in gene expression with ageing (**Fig 3 A4 and B4**). However, this increase does not appear to be linear. In specific, in the gut, the number of DEGs is fairly low until 9 months of age, after which there seems to be an inflexion point and the number of DEGs increases up to 5-fold in the oldest WT (>35 months) and 3-fold in the oldest *tert*$^{-/-}$ (22 months). In the brain, there seems to be a more gradual, consistent increase over time, both in WT and *tert*$^{-/-}$. This increase in number of DEGs with ageing may be a consequence of the known de-regulation of gene expression with ageing due to de-repression of heterochromatin and/or changes in epigenetic markers.

2.3 When does a *tert*$^{-/-}$ resemble an aged WT the most?

It is of note that, even though the *tert*$^{-/-}$ accelerates hallmarks of ageing in the gut and brain, the set of genes identified are not necessarily the same as in WT ageing. In specific, looking at the genes at the early stages of ageing, the most significant hallmark of ageing shared between the 2 months old *tert*$^{-/-}$ and WT 9 months gut is nutrient sensing, and only 2 up-regulated genes are in common, namely *lipca (Hepatic triacylglycerol lipase precursor)*

279 and *pla1a (phospholipase 1a).* Interestingly, hepatic lipase has been involved in
280 atherogenesis (Santamarina-Fojo, Gonzalez-Navarro, Freeman, Wagner, & Nong, 2004) and
281 age-related macular degeneration diseases that have been hypothesised to have a parallel
282 response to tissue injury induced by multiple factors, including impaired immune responses,
283 and oxidative stress (Neale et al., 2010). In line with the general de-regulated inflammatory
284 response we see in the gut, with ageing (**Figs. 2 and 3 and source data)**, *phospholipase 1a*
285 has been reported to be up-regulated in inflamed gut tissue of Crohn's disease patients
286 (Hong et al., 2017) and has a complex role in the regulation of immunity and inflammation
287 (recently reviewed in (Zhao, Hasse, & Bourgoin, 2021)). In the brain, there are no genes in
288 common between the 2 months old *tert*$^{-/-}$ and WT 9 months associated with altered
289 intercellular communication or genomic instability, the 2 main hallmarks of ageing shared
290 between the genotypes at the early stages of ageing. Following this type of analysis, we
291 could then ask at what age does the *tert*$^{-/-}$ best mimic naturally aged WT, at the level of
292 gene expression. For this, we considered the genes identified within the hallmarks of ageing
293 and the genes classified as "other" i.e. associated with ageing but not obviously associated
294 with the described hallmarks of ageing. We therefore analysed the overlap between all
295 DEGs identified in the old WT (35 months) and the *tert*$^{-/-}$ at the different aged (2, 9 and 22
296 months), using Venn diagrams created using the Venny 2.1 online platform (Oliveros, 2007-
297 2015) to ask this question. We observe that the 2 months old *tert*$^{-/-}$ has the most genes
298 shared with the 35 months old WT gut (59 genes in common) (**Fig. 4 A1 and Supp. Fig. 1**)
299 whereas it is the *tert*$^{-/-}$ at 22 months that has the most genes shared with the 35 months old
300 brain (112 genes in common) (**Fig. 4 B1 and Supp. Fig.1**). This similarity between the 2
301 month and 22 months *tert*$^{-/-}$, with the gut and brain, respectively, is also apparent when we
302 just look at the pattern of the main hallmarks of ageing accelerated in the *tert*$^{-/-}$ (**Fig. 3, 4**
303 **and Graphical Abstract)**. Together, these data suggest that the gut is displaying telomerase-
304 dependent hallmarks of ageing at an earlier age than the brain, consistent with what would
305 be expected for a high *versus* low proliferative tissue. Of relevance, in the brain, of the
306 genes de-regulated in the *tert*$^{-/-}$ 22 month that are in common with the WT aged at 35
307 months *ccna2* (cyclin a2), *cdk6* (cyclin-dependent kinase 6)*, chek1 (*checkpoint kinase 1*),*
308 *mad2l1(* Mitotic spindle assembly checkpoint protein mad2a*), tacc3 (*Transforming, acidic
309 coiled-coil containing protein 3*), top2a (*DNA topoisomerase ii alpha*) and mcm2 (*DNA
310 helicase, MCM2 minichromosome maintenance deficient 2*)* have also been included in the

Human Ageing Resources databases (Tacutu et al., 2018) and are mostly located within the same cluster (green), identified using *k-means* clustering in STRING analysis (Szklarczyk et al., 2021) (**Fig 4 and source data**). Of the proteins encoded by these genes, chek1 (Poehlmann et al., 2011), cdk6 (Morris, Hepburn, & Wynford-Thomas, 2002), mad2l1 (Lentini, Barra, Schillaci, & Di Leonardo, 2012) and tacc3 (Schmidt et al., 2010) have been reported to be involved in cellular senescence. Additionally, top2a has been involved in neuron proliferation (Watt & Hickson, 1994) and mcm2 depletion in mice leads to decreased proliferation in various tissue stem cell progenitors (Pruitt, Bailey, & Freeland, 2007).

2.4 Analysis of telomerase (*tert*)-dependent gene changes of old age

Once we identified what WT ageing looked like at the level of time-dependent gene expression changes over the life-course and the main biological processes affected, we sought to determine how much of these were likely to be telomerase-dependent. If a gene expression change or a biological process alteration is accelerated in the absence of telomerase (i.e becomes significant at an earlier age in the *tert*$^{-/-}$), we consider it to be telomerase dependent, as has been described before (Madalena C Carneiro et al., 2016; Henriques et al., 2013). Conversely, if none of these pre-requisites are met, we consider the gene/process alteration to be telomerase-independent. With this in mind we performed Venn diagrams (Oliveros, 2007-2015) to identify the telomerase-dependent significant gene alterations of old age, i.e, DEGs present in the WT at 35months old, when compared with the WT young control, i.e, at 2 months of age. For this, we used the DEGs identified in the STEM profiles (i.e., genes that change monotonically, in a time-dependent manner), combined with the more traditional DEG analysis, which include all gene changes, whether they change consistently in a time-dependent manner across the lifecourse or not (called "ALL DEGs" from hereafter) (**Fig. 5 and source data**). From these analyses, we identified 50 significant DEGs (out of 491; c. 10%) (present in old age (WT 35 months) that are prematurely de-regulated in the *tert*$^{-/-}$ gut (**Fig. 5 A**) and 100 genes (out of 428; c.23%) that are prematurely de-regulated in the *tert*$^{-/-}$ brain (**Fig. 5 B**). Importantly, most of these genes are directly or indirectly involved in known hallmarks of ageing (Lemoine, 2021; López-Otín et al., 2013), namely altered intercellular communication (including immunity,

342 inflammation, extra-cellular matrix), genome stability (including DNA replication and repair),
343 stem cell exhaustion and mitochondrial dysfunction. In the gut, telomerase-dependent gene
344 expression changes with ageing include up-regulation of genes involved in immune
345 response, such as *tlr18 (Toll-like receptor 18 precursor), sytl1 (Synaptotagmin-like protein 1)*
346 and down-regulation of genes involved in metabolism, such as *cyp8b1 (Cytochrome P450,*
347 *family 8, subfamily B, polypeptide 1*) and *igfbp2a* (*Insulin-like growth factor-binding protein*
348 *2*) *(Bertaggia et al., 2017).* Within the down-regulated genes, there are other well-known
349 genes such as *sox6* (involved in stem cell function), vav3b (involved in wound healing), (**Fig 5**
350 **and source data**). Additionally, we identify a number of non-annotated genes, which may
351 represent novel tert-dependent genes, that would require further investigation.
352 Importantly, amongst these *tert*-dependent genes of "old age", there are a number of
353 homolog or closely related genes which have been previously identified in ageing datasets
354 (Tacutu et al., 2018) such as several *igfbps (insulin-like growth factor binding proteins), igf*
355 *(insulin growth factor), mapks (mitogen-activated protein kinases), tlr3 (toll-like receptor 3)*
356 *and nrg1 (neuroregulin 1).*
357 In the brain, *tert*-dependent ageing gene expression changes include mostly down-
358 regulation of genes involved in cell cycle or neurogenesis, such as *aurkb (*aurora kinase B*),*
359 *chek1 (*checkpoint kinase 1*), ccnb1 (*cyclin b1*), cdk2 (*cyclin-dependent kinase 2*)* and *neurod4*
360 *(*neuronal differentiation 4*), dld (delta d), nog1* (noggin 1), respectively, as well as genome
361 stability and DNA repair, such as *rad54l (*rad54 like*), mcm5* (minichromosome maintenance
362 complex component 5) and *smc4 (*structural maintenance of chromosomes 4*)*. Up-regulated
363 genes are mostly involved in immune response or inflammation, such as *cxcl18b (*chemokine
364 (C-X-C motif) ligand 18b*), mhc2a (*major histocompatibility complex class II integral
365 membrane alpha chain gene*), socs1a (*suppressor of cytokine signalling 1a*), irf8 (*interferon
366 regulatory factor 8*)* and *csf1b (*colony stimulating factor 1b (macrophage)*)*. Of note,
367 amongst these *tert*-dependent DEGs of old age, we identified *dre-mir-29b-1*, which encodes
368 for a regulatory micro RNA 29 (mir29), widely described to be up-regulated in ageing, in
369 response to DNA damage, alongside P53 (Ugalde et al., 2011). mir29 up-regulation with
370 ageing can act as a protective response, limiting excessive iron-exposure and damage in
371 neurons (Ripa et al., 2017). As for the gut, there are a number of homolog or closely related
372 genes which have been previously identified in ageing (Tacutu et al., 2018), such as *chek1*,
373 *mad2l1 (*MAD2 mitotic arrest deficient-like 1 (yeast)*), dl3 (*delta like 3*), nog (*noggin*), ifnb1*

374 *(*interferon beta*), socs2 (*suppressor of cytokine signalling 2*,* amongst many others, which
375 can be found in the GeneAge database.

376 Since it is known that short telomeres themselves can lead to de-regulated gene
377 expression, particularly in genes near the chromosome ends due to loss of the "telomere
378 positioning effect" (TPE) (Robin et al., 2014), we proceeded to map the genes identified to
379 be de-regulated prematurely in the absence of telomerase to their chromosome location,
380 with the aim of determining whether they located to within up to 10MB of either of the
381 telomeric ends (**Fig. 5C1).** We found that whereas most tert-dependent gene changes of old
382 age do not locate to the end of chromosomes, in both gut and brain (**Fig. 5C2 and C3**).
383 However, there are significantly more *tert*-dependent genes located at the end of
384 chromosome in the gut than in the brain, (**Fig. 5 C4**). This is consistent with gut being a more
385 proliferative tissue, where telomeres are likely to shorten more, which would be particularly
386 exacerbated in the absence of *tert*, leading to a higher probability of TPE contributing to
387 gene transcription alterations.

388 **2.5 Transcriptional changes in common between the gut and brain at the early and**
389 **late stages of ageing**

390 Even though the aged phenotype is something usually observed late in life, the
391 underlying molecular and cellular mechanism driving these changes can, arguably, start
392 from the moment you are born (Gladyshev, 2021). To understand what significant
393 transcriptional changes are taking place in the gut and brain and, in particular, in common
394 between the gut and brain, we compared all DEGs and STEM DEGs in these tissues at the
395 earliest time point we detect significant changes (9 months) and at the late stages of ageing,
396 i.e, at the latest time point analysed of 35 months (**Fig 6, Supp Fig 3 and source data**). We
397 identified 7 gene changes in common between WT gut and brain at the early stages of
398 ageing (9 months of age). In specific, we identified 5 down-regulated genes, namely *ccnb1,*
399 *kif2c, serpinh1a, temem37* and *si:ch211-5k11.8*, and 2 up-regulated genes, namely *cfap45*
400 and *eif4ebp3l.* Of these, ccnb1 (G2/mitotic-specific cyclin-B1) and *kif2c* (Kinesin-like protein)
401 are both proteins involved in cell-cycle progression. Whereas *Ccnb1* has been reported to be
402 involved in normal stem-cell/progenitor maintenance in the brain (Hagey et al., 2020) and
403 gut (Basak et al., 2014); kif2c can act as a DNA damage repair protein, and, accordingly, its
404 depletion leads to significant accumulation of DNA damage (Zhu et al., 2020). STRING

analysis suggests these proteins are likely to be co-expressed in a variety of organisms including zebrafish and humans, highlighting potential functional links (**Fig 6 B and source data**). The downregulated *elf4ebp3l* gene (Eukaryotic translation initiation factor 4E-binding protein 3-like) encodes a repressor of translation and is inhibited in response to TORC1 (mammalian target of rapamycin complex 1) (Boutouja, Stiehm, & Platta, 2019).

At "late stages" of ageing, we identified 23 gene changes in common between the gut and brain. Of these, STRING analysis identified 2 main protein network interactions, namely a link between cd59 and cd99l2 and link between il2rb and b2m. In specific, cd59 and cd99l, which our data show to be downregulated in old age, in both gut and brain, have both been cited in the literature as markers of newborn neurons and oligodendrocyte progenitor cells, respectively, in a single-cell transcriptomic analysis of the adult zebrafish brain (Lange et al., 2020). In the gut, cd59, also known as protectin, has been shown to be downregulated in the gut epithelium of ulcerative colitis and Chron's disease patients and thought to render the tissue susceptible to inflammatory damage (Scheinin et al., 1999). cd99 has been shown to be a key molecule in modulating monocyte migration through endothelial junctions (Schenkel, Mamdouh, Chen, Liebman, & Muller, 2002) and monocyte differentiation to macrophages is known to be triggered by endothelial transmigration (Gerhardt & Ley, 2015; Li et al., 2020), including in the brain via migration through the blood-brain barrier (Ifergan et al., 2007). Il2rb and b2m, which our data show to be up-regulated in both gut and brain at the "late stages" of ageing, are key molecules involved in adaptive and innate immune function. Whilst the IL2R beta is important for T-cell mitogenic response to IL-2, the b2m (Beta-2-microglobulin) protein is a component of the major histocompatibility complex I (MHCI), involved in antigen presentation. Intriguingly, b2m has been shown to be present in a soluble free-form, and has been found to be systemically increased with ageing in humans and in individuals with neurodegenerative diseases (Smith et al., 2015). Importantly, heterochronic parabiosis experiments between young and old mice suggest that increased b2m with ageing leads to cognitive impairment and neurodegeneration, and has hence been identified as a systemic pro-ageing factor (Smith et al., 2015). Additionally, increased IL2 receptor expression has been shown to contribute to CD4 differentiation and exhaustion of their naïve pool and therefore ability to respond to infection with ageing, in humans (Pekalski et al., 2013; H. Zhang, Weyand, & Goronzy, 2021).

3 DISCUSSION

In this study, we used RNA sequencing analysis to determine the kinetics of telomerase-dependent and -independent transcriptomic changes occurring during WT ageing in brain and gut tissues, using the zebrafish as a model. We hypothesised that this would allow us to identify key age-associated genes and pathways that become prematurely de-regulated in both or either tissue, providing key insights into the early stages of ageing in each tissue and highlight potential interactions.

3.1 STEM versus all DEGs analysis

Ageing can be described as a time-dependent change in tissue homeostasis, that increases the probability of disease and death (Hayflick, 2007). Whether the genes and pathways driving these time-dependent changes are also consistently changing in a time-specific manner, remains unresolved. To account for both possibilities, we performed a time-series analysis (STEM) and then combined this with the more traditional differential gene expression (DEGs) analysis between young and old animals. Even though the gene changes identified with the STEM analysis were also picked up by the traditional DEGs analysis, the STEM analysis provided a much tighter, restrictive view of the transcriptomic signatures of ageing. In the gut, particularly, the main hallmarks of ageing identified using the STEM analysis are quite different from the ones using the traditional all DEGs. In particular, STEM analysis identifies mitochondrial dysfunction and de-regulated nutrient sensing as the main hallmarks of WT ageing at the earlier stages of ageing in the gut (WT 9 months), whereas all DEGs analysis identified stem cell dysfunction as the principal hallmark de-regulated at that age. The significance of these differences is difficult to judge, but it can suggest that changes in genes affecting mitochondrial function and nutrient sensing have a mostly monotonic trajectory in gut ageing, whereas the ones involved in stem cell maintenance can have different behaviours at different times throughout life. Nevertheless, this difference was not observed in the WT brain or in the *tert*$^{-/-}$ ageing, where STEM and all DEGs analysis led to very similar conclusions regarding the identity or kinetics of the main hallmarks of ageing affected. Importantly, the kinetics of gene changes and processes identified in our data match very well to the phenotypes of ageing previously reported in the *tert*$^{-/-}$ and WT ageing. In particular, in the gut, impaired cell proliferation in the gut is one

of the first *tert*-dependent phenotypes of ageing identified, followed by cellular senescence and inflammation later in life (Madalena C Carneiro et al., 2016; Henriques et al., 2013). Moreover, the recently reported *tert*-dependent changes in macrophage immune activation and increased gut permeability (Pam S. Ellis, 2022) are consistent with key tert-dependent gene changes of old age identified here. An example of such genes are *cd99*, potentially involved in macrophage differentiation via trans-endothelial migration (Gerhardt & Ley, 2015; Li et al., 2020; Schenkel et al., 2002); and *cldn5a* (claudin 5), involved in tight-junctions (Lu, Ding, Lu, & Chen, 2013).

3.2 Main hallmarks of ageing

A simplistic prediction of how the kinetics of the hallmarks of ageing would behave over the lifecourse could be that, at early ages, we would detect more changes affecting the primary hallmarks of ageing, i.e, the "causes of damage", namely genomic instability, telomere attrition, epigenetic alterations and loss of proteostasis. At the last stages, one could predict that we would detect more significant changes in the integrative hallmarks, i.e, the "culprits of the phenotype", namely stem cell exhaustion and altered intercellular communication, of which inflammation is a key component, as described in (López-Otín et al., 2013). However, either separately or combined, neither STEM nor all DEGs analysis show this. Our combined analyses show that most of the gene changes occurring at the early stages of WT ageing are observed from 9 months and are mostly involved in stem cell maintenance, in both gut and brain. One potential explanation for this observation is that our qualitative analysis was not able to distinguish between such hallmarks or is under-estimating the primary hallmarks. Another explanation comes from the relatively low sample size used for each time-point and the heterogeneity of individuals physiology amongst each population. As recently showed in (Dambroise et al., 2016), zebrafish age following the two-phase model first proposed flies (Tricoire & Rera, 2015), based on the age-related intestinal permeability assessed using the Smurf assay they previously described (Rera, Clark, & Walker, 2012). Moreover, we have recently shown that gut permeability with ageing, in zebrafish, is accelerated in the absence of telomerase (*tert*$^{-/-}$) (Pam S. Ellis, 2022). Following this model and considering the longevity curve from the population we sampled, the proportion of Smurfs might have been, approximately, <10% at 2 month, 25% at 9

month, 50% at 22 months and >80% at 35 months, based on previous work in flies (Rera et al., 2012). If we were to translate these findings to zebrafish, then the chances to have selected at least 1 Smurf by accident would be approximately 27% at 2 months, 57% at 9, 86% at 22 and 99% at 35%. This could contribute to the heterogeneity or potential bias towards having more of a specific ageing signature over another (Smurf versus non-Smurf).

At first glance, it is surprising that telomere dysfunction is not picked up as a main hallmark of ageing in the telomerase mutant model. However, this may simply be the reflection of the number of genes that have been directly implicated in each of these hallmarks. In specific, there are a lot less genes that one would identify as directly involved in telomere dysfunction, when compared to stem cell dysfunction, for example. The main culprits for telomere dysfunction would be telomerase and the shelterin components, whereas for stem cell dysfunction, all the cell cycle and DNA damage repair proteins can play a role. For example, *chek1, fxr1 and daxx*, which are all de-regulated in the old WT gut and brain (*chek1*) (**see source data**), have all been identified as potential regulators of telomeres (Nersisyan et al., 2019). Yet, because that is not the main function one would attribute to such genes, these would have been missed as part of the telomere dysfunction hallmark of ageing. Additionally, it is not possible to distinguish from just looking at lists of DEGs if such gene was de-regulated due to telomere dysfunction in the first place, or if its dysfunction will affect telomere function indirectly. These are some of the considerations that highlight the complexity of ageing and the non-linearity of how the hallmarks of ageing may drive ageing as well as each other and it is important to have them in mind when interpreting our analyses. Nevertheless, at the later stages of WT ageing (>35 months), altered intercellular communication, a previously described “culprit of the phenotype” is indeed the most significant hallmark of ageing detected in common between the gut and brain. Finally, in terms of the progression hallmarks of ageing, different hallmarks may play more prominent roles at specific times of life, and may be replaced by others at other times, explaining why we don’t necessarily always see a linear accumulation of the different hallmarks of ageing over-time. This is particularly evident in the gut, where stem cell exhaustion is the main hallmark identified by the combined STEM and ALL DEGs analysis, and it is barely represented at the later stages of life. In the brain, the progression seems to

be more linear, though, and most hallmarks present at the early time point of 9 months remain until old age, when other hallmarks are further added.

3.3 Telomerase- and/or telomere-dependent changes with ageing

If a gene expression changes or a biological process alteration is accelerated in the absence of telomerase (i.e becomes significant at an earlier age in the *tert*$^{-/-}$), we considered it to be telomerase dependent. If we take a step back and look at the processes and, in turn, hallmarks of ageing affected in the absence of telomerase, it becomes clear that the *tert*$^{-/-}$ is indeed accelerating many of these changes. It is particularly evident when we look at the pie charts depicting the different hallmarks of ageing affected at each time point in WT versus *tert*$^{-/-}$. Here, it is striking how a 2-month-old *tert*$^{-/-}$ gut resembles a WT gut at the old age of 35 months, and how a 9- and 22-month-old *tert*$^{-/-}$ brain resembles an old WT brain at 35 months. This is further highlighted by the further comparison we performed, where we asked at which age does the *tert*$^{-/-}$ share more gene expression changes with the WT old (>35 months). In this analysis, we show that the 2 months old *tert*$^{-/-}$ has the most gene expression changes shared with the 35 months old WT gut, whereas it is the *tert*$^{-/-}$ at 22 months that has the most gene expression changes shared with the 35 months old WT brain. This suggests that the gut is displaying telomerase-dependent hallmarks of ageing at an earlier age than the brain, which is consistent with what would be expected for a high *versus* low proliferative tissue. Accordingly, when looking at the specific gene changes accelerated in the absence of *tert*, i.e, we identify significantly more *tert*-dependent genes located near the ends of chromosomes in the gut than in the brain, and therefore more liable to have been altered due to telomere shortening. In the future, it would be insightful to test how many of the *tert*-dependent gene changes occur due to non-canonical functions of telomerase involved in transcriptional regulation.

Nevertheless, *tert*-dependent gene changes in both tissues are still a minority, serving as a reminder that “all roads lead to Rome”, and one should exercise caution when trying to identify genes influencing the natural process of ageing using premature ageing models. It is not necessarily the same genes influencing the processes of ageing in these models, even if the processes and phenotypes are accelerated. We should also have this in mind when we compare the sets of genes identified in this study with those previously identified as

implicated in human ageing, and not necessarily be surprised if only a small percentage of those are shared. One could argue that it is more important that the processes are shared. Nevertheless, we do identify many gene changes shared between zebrafish and human ageing databases as highlighted throughout the results' section.

3.4 Which genes to focus on if we were to target age-associated dysfunction in the gut and brain.

As a final step in our analysis, we wanted to identify common gene expression changes between the gut and the brain at the earliest stages of ageing, in our case, from the age of 9 months. We hypothesised that this may highlight potential common therapeutic targets for early intervention to prevent age-associated dysfunction in both tissues. We identified 7 significant DEGs in common between the gut and brain at 9 months of age, 5 down-regulated (*ccnb1, kif2c, serpinh1a, si:ch211-5k11.8 and tmem37*) and 2 up-regulated (*cfap45 and eif4ebp3l*). We could then hypothesise that restoring expression of these genes to youthful levels would have a positive impact on delaying or ameliorating the development of ageing phenotypes in both these tissues at the same time. Of note, we identified that, amongst these, *ccnb1* and *eif4ebp3l* were telomerase-dependent changes in the brain. If so, one could hypothesise that telomerase re-activation in the brain should restore these genes to young WT levels and potentially contribute to amelioration of *ccnb1*- and *eif4ebp3l*-dependent ageing phenotypes in the brain.

4 CONCLUSIONS

We provide the first systematic analysis of transcriptomic changes throughout the lifespan of zebrafish in the gut and brain of the same group of individuals, leading to the identification of key genes and processes likely involved in driving the ageing process in these tissues. Many of these genes have previously identified in human ageing databases and many of them are potentially new genes of ageing, which will have to be experimentally tested in relevant model organisms. This analysis and the open access availability of its source and raw data should provide a key steppingstone and framework supporting future work for understanding the ageing process and using zebrafish for studying ageing.

5 MATERIALS AND METHODS

591 **5.1 Zebrafish husbandry, genotypes and ages**

592 Zebrafish were maintained at the standard conditions of 27-28ºC, in a 14:10 hour
593 light-dark cycle, and fed twice a day with *Artemia* (live rotifers) and Sparus (dry food). All
594 the experiments were performed in the University of Sheffield. All animal work was
595 approved by local animal review boards, including the Local Ethical Review Committee at
596 the University of Sheffield (performed according to the protocols of Project Licence
597 70/8681).

598 Two strains of adult zebrafish (*Danio rerio*) were used for these studies: wild-type
599 (WT; AB strain) and $tert^{-/-}$ ($tert^{AB/hu3430}$). Wild-type fish were obtained from the Zebrafish
600 International Resource Center (ZIRC). The *telomerase* mutant line $tert^{AB/hu3430}$ was generated
601 by *N*-ethyl-nitrosourea mutagenesis (Utrecht University, Netherlands(Wienholds et al.,
602 2003)). It has a *T*→*A* point mutation in the *tert* gene and is available at the ZFIN repository,
603 ZFIN ID: ZDB-GENO-100412–50, from ZIRC. The fish used in this study are direct descendants
604 of the ones used previously[29,30], by which point it had been subsequently outcrossed five
605 times with WT AB for clearing of potential background mutations derived from the random
606 ENU mutagenesis from which this line was originated. The $tert^{hu3430/hu3430}$ homozygous
607 mutant is referred to in the paper as $tert^{-/-}$ and was obtained by incrossing
608 our $tert^{AB/hu3430}$ strain. Genotyping was performed by PCR of the *tert* gene[13,14]. In order to
609 study age-related phenotypes in the zebrafish gut and brain, we used >30 months old fish
610 for what we consider old in WT (in the last 25-30% of their lifespan), and we considered the
611 $tert^{-/-}$ old fish at the equivalent age (>22 months), which approximately corresponds to the
612 last 25-30% of their lifespan. In specific, 'Old' was defined as the age at which the majority
613 of the fish present age-associated phenotypes, such as cachexia, loss of body mass and
614 curvature of the spine. These phenotypes develop close to the time of death and are
615 observed at >30 months of age in WT and at >22 months in $tert^{-/-}$(Madalena C Carneiro et
616 al., 2016; Henriques et al., 2013).

617

618 **5.2 Tissue dissection and RNA extraction**

619 Animals from various age-groups were used for RNA-Sequencing (WT at 2, 9, 22 and
620 >30 months of age; and $tert^{-/-}$ at 2, 9 and 22 months of age). Three biological replicates were
621 used per group. Fish were culled by overdose of MS-222, followed by confirmation of death.

Then, the whole tissues were rapidly dissected in cold PBS (Sigma-Aldrich), transferred to a microcentrifuge tube containing 100 µl of Trizol (Thermo Fisher Scientific), snap-frozen in dry ice and stored at -80°. To isolate the RNA, extra 50 µl of Trizol was added to each sample, and the tissue was homogenized with a mechanical homogenizer (VWR International) and a 1.5 pestle (Kimble Chase, Vineland, NJ, USA). After 5 min incubation at room temperature (RT), 30 µl of chloroform (1:5, VWR International) was added and the samples were incubated for further 3 min at RT before centrifuged at 13,000g, for 30 min, at 4°C. Isopropanol (Thermo Fisher Scientific) was then added to the aqueous phase of the solution, and the resultant mix was incubated for 10 min at RT, before centrifuged (13,000g for 15 min at 4°C). Finally, the pellet was twice washed in 250 µl of ice cold 75% ethanol and left to air-dry, before resuspended in 14 µl of nuclease-free water. RNA integrity was assessed with the bioanalyzer Agilent 2100 Bioanalyzer (Agilent, Santa Clara, CA, USA). All the samples had a RNA integrity number (RIN) ≥9.

5.3 RNA-Sequencing

The RNA-Sequencing was performed at the *Genomics and Sequencing facility* of Sheffield Institute for Translational Neuroscience (SITraN). Library preparation was performed following the Illumina methodology. mRNA was extracted from 500 ng of RNA with oligo-dT beads, capturing poly(A) tails, using the NEBNext Poly(A) mRNA Magnetic Isolation Module (New England Biolabs Inc). cDNA libraries were made with the NEBNext® Ultra™ RNA Library Prep Kit for Illumina, following the manufacturer's instructions (New England Biolabs Inc). The samples were then sequenced on an Illumina HiSeq SQ, using a high output run and sequencing by synthesis V3 chemistry on a single 100 bp run. The data was imported into a FASTQ file format in order to perform the analysis.

5.4 RNA-Sequencing analysis

5.4.1 Data processing. Quality control was performed using MultiQC version 1.9. Cutadapt version 3.0 was used for trimming the first 13 bases from the reads to remove poor quality base pairs in the reads. Read alignment was performed as follows. Single-end reads were aligned against the reference genome Danio_rerio.GRCz11.dna. primary_assembly.fa using STAR. A bespoke alignment index was built using annotation file

653 Danio_rerio.GRCz11.103.gtf and an expected read length of 88 bp. Ht-seq count was run in
654 non-stranded mode to obtain *per* gene read counts.

655 **5.4.2 Differential expression.** To identify signatures of ageing, WT gut and brain
656 samples were subjected separately to DESEq2 analysis, comparing the time points 9, 22 and
657 >30 months with the time-point of 2 months. Then, *tert*$^{-/-}$ samples at 2, 9 and 22 months
658 were contrasted with WT at 2 months, in order to identify telomerase-dependent ageing
659 processes.

660 **5.4.3 Time-series analysis.** Short Time-series Expression Miner (STEM) software was
661 used to assign genes to predefined expression profiles genes based on their expression
662 across the time points. For this, the 2 months' time-point was used as the zero time point
663 for the analysis. The maximum number of model profiles was set to 50; the maximum unit
664 change in model profiles between time points was set to 2; and the minimum absolute log
665 ratio of expression was set to 1.0, with change based on maximum − minimum. Significance
666 level of the model profiles was set to 0.05 with Bonferroni correction. Minimum correlation
667 for profile clustering was set to 0.7. The statistically significant temporal profiles were
668 visualised as line plots using ggplot2. Median expression fold change values of the genes in
669 each profile were shown on the plots with a thicker line.

670 **5.4.4 Enrichment analysis of temporal profiles from STEM.** Enrichment analysis was
671 performed using all the differentially expressed genes (DEGs) and using the genes identified
672 in the STEM analysis, separately. Gee-set over-representation analysis (ORA) of GO
673 Biological Process (GOBP), GO Molecular Function (GOMF), and GO Cellular Compartment
674 (GOCC) terms were performed using the enrichGO function of clusterProfiler package
675 version 3.18.0. Minimum and maximum gene set sizes were set to 10 and 500, respectively.
676 Results were simplified using the simplify function of clusterProfiler with the similarity cut-
677 off set to 0.7 and minimum adjusted p-value used as the feature for selecting the
678 representative terms. Enrichment results with adjusted p-values below 0.05 and at least 3
679 core enrichment genes were considered significantly enriched. ORA of KEGG and
680 REACTOME pathway terms were performed using the enrichKEGG and enrichPathway
681 functions of clusterProfiler. Minimum and maximum gene set sizes were set to 10 and 500,
682 respectively. The same significance criteria for the enrichments were used as for the GO
683 term enrichments. Results of the enrichment results were visualised as barplots or as pie

charts. Barplots were made using the ggbarplot function of R package (R studio version 2021.09.2), ggpubr version 0.4.0, showing a maximum of five pathways with adjusted p-value below 0.05 *per* pathway category. Pie charts were made using Prism GraphPad version 9.0.

5.4.5 Protein-protein interaction network analysis

The search tool for retrieval of interacting genes (STRING) (https://string-db.org) database, which integrates both known and predicted PPIs, was used to predict functional interactions of proteins(Szklarczyk et al., 2021). Active interaction sources, including text mining, experiments, databases, and co-expression as well as species limited to "Danio rerio" and an interaction score > 0.7 (high confidence) were applied to construct the PPI networks. The network was further clustered using *k-means* clustering to a specific number of up to 3 clusters.

5.4.6 Venn Diagram analysis

Venn Diagram analysis was performed using the online tool Venny 2.1.0 – BioinfoGP (https://bioinfogp.cnb.csic.es/tools/venny/)(Oliveros, 2007-2015).

5.4.7 Statistical analysis

A chi-square test was used in **Fig 5** to compare between the relative chromosome location of telomerase-dependent genes in the gut and brain, using raw data values, even though it is the % that is represented in the graphs, to ensure accurate analysis. P value <0.05 was considered significant and denoted with a *, whereas ns was used to denote non-significant differences.

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AUTHOR CONTRIBUTIONS

RRM and CMH conceived and designed this work; RRM performed additional RNA sequencing beyond that provided by Genevia services; RRM, MR and CMH analysed and interpreted RNA sequencing results and co-wrote the manuscript; CMH designed the figures with input from co-authors.

COMPETING INTERESTS

The authors declare no competing interests

DATA AVAILABILITY STATEMENT

The RNA sequencing data from this experiment were deposited in gene expression omnibus (GEO) and will be made available when in peer review.

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SUPPORTING INFORMATION

Supporting information in the form of figures, tables and source data can be found online in the Supportive information section at the end of the article

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FIGURE LEGENDS

Graphical Abstract: Telomerase depletion accelerates the appearance of hallmark of ageing in both gut and brain. In specific, the *tert*$^{-/-}$ gut at 2 months shows similar hallmarks of ageing as the WT old gut, whereas the brain *tert*$^{-/-}$ brain only *starts* displaying similar hallmarks of ageing as the WT old brain from the age of 9 months. We identified stem cell exhaustion is the common principal hallmark of ageing at the early stages of ageing, in both tissues. Finally, we further identified altered intercellular communication as the main telomerase-dependent hallmark of ageing common between the gut and brain.

Fig 1. Summary of the experimental design and principal component analysis (PCA). (A) RNA-Sequencing was performed in whole gut and brain tissues from WT and *tert*$^{-/-}$ zebrafish, at different timepoints throughout their lifespan. Age-associated transcriptomic changes were analysed using two different methods: time-series analysis (genes that change consistently overtime) and all differentially expressed genes (ALL DEGs; genes whose expression is altered over-time, in both genotypes, as compared with the WT young baseline. (B) PCA representing the variation in the data from gut (gold) and brain (pink) tissues, in both WT and *tert*$^{-/-}$ fish. (C-D) PCA showing the variation in the data from fish at different ages (2 months, pink; 9 months, green; 22 months, blue; 35 months, purple), in WT (circle) versus *tert*$^{-/-}$ fish (triangle), in (C) gut and (D) brain samples. PCA was performed using the plotPCA function of DESeq2 and considering the top 500 genes with highest variance across the samples. (E) Summary of the number of significantly de-regulated genes at each time-point, in both genotypes, in gut (E1) and brain (E2) tissues, over-time, as compared with the WT young baseline.

Fig 2. Identification of time-dependent signatures of ageing in the WT and *tert*$^{-/-}$ zebrafish gut and brain. (A, B) Transcriptomic temporal profiles in the (A) gut and brain (B) of WT (black) and *tert*$^{-/-}$ fish (red) were identified using the Short Time-series Expression Miner (STEM) and are represented in line plots. The thicker lines on the plots represent the median fold change of each profile. For each temporal profile, enrichment analysis was

performed using the enrichGO, enrichKEGG, and enrichPathway functions of clusterProfiler package version 3.18.0. Processes and pathways from each temporal profile were further classified and grouped according to the main hallmarks of ageing (Lemoine, 2021; López-Otín et al., 2013), which is represented in pie charts. (A1, B1) A summary of the enrichment analysis of one of the profiles is represented in the bar plots, where processes and pathways are represented in different colours according to the classification into hallmarks of ageing. This summary contains the top enriched processes and pathways from each database (p-value >0.05 and at least 3 core enrichment genes; up to 5 terms *per* database: GOBP, GOCC, GOMF, KEEG, and REACTOME).

Fig 3. Qualitative changes in the hallmarks of ageing over-time, comparing WT and *tert*$^{-/-}$ zebrafish gut and brain. The age-associated enriched processes and pathways identified in the previous figures were further re-classified and grouped into the main well-known hallmarks of ageing. (A-B) Pie charts represent the hallmarks of ageing identified in the (A) gut and (B) brain, at different ages throughout WT (black) and *tert*$^{-/-}$ (red) zebrafish lifespan. This analysis was performed considering the (A1, B1) genes identified in the temporal profiles (within STEM), (A2, B2) all the genes differentially expressed at any timepoint (within ALL DEGs), and (A3, B3) STEM and ALL DEGs combined. (A4, B4) The number of transcriptomic changes increases with ageing in both (A4) gut and (B4) brain, independently of the phenotype, when considering either the genes within STEM or the genes within ALL DEGs.

Fig 4. Determining at which age *tert*$^{-/-}$ share more genes associated with the hallmarks of ageing, with the naturally aged WT. (A) Venn diagrams highlight the number of genes associated with hallmarks of ageing in common between old WT (35 months) and *tert*$^{-/-}$ at the different ages tested, in (A.1) gut and (A.2) brain tissues. Data show that the *tert*$^{-/-}$ gut at 2 months has the most number genes associated with hallmarks of ageing in common with old WT gut. The *tert*$^{-/-}$ brain at 22 months has the most number genes associated with hallmarks of ageing in common with old WT brain. The respective lists of genes shared between old WT gut and *tert*$^{-/-}$ gut at 2 months and old WT brain and *tert*$^{-/-}$ 22-month brain is shown in B (B.1 gut, B.2, brain). (B1.1, B1.2) Gene networks with the genes identified in B.1 and B.2 were performed (B1.1 and B1.2, respectively) by K-means clustering using String

software. These include the genes identified in the temporal profiles (from STEM) and in ALL DEGs.

Fig 5. Identifying tert-dependent gene changes in zebrafish gut and brain ageing and their chromosome location in relation to the telomeric end. (A-B) Gene alterations that are anticipated in the *tert*$^{-/-}$ when comparing with WT at the same age (i.e., telomerase-dependent). (A1, B1) Protein-protein interaction network of these genes highlights clusters of genes associated with (A1) metabolic processes in the gut and clusters of genes associated with (B1) cell cycle, genome instability and immune system in the brain. (C)Genes located at the end of the chromosome (i.e., <10MB from chromosome end) are likely to be directly affected by telomere shortening due to the telomere positioning effect (TPE). The data show that there is no significant difference between telomerase-dependent and -independent genes, in what concerns their proximity to the chromosome end, in either gut (C2) or brain (C3). However, there is a significantly higher number of telomerase-dependent genes located at the end of chromosomes in the gut than in the brain (C4).

Fig 6. Determining genes and pathways altered with ageing that are in common between gut and brain. (A) Graph represents WT zebrafish lifespan and highlights 'early' and 'late' stages of ageing. (A.1) All genes identified in common between gut and brain at early (9 months) and late (35 months) stages of ageing. *G and *B represent telomerase-dependent genes in the Gut or in the Brain, respectively. (B-C) Protein-protein interaction networks with the genes found in common between the gut and brain at (B) early and (C) late stages of ageing were performed using STRING software. These include the genes identified in the temporal profiles (from STEM) and in ALL DEGs.

SUPPLEMENTARY FIGURE LEGENDS

Supplementary Fig 1. Genes associated with the hallmarks of ageing that are in common between 35-months-old WT and *tert*$^{-/-}$ at different ages, in the (A1) gut and (A2) brain. The different colours represent different hallmarks of ageing: purple, loss of proteostasis; yellow, mitochondrial dysfunction; green, altered intercellular communication. These include the genes identified in the temporal profiles (from STEM) and in ALL DEGs.

Supplementary Fig 2. Protein-protein interaction network and cluster analysis of gene changes in WT at the 'origins' versus 'later' stages of ageing. (A) Genes identified in common between gut and brain at 9 and 35 months of age. B) Genes significantly altered at "early" and "late" stages of ageing in the gut (B1, 2) and brain (C1, 2), respectively. Network analysis was performed in STRING software and included the genes identified in the temporal profiles (from STEM) and in ALL DEGs. Red squares highlight the protein-protein interaction network of the gene changes that are accelerated in the *tert*$^{-/-}$ (tert-dependent).

Graphical Abstract

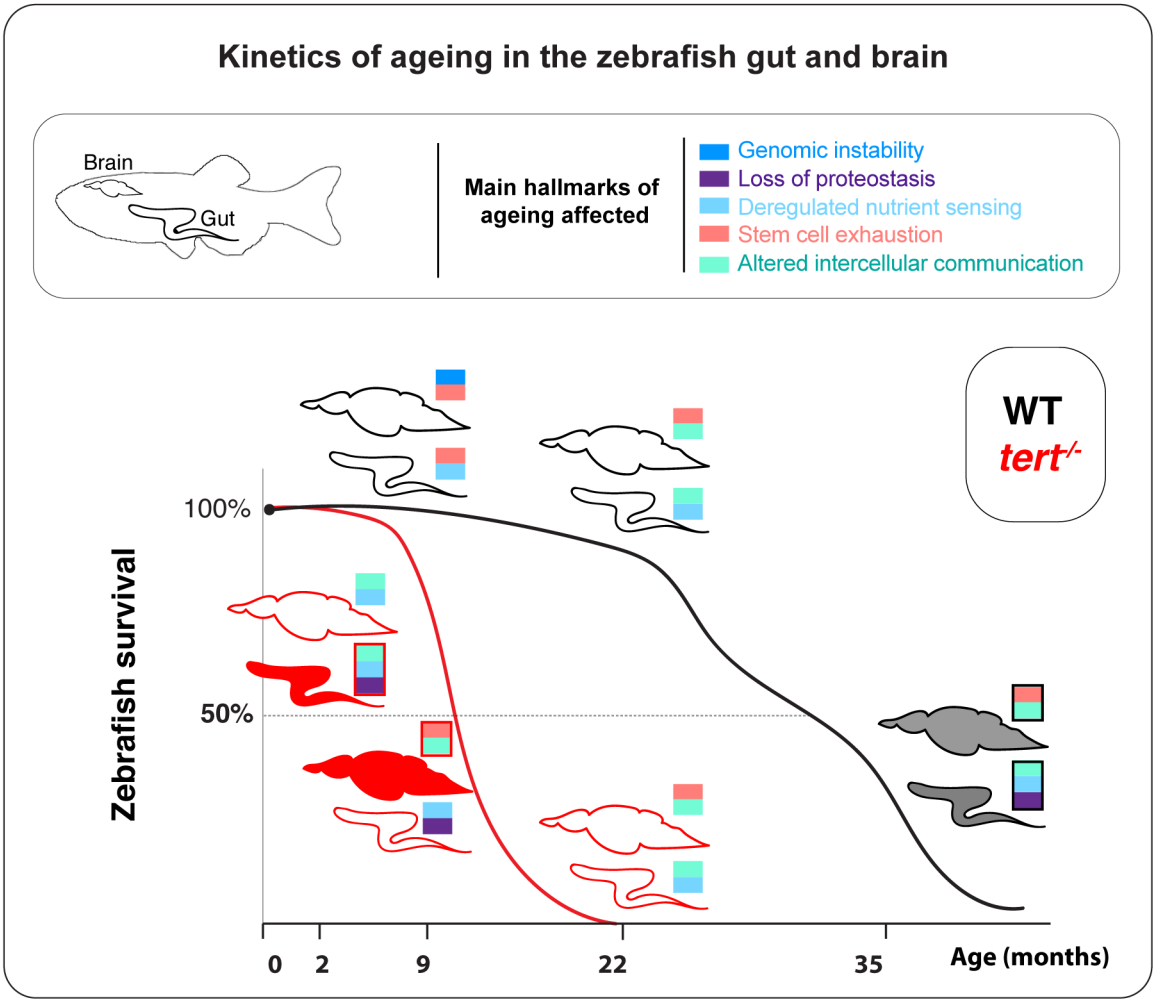
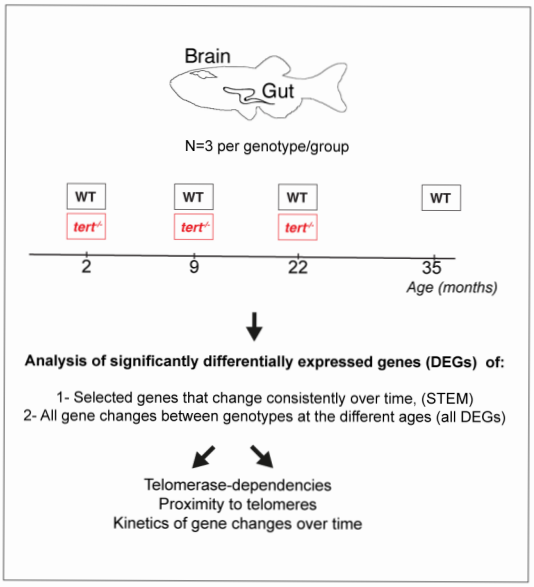
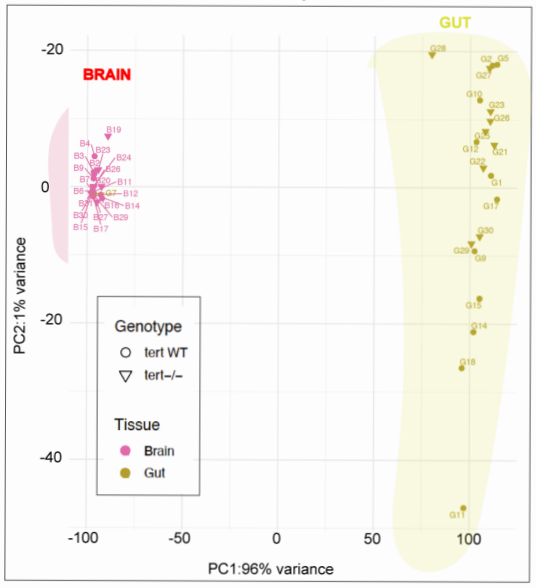


FIGURE 1

A) Study design

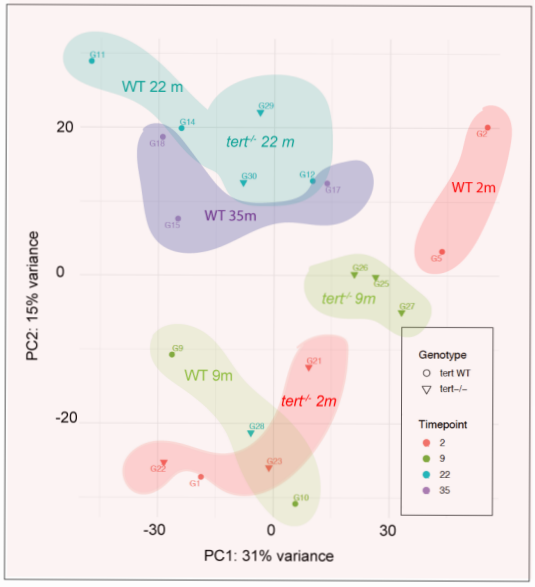


B) Principal Component Analysis (PCA) of all samples



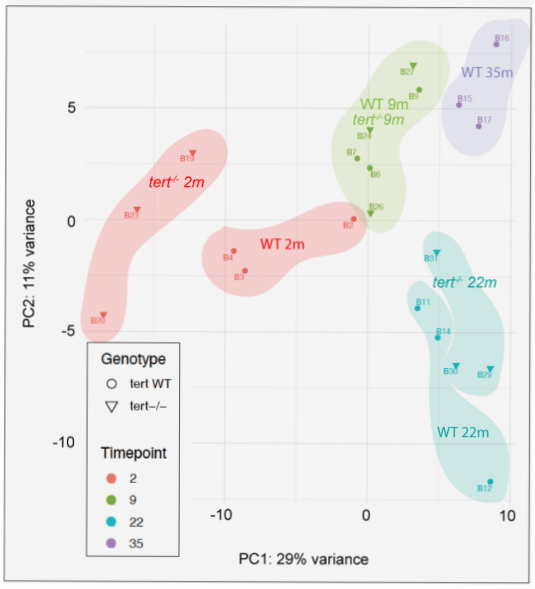
C) Gut

PCA of gut samples



D) Brain

PCA of brain samples



E) Summary of significant Differentially Expressed Genes (DEGs) numbers

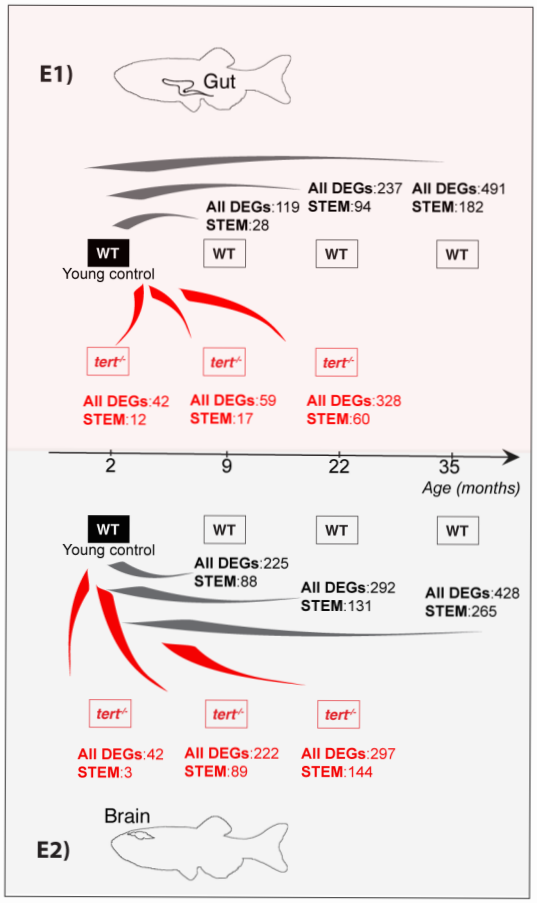


Figure 2 Signficant STEM profiles and main processes associated, categorised according to known hallmarks of ageing

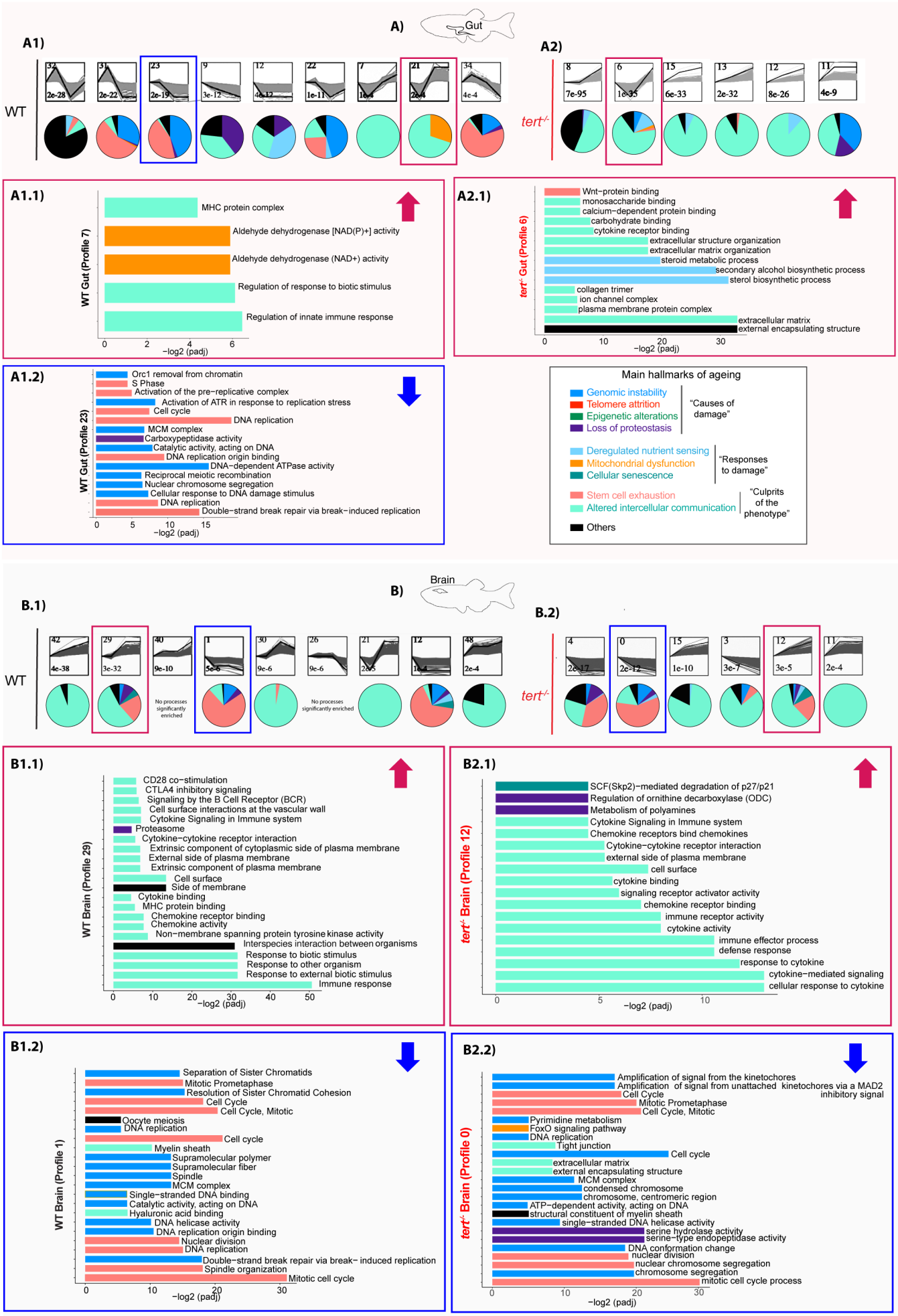


Figure 3 Kinetics of ageing in the zebrafish gut and brain

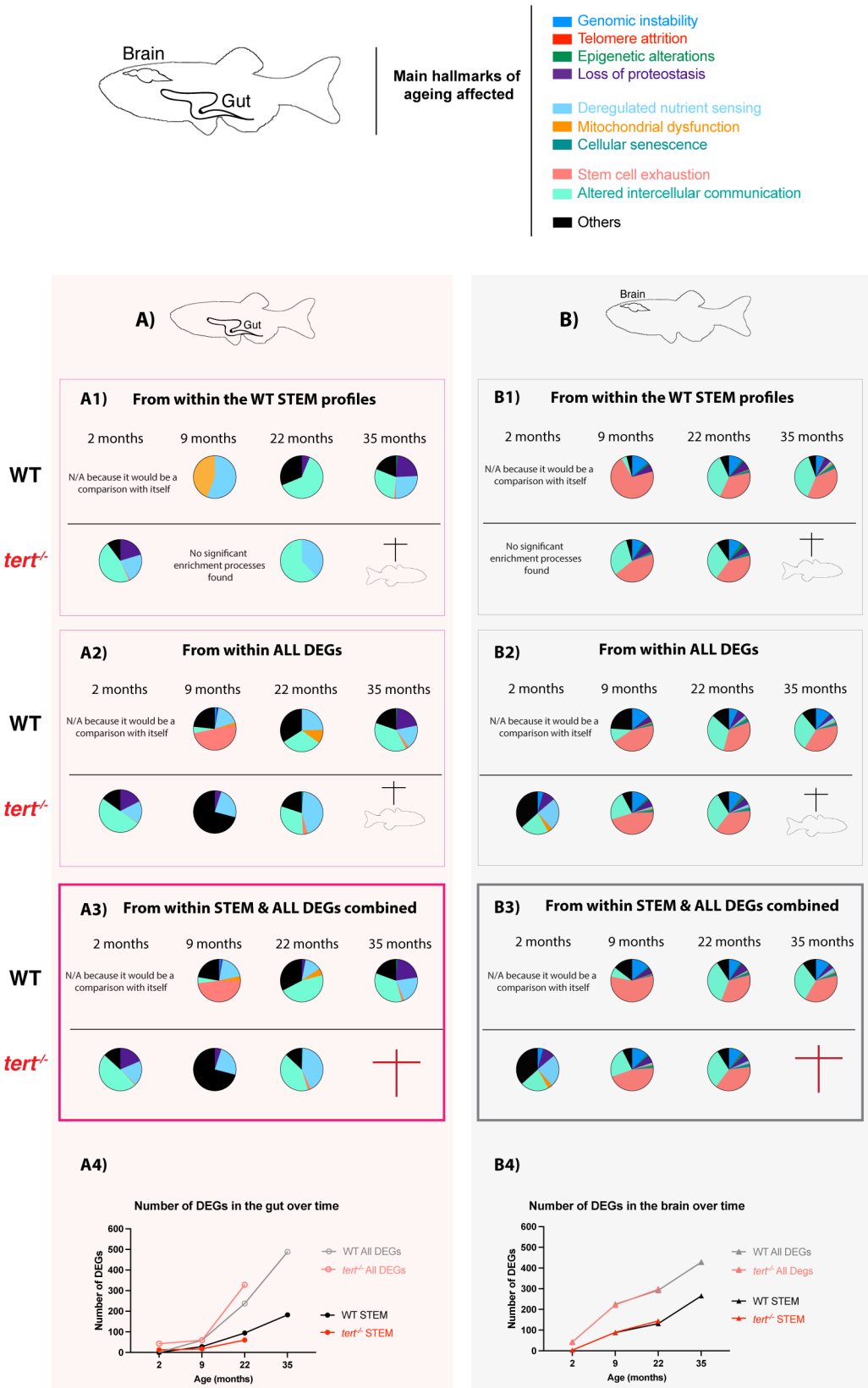
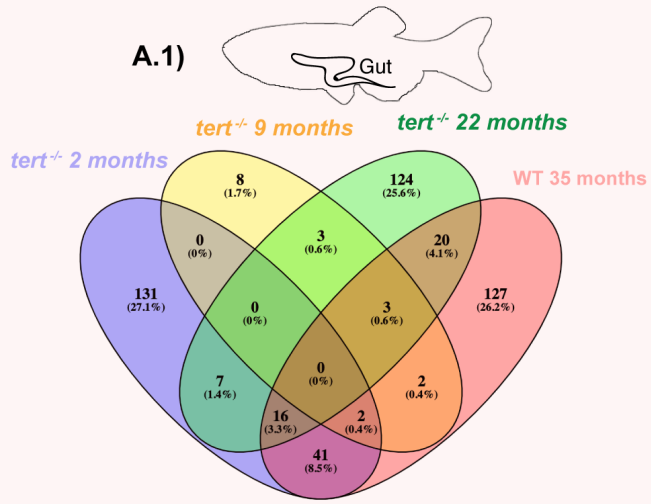
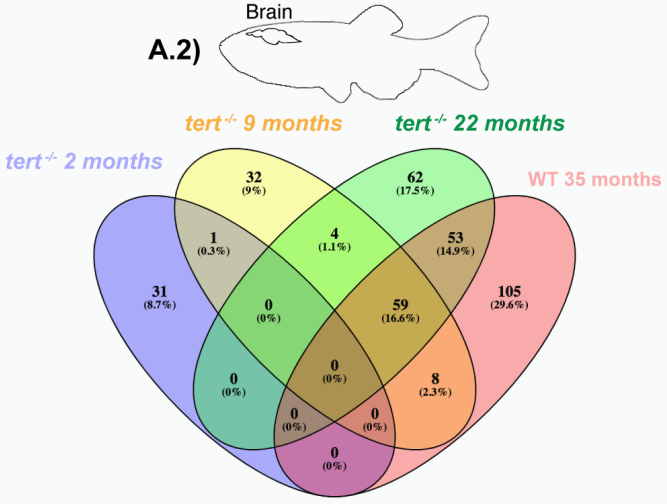


Figure 4

A) Genes associated with the hallmarks of ageing in common between *tert*$^{-/-}$ at the different ages and the aged WT at 35 months



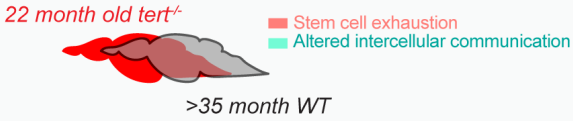


B) Most gene changes shared between *tert*$^{-/-}$ and 35 month old WT, related to the main hallmarks of ageing affected

B.1) GUT

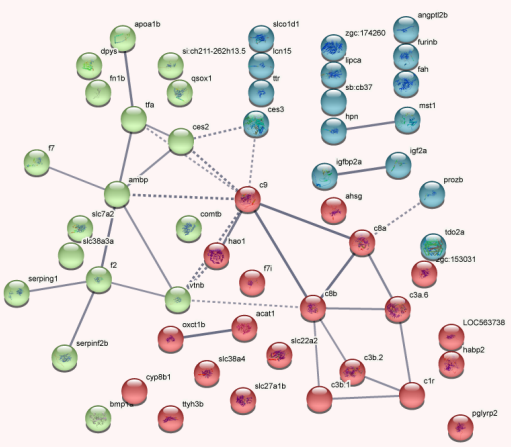
angptl2b	*acat1*	*ahsg1*
apoa1b	*ces2a*	*ambp*
c8a	*ces3*	*bmp1a*
c8b	*comtb*	*c1r*
c9	*dpys*	*c3a.6*
cyp8b1	*fah*	*c3b.1*
lcn15	*hao1*	*c3b.2*
pglyrp2	*igf2a*	*f2*
slc22a2	*igfbp2a*	*f7*
slc27a1b	*lipca*	*f7i*
slc38a3a	*pla1a*	*fn1b*
slc38a4		*furinb*
slc7a2		*habp2*
slco1d1		*hpn*
tdo2a		*mst1*
tfa		*oxct1b*
ttr		*prozb*
ttyh3b		*qsox1*
vtnb		*sb:cb37*
zgc:153031		*serpinf2b*
zgc:174260		*serping1*
		si:ch211-262h13.5
		si:ch211-284e20.8

B.2) BRAIN



aspm	*insm1a*	*kif11*	*acaa1*
aurkb	*mcm4*	*kif14*	*anln*
birc5a	*mcm5*	*kif20a*	*b2m*
ccna2	*mcm6*	*kif2c*	*batf*
ccnb1	*ncapg*	*kif4*	*ccl38.6*
ccnjl	*ncapg2*	*kifc1*	*cd40*
cdc20	*ncaph*	*lgals9l1*	*cd74a*
cdk2	*nusap1*	*mad2l1*	*ch25h*
cdk6	*prc1b*	*mcm2*	*csf1b*
cenpe	*primpol*	*ncapd3*	*cxcl18b*
chaf1a	*rad54l*	*ndc80*	*grem2b*
chek1	*rrm2*	*tuba8l3*	*h3f3b.1*
ckap2l	*sgo1*		*h3f3c*
dla	*si:ch211-244o22.2*		*hbba2*
dlb	*smc4*		*il2rb*
dld	*spc24*		*irf8*
ercc6l	*tacc3*		*lifrb*
fosab	*top2a*		*mhc2a*
gadd45gb.1	*tpx2*		*psmb8a*
h2af1al	*ttk*		*psme2*
hapln1a	*tubb5*		*pyyb*
hmmr	*uhrf1*		*rac2*
	wdhd1		*racgap1*
			si:busm1-266f07.2
			spi1a
			stat4
			tlr2
			wasb

B1.1) String Network analysis (K means clustering) GUT



B.2.1) String Network analysis (K means clustering) BRAIN

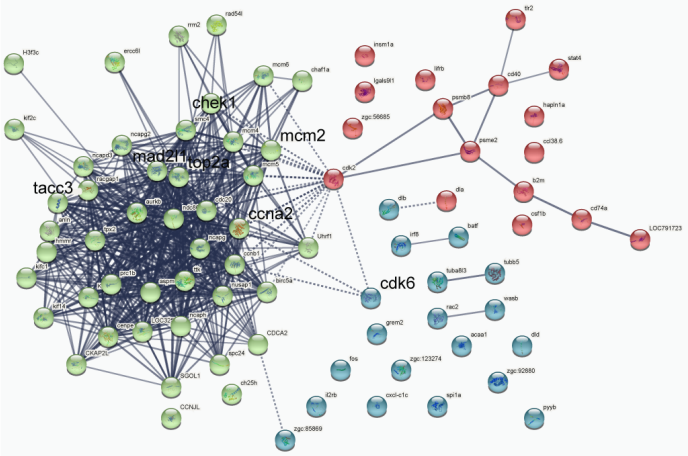


Figure 5

tert-dependent gene changes of old age

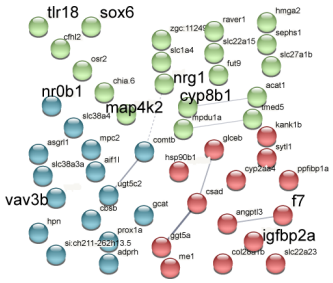
A)

↑
cbsb
col28a1b
cyp2aa4
ggt5a
kank1b
nrg1
raver1
sytl1
tlr18

↓
acat1
adprh
aif1l
angptl3
asgrl1
cfhl2
chia.6
comtb
csad
cyp8b1
f7
fut9
gcat
glceb
hbba1
hmga2
hpn
hsp90b1
igfbp2a
map4k2
me1
mpc2
mpdu1a
nr0b1
oaz2a
osr2
ppfibp1a
prox1a
sephs1
si:ch211-262h13.5
slc1a4
slc22a15
slc22a23
slc27a1b
slc38a3a
slc38a4
sox6
tmed5
ugt5c2
vav3b
zgc:112492

A1)

Protein-protein interaction (PPI) network and cluster analysis



Organic acid metabolic process
Oxoacid metabolic process
Carboxylic acid metabolic process
Metabolic pathways
Monocarboxylic acid metabolic process
Taurine and hypotaurine metabolism
PPAR signaling pathway

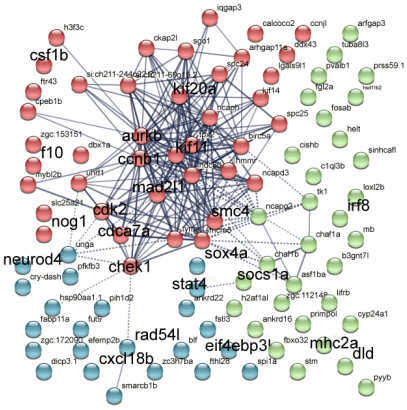
B)

↑
ankrd22
arfgap3
blf
c1ql3b
calcoco2
cishb
cpeb1b
cry-dash
csf1b
cxcl18b
cyp24a1
ddx43
dicp3.1
dre-mir-29b-1
efemp2b
eif4ebp3l
fbxo32
fcer1gl
fgl2a
fstl3
fthl28
ftr43
fut9
hsd11b2
irf8
lgals9l1
lifrb
mb
mhc2a
pfkfb3
pih1d2
pyyb
slc25a21
socs1a
spi1a
stat4
stm
zc3h7ba
zgc:112148
zgc:153151
zgc:172090

↓
ankrd16
arhgap11a
asf1ba
aurkb
b3gnt7l
birc5a
ccnb1
ccnjl
cdca7a
cdk2
chaf1a
chaf1b
chek1
ckap2l
dbx1a
dld
f10
fabp11a
fosab
h2af1al
h3f3c
helt
hmmr
hsp90aa1.1
iqgap3
kif11
kif14
kif20a
loxl2b
mad2l1
mcm5
mybl2b
ncapd3
ncapg2
ncaph
ndc80
neurod4
nog1
primpol
prss59.1
pvalb1
rad54l
sgo1
si:ch211-244o22.2
si:ch211-69g19.2
sinhcafl
smarcb1b
smc4
sox4a
spc24
spc25
tk1
tpx2
tuba8l3
tyms
uhrf1
unga
zwi

B1)

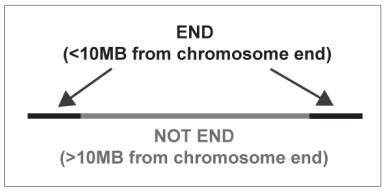
Protein-protein interaction (PPI) network and cluster analysis



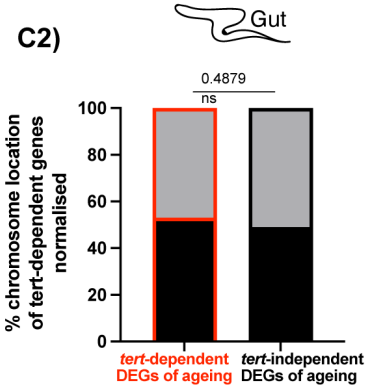
Cell cycle
genome stability
Immune system

C) Chromosome location of *tert*-dependent versus independent gene changes in the gut and brain

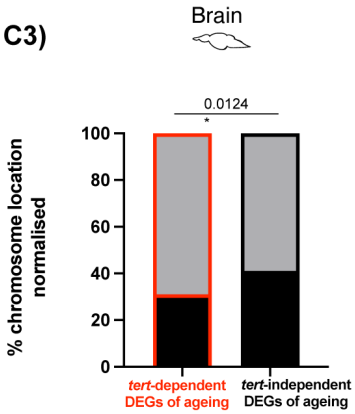
C1)



C2)



C3)



C4)

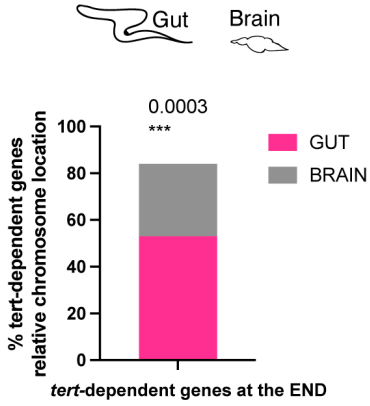
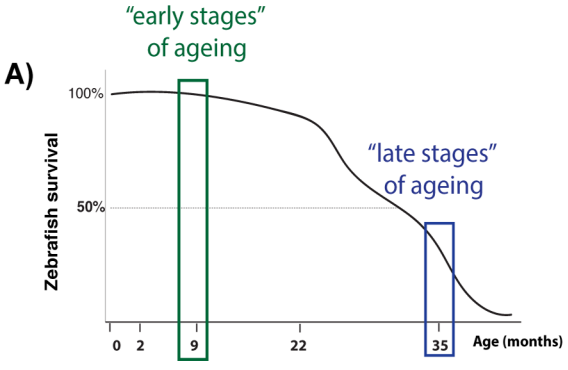
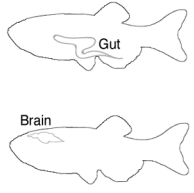


FIGURE 6

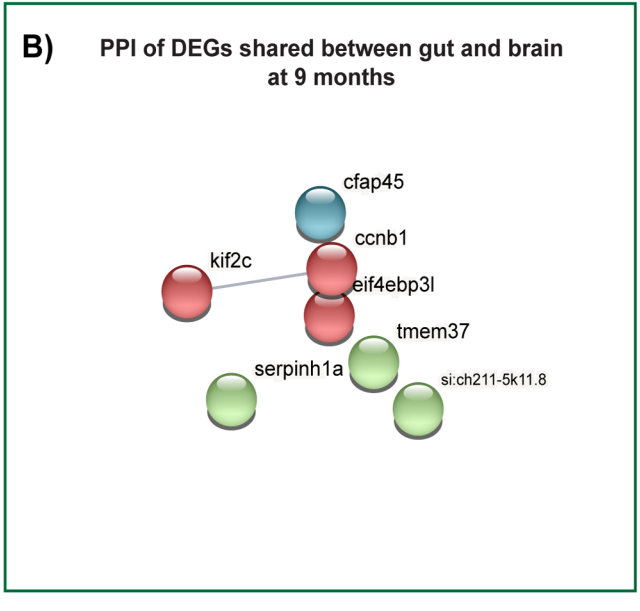
Protein-protein interaction (PPI) network and cluster analysis of changes in WT ageing (STEM and All DEGs combined)

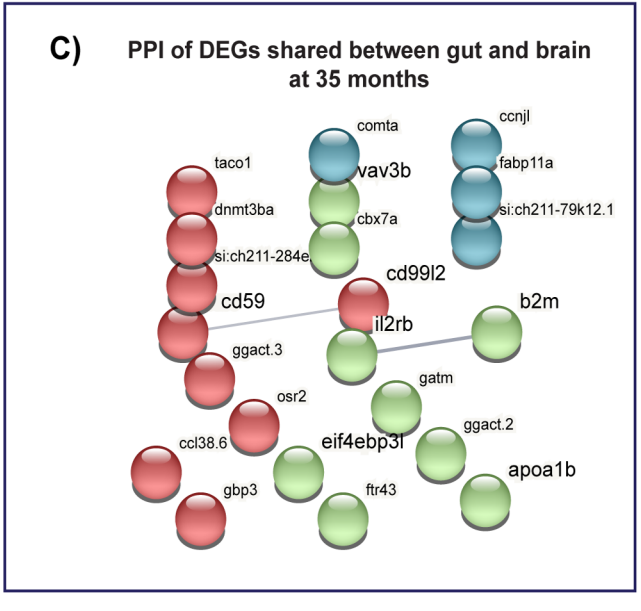


A.1) DEGs shared between gut and brain



9 months old	35 months old	
*ccnb1 (*B)*	*apoa1b*	*ftr43 (*B)*
cfap45	*b2m*	*gatm*
*eif4ebp3l (*B)*	*cbx7a*	*gbp3*
kif2c	*ccl38.6*	*ggact.2*
serpinh1a	*ccl38a.4*	*ggact.3*
si:ch211-5k11.8	*ccnjl (*B)*	*il2rb*
tmem37	*cd59*	*osr2 (*G)*
	cd99l2	*si:ch211-284e20.8*
	comta	*si:ch211-79k12.1*
	dnmt3ba	*taco1*
	*eif4ebp3l (*B)*	*vav3 (*G)*
	*fabp11a (*B)*	





Supp Fig 1

A) **Genes associated with the hallmarks of ageing in common between *tert*$^{-/-}$ at the different ages and the aged WT at 35 months**

A1)

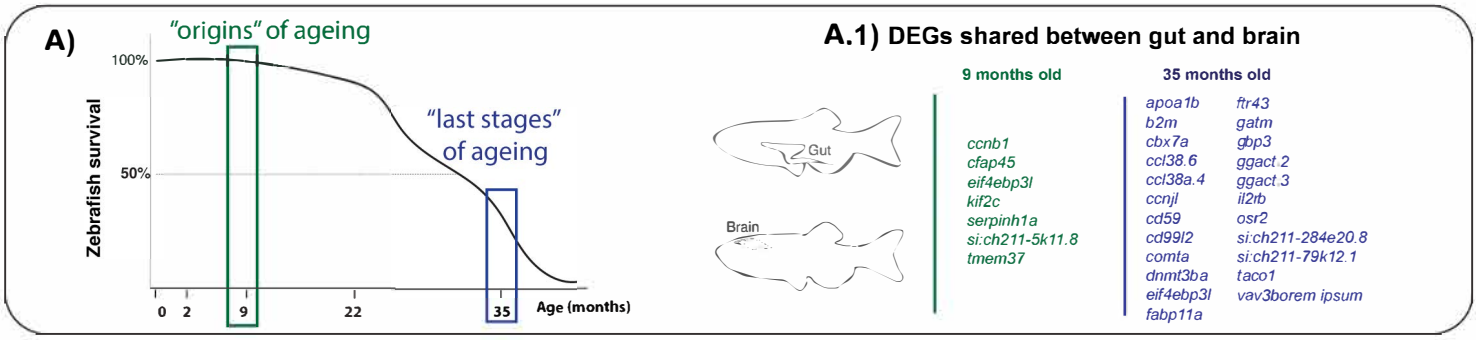
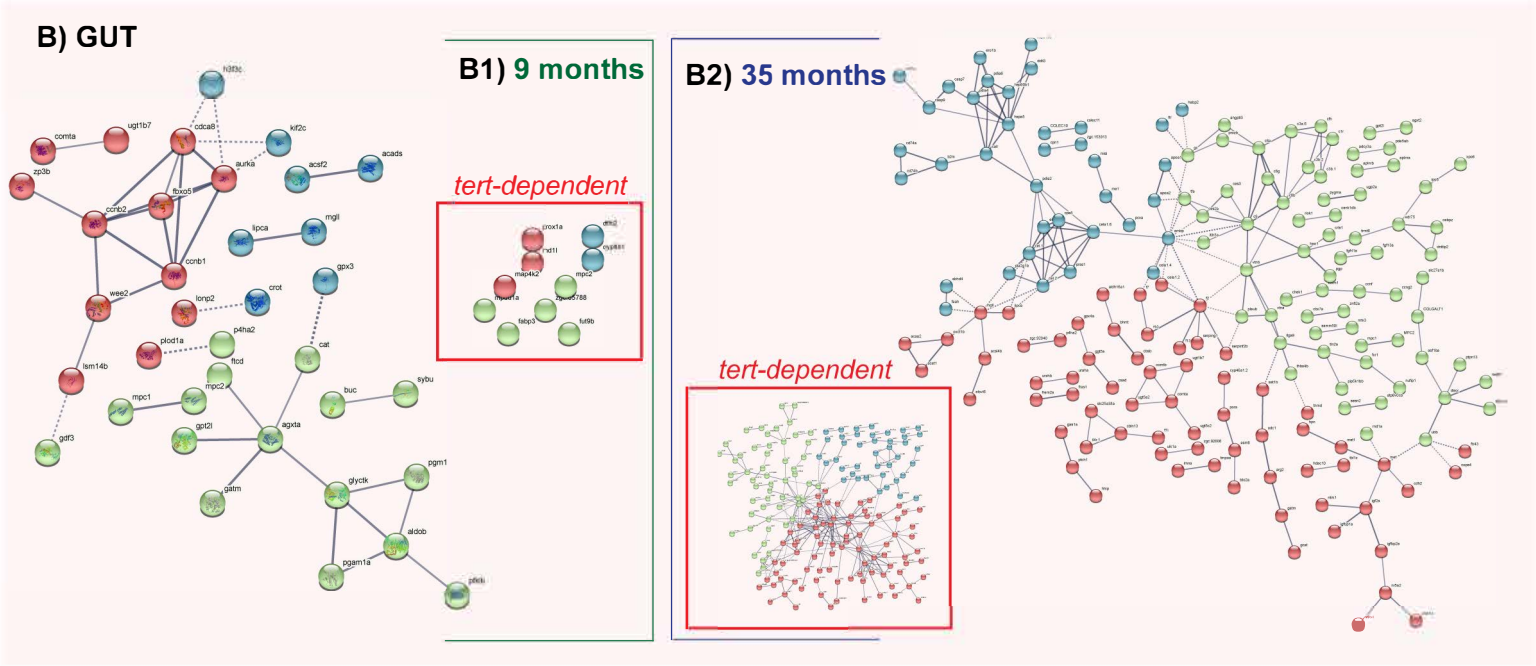
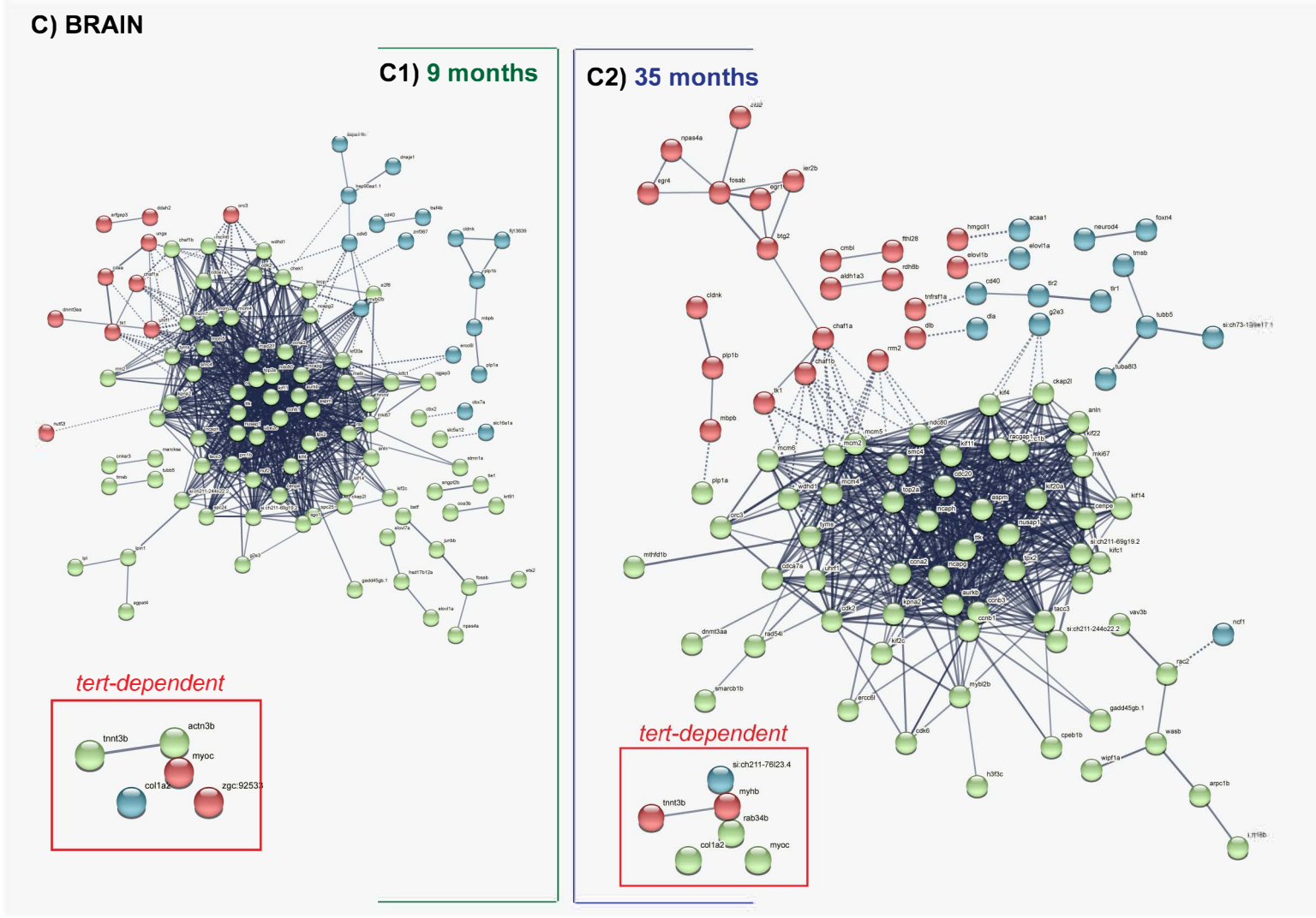
WT 35 months x

tert $^{-/-}$ 2 months		*tert* $^{-/-}$ 9 months	*tert* $^{-/-}$ 22 months
acat1	sb:cb37	acsf2	**acat1**
ahsg1	serpinf2b	**cyp8b1**	angptl3
ambp	serping1	hbaa2	comtb
angptl2b	**si:ch211-262h13.5**	mgll	f10
apoa1b	**si:ch211-284e20.8**	p4ha2	**f7**
bmp1a	slc22a2	pla1a	fam83d
c1r	**slc27a1b**	sephs1	gatm
c3a.6	**slc38a3a**		gbp3
c3b.1	**slc38a4**		gcat
c3b.2	**slc7a2**		ggt5a
c8a	**slco1d1**		glyctk
c8b	tdo2a		gpt
c9	tfa		**hpn**
ces2a	**ttr**		itln2
ces3	**ttyh3b**		me1
comtb	vtnb		mlsl
cyp8b1	zgc:153031		mpc1
dpys	zgc:174260		**nr0b1**
f2			nrg1
f7			oaz2a
f7i			pcbd1
fah			pcxa
fn1b			pfklb
furinb			**prox1a**
habp2			**si:ch211-262h13.5**
hao1			**si:ch211-284e20.8**
hbba1			slc16a6a
hpn			slc1a4
igf2a			**slc27a1b**
igfbp2a			**slc38a3a**
lcn15			**slc38a4**
lipca			slc43a1b
mst1			**slc7a2**
nr0b1			**slco1d1**
oxct1b			**ttr**
pglyrp2			**ttyh3b**
pla1a			urahb
prox1a			vkorc1
prozb			zgc:92040
qsox1			

A2)

WT 35 months x

tert $^{-/-}$ 9 months		*tert* $^{-/-}$ 22 months		
anln	**mcm6**	acaa1	h3f3b.1	**plp1b**
arhgap11a	**mybl2b**	**anln**	h3f3c	ppp1r14ba
aspm	**ncapd3**	asf1ba	**hapln1a**	prc1b
aurkb	**ncapg**	**aspm**	hbaa2	prdm8
ccna2	**ncapg2**	**aurkb**	hbba2	**primpol**
ccnb1	**ncaph**	b2m	hmmr	psmb8a
cd40	**ndc80**	batf	**igf2bp1**	psme2
cdc20	**neurod4**	birc5a	il2rb	pyyb
cdk2	**nusap1**	ccl38.6	inab	rac2
cdk6	orc3	**ccna2**	**insm1a**	racgap1
cenpe	**plp1b**	**ccnb1**	irf8	rad54l
chaf1b	**primpol**	ccnjl	**kif11**	**rrm2**
cyp24a1	**rrm2**	**cd40**	**kif14**	**sgo1**
dbx1a	**sgo1**	cd74a	**kif20a**	si:busm1-266f07.2
dld	**si:ch211-244o22.2**	**cdc20**	**kif2c**	**si:ch211-244o22.2**
dnmt3aa	**smc4**	**cdk2**	**kif4**	smarcb1b
dnmt3ba	socs1a	**cdk6**	**kifc1**	**smc4**
ercc6l	**sox4a**	**cenpe**	**lgals9l1**	sox11a
f10	**sox4b**	ch25h	**lifrb**	**sox4a**
gadd45gb.1	**spc24**	chaf1a	**mad2l1**	**sox4b**
hapln1a	spc25	chaf1b	**marcksa**	**spc24**
hsp90aa1.1	**spi1a**	chek1	**mbpb**	**spi1a**
igf2bp1	**tacc3**	ckap2l	**mcm2**	spinb
insm1a	**top2a**	csf1b	mcm4	stat4
kif11	**tpx2**	cxcl18b	**mcm5**	**tacc3**
kif14	**ttk**	dbx1a	**mcm6**	tap2a
kif20a	**tubb5**	dla	mhc2a	tlr2
kif2c	**tyms**	dlb	**mybl2b**	**top2a**
kif4	**uhrf1**	dld	**ncapd3**	**tpx2**
kifc1	unga	**dnmt3aa**	**ncapg**	**ttk**
lgals9l1		**dnmt3ba**	**ncapg2**	tuba8l3
lifrb		**ercc6l**	**ncaph**	**tubb5**
mad2l1		fosab	**ndc80**	**tyms**
marcksa		foxn4	nefmb	ube2c
mbpb		**gadd45gb.1**	**neurod4**	uhrf1
mcm2		grem2b	**nusap1**	wasb
mcm5		h2af1al	plp1a	wdhd1
				zwi

Supp Fig 2

Protein-protein interaction network and cluster analysis of changes in WT ageing (STEM and All DEGs combined)