

Revealing common lncRNAs and gene signatures: computational analysis of public RNA-seq datasets in different *in vitro* senescence models



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Background and aims

Senescence is a **cellular state** characterized by many **molecular alterations**: permanent cell cycle arrest, senescence-associated secretory phenotype (SASP), increased β -galactosidase activity (SA- β -gal) and morphological changes^{1,2}. Growing evidence suggest that **long non-coding RNAs (lncRNAs)** play **important roles in cellular senescence**³⁻⁵ and **age-related diseases**⁶⁻⁸ at multiple levels-transcriptional, post-transcriptional, translational, and post-translational. However, **their exact biological function remain largely unclear**⁹. **The aim** was the identify **differentially modulated genes/lncRNAs and pathways shared by** different type of human cells undergoing senescence, by utilizing **open access bulk RNAseq** datasets and computational analysis.

Bionformatics pipeline

RNAseq datasets were selected from the open access archive “Gene Expression Omnibus” (GEO) focusing on experiments regarding *in vitro* models of **replicative senescence (RS)** and **drug induced-senescence (Table 1)**, which were supported by published articles¹⁰⁻¹⁶. The bioinformatics pipeline was developed in R [v4.2.2, 2022-10-31] and included the following steps:

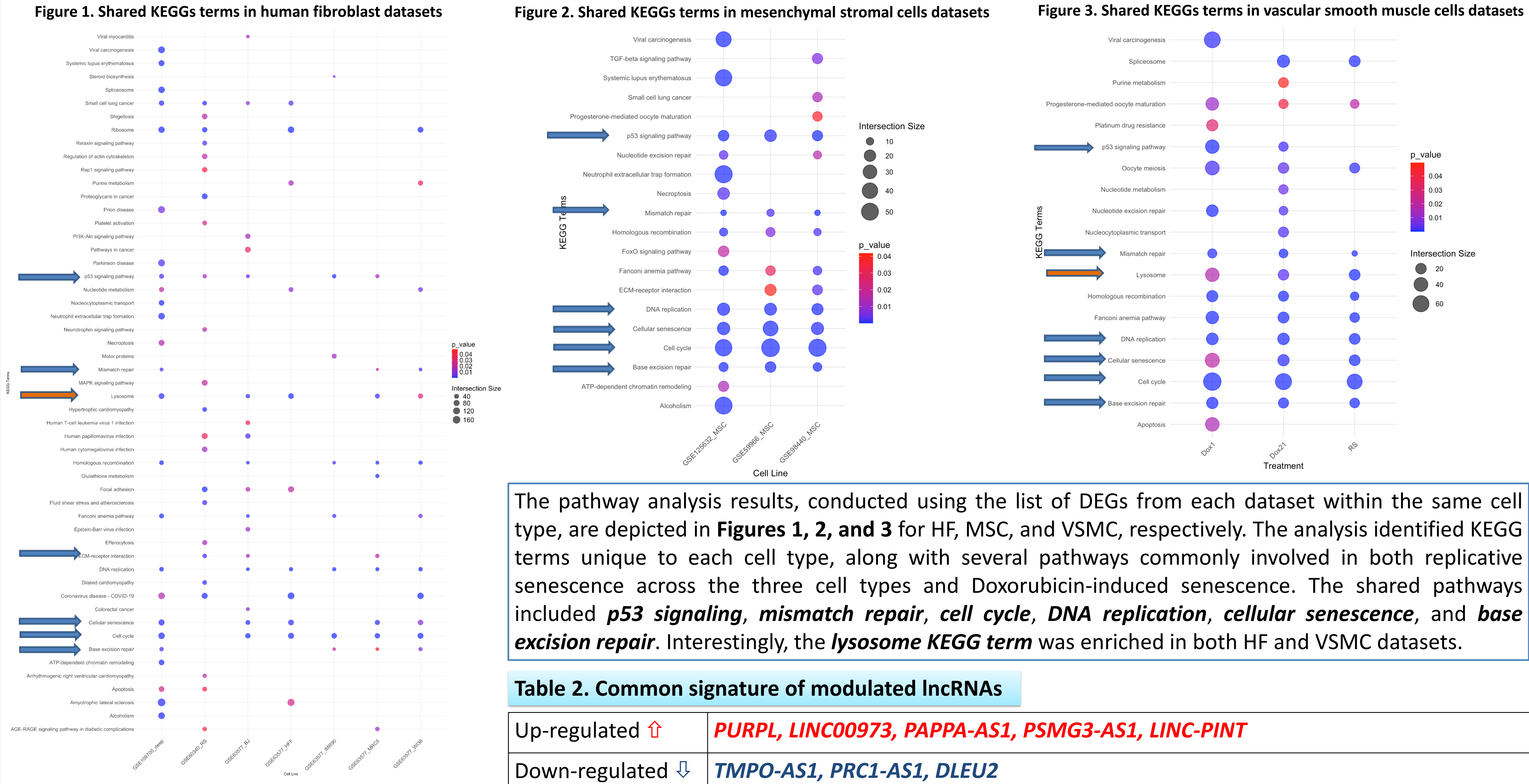
- Downloading of preprocessed raw data matrix from GEO.
- Annotation of data matrix using the human Build GRCh.38.p13.
- Data filtering and normalization using R packages Limma [v.3.54.2] and EdgeR [v.3.40.2].
- Computation of differentially expressed genes and lncRNAs (DEGs) with an adjusted p-value < 0.05 and log2 fold-change > [0.58] using DESeq2[v.1.38.3].
- Pathway enrichments analysis using the gprofiler2 [v.0.2.3].

Results

Table 1: List of selected RNAseq datasets and number of up- and down-regulated genes and lncRNAs (DEGs)

| Human fibroblast (HF) datasets | DEGs up↑/down↓ | Mesenchymal stromal cells (MSC) datasets | DEGs up↑/down↓ | Vascular smooth muscle cells (VSMC) datasets | DEGs up↑ down↓ |
|--|---|---|-------------------------------------|---|---------------------------------------|
| GSE109700 ¹⁰ GSE60340 ¹¹ GSE63577 ¹² [BJ] GSE63577 ¹² [HFF] GSE63577 ¹² [IMR90] GSE63577 ¹² [MRC5] GSE63577 ¹² [WI38] | 3255↑/3247↓ 1516↑/1645↓ 900↑/842↓ 2961↑/2316↓ 1156↑/1163↓ 1308↑/1542↓ 2534↑/2226↓ | GSE98440 ¹³ GSE59966 ¹⁴ GSE125632 ¹⁵ | 360↑/525↓ 928↑/798↓ 258↑/579↓ | GSE171663 ¹⁶ [RS] GSE171663 ¹⁶ [Dox1d] GSE171663 ¹⁶ [Dox21d] | 230↑/470↓ 1490↑/1532↓ 328↑/655↓ |

Comparison of the pathway enrichment analyses based on the list of DEGs of each dataset depicted by bubble plots



Conclusions and future plans

In summary, we identified **distinct signatures of modulated protein-coding genes**, leading to both specific and **commonly shared** pathways. These pathways further validated the **molecular signatures of the senescence models**. These pathways further corroborated the molecular signatures characteristic of the senescence models. As extensively documented, key processes such as cell cycle regulation, DNA damage repair, and p53 signaling are well-established hallmarks of senescence. Additionally, the analysis of lncRNAs confirmed **the upregulation of PURPL and uncovered novel lncRNAs** (see Table 2). Future analysis will focus on exploring the regulatory networks of these modulated lncRNAs across various cell types during senescence.

References

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