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Original article

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β2-microglobulin is overexpressed in buccal cells of elderly and correlated with expression of p16 and inflammatory genes

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Abstract

β2M (Beta 2 microglobulin) is a small protein that is found in all nucleated cells, previous finding showed that its levels increased in the serum of the elderly. Buccal cell samples are none invasive approach for assessing the expression of target genes. There was rationality to assess the expression of β 2M in buccal cells of people of a different group of ages. Indeed, the expression of B2M increased significantly with fold change 3.40, 4.80, 6.60**, 8.20*** and 12.04*** for the group of age 18-25 years, 26-35 years, 36-45 years, 46-55 years, and 56-70 years respectively. The same observation was seen with markers of biological aging (p16^{INK4a}) with fold change 3.19, 3.90, 4.80*, 8.50*** and 12.40*** for the group of age 18-25 years, 26-35 years, 36-45 years, 46-55 years, and 56-70 years respectively. As expected, there was an increase in the inflammatory genes (IL-1 β and IL-6) expression in the elderly. Moreover, there was a direct significant correlation (r =90, p < 0.001) between $\beta 2M$ expression and age (years), and the same direct significant correlation between $p16^{INK4a}$ expression and age (years) was also seen (r =90, p<0.001). In addition, a direct correlation between β 2M and p16^{INK4a} was also seen (r =0.8.3, p<0.001), there was also direct correlation between β 2M and IL-1 β and IL-6 with (r =0.5, p<0.001; r =0.68, p<0.001) respectively. This evidence showed that β 2M increased in buccal cells of the elderly compared to younger, and thereby buccal cells can be exploited to assess biological aging by measuring β 2M

levels, however, large sample size and using another assessing method such as β 2M protein levels should be performed to confirm the results.

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Background:

 β 2M is a small protein that is expressed in all nucleated cells, previous data showed that its activity increases during inflammation [1]. β 2M interplays with cytokines for instance, IL-6, IL-8 and others intracellularly to induce inflammatory responses. In addition, it can bind and modulate the activity of growth factors and hormones and receptors [2, 3]. As a cancer promoting factor, it has been shown to be a growth factors and has been associated with cancer formation[4]. β 2M high levels stimulate stem cells via promoting IL-6 activity resulting in cancer cells invasion and metastasis [5]. On another context, it has been shown to have an apoptotic role in many liquid cancers [6, 7]. β 2M is accumulated in joints of patients are under renal dialysis for long term [8], genetic mutation in β 2M is associated with amyloid fibrils [9].

 β 2M has been exploited as a biomarker for many disorders [10-19]. There are several studies that link β 2M levels in serum and presence of cancers. In multiple myeloma (MM), β 2M is used for prognostic purposes [20, 21]. In addition, β 2M can also be measured in cerebrospinal fluid (CSF) to assess nervous system diseases [22]. β 2M has been shown to be high in serum of patients with late-stage prostate cancers. Increasing of β 2M levels in these patients might be because of positive effect of androgen that stimulates β 2M secretion [23].

In addition to the previous, in inflammatory bowel diseases, $\beta 2M$ can be used a diagnostic biomarker [24]. In infectious diseases such as cytomegalovirus and human immunodeficiency virus infections, $\beta 2M$ has been shown to elevated [25], which could explain its role in immune system. In bacterial infection such as Helicobacter pylori infection, $\beta 2M$ level is elevated in gastric biopsies that again underscores infiltration of inflammatory cells [26]. In organ donation, $\beta 2M$ has been used for monitoring for successful renal transplantation [27]. These evidences are examples of using of $\beta 2M$ as a biomarkers in clinical settings of many diseases. Previous study by our group showed that β 2M expressed highly in senescent cells (old cells) [28], recently, it has been shown by our group that β 2M expressed highly in blood samples of old people comparing to youngers. Journal Pre-proofs which could underscore a potential role in oxidative stress network [29]. Therefore, there is a rationality to test the expression of β 2M across different group of age using other easier source of sample such as buccal cells.

Buccal cells are epithelial cells that is similar to brain and skin in nature. They are originated from ectodermal differentiation during embryonic development. Buccal cells can be collected easily by different method that described previously[30-33]. These methods are showed high number of cells that can be used for different biological assays [34, 35]. Comparing to other sample methods, buccal cell samples are less invasive and very easy to collect. In addition, buccal cells are very stable after isolation from mouth[36], which makes them easy to process and analyze. Moreover, buccal cells are easy to preserve in buffer [37]. These make them an easy source for diagnosis. In this study, we used buccal cells to examine the expression of β 2M in different age groups. In addition, we correlated the expression of β 2M with p16^{INK4a} a biological biomarker of aging[38-40]

Methods

2.1. The study subjects

A cross-sectional study design was used to gather data from 81 participants, who were then divided into five age groups. In the first group, participants were aged 18 to 25, in the second, 26 to 35, in the third, 36 to 45, in the fourth, 46 to 55, and in the fifth, 56 to 70. HAPO-02-K-012-2021-03-600 is the approval number for the Medical Ethics Committee of the Faculty of Medicine at Umm Al-Qura University, which approved the sampling procedures in accordance with the Declaration of Helsinki of 1975 and the written informed consent of all participants, as well as nutritional and habitat data collected in accordance with the study form item (Table 2). All subjects appear to be in good health and free of any long-term conditions.

2.2. Sampling

To preserve RNA during buccal cells collection, iSWAB RNA v2 Collection Kit (mawi, Biosamplin Reinvented) was used to collect buccal cells according to the manufacture's protocol.

iSWAB-RNA were incubated more than 3 hrs at room temperature before processing in RT-PCR. The samples were then centrifuged the for 2 minutes at 14000 rpm then 1uL of the clarified supernatants were taken and diluted in nuclease-free water to 1:16, finally taken 2 uL from the iSWAB diluted sample and applied direct on RT-PCR.

2.3. The primer designs:

Primers were ordered from (Integrated DNA Technologies). NCBI's BLAST database was used to "BLAST" generate а primer for an individual gene by using the function (https://blast.ncbi.nlm.nih.gov/blast.cgi). With amplicons shorter than 200 base pairs, primers were created (bp) (Table 1). Melting curve analysis confirmed the primer sets' specificity in this study. Endogenous GAPDH was used to maintain a constant level of gene expression during the study.

2.4. Extraction of RNA and synthesis of complementary DNA (cDNA):

The manufacturer's RNA isolation kit was used to obtain total RNA from the buccal cells (Invitrogen; Thermo Fisher Scientific, Inc., USA). A Genova Nano Micro-volume, Life Science & Standard Spectrophotometer was used to determine the concentration of RNA in the sample. The Veriti Thermal Cycler System (Applied Biosystems®, Thermo Fisher Scientific, Inc., USA) was used in accordance with the manufacturer's instructions to reverse-transcribe 500 ng of RNA into cDNA (Takara Bio, Inc.). The cDNA was synthesized from total RNA as previously described [41]. The quantity and purity of the eluted of RNA cDNA were determined by using Nano-Drop 2000 spectrophotometer-Thermo scientific[™] (thermo-fisher scientific, DE; USA). The purity of was evaluated by calculating the absorbance ratio at an optical density (OD) of 260-280 nm. A ratio between 1.7 and 2.0 was required for a suitable PCR and to indicate good purity.

2.5. RTPCR

The Fast RT-PCR 7500 System was used in accordance with the manufacturer's instructions to perform qPCR using the Applied BiosystemsTM SYBRTM Green master mix (Thermo Fisher Scientific, Inc., USA). SYBRTM Green master mix was added to 4 µl of the diluted cDNA template,

and 500 nM of each primer was mixed in equally. This is a list of the 40-cycle PCR standards. For each run, a baseline and a threshold were assigned automatically (7500 Fast Software, Version Journal Pre-proofs

(Ct). The $2^{-\Delta\Delta Ct}$ method was used to compare the relative levels of expression of various genes in this study.

2.7. Statistical analysis

The mean \pm standard deviation is used to present data. The analysis of variance (One-Way ANOVA) statistical method was applied to compare groups with assumed that the sample is drawn from the normally distributed population and the population variance is equal. Based on variance equality, the One-way ANOVA test with Tukey's HSD post-hoc tests were used to compare between the groups. Correlation coefficients between individuals (r value) were used to assess the genetic correlation with age. The P value was considered statistically significant if it was less than 0.05 in all statistical analyses (GraphPad Software Inc.).

| | NM | Gen e ID | Forward Sequence | Reverse Sequence |
|-----------------------------------|---|-------------|--|---|
| beta 2 Microglobul in (B2M) | Human qPCR Primer Pair (NM_00404 8) | 567 | CCACTGAAAAAGATGAGTATG CCT B2M F: AGCAGAGAAATGGAAAGTCAAA | CCAATCCAAATGCGGCATCT TCA B2M R: TGTTGATGTTGGATAAGAGA A |
| IL1 beta (IL1B) | Human qPCR Primer Pair (NM_00057 6) | 355 | CCACAGACCTTCCAGGAGAAT G iL: ACAGGATATGGAGCAACAAGT GG | GTGCAGTTCAGTGATCGTAC AGG IL : GGGCTTATCATCTTTCAACAC GC |
| IL6 | Human qPCR | 356 9 | AGACAGCCACTCACCTCTTCA G | TTCTGCCAGTGCCTCTTTGCT G |

| Table 1: Details | of the primer | sequences of | studied genes |
|------------------|---------------|--------------|---------------|
| | | | |

| | Primer Pair | | | |
|-----|-------------|-----|-----------------------|-----------------------|
| | (NM_00060 | | | |
| | ľ | • | Journal Pre-proofs | |
| | | | C | ATTTTCACCAGGCAAGTCTCC |
| | | | С | TC |
| | | | | |
| p16 | Human | 102 | CTCGTGCTGATGCTACTGAGG | GGTCGGCGCAGTTGGGCTCC |
| | qPCR | 9 | A | |
| | Primer Pair | | | |
| | (NM_05819 | | | |
| | 5) | | | |
| | | | | |

3. Results

Eighty-one subjects were recruited in the study and divided them into five age groups as previously described in the method. Different demographical and descriptive information were documented from the participants such as age, BMI, quantity of water drinking, smoking status, exercise and supplement & vitamins intake as shown in table 1.

3.1. β2M expression increased in buccal cells of elderly

In the study, we examined the expression of β 2M in different age groups, as depicted in the figure 1 panel A. Indeed, there was significant differences of β 2M expression in the elderly groups

comparing to the younger group, as shown in the figure 1. In the group of (36-45 years) there was a significant increase in the fold change of β 2M (6.60) comparing to the younger group (18-25

Journal Pre-proofs jeans) (5.10) p value 10.01. In addition, other group or age (10.05 years) showed more significant difference in the β 2M expression (fold change 8.20) comparing to the younger group (18-25 years) p value< 0.001^{***}. The same significant difference was also seen also in the older group of age (56-70 years) (fold change 12.04) comparing to the younger group of age p value< 0.001^{***}, as shown in the figure 1 panel A. To assess the biological aging of the participants, p16 ^{INK4} expression was used, as shown in the figure 1 panel B. The expression of p16 ^{INK4} showed a significant difference in the fold change (4.80) in the group (36-45 years) comparing to the fold change (3.19) in the younger group (18-25 years) p value^{*<} 0.05. The expression of p16 ^{INK4} was seen in the group of age (46-55 years) fold change (8.50) p value^{***<} 0.001 comparing to the younger group of age (18-25 years). The same significant difference was seen in the older group of age (56-70 years) (12.40) fold change comparing to the younger group of age (18-25 years) p value^{***<} 0.001, as shown in the figure 1 panel B.

After assessing the mRNA expression of β 2M and p16 ^{INK4} in different group of age, we sought to measure the expression of inflammatory biomarkers in these groups, as shown in the figure 1 panel C & D. Study participants aged 36 to 45 years showed a significant increase of fold change of IL-1 expression (5.10) compared to the younger group (3.60 fold change) p value**< 0.01. In addition, a significant increase also was seen in in the group age (36-45 years) with fold change of IL-1 β expression (5.10) comparing to the younger group of age (3.60 fold change) p value**< 0.01. A more significant difference was seen in the older group of age (56-70 years) with 5.70 fold change comparing to the younger group of age p value***.< 0.001. No significant difference was seen in the fold difference was seen in the figure 1 panel C.

Moreover, mRNA expression of IL-6 was also investigated. A significant fold difference was seen in the group of age (36-45 years) (5.60 fold change) comparing to the younger group of age (18-25 years) that was only 3.60 fold change, p value^{**} < 0.01. A more significant difference in the fold change was noticed in the (46-55 years) (5.80 fold change) comparing to the younger group of age (18-25 years), p value^{***} < 0.001. Same significant difference was also seen in the older group of age (56-70 years) with fold change 7.90 comparing to the younger group of age (18-25 years) with p value^{***} < 0.001, as shown in the figure 1 panel D.

3.2. β2M expression in buccal cells correlated directly with age

A for studying the surrousion of 02M p16INK4 and the inflammatory cores in different aroun of genes, we sought to assess the correlation between β 2M, p16^{INK4}, the participants' inflammatory biomarkers and age. Figure 2A shows a strong direct correlation between 2M and participant age (r= 0.90, p= 0.001). A direct correlation between the biomarker of biological aging, p16^{INK4}, and age of participants was also observed (r value= 0.77, p value <0.001) as shown in the figure 2B. In addition, inflammatory genes IL-1 β & IL-6 showed moderate direct correlation with age with r value= 0.45, p value <0.001 r value= 0.65 and p value <0.001 respectively, as shown in the figure 2C&D.

To deep illustrate the relationship between studied genes and age, color intensity of the percentage genes expression levels (Fold change) was created for each sample in relation to the age and was presented in the form of heatmap with cluster analysis as shown in the figure 3. Elderly participants showed high percentage of β 2M expression levels (Fold change) comparing to the youngers as shown in the figure 4. In addition, elderly people also showed high percentage of p16^{INK4} expression levels (Fold change) comparing to the youngers. Less intensity of IL-1b and IL-6 expression levels (Fold change) were seen in elderly people comparing to β 2M and as shown in the figure 3. p16^{INK4} and IL-1b (p<0.01) and IL-6 (p<0.05) genes showed significant differences, but p16^{INK4} and β 2M (p<0.57) showed no significant differences.

3.3. β2M expression correlated directly with expression of p16^{INK4} in buccal cells

When we looked at the gene expression of p16INK4 β 2M, p16INK4 and the genes of the inflammatory in different age groups, we found a strong correlation. β 2M expression showed a significant direct correlation with p16^{INK4} expression (r value= 0.83, p value <0.001), as shown in the figure 3A. In addition, β 2M expression also showed a significant direct correlation with inflammatory genes IL-1 β & IL-6 expressions with r value= 0.5, p value <0.001 and r value= 0.68 and p value <0.001 respectively, as shown in the figure 3B&C. The same correlation was seen between p16^{INK4} expression and inflammatory genes IL-1 β & IL-6 expressions with r value= 0.38, p value <0.001 and r value= 0.38, p value <0.001 and r value= 0.65 and p value <0.001 respectively, as shown in the figure 3. There

was significant correlation between studied genes especially in the older group of age G4 and G5 groups, as shown in the supplementary table1.

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The Pearson correlation coefficient of studied genes and subjects' age groups. Each set of age and/or gene expression data has a significant positive linear correlation (r-value, ***P<0.001, **P<0.01, and * P<0.05). It is the product of two variables' covariances and thus a normalized measurement of covariance.

Discussion

 β 2M is a small peptide that plays roles in inflammation and immunity. It has been documented that β 2M can be used in diagnosis and/ or monitoring some of infectious and malignant diseases[1, 8, 10, 42, 43]. previous work from our group showed that its blood levels is high in elderly[29]. This work revealed that β 2M levels were also high in buccal cells of old people. There was also a high correlation between β 2M expression and p16^{INK4} expression, as well as inflammatory gene expression.

p16^{INK4} expressions were found to be high in blood T cells of old people[38], in skin ageing [39] and in buccal cells of elderly people [40]. These showed that p16^{INK4} is a biomarker of aging in the studied samples or tissues. In this study p16^{INK4} has been used as a control for β 2M expression, β 2M expression is higher and more significant than the expression of p16^{INK4} in the age group (36-45 years) comparing to the younger group of age, with same significant increase for both genes in older groups. Using β 2M as a biomarker of biological ageing even in middle-aged people was supported by this study. In addition, the correlation between the β 2M and age showed that a strong direct correlation even more than the correlation between p16^{INK4} and age. Previously, β 2M has been shown to be highly expressed in the plasma membrane of senescent cells[28], later work by our group validated the potential use of β 2M as a biomarker of aging in blood [29]. Here different

sampling method was used to assess the β 2M levels, which provide a simple and non-invasive approach. This also underlined the potential ability of β 2M to be used as a biomarker for aging.

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Increasing the expression of β 2M in buccal cells of elderly people has advantages and implications. Firstly, the source of sample is not invasive comparing to other biological sources. In addition, in the future, any assessment of the biological aging and inflammation and / or chronic diseases can be evaluated in non-invasive way. Moreover, this method can be commercialized easily. However, further study should be conducted to include larger sample size of smoker and nonsmoker participants to reveal the effect such factor on the expression of β 2M.

The data showed that β 2M correlated directedly and more significantly with inflammatory genes expressions comparing to the p16^{INK4} expression. This suggested that β 2M could have a role in inflammation in elderly. Indeed, β 2M have been shown to increase in inflammation and / or inflammatory associated disorders [44-46]. Our previous work also showed the association between oxidative stress and β 2M expression [29]. The exact role and relationship between β 2M and the inflammatory and oxidative stress cascade needs to be clarified, however, in future studies

The study has limitations of being small sample size, therefore larger sample size should be included to confirm the findings. In addition, future work should also assess protein levels of β 2M and compare it to the mRNA expression to verify the findings.

In summary, the results showed that β 2M expressed highly in buccal cells in comparisons between the elderly and younger people. β 2M has a significant direct correlation with age of participants. Moreover, its expression directedly correlated with inflammatory biomarkers, these suggested the potential use of β 2M as biomarker of biological aging. However, future work should include larger sample size before translating the findings for commercial or diagnostic purposes.

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| Table 1: The demographical | and descriptive statistical dat | a of the studied subjects. |
|----------------------------|---------------------------------|----------------------------|
| | | |

| Table 1: The den | able 1: The demographical and descriptive statistical data of the studied subjects. | | | | | |
|------------------|---|------------|-------------|-----------------|-------------------------|--------------|
| | | G1 | G2 | G3 | G4 | G5 |
| Number | | 21 | 17 | 11 | 18 | 14 |
| Age (Years) | Mean ± SD | 22 ± 2.8 | 30.4 ± 3.2ª | 41.7 ± 3.35ª | 50.7 ± 2.9ª | 64.7 ± 5.8ª |
| | Range | 18-25 | 26-35 | 37-45 | 46-55 | 56-70 |
| BMI (Kg/m²) | Mean ± SD | 20.5 ± 3.2 | 22.1 ± 2.7 | 21.4 ± 2.1 | 22.2 ± 2.6 ^c | 23.8 ± 1.8 ª |
| | Range | 17.5-30 | 18-26 | 18-24 | 18-25 | 19-26 |
| Water | Mean ± SD | 3.5 ± 0.5 | 2.8 ± 0.9 | 3 ± 0.8 | 2.8 ± 0.4 | 3.2 ± 0.4 |
| drinking (L) | Range | 1.5-5 | 1-5 | 2-4 | 2-3 | 3-4 |
| Smoking | Non-smoker | 15 (71.4%) | 8 (47.1%) | 6 (54.5%) | 12 (66.7%) | 12 (85.7%) |
| | Smoker | 6 (28.6%) | 9 (52.9%) | 5(45.5%) | 6 (33.3%) | 2 (14.3%) |
| Exercise | No exercise | 19 (90.5%) | 9 (52.9%) | 6 (54.5%) | 14 (77.8%) | 9 (64.3%) |

| Cycling - 4 (23.5%) 1 (9.1%) - - Journal Free-proofs Football - 1 (5.9%) 2 (18.2%) - - Supplemental No-supplements 21 (100%) 14 (82.4%) 10 (90.9%) 18 (100%) 10 (71.4%) Intake Multi-vitamins - 2 (11.8%) 1 (9.1%) - - Vegetable No 7 (33.3%) 7 (41.2%) 10 (91.9%) 11 (61.1%) 1 (71.4%) Supplemental No 7 (33.3%) 7 (41.2%) 8 (72.7%) 3 (16.7%) 4 (86.6%) Ometimes 11 (52.4%) 8 (47.1%) 8 (72.7%) 3 (16.7%) 9 (64.3%) Sleeping 8 H 9 (%) 4 (23.5%) 1 (9.1%) 6 (33.3%) 5 (35.7%) Sleeping 8 H 1 (48.3%) 3 (17.6%) 1 (9.1%) 2 (11.1%) 5 (35.7%) Sleeping 8 H 1 (48.3%) 3 (17.6%) 1 (9.1%) 2 (11.1%) 5 (35.7%) Sleeping 8 H 1 (48.3%) 3 (17.6%) | | Walking | 2 (9.5%) | 3 (17.6%) | 2 (18.2%) | 4 (22.2%) | 5 (35.7%) |
|--|--------------|----------------|------------|----------------------------|-----------|------------|-----------|
| Football - 1 (5.9%) 2 (18.2%) - - Supplemental No-supplements 21 (100%) 14 (82.4%) 10 (90.9%) 18 (100%) 10 (71.4%) Intake Multi-vitamins - 2 (11.8%) 1 (9.1%) - 5 (28.6%) Protein - 1 (5.9%) 2 (18.2%) 1.0 (90.9%) 18 (100%) 5 (28.6%) Vegetable Multi-vitamins - 2 (11.8%) 1 (9.1%) - - Vegetable No 7 (33.3%) 7 (41.2%) 2 (18.2%) 11 (61.1%) 1 (7.1%) Sometimes 11 (52.4%) 8 (47.1%) 8 (72.7%) 3 (16.7%) 4 (86.6%) Daily 3 (14.3%) 2 (11.8%) 1 (9.1%) 4 (22.2%) 9 (64.3%) Sleeping hours 8 H 9 (%) 4 (23.5%) 1 (%) 10 (55.6%) 9 (64.3%) A 8 H 11 (52.4%) 10 (58.8%) 9 (81.8%) 6 (33.3%) 5 (35.7%) | | Cycling | 1 | | | - | - |
| Supplemental SupplementalNo-supplements $21 (100\%)$ $14 (82.4\%)$ $10 (90.9\%)$ $18 (100\%)$ $10 (71.4\%)$ Intake ProteinMulti-vitamins $ 2 (11.8\%)$ $1 (9.1\%)$ $ 5 (28.6\%)$ Protein $ 1 (5.9\%)$ $ -$ Vegetable Eating Daily $7 (33.3\%)$ $7 (41.2\%)$ $2 (18.2\%)$ $11 (61.1\%)$ $1 (7.1\%)$ Sleeping hours $8 (11 (52.4\%)$ $8 (47.1\%)$ $8 (72.7\%)$ $3 (16.7\%)$ $4 (8.6\%)$ Sleeping hours $8 H$ $9 (\%)$ $4 (23.5\%)$ $1 (\%)$ $10 (55.6\%)$ $9 (64.3\%)$ | | Football | | l Pre-proofs 1 (5.9%) | 2 (18.2%) | - | - |
| $ \begin{array}{ c c c c c c c c c c c c c c c c c c c$ | Supplemental | No-supplements | 21 (100%) | | | 18 (100%) | 10 (71.4% |
| Vegetable Eating No 7 (33.3%) 7 (41.2%) 2 (18.2%) 11 (61.1%) 1 (7.1%) Sometimes 11 (52.4%) 8 (47.1%) 8 (72.7%) 3 (16.7%) 4 (8.6%) Daily 3 (14.3%) 2 (11.8%) 1 (9.1%) 4 (22.2%) 9 (64.3%) Sleeping hours 8 H 9 (%) 4 (23.5%) 1 (%) 10 (55.6%) 9 (64.3%) | | | | | | | |
| Eating Image: constraint of the state of th | | Protein | - | 1 (5.9%) | - | - | - |
| Sometimes $11 (52.4\%)$ $8 (47.1\%)$ $8 (72.7\%)$ $3 (16.7\%)$ $4 (8.6\%)$ Daily $3 (14.3\%)$ $2 (11.8\%)$ $1 (9.1\%)$ $4 (22.2\%)$ $9 (64.3\%)$ Sleeping hours $8 H$ $9 (\%)$ $4 (23.5\%)$ $1 (\%)$ $10 (55.6\%)$ $9 (64.3\%)$ $< 8 H$ $11 (52.4\%)$ $10 (58.8\%)$ $9 (81.8\%)$ $6 (33.3\%)$ $5 (35.7\%)$ | | No | 7 (33.3%) | 7 (41.2%) | 2 (18.2%) | 11 (61.1%) | 1 (7.1%) |
| Sleeping hours 8 H 9 (%) 4 (23.5%) 1 (%) 10 (55.6%) 9 (64.3%) < 8 H | Eating | Sometimes | 11 (52.4%) | 8 (47.1%) | 8 (72.7%) | 3 (16.7%) | 4 (8.6%) |
| hours < 8 H 11 (52.4%) 10 (58.8%) 9 (81.8%) 6 (33.3%) 5 (35.7%) | | Daily | 3 (14.3%) | 2 (11.8%) | 1 (9.1%) | 4 (22.2%) | 9 (64.3%) |
| < 8 H 11 (52.4%) 10 (58.8%) 9 (81.8%) 6 (33.3%) 5 (35.7%) | | | 9 (%) | 4 (23.5%) | 1 (%) | 10 (55.6%) | 9 (64.3%) |
| > 8 H 1 (4.8%) 3 (17.6%) 1 (9.1%) 2 (11.1%) - | nours | < 8 H | 11 (52.4%) | 10 (58.8%) | 9 (81.8%) | 6 (33.3%) | 5 (35.7%) |
| | | > 8 H | 1 (4.8%) | 3 (17.6%) | 1 (9.1%) | 2 (11.1%) | - |
| | | | | | | | |

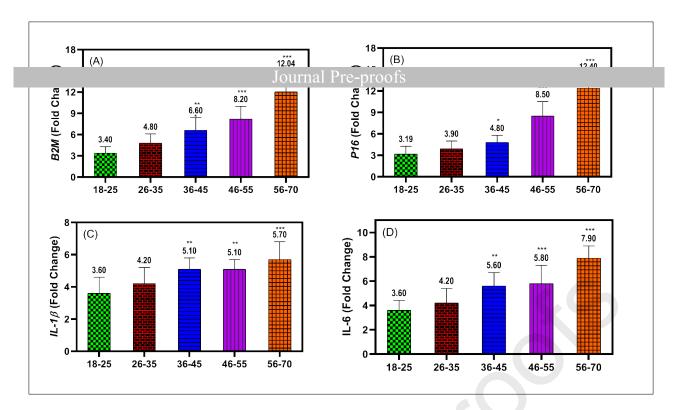


Figure 1: The fold change of the mRNA expression levels of studied genes B2M (A), P16^{INK4} (B), IL-1 β (C), and IL-6 (D). The data expressed as mean ± SD of the fold change. *, **, and *** indicate the P values<0.05, P<0.01, and P<0.001 of the significant difference levels between the age groups and the younger group (>25years) as a control, respectively.

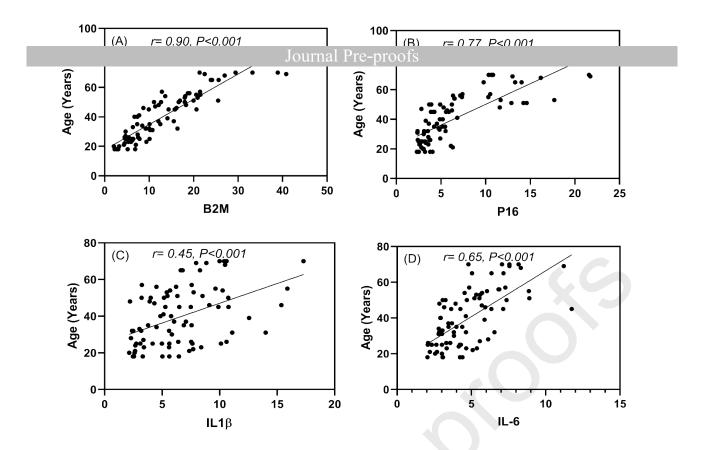


Figure 2: The Pearson correlation coefficient of studied genes and subjects' age. Each set of age and/or gene expression data has a significant positive linear correlation (r-value, P<0.001). It is the product of two variables' covariances and thus a normalized measurement of covariance.

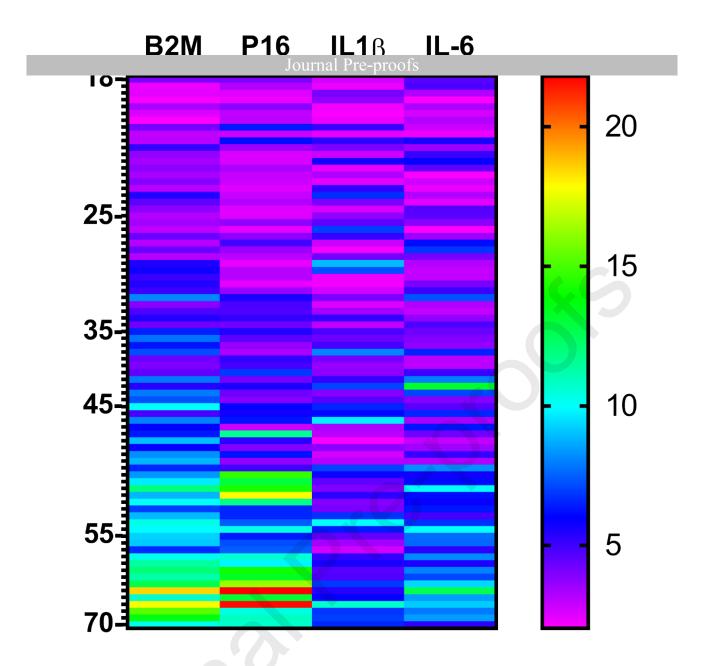


Figure 3. The Expression levels heatmap with cluster analysis. The colour intensity in each box shows the percentage of expression levels (Fold change) for each gene relative to the colour key on the right side in correspondence to the participants age in the left side. The subject sample notation in the left side. (There were significant differences between P16 "as a common aging gene" and both IL-1b (p<0.01) and IL-6 (p<0.05) genes, however non-significant differences were observed between P16 and B2MG (p<0.57).

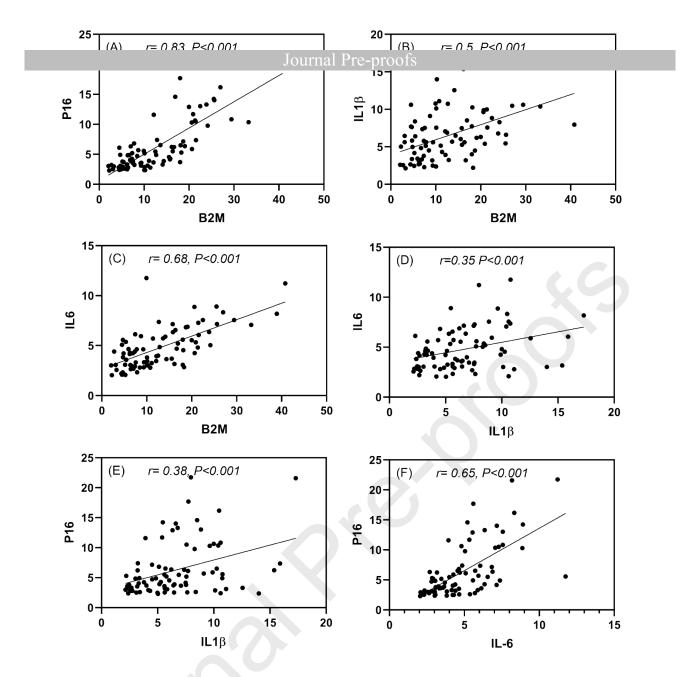


Figure 4: The Pearson correlation coefficient of studied genes. Each set of gene expression data has a significant positive linear correlation with correspondence gene (r-value, P<0.001). It is the product of two variables' covariances and thus a normalized measurement of covariance.

Table 1S:

The Pearson correlation coefficient of studied genes and subjects' age groups. Each set of age Journal Pre-proofs

**P<0.01, and * P<0.05). It is the product of two variables' covariances and thus a normalized measurement of covariance.

| | | B2M | p16 | IL-1B | IL-6 |
|-------|----|---------|---------|---------|---------|
| B2M | G1 | 1.00 | 0.20 | 0.40 | 0.05 |
| | G2 | 1.00 | 0.32 | 0.12 | 0.29 |
| | G3 | 1.00 | -0.29 | 0.37 | 0.07 |
| | G4 | 1.00 | 0.51* | 0.11 | 0.34 |
| | G5 | 1.00 | 0.83*** | 0.71*** | 0.74*** |
| P16 | G1 | 0.20 | 1.00 | 0.24 | 0.17 |
| | G2 | 0.32 | 1.00 | -0.30 | 0.43 |
| | G3 | -0.29 | 1.00 | -0.20 | 0.10 |
| | G4 | 0.51* | 1.00 | 0.02 | 0.37 |
| | G5 | 0.83*** | 1.00 | 0.63** | 0.73*** |
| IL-1B | G1 | 0.40 | 0.24 | 1.00 | 0.08 |
| | G2 | 0.12 | -0.30 | 1.00 | -0.46 |
| | G3 | 0.37 | -0.20 | 1.00 | 0.54 |
| | G4 | 0.11 | 0.02 | 1.00 | 0.22 |
| | G5 | 0.71*** | 0.63** | 1.00 | 0.31 |
| IL-6 | G1 | 0.05 | 0.17 | 0.08 | 1.00 |

| G2 | 0.29 | 0.43 | -0.46 | 1.00 |
|----|---------|-----------------|-------|------|
| | Jou | rnal Pre-proofs | • = • | |
| G4 | 0.34 | 0.37 | 0.22 | 1.00 |
| G5 | 0.74*** | 0.73*** | 0.31 | 1.00 |

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