



Prediction of medication-related osteonecrosis of the jaw (MRONJ) using automated machine learning in patients with osteoporosis associated with dental extraction and implantation: a retrospective study

Da Woon Kwack, Sung Min Park

Department of Oral and Maxillofacial Surgery, College of Dentistry, Dankook University, Cheonan, Korea

Abstract (J Korean Assoc Oral Maxillofac Surg 2023;49:135-141)

Objectives: This study aimed to develop and validate machine learning (ML) models using H₂O-AutoML, an automated ML program, for predicting medication-related osteonecrosis of the jaw (MRONJ) in patients with osteoporosis undergoing tooth extraction or implantation.

Patients and Methods: We conducted a retrospective chart review of 340 patients who visited Dankook University Dental Hospital between January 2019 and June 2022 who met the following inclusion criteria: female, age ≥ 55 years, osteoporosis treated with antiresorptive therapy, and recent dental extraction or implantation. We considered medication administration and duration, demographics, and systemic factors (age and medical history). Local factors, such as surgical method, number of operated teeth, and operation area, were also included. Six algorithms were used to generate the MRONJ prediction model.

Results: Gradient boosting demonstrated the best diagnostic accuracy, with an area under the receiver operating characteristic curve (AUC) of 0.8283. Validation with the test dataset yielded a stable AUC of 0.7526. Variable importance analysis identified duration of medication as the most important variable, followed by age, number of teeth operated, and operation site.

Conclusion: ML models can help predict MRONJ occurrence in patients with osteoporosis undergoing tooth extraction or implantation based on questionnaire data acquired at the first visit.

Key words: Medication-related osteonecrosis of the jaw, Bisphosphonate, Osteoporosis, Machine learning, Algorithm

[paper submitted 2023. 4. 18 / revised 2023. 5. 27 / accepted 2023. 6. 5]

I. Introduction

Recently, the medical field has begun developing machine learning (ML) models that can aid doctors in their decision-making regarding diagnosis, prognosis, and management of multifactorial diseases¹⁻³. The dentistry field is using ML to improve patient care^{1,3-5}. For example, a deep-learning-based convolutional neural network algorithm was used to diagnose dental caries in periapical radiographs⁴, and neural network ML considering several cephalometric variables was used to diagnose the need for extractions in the field of orthodontics⁵.

Additionally, various studies using artificial intelligence (AI) have been published in the fields of oncology and periodontics³.

As a subtype of AI, ML extrapolates results based on data and algorithms. The algorithms are trained using data with previously determined input-output relationships to infer output data with high predictability for new input data^{6,7}. As the demand for ML increases, automated ML programs that can be used effectively by non-experts are available for free or a nominal fee (e.g., DataRobot, H2O-AutoML, H2O-Driverless AI product, Darwin, Google Cloud AutoML, TPOT)⁸.

Medication-related osteonecrosis of the jaw (MRONJ) is a multifactorial disease. Although medication is one of the main risk factors, local, demographic, and systemic factors also affect its occurrence⁹. Medications that can influence MRONJ occurrence include antiresorptive therapies such as bisphosphonates (BPs), denosumab (DMB), and romosozumab, which are administered orally or parenterally to manage osteoporosis, cancer-related bone metastases, and metabolic

Sung Min Park

Department of Oral and Maxillofacial Surgery, College of Dentistry, Dankook University, 119 Dandae-ro, Cheonan 31116, Korea
TEL: +82-41-550-1996

E-mail: 12201114@dankook.ac.kr

ORCID: <https://orcid.org/0000-0001-8573-9071>

© This is an open-access article distributed under the terms of the Creative Commons Attribution Non-Commercial License (<http://creativecommons.org/licenses/by-nc/4.0/>), which permits unrestricted non-commercial use, distribution, and reproduction in any medium, provided the original work is properly cited.

Copyright © 2023 The Korean Association of Oral and Maxillofacial Surgeons.

bone diseases. MRONJ risk in patients with osteoporosis ranges from 0.02% to 0.3%, which is not higher than that in patients taking drugs for other diseases⁹. However, MRONJ can reduce quality of life, and it is even life-threatening in some cases¹⁰; hence, early diagnosis and treatment are critical. It is important to not only accurately diagnose MRONJ in a timely manner, but also to predict potential occurrence and provide appropriate follow-up during the perioperative surgical period.

The aim of this study was to develop and validate ML models that can predict MRONJ occurrence with information obtained from a questionnaire. We used an automated ML program, H2O-AutoML, to develop models for patients with osteoporosis who had undergone tooth extraction or implantation surgery.

II. Patients and Methods

1. Patients

This study included female patients aged 55 years or older who visited the Department of Oral and Maxillofacial Surgery at Dankook University Dental Hospital between January 2019 and June 2022. Medical and dental history surveys were conducted using a questionnaire presented in Table 1, with a focus on the duration and administration of antiresorptive medications for patients with osteoporosis. The inclusion criteria were as follows: female, age ≥ 55 years, osteoporosis currently or previously treated with antiresorptive therapy (e.g., BPs or DMB), and recent dental extraction or implantation (e.g., patients who were referred after surgery at the local center and patients who had undergone surgery at this hospi-

tal). The exclusion criteria were radiotherapy history for head and neck cancer and incomplete medical and dental history records. A total of 340 patients met the inclusion criteria and was included in the study. We considered medication, demographic, systemic, and dental factors. Medication factors included duration and administration of antiresorptive medications (oral [PO], intravenous [IV], or both). Demographic and systemic factors included age and history of hypertension, hyperlipidemia, diabetes mellitus, heart disease, thyroid disease, kidney disease, liver disease, rheumatoid arthritis, cancer, or dementia. Additionally, any history of anticoagulant medication was considered. We also included local factors, such as the method of operation (extraction or implantation), number of operated teeth, and operation area (maxilla, mandible, or both).

MRONJ occurrence was evaluated during post-surgical follow-up after at least six months and was diagnosed according to criteria defined in the position paper of the American Association of Oral and Maxillofacial Surgeons⁹:

- (1) Current or previous treatment with antiresorptive therapy alone or in combination with immune modulators or antiangiogenic medications.
- (2) Exposed bone or bone that can be probed through an intraoral or extraoral fistula(e) in the maxillofacial region that has persisted for more than eight weeks.
- (3) No history of radiation therapy to the jaws or metastatic disease to the jaws.

Because this study aimed to predict only MRONJ occurrence, the staging system (stages 0-3), which is helpful for doctors in determining a patient's prognosis and outcome, was not considered. Instead, the output value was dichotomized as presence or absence of MRONJ.

Table 1. Questionnaire provided to patients

Questionnaire items	Yes	No	Comment
1. Have you ever been hospitalized or undergone surgery?			Site:
2. Are you currently being treated by a doctor?			
3. What kind of medications are you currently taking?			Drug:
4. Have you ever had side effects from injections or medicines?			
5. Are you bleeding excessively or are you taking anticoagulants?			
6. Have you ever had tuberculosis or a sexually transmitted infection?			
7. Do you have hypertension?			Blood pressure: /
8. Do you have hepatitis or jaundice?			
9. Do you have heart disease?			
10. Do you have kidney disease?			
11. Do you have diabetes mellitus?			Blood sugar test
12. Are you taking osteoporosis medications?			Intravenous / oral / both
			Duration of medication:
13. Do you have trouble breathing or have stomach problems?			
14. Are you pregnant?			
15. Do you have dementia?			
16. Do you have thyroid disease?			

Da Woon Kwack et al: Prediction of medication-related osteonecrosis of the jaw (MRONJ) using automated machine learning in patients with osteoporosis associated with dental extraction and implantation: a retrospective study. J Korean Assoc Oral Maxillofac Surg 2023

Participants were divided into a control group (n=210) and MRONJ group (n=130). A comparison of the parameters between the control and MRONJ groups is presented in Table 2.

2. Automated ML

The H2O-AutoML program was used to perform hyper-parameter optimization and train the prediction models². The data were randomly divided into a training set (71.5%, n=243) and a testing set (28.5%, n=97). The training set was used to generate the prediction model, which was estimated through fivefold cross-validation. The testing set was used to estimate the accuracy of the models. The flow diagram of the proposed work is shown in Fig. 1.

ML can be classified into two categories: supervised and non-supervised. Supervised learning uses known labeled datasets to train an algorithm. Non-supervised learning uses unlabeled datasets¹¹. Among these, supervised learning, particularly classification, is suitable for disease prediction because it involves discrete output variables. Thus, many studies have used this method^{1-3,11,12}. We used the six algorithms of the generalized linear model (GLM), distributed random forest

Table 2. Parameter comparison

	MRONJ		P-value
	No (control group, n=210)	Yes (MRONJ group, n=130)	
Age (yr)	73.1±9.0	77.2±7.1	<0.001 ¹
Medication administration			
Oral	82 (39.0)	70 (53.8)	<0.001 ²
Intravenous	98 (46.7)	29 (22.3)	
Both	30 (14.3)	31 (23.8)	
Duration of medication (yr)	3.4±4.6	5.8±5.1	<0.001 ¹
Medical history			
Hypertension	104 (49.5)	85 (65.4)	0.004 ²
Hyperlipidemia	47 (22.4)	27 (20.8)	0.726
Anticoagulant	29 (13.8)	14 (10.8)	0.412
Diabetes mellitus	41 (19.5)	29 (22.3)	0.552
Heart disease	23 (11.0)	24 (18.5)	0.051
Thyroid disease	10 (4.8)	7 (5.4)	0.798
Liver disease	4 (1.9)	0 (0)	0.113
Kidney disease	10 (4.8)	4 (3.1)	0.447
Rheumatoid arthritis	4 (1.9)	7 (5.4)	0.078
Cancer	10 (4.8)	12 (9.2)	0.104
Dementia	11 (5.2)	2 (1.5)	0.084
Method of operation			
Extraction	180 (85.7)	99 (76.2)	0.026 ²
Implantation	30 (14.3)	31 (23.8)	
No. of operated teeth	2.3±1.6	1.7±1.0	<0.001 ¹
Operation area			
Maxilla	93 (44.3)	35 (26.9)	<0.001 ²
Mandible	87 (41.4)	90 (69.2)	
Both	30 (14.3)	5 (3.8)	

(MRONJ: medication-related osteonecrosis of the jaw)

Statistical analyses by ¹Student's *t*-test and ²chi-square test.

Values are presented as mean±standard deviations or number (%).

Da Woon Kwack et al: Prediction of medication-related osteonecrosis of the jaw (MRONJ) using automated machine learning in patients with osteoporosis associated with dental extraction and implantation: a retrospective study. *J Korean Assoc Oral Maxillofac Surg* 2023

(DRF), gradient boosting machine (GBM), stacked ensemble, extreme gradient boosting (XGBoost), and deep learning to generate our prediction model.

The GLM is a regression model that estimates outcomes following exponential distributions. Although this model is easy to use and interpret, it cannot predict complex problems¹³. In contrast to using a single model like GLM or decision tree (DT) algorithms, the ensemble method uses numerous models to increase forecast accuracy¹². The three combination strategies of the ensemble method are bagging, boosting, and stacking.

Bagging creates multiple classifications, trains them with different training data (bootstrapping), and then votes on different predictions for the same test data (aggregating)^{6,14}. In

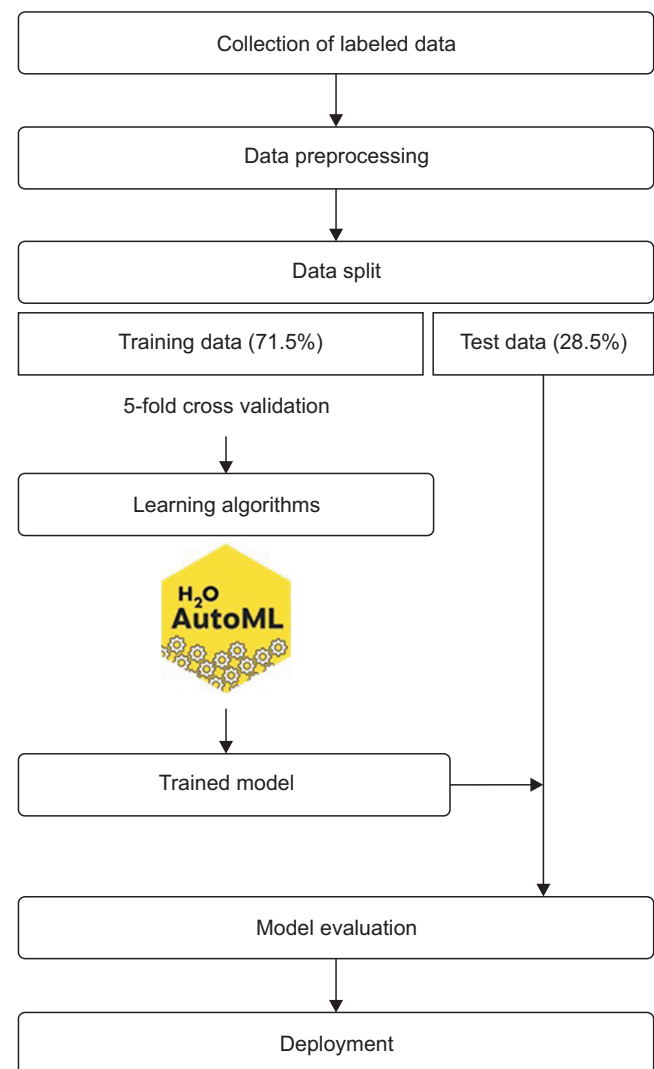


Fig. 1. Flow diagram of the proposed work.

Da Woon Kwack et al: Prediction of medication-related osteonecrosis of the jaw (MRONJ) using automated machine learning in patients with osteoporosis associated with dental extraction and implantation: a retrospective study. *J Korean Assoc Oral Maxillofac Surg* 2023

this study, we used the DRF method to bag the DT algorithm.

While bagging learns several classifiers based on different algorithms in parallel, boosting trains classifiers of the same algorithm and learns from the errors of previous iterations by increasing the importance of incorrectly predicted training instances in future iterations⁶. GBM and XGBoost are examples of such boosting ensembles.

Stacking involves two phases. The first phase trains the dataset base using different algorithms, and the output of this phase is used as a new dataset for the second phase, called meta-learning, which provides the final output^{6,14}.

Deep learning is a subtype of ML that utilizes a representation-learning method to enable a machine to find the necessary representations required for detection or classification from raw, abstract data such as pictures. This process is not designed by human engineers but learns itself from data using a common process¹⁵.

3. Statistical analyses

Statistical analyses were conducted on continuous and categorical variables between the control and MRONJ groups using Student's *t*-test and the chi-square test, respectively, as appropriate, following the Kolmogorov–Smirnov test. *P*-values <0.05 were considered statistically significant. The area under the receiver operating characteristic (ROC) curves (AUC) and F1 score were used for comparison between ML models. Statistical analyses were performed using IBM SPSS Statistics (ver. 29.0.0.0; IBM).

III. Results

Of the 29 models developed for predicting MRONJ, 20, including 14 GBM, 4 stacked ensemble models, 1 DRF model, and 1 XGBoost model, showed very good diagnostic accuracy with an AUC ≥ 0.8 . The GBM model had the highest AUC value. The training and test dataset metrics for the model are presented in Tables 3 and 4, respectively. This

Table 3. Training dataset metrics

MRONJ	Predicted		Total
	Negative	Positive	
Actual Negative	112	45	157
Actual Positive	14	72	86
Total	126	117	243

(MRONJ: medication-related osteonecrosis of the jaw)

Da Woon Kwack et al: Prediction of medication-related osteonecrosis of the jaw (MRONJ) using automated machine learning in patients with osteoporosis associated with dental extraction and implantation: a retrospective study. J Korean Assoc Oral Maxillofac Surg 2023

Table 4. Test dataset metrics

MRONJ	Predicted		Total
	Negative	Positive	
Actual Negative	28	25	53
Actual Positive	5	39	44
Total	33	64	97

(MRONJ: medication-related osteonecrosis of the jaw)

Da Woon Kwack et al: Prediction of medication-related osteonecrosis of the jaw (MRONJ) using automated machine learning in patients with osteoporosis associated with dental extraction and implantation: a retrospective study. J Korean Assoc Oral Maxillofac Surg 2023

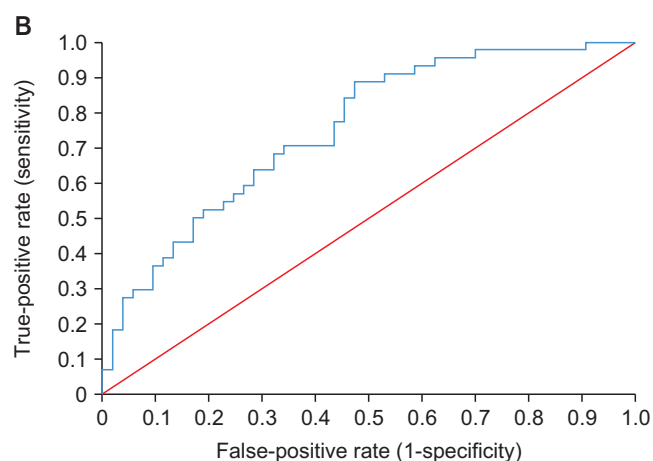
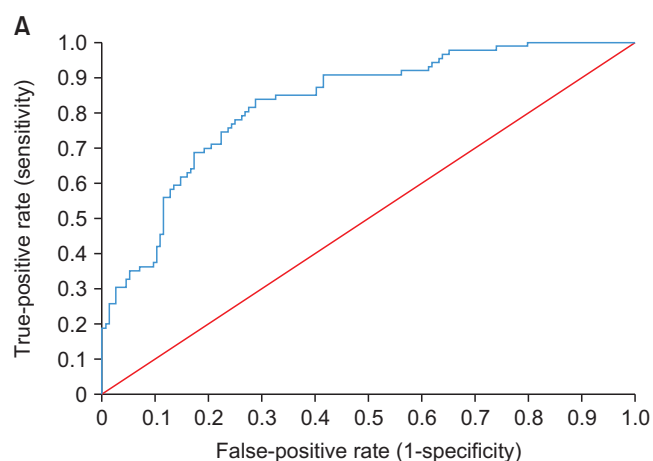


Fig. 2. Area under the receiver operating characteristic curves (AUCs) of best performed gradient boosting machine model. A. Training dataset (AUC=0.8283). B. Test dataset (AUC=0.7526).

Da Woon Kwack et al: Prediction of medication-related osteonecrosis of the jaw (MRONJ) using automated machine learning in patients with osteoporosis associated with dental extraction and implantation: a retrospective study. J Korean Assoc Oral Maxillofac Surg 2023

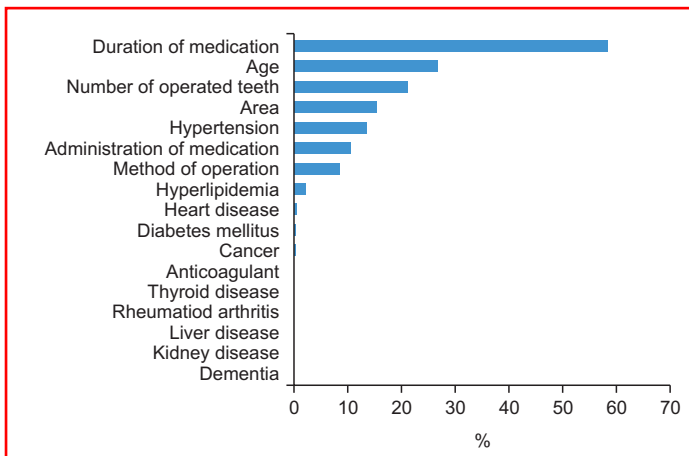


Fig. 3. Variable importance.

Da Woon Kwack et al: Prediction of medication-related osteonecrosis of the jaw (MRONJ) using automated machine learning in patients with osteoporosis associated with dental extraction and implantation: a retrospective study. *J Korean Assoc Oral Maxillofac Surg* 2023

model comprised 42 trees with a maximum depth of 6 and showed an AUC of 0.8283 on the training dataset.(Fig. 2. A) At the maximum F1 threshold, the sensitivity and specificity were 83.7% and 71.3%, respectively. For the validation dataset, the model showed a stable AUC of 0.7526.(Fig. 2. B) At the maximum F1 threshold, the sensitivity and specificity were 88.6% and 52.8%, respectively.

Variable importance analysis, shown in Fig. 3, revealed that the duration of medication was the most important variable, followed by age, number of operated teeth, operation area, hypertension, and administration of medication.

IV. Discussion

Korea has a rapidly aging population; thus, a large proportion of the population has potential for osteoporosis¹⁶. BPs are frequently prescribed for female patients diagnosed with osteoporosis after menopause because they are non-invasive and easy to take orally. However, with age, the need for tooth extraction and implantation increases. Unfortunately, life-threatening MRONJ could occur after a seemingly simple dental treatment, and it cannot be easily predicted due to limited available information. Although biomarkers such as the C-terminal telopeptide of type-I collagen obtained through blood tests have been used in MRONJ clinical decision-making, their use remains controversial. While a drug holiday can be considered to reduce MRONJ risk in patients undergoing tooth extraction or implant surgery, it is not always feasible due to potential side effects such as skeletal-related events, fragility fractures, increased rebound bone resorption, or in emergency surgery situations. Recommendations for the du-

ration of a drug holiday also remain controversial⁹. Recently, parathyroid hormone therapy, which stimulates new bone formation by osteoblasts, has been highlighted as an alternative drug. However, its use is limited to 24-month regimens due to potential side effects, and the subcutaneous daily injections required could be challenging for elderly patients¹⁷.

This study demonstrates the effectiveness of ML models for predicting MRONJ occurrence using only questionnaire-based information in patients with osteoporosis who have undergone tooth extractions or implant surgery. We objectively and consistently collected information from patients using BPs who visited our hospital in 2017 using the same questionnaire, which allowed for consistent and objective data collection. One advantage of our models is that they were developed based only on the questionnaire at the first visit, without the need for imaging or laboratory tests. These features enable surgeons to make clinical decisions and educate patients with more accurate information.

Most models in this study showed good diagnostic accuracy, with an AUC ≥ 0.8 . For the validation dataset, the model showed an AUC of 0.7526, which is slightly lower but still a good predictive value. Additionally, sensitivity increased from 83.7% to 88.6%, and specificity decreased from 71.3% to 52.8%. With reference to our model, sensitivity can be seen as a more important indicator than specificity in that our model is intended to predict the possibility of MRONJ occurrence for educating patients and for follow-up.

We also compared variable importance.(Fig. 3) The most important variable was duration of medication use, followed by age. In this study, the mean ages of the control and MRONJ groups were 73.1 and 77.2 years, respectively, and the mean durations of medication use were 3.4 and 5.8 years. Both variables were significantly higher in the MRONJ group than in the control group. Many studies have also found significant increases in incidence rates, from 0.04% to 0.21%, in patients with osteoporosis who had been using BPs for more than four years^{9,18,19}. As osteoporosis in women generally occurs after menopause, age and duration of medication are closely related.

The second most significant variable was the number of teeth that underwent operation. Some studies have found that the MRONJ incidence rate was significantly higher in cases of multiple-tooth extractions than in single-tooth extractions²⁰. However, our analysis showed that the average number of teeth extracted in the MRONJ group was lower than that in the control group, which contradicts previously reported results. In response to these results, other studies

have reported that tooth extraction does not act as a local factor alone, but it is associated with the presence or absence of inflammatory dental disease^{9,19,20}. Therefore, it is necessary to also consider the presence or absence of periodontal disease.

Several studies have reported on the surgical area associated with MRONJ, showing that MRONJ prevalence is two to three times higher in the mandible than in the maxilla^{9,19,20}. The results of our study are consistent with previous studies, in that we found 27% and 69% prevalence in the maxilla and mandible, respectively.

The most important variable in patient medical history was hypertension. Despite several studies showing that hypertension, diabetes, and anemia significantly enhance the risk of MRONJ incidence^{9,21,22}, in this study, hypertension was the most significant variable. Although the risk of overfitting in this algorithm cannot be completely ruled out because the variable distribution is not uniform, the association between MRONJ and hypertension has not been well studied, and further research is needed.

This study has several limitations. First, we collected the data from a single institution, and to include a large number of MRONJ cases (n=130, 38.2%) compared with the low incidence, we excluded drug holiday or steroid therapy from variables, which are also correlated with increased MRONJ risk⁹. Nevertheless, none of the variables showed a uniform distribution between the MRONJ and control groups. This may have caused overfitting or underfitting of the ML models. Second, the model was designed using only medical records based on patient questionnaires. Nonetheless, it is meaningful in that it created a model with a high AUC value. However, the accuracy of the model can be improved by considering variables from other records, including objective laboratory, genetics, or imaging data³. Finally, because the ML algorithm infers results by considering numerous factors at once, it is impossible to produce a single exact value for each variable.

A predictive model may help clinicians explain the possibility of MRONJ in greater depth to patients and help them actively cope with future complications before tooth extraction or implantation. Moreover, in the future, AI-based programs may support high-quality patient care and facilitate advanced decision tools. In future research, we anticipate developing a computerized system that can collect, computerize, and share big data from multiple institutions, including radiograph and laboratory data, in addition to medical history data.

V. Conclusion

We demonstrated that ML models can be used to predict the likelihood of MRONJ occurrence in patients with osteoporosis who have undergone tooth extraction or implantation surgery using information gathered from a questionnaire at their initial visit. Our findings suggest that the duration of medication use is the most significant variable associated with MRONJ risk, followed by age, number of operated teeth, and operation area.

ORCID

Da Woon Kwack, <https://orcid.org/0009-0006-6241-7065>

Sung Min Park, <https://orcid.org/0000-0001-8573-9071>

Authors' Contributions

S.M.P. and D.W.K. participated in conceptualization and writing – review and editing. D.W.K. participated in methodology, software, validation, formal analysis, investigation, resources, data curation, writing – original draft preparation, and visualization. S.M.P. participated in supervision. Both authors have read and agreed to the published version of the manuscript.

Funding

No funding to declare.

Ethics Approval and Consent to Participate

The study was conducted according to the guidelines of the Declaration of Helsinki and was approved by the Institutional Review Board (IRB) of Dankook University Dental Hospital (approval No. DKUDH IRB 2022-11-009). The requirement for patient consent was waived by the IRB of Dankook University Dental Hospital due to the retrospective nature of this study.

Conflict of Interest

No potential conflict of interest relevant to this article was reported.

References

- Kim DW, Kim H, Nam W, Kim HJ, Cha IH. Machine learning to predict the occurrence of bisphosphonate-related osteonecrosis of the jaw associated with dental extraction: a preliminary report. *Bone* 2018;116:207-14. <https://doi.org/10.1016/j.bone.2018.04.020>
- Heo J, Yoo J, Lee H, Lee IH, Kim JS, Park E, et al. Prediction of hidden coronary artery disease using machine learning in patients with acute ischemic stroke. *Neurology* 2022;99:e55-65. <https://doi.org/10.1212/wnl.0000000000200576>
- Chen YW, Stanley K, Att W. Artificial intelligence in dentistry: current applications and future perspectives. *Quintessence Int* 2020;51:248-57. <https://doi.org/10.3290/j.qi.a43952>
- Lee JH, Kim DH, Jeong SN, Choi SH. Detection and diagnosis of dental caries using a deep learning-based convolutional neural network algorithm. *J Dent* 2018;77:106-11. <https://doi.org/10.1016/j.jdent.2018.07.015>
- Jung SK, Kim TW. New approach for the diagnosis of extractions with neural network machine learning. *Am J Orthod Dentofacial Orthop* 2016;149:127-33. <https://doi.org/10.1016/j.jado.2015.07.030>
- Ben-Israel D, Jacobs WB, Casha S, Lang S, Ryu WHA, de Lotbiniere-Bassett M, et al. The impact of machine learning on patient care: a systematic review. *Artif Intell Med* 2020;103:101785. <https://doi.org/10.1016/j.artmed.2019.101785>
- Jordan MI, Mitchell TM. Machine learning: trends, perspectives, and prospects. *Science* 2015;349:255-60. <https://doi.org/10.1126/science.aaa8415>
- Truong A, Walters A, Goodsitt J, Hines K, Bruss CB, Farivar R. Towards automated machine learning: evaluation and comparison of AutoML approaches and tools. 2019 IEEE 31st International Conference on Tools with Artificial Intelligence (ICTAI) 2019:1471-9. <https://doi.org/10.1109/ICTAI.2019.00209>
- Ruggiero SL, Dodson TB, Aghaloo T, Carlson ER, Ward BB, Kademani D. American Association of Oral and Maxillofacial Surgeons' position paper on medication-related osteonecrosis of the jaws-2022 update. *J Oral Maxillofac Surg* 2022;80:920-43. <https://doi.org/10.1016/j.joms.2022.02.008>
- Yamagata K, Nagai H, Baba O, Uchida F, Kanno N, Hasegawa S, et al. A case of brain abscess caused by medication-related osteonecrosis of the jaw. *Case Rep Dent* 2016;2016:7038618. <https://doi.org/10.1155/2016/7038618>
- Uddin S, Khan A, Hossain ME, Moni MA. Comparing different supervised machine learning algorithms for disease prediction. *BMC Med Inform Decis Mak* 2019;19:281. <https://doi.org/10.1186/s12911-019-1004-8>
- Sahin EK. Assessing the predictive capability of ensemble tree methods for landslide susceptibility mapping using XGBoost, gradient boosting machine, and random forest. *SN Appl Sci* 2020;2:1308. <https://doi.org/10.1007/s42452-020-3060-1>
- Li D, Liu Z, Armaghani DJ, Xiao P, Zhou J. Novel ensemble intelligence methodologies for rockburst assessment in complex and variable environments. *Sci Rep* 2022;12:1844. <https://doi.org/10.1038/s41598-022-05594-0>
- Graczyk M, Lasota T, Trawiński B, Trawiński K. Comparison of bagging, boosting and stacking ensembles applied to real estate appraisal. In: Nguyen NT, Le MT, Świątek J, eds. *ACIIDS 2010: intelligent information and database systems. Lecture Notes in Computer Science*, Vol. 5991. Springer; 2010:340-50. https://doi.org/10.1007/978-3-642-12101-2_35
- LeCun Y, Bengio Y, Hinton G. Deep learning. *Nature* 2015;521:436-44. <https://doi.org/10.1038/nature14539>
- Lee JK, Kim KW, Choi JY, Moon SY, Kim SG, Kim CH, et al. Bisphosphonates-related osteonecrosis of the jaw in Korea: a preliminary report. *J Korean Assoc Oral Maxillofac Surg* 2013;39:9-13. <https://doi.org/10.5125/jkaoms.2013.39.1.9>
- Sim IW, Borromeo GL, Tsao C, Hardiman R, Hofman MS, Papatziomos HJelle C, et al. Teriparatide promotes bone healing in medication-related osteonecrosis of the jaw: a placebo-controlled, randomized trial. *J Clin Oncol* 2020;38:2971-80. <https://doi.org/10.1200/jco.19.02192>
- Lo JC, O'Ryan FS, Gordon NP, Yang J, Hui RL, Martin D, et al.; Predicting Risk of Osteonecrosis of the Jaw with Oral Bisphosphonate Exposure (PROBE) Investigators. Prevalence of osteonecrosis of the jaw in patients with oral bisphosphonate exposure. *J Oral Maxillofac Surg* 2010;68:243-53. <https://doi.org/10.1016/j.joms.2009.03.050>
- Aljohani S, Fliefel R, Ihbe J, Kühnisch J, Ehrenfeld M, Otto S. What is the effect of anti-resorptive drugs (ARDs) on the development of medication-related osteonecrosis of the jaw (MRONJ) in osteoporosis patients: a systematic review. *J Craniomaxillofac Surg* 2017;45:1493-502. <https://doi.org/10.1016/j.jcms.2017.05.028>
- Buchbender M, Bauerschmitz C, Pirkel S, Kesting MR, Schmitt CM. A retrospective data analysis for the risk evaluation of the development of drug-associated jaw necrosis through dentoalveolar interventions. *Int J Environ Res Public Health* 2022;19:4339. <https://doi.org/10.3390/ijerph19074339>
- Wick A, Bankosegger P, Otto S, Hohlweg-Majert B, Steiner T, Probst F, et al. Risk factors associated with onset of medication-related osteonecrosis of the jaw in patients treated with denosumab. *Clin Oral Investig* 2022;26:2839-52. <https://doi.org/10.1007/s00784-021-04261-4>
- McGowan K, McGowan T, Ivanovski S. Risk factors for medication-related osteonecrosis of the jaws: a systematic review. *Oral Dis* 2018;24:527-36. <https://doi.org/10.1111/odi.12708>

How to cite this article: Kwack DW, Park SM. Prediction of medication-related osteonecrosis of the jaw (MRONJ) using automated machine learning in patients with osteoporosis associated with dental extraction and implantation: a retrospective study. *J Korean Assoc Oral Maxillofac Surg* 2023;49:135-141. <https://doi.org/10.5125/jkaoms.2023.49.3.135>