

Identification of several hub-genes associated with periodontitis using integrated microarray analysis

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Abstract. The aim of the present study was to identify differentially expressed genes and biological processes associated with periodontitis. In this study, the most significant 200 differentially expressed genes associated with periodontitis were identified using integrated analysis of multiple microarray data in combination with screening for genome-wide relative significance and genome-wide global significance. Gene Ontology (GO) enrichment analysis and pathway analysis were performed using the GO website and Kyoto Encyclopedia of Genes and Genomes (KEGG). A protein-protein interaction (PPI) network was constructed based on the Search Tool for the Retrieval of Interacting Genes/Proteins. The top 200 differentially expressed genes were found to be highly associated with periodontitis. GO enrichment analyses revealed that the identified genes were significantly enriched in terms of response to organic substance, response to wounding and cell migration. The most common term of the KEGG pathway was cytokine-cytokine receptor interaction. PPI network analysis indicated that interleukin (IL)8, IL1 β , vascular endothelial growth factor A, intercellular adhesion molecule 1, PTGS2 and CXCL10 were hub genes, which formed numerous interactions with several genes. In conclusion, the present study identified numerous genes that were differentially expressed in periodontitis, as well as determined the biological pathways and PPI network associated with those genes.

Introduction

Periodontitis is a common chronic infection of the supporting tissues of teeth, which has been epidemiologically associated

with cardiovascular diseases (1). Previous studies have indicated that periodontitis is more than a localized oral infection and it may have marked effects on systemic health (2,3).

Numerous clinical and scientific studies have provided evidence that the risk of adult periodontitis has a genetic component (4-6). Genetic factors are important for assessing the susceptibility of patients to periodontitis and the rate of its progression, as demonstrated by studies in humans and animals, which have indicated that genetic factors influence general inflammatory and immune responses (7,8). Individuals have differential responses to common environmental stimuli, which is dependent on the genetic profile of an individual (9). Different forms of genes (allelic variants) may induce variations in tissue structure (innate immunity), antibody responses (adaptive immunity) and inflammatory mediators (non-specific inflammation) (10). In order to elucidate the potential effect of a patient's genetic profile on periodontitis, the contribution of different genes to the pathogenesis of periodontitis must be understood. Therefore, identification of the most important genes that contribute to the development of this disease is essential.

Analyses of differential gene expression, performed using high-throughput experimental methods, such as microarray analysis, have been used in an increasing number of studies in recent years (11,12). At present, a vast quantity of microarray data have been produced and deposited in publically-available data repositories, including Gene Expression Omnibus (GEO; <http://www.ncbi.nlm.nih.gov/geo/>) (13) and Array Express Archive (<http://www.ebi.ac.uk/arrayexpress/>) (14). These repositories allow researchers to advance the identification of genetic and diagnostic signatures using data integration and bioinformatics analysis. This may therefore provide insights into the biological mechanisms underlying the development of periodontitis. Demmer *et al* (15) revealed that 5,295 genes were upregulated and 7,449 genes were downregulated in periodontitis using transcriptome analysis. Gene ontology (GO) analysis identified groups of differently expressed genes (DEGs) in periodontitis, including those involved in apoptosis, antimicrobial humoral responses and antigen presentation. Becker *et al* (16) also presented a disease-associated mRNA profile, which displayed the potential underlying mechanisms of periodontitis. GO analysis revealed that the regulation of transcripts associated with bacterial response systems were dominant in periodontitis

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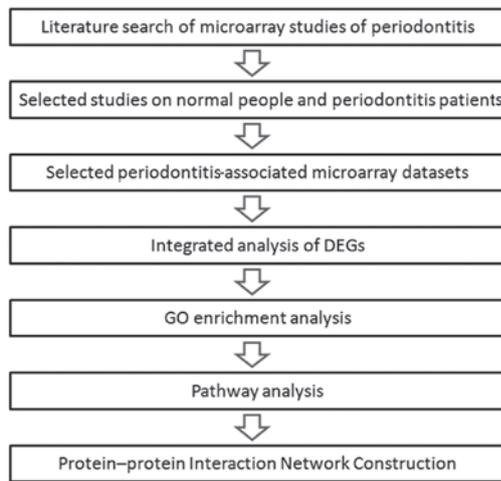


Figure 1. Experimental protocol of the present study. DEG, differentially expressed genes; GO, gene ontology.

tissues. However, microarray-based biomarkers have had poor translation into clinical practice; furthermore, the results were non-conformal among studies due to small sample sizes.

A robust genetic marker signature may be beneficial to the diagnosis and targeted treatment of periodontitis in clinical practice. In the present study, in order to identify a more credible gene biomarker signature for periodontitis, an integrated analysis of microarray data was performed using a novel model that screened for genome-wide relative significance (GWRS) and genome-wide global significance (GWGS) (17). The present study also aimed to build a more precise target network for periodontitis using the selected biomarkers and then further investigate the identified DEGs through functional enrichment analysis, pathway enrichment analysis and protein-protein interaction (PPI) networks.

Materials and methods

Identification of gene expression datasets. In the present study, DEGs were identified between normal healthy volunteers and periodontitis patients. Between January 2008 and January 2014, three microarray datasets were extracted from the National Center for Biotechnology Information GEO database: GEO access nos. GSE10334 (15), GSE27993 (18) and GSE33774 (16). The experimental protocol for the present study is shown in Fig. 1. The following information was extracted from each identified study: GEO accession number, sample type, platform, number of cases and controls, and gene expression data. Animal studies and studies in which microarray data were uncertain were excluded.

Integrated analysis of DEGs from multiple microarray data. Data pre-processing was performed using the Bioconductor XPS package v1.24.1 (www.bioconductor.org/packages/release/bioc/html/xps.html) (19), which was based on the data analysis framework ROOT v5 (<http://root.cern.ch>). Probes were excluded when they did not match a corresponding gene symbol; when a probe-set was mapped to multiple genes, all of these genes were given the expression of the probe-set.

Maxim-based methods were used to deal with the situations where multiple probe-sets were associated with a gene. Microarray Suite 5 v7.3-35 (Affymetrix, Inc., Santa Clara, CA, USA) was used to compute expression levels. For each dataset, $P < 0.05$ and a false discovery rate value < 0.05 were set as the parameters for DEGs.

Integrated analysis of DEGs identified in the three microarray datasets using GWRS and GWGS. The gene signatures were identified using a novel model which measured the GWRS and GWGS of gene expression. The GWRS of a gene was measured using its ranking position on a genome-wide scale based on the differential expression measured in each individual microarray dataset. The number of datasets was denoted by n , the number of unique genes across n datasets was denoted by m ; GWRS of the i -th gene in the j -th dataset was measured by: $s_{ij} = -2(\log \frac{r_{ij}}{m})$, where r_{ij} , $i=1-m$, $j=1-n$, was the rank number of the i -th gene in the j -th study. GWGS was applied to synthetically analyze multiple microarray studies. The GWGS of a gene was estimated based on its corresponding GWRS across n datasets. GWGS of a gene was calculated by: $s_i^r = \sum_{j=1}^n \omega_j s_{ij}$, where w_j represented the relative weight of the j -th dataset. These calculations were performed as previously described (17). A gene with a large GWGS value was considered to be globally significant across multiple independent studies.

In the present study, the degree of differential gene expression was measured by a fold-change based algorithm, which was found to be more suited to measure the significance of differential expression compared with that of other statistical tests which give P-values and the significance analysis of microarray (SAM). The fold-change-based differential expression analysis was conducted using the Linear Models for Microarray Data (LIMMA) package (<http://bioinf.wehi.edu.au/limma/>) (20). A rank number was assigned to each gene according to its degree of differential expression (in descending order from 1 to m) and equal weight was assigned to each data. The smallest number of genes that may achieve a suitable level of classification performance was 200; therefore, the top 200 genes were selected for further analysis.

Functional and pathway enrichment analysis. In order to further investigate the functions of the DEGs, GO-categories were organized based on the GO database (<http://www.geneontology.org/>). The enrichment analyses required > 5 genes to be present and $P < 0.01$ for a category to be considered significant. In order to further investigate the signaling pathway of the DEGs, pathway analysis based on the Kyoto Encyclopedia of Genes and Genomes (KEGG) database (www.genome.jp/kegg/) was performed. The two analyses were performed using the Database for Annotation, Visualization and Integrated Discovery (DAVID; <http://david.abcc.ncifcrf.gov/home.jsp>) (21). The significant categories were identified by expression analysis systematic explorer (EASE) scores, a modified Fisher Exact P-value. The threshold for significance for a category was $P < 0.05$, with > 2 genes for the corresponding term.

PPI network construction. The PPI network is a useful research tool for investigating the cellular networks of protein interactions. A PPI network was constructed for the top 200 DEGs.

Table I. Top 200 differentially expressed genes identified by the integrated analysis of three microarray datasets.

No.	Gene	No.	Gene	No.	Gene	No.	Gene
1	SLAMF7	51	ENTPD1	101	BNC1	151	CNIH4
2	DDX3Y	52	STEAP4	102	KRT19	152	SFRP4
3	CD79A	53	CLDN20	103	FCGR2A	153	ANKRD22
4	MMP12	54	RGS2	104	STK39	154	EXOC5
5	FAM46C	55	SPIN4	105	IL1A	155	CORIN
6	IL8	56	SORBS2	106	CTGF	156	MS4A6A
7	SELE	57	VPS41	107	SLC7A5	157	MYO5B
8	SEL1L3	58	BAMBI	108	EPCAM	158	ANXA3
9	APOD	59	PCDH18	109	ZNF273	159	ZFAND2A
10	EIF1AY	60	BBS10	110	DUSP5	160	BGN
11	SLC7A11	61	WDR74	111	MMP1	161	PLP1
12	CD27	62	CXCL6	112	HLA-DQB1	162	GLT8D2
13	MMP7	63	GPR110	113	GATM	163	SERPINE1
14	ADAM28	64	MALAT1	114	F13A1	164	CRIP2
15	IL1 β	65	CXCR4	115	TNFRSF19	165	PALMD
16	LAX1	66	TXNDC11	116	ADAM12	166	TMEM117
17	IGLJ3	67	PLA2G7	117	STXBP5	167	JUND
18	CSF3	68	KCNJ2	118	LYRM2	168	CISD2
19	TNFRSF17	69	EVI2B	119	FOSB	169	RNASE1
20	CD36	70	IFI44L	120	IFIT1	170	FAM103A1
21	SOCS3	71	SELL	121	PVRL4	171	CXCL12
22	RPS4Y1	72	NRCAM	122	MAN1A1	172	STK17A
23	PIM2	73	NR4A2	123	CDH19	173	CORO1A
24	FLG	74	GLCCI1	124	TTC39A	174	IGJ
25	FLG2	75	PANX1	125	NEFM	175	MMP13
26	SEC11C	76	BTG2	126	PIGW	176	CALCRL
27	PTGS2	77	CRISP3	127	EIF2AK2	177	A2M
28	MMP3	78	ABCA8	128	CFB	178	CORO1C
29	RGS5	79	MYADM	129	STX11	179	TGFBR3
30	CCL18	80	PDK4	130	GBP1	180	GSTM1
31	RGS1	81	PLAUR	131	SLC39A8	181	TRIM5
32	SFRP2	82	BPGM	132	DDX60L	182	FGFBP1
33	ZNF215	83	LAMC2	133	ELTD1	183	BMP2K
34	YOD1	84	ROR1	134	CYP4X1	184	OLFML1
35	CXCL1	85	PRDM1	135	ACTA2	185	UFM1
36	AQP9	86	ICK	136	GBP2	186	SERPINB9
37	KDM5D	87	IFIT5	137	DLX5	187	PMP22
38	SLC2A3	88	FMO1	138	SOSTDC1	188	FMO2
39	ODAM	89	ABCA12	139	OAS1	189	TRIM13
40	ALOX15B	90	STARD5	140	TSPAN11	190	HIST2H2BE
41	HOOK1	91	AVPI1	141	TNFRSF12A	191	C15orf41
42	SEL1L	92	ARG1	142	ALYREF	192	PHACTR2
43	CXCL10	93	VEGFA	143	NID1	193	TCEA3
44	SGMS2	94	SELK	144	SPAG17	194	CDK8
45	IL10RA	95	CEACAM7	145	LCE2B	195	CD177
46	TP53AIP1	96	PSPH	146	PNLIPRP3	196	CACHD1
47	XDH	97	HBB	147	PLEKHM3	197	SELP
48	COL1A1	98	ALDH1L2	148	CHD9	198	ADH7
49	CYP1B1	99	BHLHE41	149	PARM1	199	EEA1
50	ICAM1	100	SLC16A9	150	EPHA4	200	HSPA4L

IL, interleukin; VEGFA, vascular endothelial growth factor A; ICAM1, intercellular adhesion molecule 1.

Table II. Pathway analysis based on the Kyoto Encyclopedia of Genes and Genomes.

Term	P-value	Count
Cytokine-cytokine receptor interaction	2.60E-06	16
Cell adhesion molecules	1.10E-02	7
Complement and coagulation cascades	1.60E-02	5
Intestinal immune network for Immunoglobulin A production	3.20E-02	4
Chemokine signaling pathway	4.80E-02	7
Drug metabolism	5.70E-02	4

Table III. Top 10 GO terms of the top 200 differentially expressed genes.

ID	Term	P-value	Count
GO:0006955	Immune response	6.60E-09	29
GO:0010033	Response to organic substance	1.10E-05	24
GO:0009611	Response to wounding	1.30E-05	20
GO:0016477	Cell migration	9.60E-05	13
GO:0006935	Chemotaxis	1.10E-04	10
GO:0042330	Taxis	1.10E-04	10
GO:0051674	Localization of cell	2.60E-04	13
GO:0048870	Cell motility	2.60E-04	13
GO:0032963	Collagen metabolic process	2.90E-04	5
GO:0042127	Regulation of cell proliferation	3.30E-04	22

Table IV. Genes with degrees >10 in the protein-protein interactions network.

Gene	Degree
IL8	28
IL1 β	25
VEGFA	24
ICAM1	22
PTGS2, CXCL10	21
MMP1, CXCR4	17
CXCL12	16
COL1A1, CSF3	15
IL1A, SERPINE1	14
CD79A, SELL, PLAUR, MMP3	13
CXCL1, CTGF	12
SELP, SELE, MMP12, A2M	11

IL, interleukin; VEGFA, vascular endothelial growth factor A; ICAM1, intercellular adhesion molecule 1.

Interaction data were obtained from the Search Tool for the Retrieval of Interacting Genes/Proteins (STRING; <http://string.embl.de/>). DEGs were then imported into the interaction network and interactions were screened with both end nodes having DEGs. All nodes with the degree ≥ 1 were reserved. The networks were constructed using Cytoscape 3.1.0 (<http://www.cytoscape.org/>).

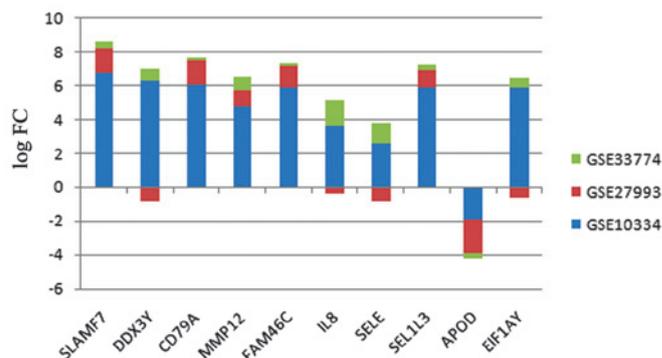


Figure 2. Expression pattern of the top 10 ranked differentially expressed genes identified by the integrated analysis of three microarray databases. FC, fold change.

Results

Integrated analysis of DEGs in multiple studies. Information was extracted from the microarray datasets GSE10334, GSE27993 and GSE33774 for 9,925, 8,322 and 6609 genes, respectively, following the exclusion of animal and uncertain data as well as duplicated genes. Following intersection of the microarray datasets, 2,754 common genes were identified. The GWGS values of genes were then applied to the integrated independent microarray data. Genes with large GWGS values were considered to be globally significant across multiple independent studies as genes with larger fold-changes were ranked highly for individual microarray.

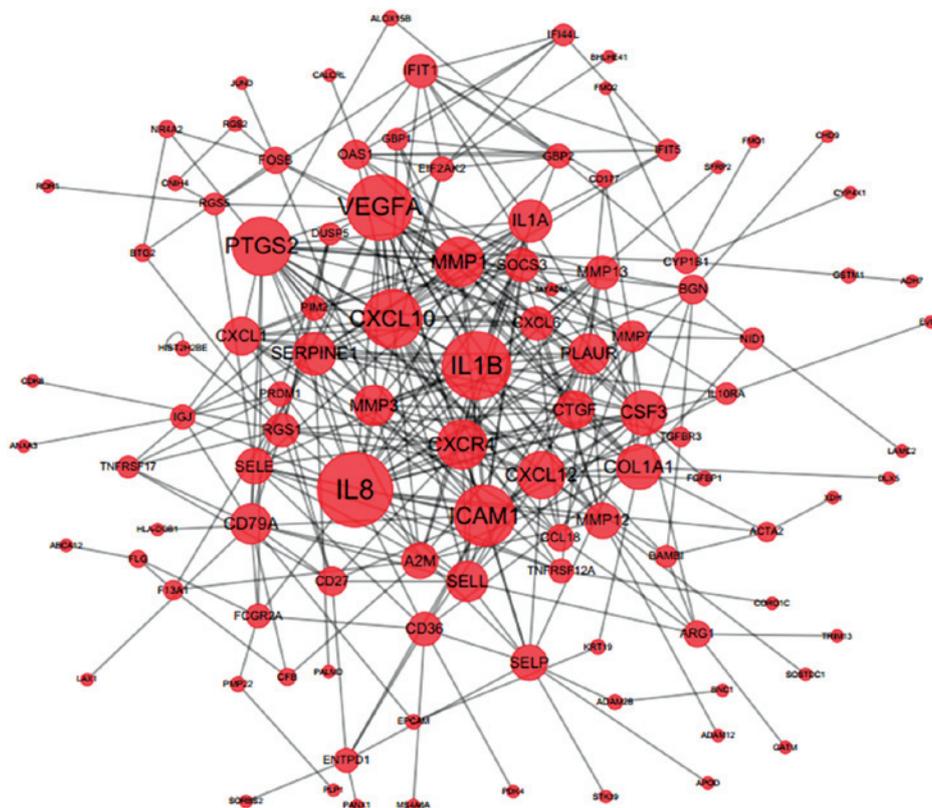


Figure 3. Protein-protein interaction networks of the top 200 differentially expressed genes identified by the integrated analysis of three microarray databases. Each biological relationship (an edge) between two genes (nodes) is supported by at least one reference from the literature or information stored in the Search Tool for the Retrieval of Interacting Genes/Proteins Base. The degree for each gene is represented by the size of the gene node.

Following the intersection of the three microarray datasets, the top 200 DEGs were identified (Table I). The expression pattern of the top 10 genes is presented in Fig. 2.

Functional and pathway enrichment analysis. In order to further investigate the functions of the identified top 200 genes, pathway and GO analysis were performed. A total of 135 terms were retrieved from the DAVID database.

Pathway analysis revealed genes that were significantly enriched in six terms; the most common three terms were those involved in cytokine-cytokine receptor interactions ($P=2.6 \times 10^{-6}$), cell adhesion molecules ($P=1.1 \times 10^{-2}$) and complement and coagulation cascades ($P=1.6 \times 10^{-2}$) (Table II).

GO analysis revealed that the identified genes were predominantly involved in biological processes of immune response ($P=6.6 \times 10^{-9}$), response to organic substances ($P=1.1 \times 10^{-5}$) and response to wounding ($P=1.3 \times 10^{-5}$). The top 10 GO terms are shown in Table III.

Structure interaction network of the DEGs. Based on the STRING database, the interaction networks of the identified DEGs were constructed, which consisted of 332 edges (biological associations) and 106 nodes (genes that form associations) (Fig. 3). Genes with a high degree of association included interleukin (IL)8, IL1 β , vascular endothelial growth factor A (VEGFA), intercellular adhesion molecule 1 (ICAM1), PTGS2 and CXCL10. Genes with degrees >10 in the PPI network are shown in Table IV.

Discussion

GWGS has been demonstrated to be a reliable method for comparing microarray data from different sources and studies (17). Based on this strategy, the analysis of the present study focused on significantly DEGs in order to reveal the transcriptional responses of each periodontitis sample. The results of this analysis indicated that these genes were enriched in cytokine-cytokine receptor interactions, cell adhesion molecules as well as complement and coagulation cascades.

Hub nodes are genes highly connected with other genes and have been identified to have important roles in numerous networks. In addition, highly connected hub genes were proposed to have an important role in biological development (22). Hub nodes have more complex interactions compared with those of other genes, which indicates that they have pivotal roles in the underlying mechanisms of disease. Identification of the hub genes involved in periodontitis may lead to the development of effective diagnostic and therapeutic approaches for periodontitis. In the present study, DEGs, which were algorithmically predicted to be biomarkers for periodontitis using GWGS, were found in high levels within patients with periodontitis. In addition, certain identified biomarkers, including IL8, IL1 β , ICAM1 and VEGFA, were hub genes, therefore suggesting that they may be useful diagnostic markers for periodontitis.

In the present study, the IL8 gene had a degree of 28, which was the highest degree in the PPI network. IL8 is a member

of the family of chemokines that mediates the activation and migration of neutrophils from peripheral blood into tissues (23). IL8 expression may be induced by IL18 in natural killer cells, which was reported to be inhibited by tumor necrosis factor binding protein (24). Kimata *et al* (25) demonstrated that IL8 selectively inhibited immunoglobulin (Ig)E and IgG4, which were induced by IL4; however, IL8 exerted no effects on the production of IgM, IgG1, IgA, IgG2 and IgG3. A previous study suggested that chronic stress was associated with increased IL8 levels (26). This was consistent with another study, which reported that individuals with high levels of trait anxiety were more prone to periodontal disease (27).

IL1 β is also a potent pro-inflammatory cytokine, which is upregulated by radiation and is involved in the regulation of other inflammation-associated molecules (28). In cultured human umbilical vein endothelial cells, IL1 β activated VCAM-1 gene expression (29). Several previous studies have demonstrated strong associations between the high frequency of genetic polymorphism in IL1 β and the development of severe chronic periodontitis (30-32). Gursoy *et al* (33) revealed that IL1 β was the only biomarker associated with periodontitis among the salivary cytokines and enzymes tested. In addition, IL1 β polymorphisms were reported to be associated with an increased risk of cancer (34,35).

In the present study, the VEGFA gene was also found to have a high degree of 24 in the PPI network. VEGFA is the predominant mediator of angiogenesis in the VEGF family (36) and is essential for chondrocyte survival during bone development (37). A previous study demonstrated that VEGFA stimulated lymphangiogenesis and hemangiogenesis in inflammatory neovascularization via macrophage recruitment (38). In a study by Kasprzak *et al* (39) VEGF was overexpressed in patients with chronic periodontitis, suggesting its significance in protracting the inflammatory process or periodic exacerbations of the process and the destruction of the periosteum. In addition, Tian *et al* (40) reported that the genotypes of VEGF and its protein production were associated with chronic periodontitis in a Chinese population. Furthermore, VEGFs were also found to have an important role in neurodegeneration (41).

ICAM1 was reported to have a role in the onset and manifestation of the inflammatory responses. ICAM1 was identified as a co-stimulatory ligand, which bound to lymphocyte function-associated antigen-1 (42); in addition, ICAM1 was reported to contribute to the adhesion of T lymphocytes to chondrocytes (43). A previous study demonstrated that ICAM1 was highly expressed in infected gingival cells as well as in tissues from periodontitis patients compared with those from healthy controls (44). Nedbal *et al* (45) suggested that the local topical applications of ICAM1-directed antisense oligonucleotides may be used as an effective therapeutic tool against the inflammatory processes of the human gingiva. Furthermore, ICAM1 expression has been reported to be associated with other diseases including Alzheimer's Disease (46) and cancer (47).

In conclusion, the present study identified several DEGs associated with periodontitis, as well as the functions and signaling pathways in which these genes were involved. Comprehensive network analyses of the dysregulated gene expression in periodontitis identified a series of hub genes that had high degrees of PPI in this network. This PPI network was

therefore proposed to reflect the interaction of genes in the periodontitis-specific micro-environment. In addition, the hub genes identified may be potential useful diagnostic markers and novel therapeutic targets for periodontitis.

References

- Loos BG, Craandijk J, Hoek FJ, *et al*: Elevation of systemic markers related to cardiovascular diseases in the peripheral blood of periodontitis patients. *J Periodontol* 71: 1528-1534, 2000.
- Iacopino AM and Cutler CW: Pathophysiological relationships between periodontitis and systemic disease: recent concepts involving serum lipids. *J Periodontol* 71: 1375-1384, 2000.
- Iacopino AM: Periodontitis and diabetes interrelationships: role of inflammation. *Ann Periodontol* 6: 125-137, 2001.
- Michalowicz BS, Diehl SR, Gunsolley JC, *et al*: Evidence of a substantial genetic basis for risk of adult periodontitis. *J Periodontol* 71: 1699-1707, 2000.
- Divaris K, Monda KL, North KE, *et al*: Exploring the genetic basis of chronic periodontitis: a genome-wide association study. *Hum Mol Genet* 22: 2312-2324, 2013.
- Grigoriadou ME, Koutayas SO, Madianos PN and Strub JR: Interleukin-1 as a genetic marker for periodontitis: review of the literature. *Quintessence Int* 41: 517-525, 2010.
- Scapoli L, Girardi A, Palmieri A, Carinci F, Testori T, Zuffetti F, Monguzzi R and Lauritano D: IL6 and IL10 are genetic susceptibility factors of periodontal disease. *Dent Res J (Isfahan)* 9 (Suppl 2): S197-S201, 2012.
- Zhang N, Xu Y, Zhang B, Zhang T, Yang H, Feng Z and Zhong D: Analysis of interleukin-8 gene variants reveals their relative importance as genetic susceptibility factors for chronic periodontitis in the Han population. *PLoS One* 9: e104436, 2014.
- Mercado FB, Marshall RI, Klestov AC and Bartold PM: Relationship between rheumatoid arthritis and periodontitis. *J Periodontol* 72: 779-787, 2001.
- Kinane DF and Hart TC: Genes and gene polymorphisms associated with periodontal disease. *Crit Rev Oral Biol Med* 14: 430-449, 2003.
- Wray CJ, Ko TC and Tan FK: Secondary use of existing public microarray data to predict outcome for hepatocellular carcinoma. *J Surg Res* 188: 137-42, 2014.
- Lee S, Stewart S, Nagtegaal I, Luo J, Wu Y, Colditz G, Medina D and Allred DC: Differentially expressed genes regulating the progression of ductal carcinoma in situ to invasive breast cancer. *Cancer Res* 72: 4574-4586, 2012.
- Barrett T, Troup DB, Wilhite SE, *et al*: NCBI GEO: archive for functional genomics data sets - 10 years on. *Nucleic Acids Res* 39: D1005-D1010, 2011.
- Parkinson H, Kapushesky M, Shojatalab M, *et al*: Array Express - a public database of microarray experiments and gene expression profiles. *Nucleic Acids Res* 35: D747-D750, 2007.
- Demmer RT, Behle JH, Wolf DL, *et al*: Transcriptomes in healthy and diseased gingival tissues. *J Periodontol* 79: 2112-2124, 2008.
- Becker ST, Beck-Broichsitter BE, Graetz C, Dörfer CE, Wiltfang J and Häslner R: Peri-implantitis versus periodontitis: functional differences indicated by transcriptome profiling. *Clin Implant Dent Relat Res* 16: 401-411, 2014.
- Liu W, Peng Y and Tobin DJ: A new 12-gene diagnostic biomarker signature of melanoma revealed by integrated microarray analysis. *Peer J* 1: e49, 2013.
- Miosge N: Expression data from human periodontal ligament. *Gene Expression Omnibus [Database]*, 2011.
- Stratowa C: Introduction to the xps Package: Overview, 2013.
- Smyth GK: Linear models and empirical bayes methods for assessing differential expression in microarray experiments. *Stat Appl Genet Mol Biol* 3: Article3, 2004.
- Huang da W, Sherman BT and Lempicki RA: Systematic and integrative analysis of large gene lists using DAVID bioinformatics resources. *Nat Protoc* 4: 44-57, 2009.
- Langfelder P, Mischel PS and Horvath S: When is hub gene selection better than standard meta-analysis? *PLoS One* 8: e61505, 2013.
- Puthothu B, Krueger M, Heinze J, Forster J and Heinzmann A: Impact of IL8 and IL8-receptor alpha polymorphisms on the genetics of bronchial asthma and severe RSV infections. *Clin Mol Allergy* 4: 2, 2006.
- Puren AJ, Fantuzzi G, Gu Y, Su MS and Dinarello CA: Interleukin-18 (IFN γ -inducing factor) induces IL8 and IL1 β via TNF α production from non-CD14 $^{+}$ human blood mononuclear cells. *J Clin Invest* 101: 711-721, 1998.

25. Kimata H, Yoshida A, Ishioka C, Lindley I and Mikawa H: Interleukin 8 (IL8) selectively inhibits immunoglobulin E production induced by IL4 in human B cells. *J Exp Med* 176: 1227-1231, 1992.
26. Shahzad MM, Arevalo JM, Armaiz-Pena GN, *et al*: Stress effects on FosB-and interleukin-8 (IL8)-driven ovarian cancer growth and metastasis. *J Biol Chem* 285: 35462-35470, 2010.
27. Vettore MV, Leão AT, Monteiro Da, Silva AM, Quintanilha RS and Lamarca GA: The relationship of stress and anxiety with chronic periodontitis. *J Clin Periodontol* 30: 394-402, 2003.
28. Liu W, Ding I, Chen K, *et al*: Interleukin 1 β (IL1 β) signaling is a critical component of radiation-induced skin fibrosis. *Radiat Res* 165: 181-191, 2006.
29. Marui N, Offermann MK, Swerlick R, *et al*: Vascular cell adhesion molecule-1 (VCAM-1) gene transcription and expression are regulated through an antioxidant-sensitive mechanism in human vascular endothelial cells. *J Clin Invest* 92: 1866-1874, 1993.
30. Tawil G, Akl FA, Dagher MF, Karam W, Abdallah Hajj Hussein I, Leone A and Jurjus AR: Prevalence of IL-1 β +3954 and IL-1 α -889 polymorphisms in the Lebanese population and its association with the severity of adult chronic periodontitis. *J Biol Regul Homeost Agents* 26: 597-606, 2012.
31. Tsarev VN and Nikolaeva EN: Polymorphism of IL1 α and IL1 β genes and bacterial invasion in patients with chronic generalized periodontitis. *Stomatologiia (Mosk)* 89: 19-23, 2010 (In Russian).
32. Gayathri R, Saadi AV, Bhat KM, Bhat SG and Satyamoorthy K: Allele, genotype, and composite genotype effects of IL-1A +4845 and IL-1B +3954 polymorphisms for chronic periodontitis in an Indian population. *Indian J Dent Res* 22: 612, 2011.
33. Gursoy UK, Könönen E, Uitto VJ, Pussinen PJ, Hyvärinen K, Suominen-Taipale L and Knuutila M: Salivary interleukin-1 β concentration and the presence of multiple pathogens in periodontitis. *J Clin Periodontol* 36: 922-927, 2009.
34. Al-Moundhri MS, Al-Nabhani M, Al-Bahrani B, *et al*: Interleukin-1 β gene (IL1B) and interleukin 1 receptor antagonist gene (IL1RN) polymorphisms and gastric cancer risk in an Omani Arab population. *Gastric Cancer* 9: 284-290, 2006.
35. Xu J, Yin Z, Cao S, *et al*: Systematic review and meta-analysis on the association between IL1B polymorphisms and cancer risk. *PLoS One* 8: e63654, 2013.
36. Ghosh R, Lipson KL, Sargent KE, *et al*: Transcriptional regulation of VEGF-A by the unfolded protein response pathway. *PLoS One* 5: e9575, 2010.
37. Zelzer E, Mamluk R, Ferrara N, Johnson RS, Schipani E and Olsen BR: VEGFA is necessary for chondrocyte survival during bone development. *Development* 131: 2161-2171, 2004.
38. Cursiefen C, Chen L, Borges LP, *et al*: VEGF-A stimulates lymphangiogenesis and hemangiogenesis in inflammatory neovascularization via macrophage recruitment. *J Clin Invest* 113: 1040-1050, 2004.
39. Kasprzak A, Surdacka A, Tomczak M, Przybyszewska W, Seraszek-Jaros A, Malkowska-Lanzafame A, Siodla E and Kaczmarek E: Expression of angiogenesis-stimulating factors (VEGF, CD31, CD105) and angiogenic index in gingivae of patients with chronic periodontitis. *Folia Histochem Cytobiol* 50: 554-564, 2012.
40. Tian Y, Li JL, Hao L, *et al*: Association of cytokines, high sensitive C-reactive protein, VEGF and beta-defensin-1 gene polymorphisms and their protein expressions with chronic periodontitis in the Chinese population. *Int J Biol Markers* 28: 100-107, 2013.
41. Storkebaum E and Carmeliet P: VEGF: a critical player in neurodegeneration. *J Clin Invest* 113: 14-18, 2004.
42. Lebedeva T, Dustin ML and Sykulev Y: ICAM-1 co-stimulates target cells to facilitate antigen presentation. *Curr Opin Immunol* 17: 251-258, 2005.
43. Kienzle G and von Kempis J: Vascular cell adhesion molecule 1 (CD106) on primary human articular chondrocytes: functional regulation of expression by cytokines and comparison with intercellular adhesion molecule 1 (CD54) and very late activation antigen 2. *Arthritis Rheum* 41: 1296-1305, 1998.
44. Umeda JE, Demuth DR, Ando ES, Faveri M and Mayer MP: Signaling transduction analysis in gingival epithelial cells after infection with *Aggregatibacter actinomycetemcomitans*. *Mol Oral Microbiol* 27: 23-33, 2012.
45. Nedbal W, Tomakidi P, Lehmann MJ, Dorfer C, Kohl A and Sczakiel G: Antisense-mediated inhibition of ICAM-1 expression: a therapeutic strategy against inflammation of human periodontal tissue. *Antisense Nucleic Acid Drug Dev* 12: 71-78, 2002.
46. Mattila KM, Hiltunen M, Rinne JO, *et al*: Absence of association between an intercellular adhesion molecule 1 gene E469K polymorphism and Alzheimer's disease in Finnish patients. *Neurosci Lett* 337: 61-63, 2003.
47. Usami Y, Ishida K, Sato S, *et al*: Intercellular adhesion molecule-1 (ICAM-1) expression correlates with oral cancer progression and induces macrophage/cancer cell adhesion. *Int J Cancer* 133: 568-578, 2013.