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Original Article

Quantitative analysis of mandibular cortical morphology using artificial intelligencebased computer assisted diagnosis for panoramic radiography on underlying diseases and dental status in women over 20 years of age



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KEYWORDS

Artificial intelligence; Computer-assisted diagnosis; Panoramic radiography; Mandibular cortical index; Women; Osteoporosis Abstract Background/purpose: Recently, an artificial intelligence-based computer-assisted diagnosis (AI-CAD) for panoramic radiography was developed to scan the inferior margin of the mandible and automatically evaluate mandibular cortical morphology. The aim of this study was to analyze quantitatively the mandibular cortical morphology using the AI-CAD, especially focusing on underlying diseases and dental status in women over 20 years of age. *Materials and methods:* 419 patients in women over 20 years of age who underwent panoramic radiography were included in this study. The mandibular cortical morphology was analyzed with an AI-CAD that evaluated the degree of deformation of the mandibular inferior cortex (MIC) and mandibular cortical index (MCI) automatically. Those were analyzed in relation to underlying diseases, such as diabetes, hypertension, dyslipidemia, rheumatism and osteoporosis, and dental status, such as the number of teeth present in the maxilla and mandible. *Results:* The degree of deformation of MIC in women under 51 years of age (51–90 years; n = 190, 45.1 ± 23.0), and the MCI was a significant difference for the different age group. Regarding the degree of deformation of MIC and MCI in women over 50 years of age,

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osteoporosis and number of total teeth present in the maxilla and mandible were significant differences.

Conclusion: The results of this study indicated that the mandibular cortical morphology using the AI-CAD is significantly related to osteoporosis and dental status in women over 50 years of age.

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Introduction

The mandibular cortical index (MCI) is a qualitative index that assesses the appearance of the mandibular endosteum.¹ It classifies the mandibular cortex as Class 1, where the mandibular cortical bone has a normal shape, indicating that the patient is not at risk; Class 2, where the mandibular cortical bone presents few radiolucent areas with semi-circular or linear shape; and Class 3, where the mandibular cortical bone presents several radiolucent areas, indicating that the patient is at high risk of systemic osteoporosis. The MCI is useful as an auxiliary tool for identifying postmenopausal females at risk of low bone mineral density (BMD), as it correlates with skeletal BMD measured by dual X-ray absorptiometry (DXA).²

The computer-assisted diagnosis (CAD) systems that may help differential diagnosis by using medical images take place in the literature day by day along with the recent development of artificial intelligence (AI) methods and computer hardware.³ There are studies on the prediction and clarification of osteoporosis with machine learning and fuzzy systems by obtaining features from panoramic images in the literature.^{4,5} Recently, an AI-CAD system for panoramic radiography "PanoSCOPE" was developed to scan the inferior margin of the mandible and automatically evaluate mandibular cortical morphology, such as the MCI, and the AI-CAD system is an effective tool for pre-screening of osteoporosis by panoramic radiography.^{6,7} The AI-CAD system has been reported on the mandibular cortical morphology in medication-related osteonecrosis of the jaw with osteoporosis, in relation to age and gender, and preand post-dental implant operations. $^{8-10}$ However, no studies were found for mandibular cortical morphology in relation to underlying diseases and dental status using the AI-CAD by PubMed. We analyzed quantitatively mandibular cortical morphology using the AI-CAD for panoramic radiography, especially focusing on underlying diseases and dental status in women over 20 years of age.

Materials and methods

Patients

This prospective study was approved by the Ethics Committee of The Nippon Dental University School of Life Dentistry at Niigata (ECNG-R-318), and all patients gave written informed consent. 419 patients in women over 20 years of age (mean age, 49.3 years; range, 21–90 years) who underwent panoramic radiography at our university hospital from April 2022 to September 2022 were included in this study. The patients with dental implants and jaw lesions, such as cysts and tumors, were excluded. The underlying diseases, such as diabetes, hypertension, dyslipidemia, rheumatism and osteoporosis, was recorded from the electronic medical chart of the university hospital.

Panoramic imaging and data analysis

Panoramic radiography was performed with panoramic Xray unit (Veraviewepocs; MORITA, Kyoto, Japan), following our hospital maxillofacial protocol.^{8–10} Regarding dental status, the numbers of total anterior and posterior teeth present were assessed from the panoramic radiography.^{11,12} Anterior teeth were defined as central and lateral incisors and canines, and posterior teeth as premolars and molars.¹³ Third molars were excluded. Root remnants were counted as teeth except for the ones which were totally embedded in bone.¹⁴

The mandibular cortical morphology on panoramic radiographs was analyzed with an AI-CAD (PanoSCOPE; MEDIA, Tokyo, Japan) that evaluated the degree of deformation of the mandibular inferior cortex (MIC) and the MCI automatically, following our hospital maxillofacial protocol.^{8–10} The degree of deformation of MIC was calculated using the thickness and roughness of the mandibular cortex with AI-CAD. The MCI classification was as follows: normal (Class 1, Fig. 1), mildly to moderately eroded (Class 2, Fig. 2), or severely eroded (Class 3, Fig. 3) with AI-CAD.

Statistical analysis

The degree of deformation of MIC at different ages was performed by one-way measures analysis of variance with Tukey's honestly significant difference test and Mann—Whitney U-test. The degree of deformation of MIC in relation to underlying diseases and dental status was performed by Mann—Whitney U-test. MCI classifications of all patients were performed by Pearson chi-square test. Those statistical analyses were evaluated with the statistical package SPSS Statistics (version 26, IBM Japan, Tokyo, Japan) using a 5% significance level.

Results

The degree of deformation of MIC using AI-CAD at different ages in women over 20 years of age was shown in Table 1. The degree of deformation of MIC in over 20-years group $(21-30 \text{ years}, 15.1 \pm 13.2)$ was lower than those in over 50-



Figure 1 PanoSCOPE software in a 73-year-old woman indicates that the degree of deformation and MCI are 29/100 and Class 1, respectively.



Figure 2 PanoSCOPE software in a 62-year-old woman indicates that the degree of deformation and MCI are 64/100 and Class 2, respectively.

years group (51–60 years, 31.7 ± 22.5 , P < 0.001), over 60years group (61–70 years, 48.1 ± 21.7 , P < 0.001), over 70years group (71–80 years, 48.9 ± 19.4 , P < 0.001) and over 80-years group (81–90 years, 60.1 ± 19.5 , P < 0.001), although those in over 20-years group was not a significant difference in over 30-years group (31–40 years, 15.8 \pm 11.9, P= 1.000) and over 40-years group (41–50 years, 17.9 \pm 12.7, P= 0.951). Furthermore, the degree of deformation of MIC in women over 20 years of age (21–50 years; n = 229, 16.0 \pm 12.7) was significantly lower than those of over 50 years of age (51–90 years; n = 190, 45.1 \pm 23.0, P< 0.001).



Figure 3 PanoSCOPE software in a 56-year-old woman indicates that the degree of deformation and MCI are 71/100 and Class 3, respectively.

The mandibular cortical index (MCI) using AI-CAD at different ages in women under 51 and over 50 years of age was shown in Table 2. The MCI was a significant difference for the different age groups (P < 0.001). Furthermore, the MCI under 51 years of age was not significant difference for different age groups (P = 0.967), and the MCI over 50 years of age was a significant difference for different age groups (P < 0.001).

The degree of deformation of MIC using AI-CAD in relation to underlying diseases in women under 51 and over 50 years of age was shown in Table 3. Regarding those in relation to underlying diseases in women over 50 years of

Table	1	Degree	of	deformation	of	mandibular	inferior
cortex	usir	ng artific	ial	intelligence-b	base	d computer-	assisted
diagno	sis a	t differe	ent	ages in wome	en o	ver 20 years	of age.

Age (years)	Cases	Degree of deformation of MIC using AI-CAD					
		$\text{Mean} \pm \text{SD}$	Range	P-value			
21-30	100	$\textbf{15.1} \pm \textbf{13.2}$	0–66	_			
31-40	72	$\textbf{15.8} \pm \textbf{11.9}$	1-71	1.000			
41-50	57	$\textbf{17.9} \pm \textbf{12.7}$	0-65	0.951			
51-60	58	$\textbf{31.7} \pm \textbf{22.5}$	1-78	<0.001			
61-70	50	$\textbf{48.1} \pm \textbf{21.7}$	6-86	<0.001			
71-80	54	$\textbf{48.9} \pm \textbf{19.4}$	6-90	<0.001			
81-90	28	$\textbf{60.1} \pm \textbf{19.5}$	16-91	<0.001			

Abbreviations: MIC, mandibular inferior cortex; AI-CAD, artificial intelligence-based computer-assisted diagnosis; SD, standard deviation.

Under 51 years of age (21–50 years; n = 229, 16.0 \pm 12.7) vs. over 50 years of age (51–90 years; n = 190, 45.1 \pm 23.0, P<0.001).

age, osteoporosis was a significant difference (P = 0.006). Furthermore, the MCI using AI-CAD in relation to underlying diseases in women under 51 and over 50 years of age was shown in Table 4. Regarding those in relation to underlying diseases in women over 50 years of age, osteoporosis was a significant difference (P = 0.013).

The degree of deformation of MIC using AI-CAD in relation to dental status in women under 51 and over 50 years of age was shown in Table 5. Regarding those in relation to

Table 2Mandibular cortical index using artificialintelligence-based computer-assisted diagnosis at differentages in women under 51 and over 50 years of age.

Age (years)	Cases		Mandibular index using	cortical AI-CAD			
		Class 1 (%)	Class 2 (%)	Class 3 (%)	P-value		
Under 51 years of age	229	212 (92.6%)	14 (6.1%)	3 (1.3%)	0.967		
21-30	100	94 (94.0%)	5 (5.0%)	1 (1.0%)			
31-40	72	66 (91.7%)	5 (6.9%)	1 (1.4%)			
41-50	57	52 (91.2%)	4 (7.0%)	1 (1.8%)			
Over 50 years	190	68 (35.8%)	79 (41.6%)	43 (22.6%)	<0.001		
of age							
51-60	58	38 (65.5%)	14 (24.1%)	6 (10.3%)			
61-70	50	17 (34.0%)	20 (40.0%)	13 (26.0%)			
71-80	54	11 (20.4%)	32 (59.3%)	11 (20.4%)			
81-90	28	2 (7.1%)	13 (46.4%)	13 (46.4%)			

Abbreviations: AI-CAD, artificial intelligence-based computerassisted diagnosis. Under 51 years of age vs. over 50 years of age (P < 0.001).

Underlying		Degree of deformation of mandibular inferior cortex using AI-CAD								
diseases	Unde	er 51 years of age ((21—50 years	, n = 229)	n = 229) Over 50 years of age (51–90 years, n =					
	Cases	$\text{Mean} \pm \text{SD}$	Range	P-value	Cases	$\text{Mean} \pm \text{SD}$	Range	P-value		
Diabetes				_				0.092		
Presence	0				10	$\textbf{56.5} \pm \textbf{13.6}$	33—81			
Absence	229	$\textbf{16.0} \pm \textbf{12.7}$	0-71		180	$\textbf{44.5} \pm \textbf{23.3}$	1-91			
Hypertension				0.541				0.691		
Presence	1	8	8		60	$\textbf{44.5} \pm \textbf{20.5}$	5—89			
Absence	228	$\textbf{16.1} \pm \textbf{12.7}$	0-71		130	$\textbf{45.4} \pm \textbf{24.2}$	1-91			
Dyslipidemia				0.122				0.643		
Presence	1	39	39		34	$\textbf{46.3} \pm \textbf{23.7}$	5-85			
Absence	228	$\textbf{15.9} \pm \textbf{12.6}$	0-71		156	$\textbf{44.9} \pm \textbf{22.9}$	1-91			
Rheumatism				0.048				0.648		
Presence	2	$\textbf{38.5} \pm \textbf{17.7}$	26-51		6	$\textbf{48.3} \pm \textbf{18.9}$	11–62			
Absence	227	$\textbf{15.8} \pm \textbf{12.5}$	0-71		184	$\textbf{45.0} \pm \textbf{23.2}$	1-91			
Osteoporosis				_				0.006		
Presence	0				35	$\textbf{54.6} \pm \textbf{18.4}$	17—89			
Absence	229	$\textbf{16.0} \pm \textbf{12.7}$	0-71		155	$\textbf{43.0} \pm \textbf{23.5}$	1–91			

Table 3 Degree of deformation of mandibular inferior cortex using artificial intelligence-based computer-assisted diagnosis in relation to underlying diseases in women under 51 and over 50 years of age.

Abbreviations: AI-CAD, artificial intelligence-based computer-assisted diagnosis; SD, standard deviation.

dental status in women over 50 years of age, the number of total teeth present in the maxilla and mandible was a significant difference (P < 0.001 and P < 0.001, respectively). Furthermore, the MCI using AI-CAD in relation to dental status in women under 51 and over 50 years of age was shown in Table 6. Regarding those in relation to dental status in women over 50 years of age, the number of total teeth present in the maxilla and mandible was a significant difference (P = 0.003 and P = 0.002, respectively).

Discussion

The MCI is an easy-application index that can be broadly implemented by dentists.² Taguchi¹⁵ showed that the MCI seem to be appropriate index for triaging individuals with undiagnosed low skeletal BMD or osteoporosis. Furthermore, Taguchi et al.¹⁶ suggest that the MCI, when used for identifying asymptomatic postmenopausal female patients at risk of having osteoporosis in general dental practice,

Mandibular cortical index using artificial intelligence-based computer-assisted diagnosis in relation to underlying Table 4 diseases in women under 51 and over 50 years of age.

Underlying	Mandibular cortical index using AI-CAD										
diseases	Un	der 51 years of	f age (21—50	years, n =	= 229)	0\	ver 50 years o	of age (51—90) years, $n =$	190)	
	Cases	Class 1	Class 2	Class 3	P-value	Cases	Class 1	Class 2	Class 3	P-value	
Diabetes					_					0.215	
Presence	0	0 (0%)	0 (0%)	0 (0%)		10	1 (10.0%)	6 (60.0%)	3 (30.0%)		
Absence	229	212 (92.6%)	14 (6.1%)	3 (1.3%)		180	67 (37.2%)	73 (40.6%)	40 (22.2%)		
Hypertension					0.961					0.269	
Presence	1	1 (100%)	0 (0%)	0 (0%)		60	19 (31.7%)	30 (50.0%)	11 (18.3%)		
Absence	228	211 (92.5%)	14 (6.1%)	3 (1.3%)		130	49 (37.7%)	49 (37.7%)	32 (24.6%)		
Dyslipidemia		· · ·	, , , , , , , , , , , , , , , , , , ,	, ,	<0.001		. ,	. ,	· · ·	0.415	
Presence	1	0 (0%)	1 (100%)	0 (0%)		34	13 (38.2%)	11 (32.4%)	10 (29.4%)		
Absence	228	212 (93.0%)	13 (5.7%)	3 (1.3%)		156	55 (35.3%)	68 (43.6%)	33 (21.2%)		
Rheumatism					0.034					0.099	
Presence	2	1 (50.0%)	1 (50.0%)	0 (0%)		6	1 (16.7%)	5 (83.3%)	0 (0%)		
Absence	227	211 (93.0%)	13 (5.7%)	3 (1.3%)		184	67 (36.4%)	74 (40.2%)	43 (23.4%)		
Osteoporosis				. ,	_		. ,		. ,	0.013	
Presence	0	0 (0%)	0 (0%)	0 (0%)		35	5 (14.3%)	19 (54.3%)	11 (31.4%)		
Absence	229	212 (92.6%)	14 (6.1%)	3 (1.3%)		155	63 (40.6%)	60 (38.7%)	32 (20.6%)		
Abbreviations	ALCAD :	artificial intellig	ionco-based a	computor-as	sisted diagr	ocic					

artificial intelligence-based computer-assiste

Number of		Degre	libular inferi	oular inferior cortex using AI-CAD				
teeth present	Un	der 51 years of age (21–50 years	s, n = 229)	n = 229) Over 50 years of age (51–90 years			, n = 190)
	Cases	$\text{Mean} \pm \text{SD}$	Range	P-value	Cases	$\text{Mean} \pm \text{SD}$	Range	P-value
Maxilla								
Total				0.838				<0.001
0 - 13	21	$\textbf{16.8} \pm \textbf{14.4}$	0-65		124	$\textbf{49.5} \pm \textbf{22.1}$	3–91	
14	208	$\textbf{16.0} \pm \textbf{12.5}$	0-71		66	$\textbf{36.9} \pm \textbf{22.6}$	1-86	
Anterior				0.436				0.023
0 - 5	9	$\textbf{22.1} \pm \textbf{20.4}$	0-65		63	$\textbf{50.6} \pm \textbf{22.3}$	3-90	
6	220	$\textbf{15.8} \pm \textbf{12.3}$	0-71		127	$\textbf{42.4} \pm \textbf{23.0}$	1-91	
Posterior				0.830				<0.001
0 - 7	17	$\textbf{17.2} \pm \textbf{15.5}$	0-65		117	$\textbf{49.9} \pm \textbf{22.3}$	3–91	
8	212	$\textbf{15.9} \pm \textbf{12.5}$	0-71		73	$\textbf{37.4} \pm \textbf{22.1}$	1-86	
Mandible								
Total				0.886				<0.001
0 - 13	15	$\textbf{14.9} \pm \textbf{11.0}$	0-43		128	$\textbf{49.3} \pm \textbf{22.0}$	1-91	
14	214	$\textbf{16.1} \pm \textbf{12.8}$	0-71		62	$\textbf{36.5} \pm \textbf{22.8}$	2-81	
Anterior				0.638				0.002
0 - 5	1	18	18		21	$\textbf{59.1} \pm \textbf{21.7}$	11-91	
6	228	$\textbf{16.0} \pm \textbf{12.7}$	0-71		169	$\textbf{43.4} \pm \textbf{22.6}$	1-90	
Posterior				0.748				0.001
0 - 7	14	14.6 ± 11.4	0-43		127	$\textbf{49.1} \pm \textbf{21.9}$	1-91	
8	215	$\textbf{16.1} \pm \textbf{12.8}$	0-71		63	$\textbf{37.2} \pm \textbf{23.3}$	2-81	

Table 5Degree of deformation of mandibular inferior cortex using artificial intelligence-based computer-assisted diagnosis inrelation to dental status in women under 51 and over 50 years of age.

Abbreviations: AI-CAD, artificial intelligence-based computer-assisted diagnosis; SD, standard deviation.

may be helpful in reducing the incidence of first fractures, with a consequent reduction in the secondary fractures, medical costs, and mortality associated with osteoporotic fragility fractures, without incurring any additional cost. Therefore, we analyzed quantitatively the mandibular cortical morphology, such as the MCI, using an AI-CAD for panoramic radiography, especially focusing on underlying diseases and dental status in women over 20 years of age.

In terms of the correlation between the MCI classification and age, Ledgerton et al.¹¹ indicated that the MCI showed an age-related distribution in a British female population. Gulsahi et al.¹² suggested that logistic regression analysis clearly showed that age was the most important parameter for the MCI in Turkish patients, and as age increased, the likelihood of being in the Class 3 category also increased. Ogawa et al.⁹ showed that the MCI of Japanese women was a significant difference for aging, although those of men was not significant difference for aging. In this study, the degree of deformation of MIC in women under 51 years of age was significantly lower than those of over 50 years of age, and the MCI was a significant difference for the different age groups. Munhoz et al.² concluded that this correlation is expected, as the risk for osteoporosis increases with age, and systemic BMD alterations due to osteoporosis affect the mandible, similar to how they affect other major bones. We also support those hypotheses, and consider that this correlation is expected, as the patients over 50 years of age include postmenopausal women.

In terms of the correlation between the MCI classification and underlying disease excluding osteoporosis, Munhoz et al.¹⁷ investigated type 2 diabetes and its association with the MCI, and suggest that type 2 diabetes might increase BMD in postmenopausal females. Tenório et al.¹⁸ concluded that the MCI suggestive of reduced BMD is more likely to be identified in panoramic radiographs of hepatic cirrhotic individuals than of healthy ones. In this study, the degree of deformation of MIC and the MCI in women over 50 years of age was a significant difference for only osteoporosis not diabetes, hypertension, dyslipidemia and rheumatism. Further research is necessary to validate these results and the relationship between MCI and underlying diseases.

Regarding the MCI in relation to dental status, Gulsahi et al.¹² indicated that compared with fully dentate patients, the likelihood of being in the Class 3 category for edentulous patients was 27.30 times higher and in partially dentate patients it was 2.68 times higher, and concluded that masticatory muscle atrophy may be one of the factors resulting in a change in the quality of mandibular bone. Taguchi et al.¹³ suggest that the loss of posterior teeth may be associated with a decrease not only in alveolar bone height, but also in alveolar BMD. However, Yaşar et al.¹⁴ were unable to show the relationship between osteoporosis and the number of teeth present. In this study, the degree of deformation of MIC and the MCI in women over 50 years of age, and the number of total teeth present in the maxilla and mandible was a significant difference. We consider that those results may indicate that local factors such as dental status affect mandibular bone quality, and patients of dental treatments may include osteoporosis who are not diagnosed and treated. Our results recommend that women with tooth loss over 50 years of age should

Number of	Mandibular cortical index using AI-CAD										
teeth present	Un	der 51 years o	of age (21—5	0 years, n =	Over 50 years of age (51–90 years, $n = 190$)						
	Cases	Class 1	Class 2	Class 3	P-value	Cases	Class 1	Class 2	Class 3	P-value	
Maxilla											
Total					0.336					0.003	
0 - 13	21	19 (90.5%)	1 (4.8%)	1 (4.8%)		124	34 (27.4%)	56 (45.2%)	34 (27.4%)		
14	208	193 (92.8%)	13 (6.3%)	2 (1.0%)		66	34 (51.5%)	23 (34.8%)	9 (13.6%)		
Anterior					0.024					0.032	
0 - 5	9	7 (77.8%)	1 (11.1%)	1 (11.1%)		63	17 (27.0%)	25 (39.7%)	21 (33.3%)		
6	220	205 (93.2%)	13 (5.9%)	2 (0.9%)		127	51 (40.2%)	54 (42.5%)	22 (17.3%)		
Posterior					0.227					0.004	
0 - 7	17	15 (88.2%)	1 (5.9%)	1 (5.9%)		117	32 (27.4%)	52 (44.4%)	33 (28.2%)		
8	212	197 (92.9%)	13 (6.1%)	2 (0.9%)		73	36 (49.3%)	27 (37.0%)	10 (13.7%)		
Mandible											
Total					0.896					0.002	
0 - 13	15	14 (93.3%)	1 (6.7%)	0 (0%)		128	35 (27.3%)	59 (46.1%)	34 (26.6%)		
14	214	198 (92.5%)	13 (6.1%)	3 (1.4%)		62	33 (53.2%)	20 (32.3%)	9 (14.5%)		
Anterior					0.961					0.002	
0 - 5	1	1 (100%)	0 (0%)	0 (0%)		21	4 (19.0%)	6 (28.6%)	11 (52.4%)		
6	228	211 (92.5%)	14 (6.1%)	3 (1.3%)		169	64 (37.9%)	73 (43.2%)	32 (18.9%)		
Posterior					0.895					0.003	
0 - 7	14	13 (92.9%)	1 (7.1%)	0 (0%)		127	35 (27.6%)	59 (46.5%)	33 (26.0%)		
8	215	199 (92.6%)	13 (6.0%)	3 (1.4%)		63	33 (52.4%)	20 (31.7%)	10 (15.9%)		
Abbreviations:		artificial intellio	ance-based	computer-ass	isted diago	osis					

Table 6 Mandibular cortical index using artificial intelligence-based computer-assisted diagnosis in relation to dental status in women under 51 and over 50 years of age.

undergo BMD testing, such as DXA, because of a risk factor for osteoporosis.

There were several limitations of this study. Although the patients in women over 20 years of age who underwent panoramic radiography were included in this study, those patients were not investigated about pre- or postmenopause and BMD testing, such as DXA, because of dental treatments not osteoporosis. We consider that the most interesting point of the PanoSCOPE is how early to detect and diagnose osteoporosis, and the MCI may differ depending on the stages of osteoporosis. Further research on osteoporosis is necessary to validate these results.

In conclusion, this study analyzed quantitatively the mandibular cortical morphology using the AI-CAD for panoramic radiography, especially focusing on underlying diseases and dental status in women over 20 years of age. The results of this study indicated that the mandibular cortical morphology using the AI-CAD is significantly related to osteoporosis and dental status in women over 50 years of age.

Declaration of competing interest

The authors have no conflicts of interest relevant to this article.

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