



Medication-Related Osteonecrosis of the Jaw: A Systematic Review and a Bioinformatic Analysis

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Abstract: The objective was to evaluate the current evidence regarding the etiology of medicationrelated osteonecrosis of the jaw (MRONJ). This study systematically reviewed the literature by searching PubMed, Web of Science, and ProQuest databases for genes, proteins, and microRNAs associated with MRONJ from the earliest records through April 2023. Conference abstracts, letters, review articles, non-human studies, and non-English publications were excluded. Twelve studies meeting the inclusion criteria involving exposure of human oral mucosa, blood, serum, saliva, or adjacent bone or periodontium to anti-resorptive or anti-angiogenic agents were analyzed. The Cochrane Collaboration risk assessment tool was used to assess the quality of the studies. A total of 824 differentially expressed genes/proteins (DEGs) and 22 microRNAs were extracted for further bioinformatic analysis using Cytoscape, STRING, BiNGO, cytoHubba, MCODE, and ReactomeFI software packages and web-based platforms: DIANA mirPath, OmicsNet, and miRNet tools. The analysis yielded an interactome consisting of 17 hub genes and hsa-mir-16-1, hsa-mir-21, hsa-mir-23a, hsa-mir-145, hsa-mir-186, hsa-mir-221, and hsa-mir-424. A dominance of cytokine pathways was observed in both the cluster of hub DEGs and the interactome of hub genes with dysregulated miRNAs. In conclusion, a panel of genes, miRNAs, and related pathways were found, which is a step toward understanding the complexity of the disease.

Keywords: osteonecrosis; systematic review; computational biology

1. Introduction

Medication-related osteonecrosis of the jaw (MRONJ) encompasses osteonecrosis of the jaw associated with bisphosphonates, denosumab, and anti-resorptive agents [1–4]. It is a rare but serious drug reaction associated often with receiving high doses of anti-angiogenic and anti-resorptive medication, including mTOR inhibitors [5–7]. Anti-resorptive drugs, namely bisphosphonate and denosumab, are monoclonal antibodies that target the receptor activator of the tumor necrosis factor ligand superfamily member 11 [8,9]. Under conditions of accelerated skeletal turnover, bisphosphonates are selectively absorbed at sites of active bone remodeling [10]. Anti-angiogenic drugs, namely sunitinib and bevacizumab, are humanized monoclonal antibodies that impede the creation of novel blood vessels by suppressing the function of tyrosine kinases and vascular endothelial growth factor (VEGFA) [11].

MRONJ is a recognized phenomenon in nearly 1% of cancer patients and in 0.1% of those suffering from metabolic bone diseases [12]. The prevalence of MRONJ reported in studies varies widely, with incidence rates ranging from as low as 0.01% following low-dose oral bisphosphonate therapy to as high as 14.4% in high dose intravenous bisphosphonate treatment [13]. Dental extractions, implant procedures, oral and maxillofacial surgeries, periodontal disease, and invasive periodontal procedures have been identified as risk



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Copyright: © 2023 by the authors. Licensee MDPI, Basel, Switzerland. This article is an open access article distributed under the terms and conditions of the Creative Commons Attribution (CC BY) license (https:// creativecommons.org/licenses/by/ 4.0/). factors for MRONJ, with local inflammation being of greatest importance [14]. The etiology of MRONJ is multifactorial, encompassing multiple deficiencies that synchronize to result in bone resorption suppression [15], infection/inflammation [16], immune system dysfunction [17], angiogenesis inhibition [18], soft tissue toxicity [19], and systemic disorders related to conditions such as rheumatoid arthritis or diabetes mellitus [20]. Although there is no conclusive pathophysiology supported by scientific data, numerous fundamental queries persist. Even though there has been a decades-long investigation, the exact reason MRONJ is more frequently observed in the jawbone remains unclear.

Oral disorders encompass a range of conditions with worldwide prevalence and clinical significance. These ailments can have mutilating effects and significantly diminish the quality of life, as they affect a restricted area with critical physiological and social functions. Although crucial, several oral illnesses remain inadequately understood and frequently receive ineffective treatment. It is crucial to have a comprehensive understanding of the mechanisms at play in oral diseases to identify dependable, mechanistic indicators of clinical results, establish targeted therapeutic strategies, and customize prevention and treatment techniques. Conventional analysis of diseases only provides surface-level interpretations. Gaining a comprehensive knowledge of complex human disorders necessitates collating all pertinent data and scrutinizing biomarkers that are genetically associated with disease susceptibility. To facilitate the identification of innovative underlying molecular disease mechanisms, unbiased screening methods have been employed at various molecular levels to produce large-scale datasets. Numerous biological research areas, such as the investigation of oral conditions, gain advantages from recognizing these processes and biomarkers in single-omics analysis [21]. The application of multi-omics approaches, including genomics, transcriptomics, proteomics, and metabolomics, provides comprehensive molecular insights beyond single-omics methods. Therefore, the analysis aimed to pool existing data on the pathophysiological processes of MRONJ in humans, provided by multiomics techniques such as high-throughput sequencing, gene expression arrays, and mass spectroscopy, to identify groups of biomarkers differentially expressed between cohorts and worthy of further investigation. At the same time, the objective of the work was to reveal altered signaling pathways and to create a multidimensional, layered configuration of MRONJ that would provide new insights into its pathobiology.

2. Materials and Methods

The results presented in this systematic analysis followed the Preferred Reporting Items for Systematic Reviews and Meta-Analyses (PRISMA 2020) guidelines [22].

2.1. Study Selection

2.1.1. Inclusion Criteria

To be included, studies had to involve human oral mucosa, blood, serum, saliva, or adjacent bone or periodontium exposed to anti-resorptive or anti-angiogenic agents. The date of publication was not restricted, but only English language articles were considered.

2.1.2. Exclusion Criteria

Studies that have been retracted previously, along with reviews, conference abstracts, case reports, letters to the editor, case studies, and studies involving non-human animal studies, have been excluded.

2.1.3. Screening Process

The screening process was conducted with predetermined, objective inclusion criteria after the completion of a literature search. Figure 1 represents the procedure outlined. The survey was conducted between 7 and 10 March 2023. From inception until April 2023, two evaluators independently searched the databases of PubMed, Web of Science (WoS), and ProQuest. The data were pooled from those source articles that were related to mRNAs, microRNAs (miRNAs), proteins, and metabolites. The study merged key phrases with the

logical operator "OR" and the results with the logical operator "AND". The following terms were used in the search strategy: "mronj", "bronj", "aronj", "dronj", "medication-related osteonecrosis of the jaw", "bisphosphonate-related osteonecrosis of the jaw", "antiresorptive agent-related osteonecrosis of the jaw", and "denosumab-related osteonecrosis of the jaw". In addition, the search terms for transcriptomics were "microRNA" or "miRNA" and "transcriptome", "transcriptomics" or "mRNA". Proteomics was searched using the terms "proteomics" or "proteome", while metabolomics was obtained by searching for "metabolome", "metabolome", "metabolome". No year restrictions were applied for article publication.

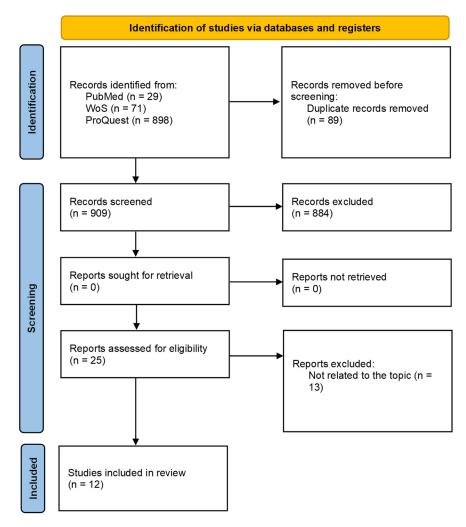


Figure 1. PRISMA flow diagram of literature for a systematic review on the screening of medicationinduced osteonecrosis of the jaw.

Further, the screening process involved manually removing duplicate results from the analysis. Two reviewers independently analyzed the titles and abstracts of the papers and evaluated the remaining articles to determine their eligibility. If a study's suitability could not be determined solely from its title and abstract, its full text was examined. Citations for the included papers were tracked using Google Scholar or PubMed. A manual review of the reference lists of the included articles was conducted to select relevant articles. The results were then summarized from the articles that satisfied the inclusion criteria. The Revised Cochrane risk-of-bias tool for randomized trials (RoB 2, https://www.riskofbias. info/welcome/rob-2-0-tool/current-version-of-rob-2, accessed on 14 November 2023) was used to assess the risk of bias in domains related to the randomization process, deviations

from the intended interventions, missing outcome data, measurement of the outcome, and selection of the reported result.

2.2. Data Analysis

2.2.1. Gene Ontology Enrichment Analysis

The GeneCards human gene database (https://www.genecards.org, accessed on 10 July 2023) was used to verify and revise the list of differentially expressed genes (DEGs) compiled from eligible articles. Protein accession numbers were mapped to genes using the UniProt mapping tool (https://www.uniprot.org, accessed on 14 July 2023) [23]. The downloaded matching genes were then used for further investigation.

To investigate DEGs for overrepresentation in the hierarchical gene ontology (GO), the extensions of Cytoscape 3.10.0 [24] and BiNGO 3.0.5 [25] were used. The enrichment analysis for cell components, molecular function, and biological process terms was performed using the Benjamini and Hochberg multiple testing procedure. The significance level was set at 0.05 (p < 0.05).

2.2.2. Protein–Protein Interaction Network and Module Analysis

The network of DEGs was created using the Search Tool for the Retrieval of Interacting Genes (STRING) database, which presents both insights and predictions regarding protein–protein interactions (PPIs). The network itself was constructed with the help of StringApp, Version 2.0.1 [26].

CytoHubba 0.1 [27] and Molecular Complex Detection (MCODE) 2.0.3 [28] allowed the exploration of hub genes and clusters within the network. All CytoHubba plugin algorithms, including Maximum Neighborhood Component (MNC), Maximal Clique Centrality (MCC), Density of Maximum Neighborhood Component (DMNC), Degree (Deg), Betweenness (BC), Bottleneck (BN), Closeness (Clo), EcCentricity (EC), Edge Percolated Component (EPC), Stress (Str), and Radiality (Rad), were used to detect the hub genes [27]. The MCODE clustering was performed with a degree cutoff of 2, a node score cutoff of 0.2, a maximum depth of 20, and a k-score of 5.

2.2.3. Pathway Enrichment Analysis

The ReactomeFI plug-in pathway database version 8.0.6 [29] was used for pathway enrichment analysis.

The mirPath v.4 database from DIANA Tools (https://diana-lab.e-ce.uth.gr/app/miRPathv4, accessed on 17 July 2023) was utilized to identify genes that could serve as miRNA target candidates.

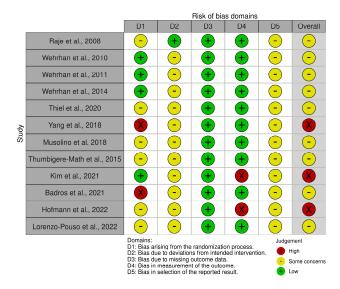
2.2.4. Multi-Omics Network

The multi-omics data were integrated using web-based platforms such as OmicsNet (https://www.omicsnet.ca, accessed on 19 July 2023) [30] and miRNet (https://www.mirnet.ca, accessed on 21 July 2023) [31]. If the multi-omics network consisted of more than 3000 nodes, we implemented the minimum network setting, i.e., the algorithms that generate the minimum network connecting all specified nodes.

3. Results

3.1. Systematic Review of Screening for MRONJ

The search strategy produced 998 articles. Twenty-five articles underwent full-text review following a screening of their titles and abstracts. Twelve articles were ultimately included in the library after thirteen articles were excluded following a thorough examination of their full text (Figure 2). Exclusion criteria comprised such items as conference abstracts, letters, and review articles; non-human studies; and publications in languages other than English. Table 1 summarizes the characteristics of the studies on MRONJ that were included. Nine of the twelve studies that met the inclusion criteria were found to



have an overall risk of bias of some concern, and three were found to have a high overall risk of bias.

Figure 2. Risk of bias assessment [32–43]. According to the Revised Cochrane risk-of-bias tool for randomized trials (RoB 2, https://www.riskofbias.info/welcome/rob-2-0-tool/current-version-of-rob-2, accessed on 14 November 2023).

mRNA Reference Sample Type Method Peripheral blood mononuclear cells. Raje et al., 2008, Patients: MM patients with ONJ (n = 8). Affymetrix U133Plus 2.0 Gene Chip 10.1158/1078-0432.CCR-07-1430 [32] Controls: MM patients without ONJ (Affymetrix, Santa Clara, CA, USA). (n = 10), healthy volunteers (n = 5). Microfluid Lab-on-a-Chip technology (Agilent RNA 6000 Pico Kit and the Agilent 2100 Bioanalyzer, Agilent, Waldbronn, Periodontal samples. Germany). The cDNAs from total RNA were Wehrhan et al., 2010, Patients: patients with BRONJ (n = 20). synthesized using the High-Capacity cDNA 10.1186/1479-5876-8-96 [33] Controls: non-BP exposed periodontal Archive Kit (Cat. 4322171; Applied samples (n = 20). Biosystems, Foster City, CA, USA). Real-time RT qPCR (QuantiTect Primer Assay; Qiagen, Hilden, Germany). Microfluid Lab-on-a-Chip technology (Agilent RNA 6000 Pico Kit and the Agilent 2100 Bioanalyzer, Agilent, Waldbronn, Periodontal samples. Germany). The cDNAs from total RNA were Wehrhan et al., 2011, Patients: patients with BRONJ (n = 20). synthesized using the High-Capacity cDNA 10.1111/j.1601-0825.2010.01778.x [34] Controls: non-BP exposed periodontal Archive Kit (Cat. 4322171; Applied samples (n = 20). Biosystems, Foster City, CA, USA). Real-time RT qPCR (QuantiTect Primer Assay; Qiagen, Hilden, Germany).

Table 1. Characteristics of included studies on medication-induced osteonecrosis of the jaws.

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	mRNA	
Reference	Sample Type	Method
Wehrhan et al., 2014 10.1007/s00784-014-1354-7 [35]	Jawbone samples. Patients: patients with BRONJ ($n = 15$). Controls: non-BP exposed samples ($n = 20$).	Total RNA extraction (RNeasy Kit, Qiagen, Hilden, Germany). Microfluid Lab-on-a-Chip technology (Agilent RNA 6000 Pico Kit and the Agilent 2100 Bioanalyzer, Agilent, Waldbronn, Germany). High-capacity cDNA Archive Kit (Cat. No. 4322171; Applied Biosystem, Foster City, CA, USA). Real-time RT quantitative PCR analyses: Hs_SPP1_1_SGQuantiTect Primer Assay (200) on the ABI Prism 7300 Sequence Detection System (Applied Biosystems, Waltham, MA, USA). PCR amplification: the QuantiTect TM SYBR [®] green PCR kit (Cat. No. 204143; Qiagen, Hilden, Germany).
Thiel et al., 2020 10.1016/j.prp.2020.153245 [<mark>36</mark>]	Jawbone samples. Patients: diagnosed with MRONJ ($n = 12$). Controls: subjects without MRONJ ($n = 6$).	RNA extraction kit (miRNeasy Mini Kit; Qiagen, Hilden, Germany). The total RNA was reverse transcribed into cDNA using the iScript [™] cDNA Synthesis Kit (Bio-Rad, Hercules, CA, USA). PCR amplification: SsoAdvanced [™] Universal SYBR [®] Green Supermix (Bio-Rad, Hercules, CA, USA) Amplification was conducted on the CFX Connect Real-Time PCR System (Bio-Rad, Hercules, CA, USA).
	miRNA	
Reference	Sample Type	Method
Raje et al., 2008, 10.1158/1078-0432.CCR-07-1430. [32]	Peripheral blood mononuclear cells. Patients: MM patients with ONJ ($n = 8$). Controls: MM patients without ONJ ($n = 10$), healthy volunteers ($n = 5$).	Affymetrix U133Plus 2.0 Gene Chip (Affymetrix, Santa Clara, California, USA).
Yang et al., 2018, 10.7150/ijms.27593 [37]	Serum. Patients: patients with BRONJ ($n = 6$). Controls: non-BP healthy individuals ($n = 11$).	RNA extraction: mirVana Paris Kit (Ambion, Huntingdon, Cambridgeshire, United Kingdom). The microRNAs were reversed to cDNA using the miScript II RT Kit (Qiagen, Hilden, Germany). Q-RT-PCR analysis was conducted using the miScript SYBR Green PCR Kit (Qiagen, Hilden, Germany) with a 7500 Real-Time PCR System (Applied Biosystems, Foster City, CA, USA).
Musolino et al. 2018, 10.1007/s00277-018-3296-7 [38]	Peripheral blood. Patients: MM patients with ONJ ($n = 5$). Controls: healthy volunteers ($n = 5$).	RNA extraction: the Total Purification Plus Kit (Norgen Biotek Corporation, Thorold, ON, Canada). Total RNA was transcribed into cDNA through an All-in-One miRNA first-strand cDNA synthesis kit (GeneCopoeia Inc., Rockville, MD, USA). Real-Time qPCR employed a 7500 Real-Time PCR System (Applied Biosystems, Foster City, CA, USA).

Table 1. Cont.

	Proteins	
Reference	Sample Type	Method
Thumbigere-Math et al., 2015, 10.1111/odi.12204 [39]	Saliva. Patients: BRONJ (<i>n</i> = 20), high- and low-infusion groups. Controls: non-BRONJ patients (<i>n</i> = 20).	iTRAQ labeling was followed by fractionation using strong cation exchange chromatography, and fractions were analyzed by reversed-phase microcapillary LC-S (LTQ-Orbitrap).
Kim et al., 2021, 10.7150/ijms.61552 [40]	MG-63, SCC-9, SCC-15, and HUVEC cells. ALN-treated and non-ALN control groups.	2D-DIGE, followed by MALDI TOF/TOF MS (4800 Plus, Applied Biosystems, Foster City, CA, Life Sciences, USA).
Badros et al., 2021, 10.3389/fonc.2021.704722 [41]	Saliva, serum. Patients: MM patients who underwent intravenous BP therapy and developed BRONJ ($n = 14$). Controls: non-BRONJ MM patients ($n = 96$).	Luminex™ technology (EMD Millipore, Burlington, MA, USA).
Hofmann et al., 2022, 10.1007/s10266-022-00691-y [42]	HAOB cells. BEV/SUN-treated and non-BEV/SUN control groups.	ELISA
Lorenzo-Pouso et al., 2022, 10.1111/odi.14201 [43]	Saliva. Patients: Group 1—MRONJ cases ($n = 18$). Controls: Group 2—individuals undergoing treatment with BMAs for more than 24 months without MRONJ ($n = 10$). Group 3—healthy volunteers ($n = 10$).	SDS-PAGE, shotgun DDA by micro-flow LC-MS/MS, a quadrupole-TOF mass spectrometer (Triple TOF 6600 [SCIEX, Framingham, MA, USA]) working in ESI + performed DDA analysis.

ALN—alendronate, BEV—bevacizumab, BP—bisphosphonate; BMAs—bone-modifying agents, BRONJ—bisphosphonate-related osteonecrosis of the jaw, DDA—data-dependent acquisition, DEP—differentially expressed proteins, ELISA—enzyme-linked immunoabsorbance assay, iTRAQ—isobaric tags for relative and absolute quantitation, LC—liquid chromatography, MM—multiple myeloma, MS—mass spectrometry, ONJ—osteonecrosis of the jaw, SDS-PAGE—sodium dodecyl-sulfate polyacrylamide gel electrophoresis, SUN—sunitinib, TOF—timeof-flight.

No relevant studies have been found regarding the metabolomics of MRONJ.

3.2. Network Analysis of Protein Interaction Data

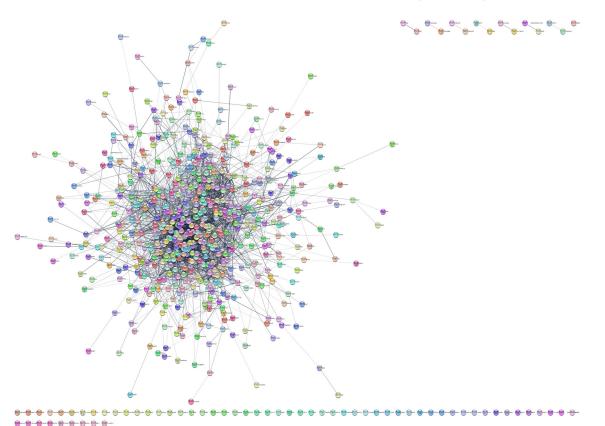
To investigate the protein interactions involved in MRONJ pathogenesis, we utilized STRING databases to analyze the 824 identified genes/proteins and created an interactive network via Cytoscape. Figure 3 depicts the resulting network consisting of 701 genes and 3993 edges, while Table 2 summarizes the network topology.

Table 2. The most relevant topological parameters of the STRING network.

Topological Parameters	Values
The average number of neighbors	12.755
Clustering coefficient	0.264
Characteristic path length	3.294
Network diameter	9
Number of edges	3993
Number of nodes	701

The network topology was analyzed using the cytoHubba (Version 0.1) extension of Cytoscape. The highly linked hub genes were extracted from the main complex network of DEGs by using the algorithms of cytoHubba. Subsequently, 24 genes were extracted that occurred at the intersection of at least three methods: *ALB*, *ANXA5*, *ATM*, *CCL2*, *CD44*,

Table 1. Cont.



CXCL8, CXCR4, EEF2, EGF, GART, HSP90AB1, HSPA4, IGF1, IL1B, IL6, ITGB1, JUN, LMNA, MMP9, PTPRC, RAB5A, RHOA, TNF, and VEGFA (see Table 3).

Figure 3. The STRING network formed from the dysregulated genes/proteins in MRONJ.

Table 3. Hub genes identified by the cytoHubba plug-in algorithms. The crucial hub genes that occurred at the intersection of at least three methods are highlighted in red.

MCC	DMN	MNC	Degree	FPC	Bottleneck	EcCentricity	Closeness	Radiality	Betweenness	Stress	Clustering Coefficient
ALB	A1BG	ALB	ALB	ALB	ALB	ARF1	ALB	ALB	ALB	ALB	A2ML1
ANXA5	AGER	ANXA5	ANXA5	ANXA5	CAT	ARHGDIA	ANXA5	ANXA5	ATM	ATM	BANK1
CCL2	ANGPT1	B2M	CCL2	CCL2	CD44	ARRB1	ATM	ATM	CD44	CD44	CFHR5
CD44	CD83	CCL2	CD44	CD44	CXCL8	BCL2L11	CCL2	CASP8	CLTC	CXCL8	DPT
CSF3	CXCL1	CD44	CXCL8	CXCL8	EEF1A1	BTK	CD44	CD44	EEF2	EEF2	ENSP00000330898
CXCL8	CXCL2	CXCL8	CXCR4	CXCR4	EGF	CD83	CXCL8	CXCL8	EGF	EGF	ENSP00000377747
CXCR4	EEF1B2	CXCR4	EEF2	EGF	FUS	COL1A1	CXCR4	CXCR4	GART	GART	FAM213A
EGF	EIF2S3	EGF	EGF	HSPA4	GART	CYCS	EGF	CYCS	HIST1H4F	HSP90AB1	GTF3C4
IGF1	EIF5A2	HSP90AB1	HSP90AB1	IGF1	HIST1H4F	DNAJB1	HSP90AB1	EGF	HSP90AB1	HSPA4	CHI3L1
IL1B	CHI3L1	HSPA4	HSPA4	IL1B	HSP90AB1	FAS	HSPA4	HSP90AB1	HSPA4	IL1B	IL36A
IL6	LRG1	IGF1	IGF1	IL6	HSPA4	FCGR3A	IGF1	HSPA4	IL6	IL6	KRT76
JUN	MMP1	IL1B	IL1B	ITGB1	IGF1	GART	IL1B	IGF1	JUN	JUN	LMF1
KDR	ORM1	IL6	IL6	JUN	JUN	IL6	IL6	IL1B	LMNA	LMNA	ME1
MMP9	ORM2	ITGB1	ITGB1	MMP9	KRT14	ITGB1	ITGB1	IL6	PTPRC	PTPRC	NOV
PTGS2	PSMC1	JUN	JUN	PTGS2	LMNA	KRT19	JUN	JUN	RAB5A	RAB5A	POLR2J3
PTPRC	RPLP1	MMP9	MMP9	PTPRC	PTPRC	NR4A2	MMP9	MMP9	RHOA	RHOA	SEL1L3
SPP1	RPLP2	PTPRC	PTPRC	RHOA	RAB5A	PPP2CB	PTPRC	PTPRC	SRSF1	TFRC	SELM
TGFB1	SAA4	RHOA	RHOA	SPP1	RHOA	SAA4	RHOA	RHOA	TFRC	TNF	SERPIND1
TNF	SERPIND1	TNF	TNF	TNF	SRSF1	TXN	TNF	TNF	TNF	VEGFA	TNN
VEGFA	TNFRSF11B	VEGFA	VEGFA	VEGFA	TNF	VEGFA	VEGFA	VEGFA	VEGFA	YWHAZ	VPS36

In complement to the cytoHubba algorithms, MCODE clustering was employed to detect the molecular complexes and the seeds—the hub genes with a high degree of connectivity. In a complex PPI network of DEGs, MCODE identified five of the intra-

connected regions/clusters and five hub genes/seeds with a high degree of connectivity (see Table 4).

Table 4. MCODE-interconnected clusters generated from the Cytoscape-derived gene interaction network. The seed node with the highest score within the cluster is marked with an asterisk.

Cluster	Score (Density * Nodes)	Nodes	Edges	Node IDs
1	24.794	64	781	BGLAP, BMP2, CAT, CCL4, CCT2, CD44, COL1A1, CSF3, CXCL1, CXCL2, CXCL8, CXCR4, CYCS, EEF1A1, EEF1B2, EEF1D, EEF1G, EEF2, EGF, EIF2S3, EIF5A, EIF5A2, FGG, IGF1, IL1B, IL6, ITGB1, JUN, KDR, MARS, MMP1, MMP8, MMP9, NT5E, PSMC1, PTGS2, PTPRC, RHOA, RPL10, RPL12, RPL27A, RPL4, RPLP1, RPLP2, RPS10, RPS12, RPS16, RPS23, RPS25, RPSA, RUNX2, SERPINA1, SERPINC1, SOD2, SPP1, TGFB1, TNF *, TNFRSF11B, TNFSF11, TPT1, VEGFA
2	13.429	43	282	A1BG *, A2M, AGER, AMBP, ANXA5, APOA2, APOB, APOH, ATM, AZGP1, BCL2L11, C3, C4B, CASP8, CCL2, CP, CREB1, FAS, FCGR3A, FOXO1, GART, GIG25, HP, HPX, HSP90AB1, HSPA4, HSPB1, ITIH2, ITIH4, JAK1, KLRK1, LCK, LCN2, LRG1, MCL1, NFATC1, ORM1, ORM2, PDGFB, TF, TFRC, TTR, TXN
3	10.133	16	76	DSG1, DSP, IVL, KRT14, KRT15, KRT16, KRT17, KRT4, KRT5, KRT6B, KRT6C, SCEL, SPRR1A *, SPRR1B, SPRR3, TGM1
4	6.933	31	104	ACTG2, ALAS2, ATRX, CA2, CBFB, DDX3X, ETS1, GATA2, H2AFJ, HBA1, HBA2, HBB *, HBD, HBG1, HBG2, HIST1H1B, HIST1H1E, HIST1H2AB, HIST1H2AC, HIST1H3J, KMT2A, MYL12A, MYL6, SLC25A37, SLC4A1, SRSF1, SUPT16H, TAL1, TPM2, TPM3, TPM4
5	6	6	16	CELF1, FUS, HNRNPK *, MBNL1, SRSF10, SRSF3

In conjunction with cluster 1, the highest scoring MCODE clustering module, with the cytoHubba analysis results, a total of 17 hub genes were retrieved, comprising *ALB*, *CD44*, *CXCL2*, *CXCL8*, *CXCR4*, *EEF2*, *EGF*, *IGF1*, *IL1B*, *IL6*, *ITGB1*, *JUN*, *MMP9*, *PTPRC*, *RHOA*, *TNF*, and *VEGFA*.

3.3. GO Enrichment Analysis

A survey of how genes and gene products are represented in the biological domains concerning three aspects of molecular biology was conducted using Gene Ontology (GO). To associate GO terms with gene and protein sets, a series of enrichment analyses were performed in Cytoscape with the help of the BiNGO extension.

The dysregulated genes within the complex gene panel were linked to 422 GO terms, consisting of 333 biological process terms, 23 molecular function terms, and 66 cell component terms, as identified by GO enrichment analysis. The top GO enrichment terms associated with DEGs by *p*-values are shown in Table 5. The key molecular biological processes identified among these genes are those involved in regulating the immune system and the organism's immune response. Numerous genes have been implicated in protein binding. DEGs were predominantly linked to the extracellular region and the extracellular space.

A total of 918 GO terms were obtained from the analysis of seventeen hub genes. Of these GO terms, 893 were related to biological processes, 22 were associated with molecular function, and 23 were linked to cell component terms. Protein phosphorylation of amino acids constitutes a central molecular biological process. While protein binding was a common association among all hub genes, the top-ranked molecular function was cytokine receptor binding. DEGs were predominantly located in the extracellular space and region, as indicated in Table 6.

Table 5. Gene ontology enrichment analysis performed in Cytoscape using the BiNGO extension. The full set of dysregulated genes/proteins was considered in the analysis. The most enriched gene ontology terms based on the respective *p*-values are depicted.

GO-ID	Description	<i>p</i> -Value	Corr <i>p</i> -Value	x	n	X	Ν
		Biological Pro	cess				
2376	immune system process	$2.9469 imes 10^{-15}$	$1.0179 imes 10^{-11}$	97	947	631	14,265
6950	response to stress	$1.7284 imes 10^{-13}$	$2.9850 imes 10^{-10}$	143	1771	631	14,265
9611	response to wounding	$4.6100 imes 10^{-13}$	$5.3076 imes 10^{-10}$	64	541	631	14,265
6955	immune response	$2.5317 imes 10^{-12}$	2.1861×10^{-9}	68	618	631	14,265
6952	defense response	$8.4706 imes 10^{-12}$	$5.1466 imes 10^{-9}$	67	620	631	14,265
42221	response to a chemical stimulus	$8.9402 imes 10^{-12}$	$5.1466 imes 10^{-9}$	120	1462	631	14,265
48513	organ development	$1.7962 imes 10^{-11}$	8.8627×10^{-9}	138	1792	631	14,265
48583	regulation of response to stimulus	$3.1299 imes 10^{-11}$	1.3514×10^{-8}	59	524	631	14,265
9888	tissue development	$1.3146 imes 10^{-10}$	4.9607×10^{-8}	73	750	631	14265
6954	inflammatory response	$1.4362 imes 10^{-10}$	4.9607×10^{-8}	42	315	631	14,265
		Molecular Fun	ction				
5515	protein binding	$2.9969 imes 10^{-19}$	$2.6493 imes 10^{-16}$	462	8106	667	15,404
5198	structural molecule activity	$2.5854 imes 10^{-13}$	$1.1427 imes 10^{-10}$	68	600	667	15,404
5488	binding	$4.2458 imes 10^{-11}$	1.2511×10^{-8}	596	12,340	667	15,404
5200	structural constituent of the cytoskeleton	8.2347×10^{-8}	1.8199×10^{-5}	16	74	667	15,404
3823	antigen binding	1.5501×10^{-7}	$2.3636 imes 10^{-5}$	14	59	667	15,404
4857	enzyme inhibitor activity	1.6043×10^{-7}	2.3636×10^{-5}	33	279	667	15,404
3746	translation elongation factor activity	9.3982×10^{-7}	1.1869×10^{-4}	8	20	667	15,404
4866	endopeptidase inhibitor activity	$1.2753 imes 10^{-6}$	1.4031×10^{-4}	21	146	667	15,404
61135	endopeptidase regulator activity	$1.4285 imes 10^{-6}$	$1.4031 imes10^{-4}$	21	147	667	15,404
30414	peptidase inhibitor activity	$3.4059 imes10^{-6}$	$3.0108 imes 10^{-4}$	21	155	667	15,404
		Cell Compon	ent				
5615	extracellular space	$4.5890 imes 10^{-14}$	$2.1385 imes 10^{-11}$	78	748	680	16,336
5576	extracellular region	$1.9191 imes 10^{-13}$	$4.4715 imes 10^{-11}$	151	2022	680	16,336
44421	extracellular region part	2.5524×10^{-12}	$3.9647 imes 10^{-10}$	89	985	680	16,336
43228	non-membrane-bounded organelle	$6.4518 imes 10^{-10}$	6.0131×10^{-8}	160	2425	680	16,336
43232	intracellular non-membrane-bounded organelle	$6.4518 imes 10^{-10}$	6.0131×10^{-8}	160	2425	680	16,336
5737	cytoplasm	$2.2730 imes 10^{-9}$	1.7654×10^{-7}	393	7634	680	16,336
5856	cytoskeleton	3.1182×10^{-9}	2.0758×10^{-7}	104	1399	680	16,336
1533	cornified envelope	1.0235×10^{-8}	$5.9620 imes10^{-7}$	10	23	680	16,336
31983	vesicle lumen	$2.4577 imes 10^{-8}$	$1.2725 imes 10^{-6}$	12	38	680	16,336
31093	platelet alpha granule lumen	1.0035×10^{-7}	4.6761×10^{-6}	11	35	680	16,336

Table 6. Gene ontology enrichment analysis in Cytoscape using the BiNGO extension. A set of seventeen hub-dysregulated genes/proteins was considered in the analysis. The most enriched gene ontology terms based on the respective *p*-values are depicted.

GO-ID	Description	<i>p</i> -Value	Corr <i>p</i> -Value	x	n	x	Ν
		Biological Process					
1932	regulation of protein amino acid phosphorylation	5.3257×10^{-11}	$1.6700 imes 10^{-8}$	8	217	17	14,306
42325	regulation of phosphorylation	$5.5034 imes 10^{-11}$	1.6700×10^{-8}	10	518	17	14,306
42327	positive regulation of phosphorylation	$8.2778 imes 10^{-11}$	$1.6700 imes 10^{-8}$	7	131	17	14,306
19220	regulation of the phosphate metabolic process	$8.5947 imes 10^{-11}$	1.6700×10^{-8}	10	542	17	14,306
51174	regulation of the phosphorus metabolic process	$8.5947 imes 10^{-11}$	1.6700×10^{-8}	10	542	17	14,306
10562	positive regulation of the phosphorus metabolic process	$9.7175 imes 10^{-11}$	$1.6700 imes 10^{-8}$	7	134	17	14,306
45937	positive regulation of the phosphate metabolic process	$9.7175 imes 10^{-11}$	$1.6700 imes 10^{-8}$	7	134	17	14,306
35468	positive regulation of the signaling pathway	1.2071×10^{-10}	1.8152×10^{-8}	9	380	17	14,306
48661	positive regulation of smooth muscle cell proliferation	$1.4481 imes 10^{-10}$	$1.9356 imes 10^{-8}$	5	29	17	14,306
10647	positive regulation of cell communication	$2.5305 imes 10^{-10}$	3.0442×10^{-8}	9	413	17	14,306
		Molecular Function					
5126	cytokine receptor binding	3.1215×10^{-8}	3.3196×10^{-6}	6	186	17	15,443
5125	cytokine activity	4.3968×10^{-8}	3.3196×10^{-6}	6	197	17	15,443
8083	growth factor activity	6.2752×10^{-7}	3.1585×10^{-5}	5	160	17	15,443
70851	growth factor receptor binding	1.6672×10^{-6}	$6.2937 imes 10^{-5}$	4	82	17	15,443
5102	receptor binding	$2.3472 \text{ x} \times 10^{-6}$	$7.0887 imes 10^{-5}$	8	922	17	15,443
5515	protein binding	1.7881×10^{-5}	$4.4999 imes 10^{-4}$	17	8122	17	15,443
17022	myosin binding	2.3660×10^{-4}	5.1037×10^{-3}	2	21	17	15,443
5518	collagen binding	7.8337×10^{-4}	1.4786×10^{-2}	2	38	17	15,443
8009	chemokine activity	1.1976×10^{-3}	2.0093×10^{-2}	2	47	17	15,443
42379	chemokine receptor binding	1.4643×10^{-3}	2.1505×10^{-2}	2	52	17	15,443
		Cell Component					
5615	extracellular space	$5.3421 imes 10^{-10}$	4.8078×10^{-8}	10	747	17	16,377
44421	extracellular region part	7.8189×10^{-9}	3.5185×10^{-7}	10	985	17	16,377
31093	platelet alpha granule lumen	4.0780×10^{-8}	1.0316×10^{-6}	4	35	17	16,377
60205	cytoplasmic membrane-bounded vesicle lumen	$4.5849 imes 10^{-8}$	1.0316×10^{-6}	4	36	17	16,377
31983	vesicle lumen	$5.7381 imes10^{-8}$	1.0329×10^{-6}	4	38	17	16,377
31091	platelet alpha granule	$1.9266 imes10^{-7}$	2.8900×10^{-6}	4	51	17	16,377
9986	cell surface	$7.8170 imes 10^{-7}$	$1.0050 imes 10^{-5}$	6	340	17	16,377
30141	stored secretory granule	$9.9195 imes 10^{-7}$	1.1159×10^{-5}	5	186	17	16,377
16023	cytoplasmic membrane-bounded vesicle	2.0006×10^{-6}	2.0006×10^{-5}	7	647	17	16,377
31988	membrane-bounded vesicle	2.4025×10^{-6}	2.1623×10^{-5}	7	665	17	16,377

3.4. Multiomics Networks in MRONJ

To investigate the fundamental mechanisms of MRONJ regulation, OmicsNet tools to visualize multi-layered networks with a 3D-based layered layout were used. We detected a complex intrinsic network that was eventually reduced to a minimally connected network consisting of 1300 nodes (1289 genes/proteins and 11 miRNAs) and 7816 edges after merging the initial set of 22 miRNAs, 550 genes, and 292 proteins (see Figure 4).

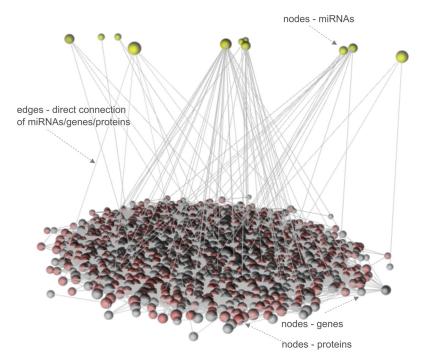


Figure 4. The multi-omics 3D layered network of MRONJ.

Next, we integrated the 17 genes shared between the MCODE cluster and cytoHubba analysis with the 22 miRNAs that were differentially expressed, ultimately producing a highly interconnected new network. This network produced an interactome of seventeen input genes, including *ALB*, *CD44*, *CXCL2*, *CXCL8*, *CXCR4*, *EEF2*, *EGF*, *IGF1*, *IL1B*, *IL6*, *ITGB1*, *JUN*, *MMP9*, *PTPRC*, *RHOA*, *TNF*, and *VEGFA*, and seven input miRNAs (hsa-mir-16-1, hsa-mir-21, hsa-mir-135, hsa-mir-186, hsa-mir-221, and hsa-mir-424) with connector genes/miRNAs using 1693 edges (Figure 5).

3.5. Pathway Enrichment Analysis

To identify the pathways involved in medication-induced osteonecrosis of the jaw, we analyzed the associated genes/proteins of the complex panel of DEGs using the ReactomeFI tool within Cytoscape. Table 7 and Figure 6a indicate that DEGs were notably enriched in the top ten signaling pathways, specifically in the Innate immunity system pathways. These pathways form the nonspecific part of immunity and include functions such as Neutrophil degranulation (R-HSA-6798695) and regulation of the complement cascade (R-HSA-6803157). Similarly, pathway enrichment analysis was performed on the significantly dysregulated Reactome signaling pathways using the set of 17 hub genes/proteins identified in MRONJ (Table 8, Figure 6b).

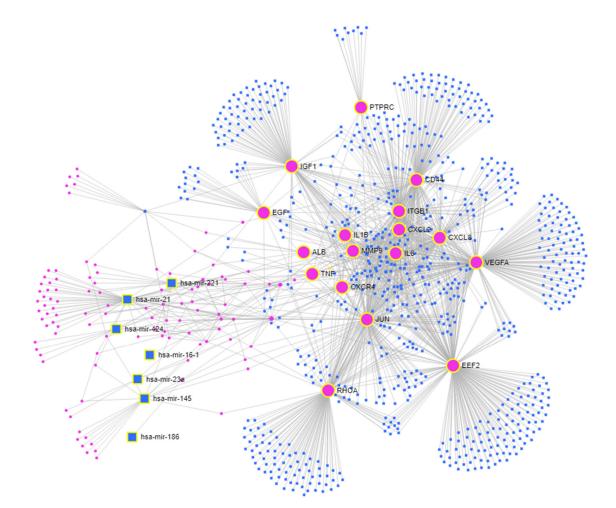


Figure 5. The multi-omics network of MRONJ. Input genes—large pink circles; input miRNAs—large blue squares; connector genes—small pink circles; connector miRNAs—small blue squares.

Table 7. Pathway enrichment analysis of the Reactome signaling pathways dysregulated in MRONJ ranked according to their *p*-values. The full set of dysregulated genes/proteins was considered in the analysis.

Reactome Pathway ID	Name	FDR	<i>p</i> -Value	Number of Proteins in Pathway	Proteins from Gene Set
R-HSA-168249	Innate immune system	$3.90 imes 10^{-13}$	$3.33 imes 10^{-16}$	1155	120
R-HSA-6798695	Neutrophil degranulation	$7.95 imes 10^{-8}$	$1.82 imes 10^{-10}$	479	58
R-HSA-977606	Regulation of complement cascade	$7.95 imes 10^{-8}$	$2.62 imes 10^{-10}$	127	27
R-HSA-2168880	Scavenging of heme from plasma	$7.95 imes 10^{-8}$	$2.72 imes 10^{-10}$	92	23
R-HSA-2173782	Binding and uptake of ligands by scavenger receptors	$1.25 imes 10^{-7}$	5.35×10^{-10}	122	26
R-HSA-114608	Platelet degranulation	$2.54 imes10^{-7}$	$1.43 imes 10^{-9}$	128	26
R-HSA-166658	Complement cascade	$2.54 imes10^{-7}$	$1.52 imes 10^{-9}$	138	27
R-HSA-5690714	CD22-mediated BCR regulation	$3.74 imes10^{-7}$	$2.64 imes 10^{-9}$	70	19
R-HSA-76005	Response to elevated platelet cytosolic Ca2+	$3.74 imes 10^{-7}$	$3.12 imes 10^{-9}$	133	26
R-HSA-2029482	Regulation of actin dynamics for phagocytic cup formation	$3.74 imes 10^{-7}$	$3.20 imes 10^{-9}$	143	27

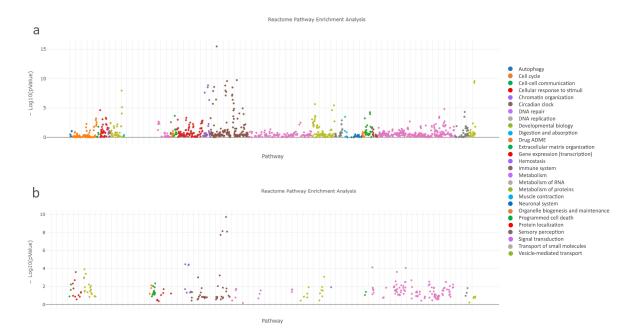


Figure 6. Pathway enrichment analysis of significantly dysregulated Reactome signaling pathways ranked according to the *p*-values; (**a**) the full set of dysregulated genes/proteins, (**b**) the set of hub genes/proteins of MRONJ.

Table 8. Pathway enrichment analysis of Reactome signaling pathways dysregulated in MRONJ ranked according to their *p*-values. The set of seventeen hub-dysregulated genes/proteins was considered in the analysis.

Reactome Pathway ID	Name	FDR	<i>p</i> -Value	Number of Proteins in Pathway	Proteins from Gene Set
R-HSA-6785807	Interleukin-4 and interleukin-13 signaling	$4.49 imes 10^{-8}$	$\begin{array}{c} 1.83 \times \\ 10^{-10} \end{array}$	112	7
R-HSA-449147	Signaling by interleukins	$6.39 imes10^{-7}$	$6.95 imes 10^{-9}$	466	9
R-HSA-6783783	Interleukin-10 signaling	$6.39 imes10^{-7}$	$7.88 imes 10^{-9}$	47	5
R-HSA-1280215	Cytokine signaling in the immune system	$1.12 imes 10^{-6}$	$1.83 imes 10^{-8}$	730	10
R-HSA-76002	Platelet activation, signaling, and aggregation	$1.47 imes 10^{-3}$	3.37×10^{-5}	260	5

To map the signaling pathways of MRONJ and to identify potential Reactome molecular pathway targets of miRNAs, the associated miRNAs (hsa-mir-16-1, hsa-mir-21, hsa-mir-23a, hsa-mir-145, hsa-mir-186, hsa-mir-221, and hsa-mir-424), obtained with the OmicsNet tools (Figure 5), underwent an in silico analysis with the DIANA Tools mirPath v.4 database. MiRNA-centric analysis of hsa-mir-16-1, hsa-mir-21, hsa-mir-23a, hsa-mir-145, hsa-mir-186, hsa-mir-221, and hsa-mir-221, and hsa-mir-186, hsa-mir-221, and hsa-mir-424 was carried out with the help of the TarBase v8.0 database and miTarBase2022 as a secondary target source, searching for the direct miRNA target genes. The union of Reactome pathways option was used in the analysis. The results of the analysis are shown in Table 9. The Reactome pathways Interleukin-4 and interleukin-13 signaling (R-HSA-6785807) and signaling by interleukins (R-HSA-449147) show the most comprehensive association of miRNAs with gene targets consisting of *IL1B*, *VEGFA*, and *CXCL8*.

Table 9. Pathway enrichment analysis of significantly dysregulated Reactome signaling pathways conducted on the set of seven miRNAs (hsa-mir-16-1, hsa-mir-21, hsa-mir-23a, hsa-mir-145, hsa-mir-186, hsa-mir-221, and hsa-mir-424) of MRONJ based on the *p*-value. Direct target genes of the set of seventeen input genes (*ALB, CD44, CXCL2, CXCL8, CXCR4, EEF2, EGF, IGF1, IL1B, IL6, ITGB1, JUN, MMP9, PTPRC, RHOA, TNF,* and *VEGFA*) are displayed.

Reactome Pathway ID	Name	Merged <i>p</i> -Value	Merged FDR	Term Genes	miRNAs	Direct Targe Genes
					hsa-miR-21-5p	IL1B, VEGFA
	Interleukin-4 and interleukin-13 signaling				hsa-miR-23a-3p	CXCL8
R-HSA-6785807		$5.9652 imes 10^{-33}$	$1.2229 imes 10^{-30}$	122	hsa-miR-145-5p	VEGFA
					hsa-miR-186-5p	VEGFA
					hsa-miR-16-1-3p	VEGFA
					hsa-miR-21-5p	IL1B, VEGFA
					hsa-miR-23a-3p	CXCL8
R-HSA-449147	Signaling by interleukins	$7.6257 imes 10^{-24}$	$7.8163 imes 10^{-22}$	512	hsa-miR-145-5p	VEGFA
					hsa-miR-186-5p	VEGFA
					hsa-miR-16-1-3p	VEGFA
R-HSA-1643685	D:	0.1000 10-15	(1002 10-14	1010	hsa-miR-21-5p	IL1B, VEGFA
K115/1 1045005	Diseases	2.1882×10^{-15}	$6.4083 imes 10^{-14}$	1819	hsa-miR-145-5p	VEGFA
					hsa-miR-21-5p	IL1B, VEGFA
R-HSA-1280215	Cytokine signaling in the immune system	1.3377×10^{-13}	$2.1094 imes 10^{-12}$	10501	hsa-miR-145-5p	CD44, VEGE
					hsa-miR-16-1-3p	VEGFA
R-HSA-74160	Gene expression (transcription)	$2.3944 imes 10^{-13}$	$3.2723 imes 10^{-12}$	1661	hsa-miR-21-5p	VEGFA
D LIC A 000/024	Signaling by receptor tyrosine kinases	4 505 (2 10-13	< 14441 10 ⁻¹²	529	hsa-miR-21-5p	VEGFA
R-HSA-9006934	Signaling by receptor tyrosine kinases	4.79563×10^{-13}	6.14441×10^{-12}	528	hsa-miR-145-5p	VEGFA
D LICA 010404	Generic transcription pathway	0.7(14010-13	1.05653×10^{-11}	1070	hsa-miR-21-5p	VEGFA
R-HSA-212436	Generic transcription pattway	8.76149×10^{-13}	1.05653 × 10	1372	hsa-miR-145-5p	VEGFA
R-HSA-195258	RHO GTPase effectors	$3.07483 imes 10^{-7}$	$1.40076 imes 10^{-6}$	333	hsa-miR-186-3p	ITGB1
D LICA 8967010	TFAP2 (AP-2) family regulates the	0.00607 10-6	1 11204 10-5	15	hsa-miR-21-5p	VEGFA
R-HSA-8866910	transcription of growth factors and their receptors	2.98627×10^{-6}	1.11306×10^{-5}	15	hsa-miR-145-5p	VEGFA
R-HSA-168256	Immune system	1.30146×10^{-5}	$3.75774 imes 10^{-5}$	2755	hsa-miR-21-5p	IL1B, VEGFA
D LICA 1/0500	Cional transduction	1.05050 10-5	0.01000 10-5	2120	hsa-miR-21-5p	VEGFA
R-HSA-162582	Signal transduction	1.37359×10^{-5}	3.91092×10^{-5}	3138	hsa-miR-145-5p	VEGFA
R-HSA-8864260	Transcriptional regulation by the AP-2 (TFAP2) family of transcription factors	2.69119×10^{-5}	$6.89618 imes 10^{-5}$	38	hsa-miR-145-5p	VEGFA
R-HSA-446652	Interleukin-1 family signaling	$4.46946 imes 10^{-4}$	$5.51562 imes 10^{-4}$	165	hsa-miR-21-5p	IL1B
R-HSA-6783783	Interleukin-10 signaling	$4.49321 imes 10^{-4}$	$5.51562 imes 10^{-4}$	59	hsa-miR-21-5p	IL1B
R-HSA-1474244	Extracellular matrix organization	$8.44467 imes 10^{-4}$	$9.15957 imes 10^{-4}$	318	hsa-miR-145-5p	CD44
R-HSA-5660668	CLEC7A/inflammasome pathway	$1.724928 imes 10^{-3}$	1.724928×10^{-3}	6	hsa-miR-21-5p	IL1B

4. Discussion

To explore the pathological mechanisms of MRONJ, we utilized gene profiling datasets, proteins, and miRNAs. A variety of analytical strategies were employed to investigate the molecular mechanisms underlying MRONJ, including PPI network analysis, GO enrichment, and Reactome pathway enrichment analysis.

Antiresorptive therapy, including bisphosphonates, denosumab, and angiogenesis inhibitors, may trigger MRONJ, which can affect both the maxilla and mandible [44,45]. Several hypotheses have been developed regarding the pathophysiology of MRONJ: (1) suppression of bone resorption; (2) inflammation and oral microbial infection; (3) inhibition of angiogenesis and anti-lymphangiogenesis; (4) dysfunction of innate or acquired immunity (T and B cells, macrophages, DCs, and natural killer cells); and (5) soft tissue toxicity are all potential adverse effects [46,47].

Pathway enrichment analysis is a valuable tool for gaining a mechanistic comprehension of the intricate gene, miRNA, and protein inventories resulting from omics experiments. It assists in the interpretation of biomedical data to reveal the molecular basis of disease [48]. Our analysis identified immune dysfunction-related pathways associated with MRONJ as the main reason for developing and progressing osteonecrosis. When analyzing the entire pool of genes and proteins, it is evident that the Reactome signaling pathways that are significantly dysregulated are primarily dominated by the Innate immune system and Neutrophil degranulation pathways, as indicated by the *p*-values. Reducing the set of hub genes to 17, which includes *ALB*, *CD44*, *CXCL2*, *CXCL8*, *CXCR4*, *EEF2*, *EGF*, *IGF1*, *IL1B*, *IL6*, *ITGB1*, *JUN*, *MMP9*, *PTPRC*, *RHOA*, *TNF*, and *VEGFA*, resulted in the dominance of cy-tokine signaling pathways in the immune system. Moreover, simultaneously analyzing hub genes with miRNAs (hsa-mir-16-1, hsa-mir-21, hsa-mir-23a, hsa-mir-145, hsa-mir-186, hsa-mir-221, and hsa-mir-424) using network-based multiomics analysis revealed dysregulated pathways in the immune system's cytokine signaling, specifically signaling by interleukins.

There is increasing evidence that inflammatory osteoimmunology is critical to the development of osteonecrosis [49]. Cytokines that regulate inflammatory responses contribute to the onset and progression of osteonecrosis [17,50,51]. In individuals with osteonecrosis, necrotic cells produce cytokines that attract inflammatory cells, triggering both local and systemic immune responses [17,52,53].

The function of immune cells and bone-forming cells, especially osteoblasts and osteoclasts, is regulated by cytokines, inflammatory chemokines, and growth factors [54]. Research confirms the involvement of cytokine networks in osteoclast differentiation and regulation. Cytokines, such as tumor necrosis factor-alpha, interleukins 1, -6, -7, -8, -11, -15, -17, -23, and -34, facilitate the process of osteoclast differentiation. By contrast, anti-osteoclastogenic cytokines, namely interferons alpha, beta, and gamma and interleukins 3, -4, -10, -12, -27, and -33, suppress osteoclasts. [55].

The pathogenesis of inflammatory bone disease is significantly influenced by T cells and B cells [56]. There have been intense discussions regarding new roles for B cells and a potential role for peripheral blood $\gamma\delta$ T cells [57,58]. $\Gamma\delta$ T cells are innate lymphocytes with a crucial role in regulating immune homeostasis [59]. Kalyan et al. investigated the potential predictive role of peripheral blood $\gamma\delta$ T cells in osteonecrosis of the jaw. The authors propose that the loss of $\gamma\delta$ T cells caused by bisphosphonates may be involved in the development of osteonecrosis [57]. Moreover, the proliferation of macrophages and $\gamma\delta$ T cells promotes inflammation in zoledronic acid-induced jaw necrosis, as the authors of the study [60] concluded.

The understanding of biological systems is facilitated by the objective study of PPIs. An effective approach to characterizing system-wide PPIs is the use of PPI networks. These networks are constructed from pairwise protein interactions and serve as an efficient tool for describing PPI landscapes [61]. To investigate protein functions and biological processes based on predicted PPIs and to gain new insights into diseases, the DEG PPI network was analyzed in this study. For osteonecrosis of the jaw, 17 hub genes with aberrant expression were selected. They included *ALB*, *CD44*, *CXCL2*, *CXCL8*, *CXCR4*, *EEF2*, *EGF*, *IGF1*, *IL1B*, *IL6*, *ITGB1*, *JUN*, *MMP9*, *PTPRC*, *RHOA*, *TNF*, and *VEGFA*.

Various signaling molecules, such as *VEGFA*, *EGF*, *MMP9*, and *TNF*, contribute to angiogenesis by stimulating the proliferation and migration of vascular endothelial cells [62]. *VEGFA* is a highly potent pro-angiogenic factor that plays a critical role in the healing of microvascular wounds associated with bisphosphonate administration [63]. A statistically significant increase in *VEGFA* gene expression was also demonstrated in response to zoledronic acid [64]. The physiological processes of angiogenesis and vascular remodeling involve the regulation of non-coding RNAs, specifically miRNA-based regulation (as noted by reference [65]). Objective evaluation of these processes is necessary for accurate understanding.

MicroRNAs are small endogenous RNA molecules (\sim 22 nt) that were recently discovered. Disorders such as cancer or heart disease have demonstrated the diagnostic potential of circulating miRNAs [66,67]. MiRNA-mediated RNA interference, a unique

mechanism that binds miRNAs to different direct targets, controls both post-transcriptional gene expression and protein expression [68].

Our OmicsNet network analysis generated an interactome of input genes including *ALB*, *CD44*, *CXCL2*, *CXCL8*, *CXCR4*, *EEF2*, *EGF*, *IGF1*, *IL1B*, *IL6*, *ITGB1*, *JUN*, *MMP9*, *PTPRC*, *RHOA*, *TNF*, and *VEGFA* with input miRNAs comprising hsa-mir-16-1, hsa-mir-21, hsa-mir-23a, hsa-mir-145, hsa-mir-186, hsa-mir-221, and hsa-mir-424. Further analysis focused on miRNA identified the most extensively linked miRNAs targeting *IL1B*, *VEGFA*, *CXCL8*, and *CD44* directly (Table 9).

MiR-145 targets mainly the interleukins [38] and is also implicated in M2 macrophage polarization [69]. Silencing miR-145 leads to the advancement of femoral head regeneration by upregulating *VEGFA* [70].

The PPI network revealed that *CXCL8* [71] is among the factors related to the progression of osteonecrosis of the jaw, and our analysis identified it as a direct target of miR-23a. Earlier studies reported a significant reduction of miR-23a-3p during osteogenic differentiation of human mesenchymal stem cells of bone marrow origin [72]. Furthermore, inhibition of miR-23a in a rat model resulted in a lower incidence of osteonecrosis [73].

Serum microRNAs, including miR-21, miR-23a, and miR-145, were observed to be dysregulated in BRONJ rats [37]. Furthermore, candidate microRNA expressions were confirmed in human samples. During the progression of BRONJ, there was an upregulation of circulating miR-21, which corresponds to the alteration of miR-21 in proosteoclastogenesis [37].

5. Conclusions

The emergence of high-throughput platforms for the comprehensive analysis of genes, proteins, and other biological molecules has afforded an exceptional capability for the recognition of novel, valid signatures of disease-related processes.

In conclusion, our systematic review study indicates specific alterations in proteins, genes, and microRNAs and thus unravels novel insights into the molecular mechanism behind the MRONJ disease. The identified dysregulated genes in MRONJ are mostly linked to the regulation of immune system processes and the immune response of the organism. These dysregulated genes significantly enrich pathways related to the Innate immunity system, a crucial component of the nonspecific part of immunity. Particularly important are the 17 hub genes, which exert dominance in the cytokine signaling pathways within the immune system. Additionally, the interaction network between these hub genes and dysregulated miRNAs uncovered pathways associated with the cytokine signaling in the immune system, particularly the signaling by interleukins pathway. Subsequent miRNA analysis showed a set of highly connected miRNAs with direct targeting of multiple genes such as *IL1B*, *VEGFA*, *CXCL8*, and *CD44*.

This study has potential limitations that should be noted. There was considerable heterogeneity observed between studies, which may impact the interpretation of the results. Factors such as patient selection variability, differences in the origins and causes of MRONJ, and variations in the material and methodology used could contribute to this heterogeneity. To ensure more reliable results, it would be beneficial to establish strict inclusion/exclusion criteria based on the disease state and MRONJ treatment in future studies. Additionally, our objective was to present a comprehensive overview of the pathophysiological processes of MRONJ in humans using multi-omics techniques. Expanding the research to encompass other types of non-coding RNAs could prove helpful in filling this gap. Furthermore, there is a shortage of data for analysis due to the limited amount of metabolomics research.

Despite the limitations of this study, the panel of proteins, genes, and microRNAs presented, along with their associated pathways, constitutes a significant advancement toward comprehending the intricate cause of MRONJ.

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