



Segmentation-based 3D volumetry and linear regression modeling for assessing the vertebral bone loss in pyogenic vertebral osteomyelitis

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Abstract

Background Pyogenic vertebral osteomyelitis (PVO) presents an escalating clinical challenge due to rising incidence, high mortality, and significant bone destruction. Objective quantification of vertebral body (VB) bone loss for assessing the disease severity and guiding therapeutic decisions is yet to be established.

Methods We retrospectively identified patients with confirmed PVO between 2010 and 2020. Volumetric assessments of VBs were performed using 3D Slicer, and pre-infection volumes were estimated by linear regression based on adjacent, non-infected vertebrae. A “Destruction Quotient” (DQ) was calculated (measured volume/estimated original volume) to quantify VB loss. In a subgroup analysis VB bone loss was evaluated, depending on sex, spinal location and pathogen group.

Results Thirty-one patients met the inclusion criteria for 3D volumetry (16 males, 15 females; mean age: 67.0 ± 9.2 years; mean BMI 32.4 kg/m^2). In total, $n=267$ VBs were segmented. Linear regression models demonstrated a high mean coefficient of determination ($R^2 > 0.95$), with mean slopes of $m=2.3$ (95% CI = 1.94–2.75) in males and $m=1.8$ (95% CI = 1.46–2.19) in females. The mean measured volume of infected VBs ($17.8 \pm 9.3 \text{ cm}^3$) was significantly lower than the estimated original volume ($24.1 \pm 10.5 \text{ cm}^3$; $p < 0.001$). VBs at the lumbar spine experienced a median volume loss of 30%, whereas thoracic VBs showed 18% loss of volume. Female patients demonstrated a significantly higher median VB loss (32%) than males (12%; $p < 0.05$). No significant variation in DQs was observed among different pathogen groups, with *Staphylococcus aureus* being the most prevalent; however, within the *Staphylococcus aureus* subgroup, the measured VB volume was significantly smaller than the original estimated volume with a mean difference of $6.13 \pm 4.9 \text{ cm}^3$ ($p < 0.01$).

Conclusion A 3D-volumetric approach and linear regression modeling offers an individualized method for quantifying VB destruction in PVO. Integrating automated segmentation and densitometric data may further enhance predictive accuracy and improve patient-specific treatment strategies.

Keywords Spinal infections · Vertebral osteomyelitis · Spondylodiscitis · 3D-Volumetry · Segmentation

Introduction

Spinal infections, first and foremost pyogenic vertebral osteomyelitis (PVO) demonstrates a troubling increase in the incidence and mortality of vertebral osteomyelitis in Germany, with a significant impact on the aging population [1, 2]. The high in-hospital mortality rates of 2–13% pose a significant clinical challenge due to the potential for severe morbidity and spinal column instability [3, 4]. In addition to the impact of age, comorbidities, such as rheumatoid arthritis, congestive heart failure and kidney disease, have

been strongly associated with an increased risk of mortality [4–6]. Beyond physical impairment, vertebral osteomyelitis can also lead to substantial mental disorders [7].

PVO is marked by the destruction of vertebral bodies (VB), potentially resulting in spinal instability, neurological deficits, and chronic pain [8]. Traditional PVO assessment depends on clinical evaluation, MRI, and CT, but lacks precise quantification of bone loss, limiting severity assessment and instability prediction. Several studies stress the need for accurate vertebral bone loss quantification. For instance, Limthongkul et al. demonstrated a quantitative

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approach to vertebral body volume assessment using CT scans, showing significant variations in VB volumes along the thoracic and lumbar spine [9]. However, these studies lack a standardized method for estimating the original, pre-infection volumes of VB, which is essential for calculating the true extent of bone destruction in infected vertebrae [9, 10]. Advancements in three-dimensional (3D) imaging and open-source software tools, such as 3D Slicer, have opened new avenues for volumetric analysis of VB. These tools allow for the creation of detailed 3D models from CT scans, enabling more precise measurements of VB volumes and the ability to differentiate between infected and non-infected vertebrae [11, 12]. Accurate quantification of vertebral bone loss is clinically relevant, as it may reflect disease severity, support surgical decision-making, and aid in predicting spinal instability, particularly in infections involving weight-bearing segments.

This study aimed to use 3D volumetric analysis and linear regression modeling to quantify vertebral body destruction in PVO, establishing a reliable destruction quotient to assess disease severity, predict spinal instability, and guide surgical decisions. The goal is to create a robust framework for quantitative evaluation of bone loss in PVO.

Methods

Data analysis of pyogenic vertebral osteomyelitis patients

Patients diagnosed with PVO (ICD10: M46.1-4), treated conservatively or surgically at the University Hospital Regensburg between 2004 and 2020 were eligible for inclusion. Subsequently, adult patients' medical records, surgical protocols, laboratory results, and microbiological and histopathological reports were reviewed for criteria indicative of vertebral osteomyelitis (VO). The diagnosis of VO was established if at least two of the following criteria were met: (1) clinical features consistent with VO; (2) radiological evidence of vertebral osteomyelitis on CT and/or MRI [13, 14]; or (3) microbiological identification of bacterial pathogens, either directly from the infection site (e.g., abscess, intervertebral disc, or vertebral bone) or from blood cultures [15]. Furthermore, the records needed to contain a CT scan of the thoracolumbar spine.

CT-DICOM data set selection criteria

CT images were obtained from Siemens Healthineers' SOMATOM Definition Flash and Go Top systems. The Flash offers high-speed, low-radiation dual-source imaging, while the Go Top captures 128-slice scans, ideal for

detail. For the logistic model, eligible scans had clear views of the affected vertebrae and four non-infected adjacent vertebrae, with minimal noise and slice thickness up to three millimeters.

3D slicer software

We utilized 3D Slicer, a robust, open-source platform for medical image analysis and visualization, to perform volumetric analysis of vertebral structures. 3D Slicer has become a standard tool in clinical research, medical education, and image-guided therapy due to its support for multimodal imaging (CT, MRI, PET) and extensive modular capabilities [16]. The software's accuracy in volume measurement is well-documented in peer-reviewed studies, confirming its reliability for clinical research [17, 18].

Application of 3D slicer for volume measurement

Initially, anonymized DICOM data was imported into 3D Slicer. The software's modular workflow facilitates a range of analyses, from preprocessing to quantitative assessment, and is optimized for reproducibility and customization. For this study, key modules were utilized to ensure precise volumetric analysis of vertebral structures:

Preprocessing and region of interest (ROI) definition

The “Crop Volume” module isolated the target vertebra, excluding posterior structures from the pedicles onward, ensuring efficient processing and precise segmentation.

Segmentation workflow

Using the Segment Editor module, individual vertebral bodies were segmented, with separate segments assigned for infected and non-infected vertebrae. All segmentations were performed by a single trained annotator under supervision of a board-certified spine surgeon. Random samples were reviewed for anatomical plausibility. Formal inter-observer variability assessment was not conducted and is acknowledged as a limitation. The segmentation process combined automated and manual tools for optimal precision:

- Thresholding: Hounsfield Unit (HU) thresholds, ranging from 150 to 250 HUs, were applied to isolate bone tissue from surrounding soft tissues. Threshold adjustments were subjectively fine-tuned for each scan to enhance contrast (Fig. 1).
- Grow from Seeds: This semi-automated tool extrapolated initial segmentation across the entire vertebral

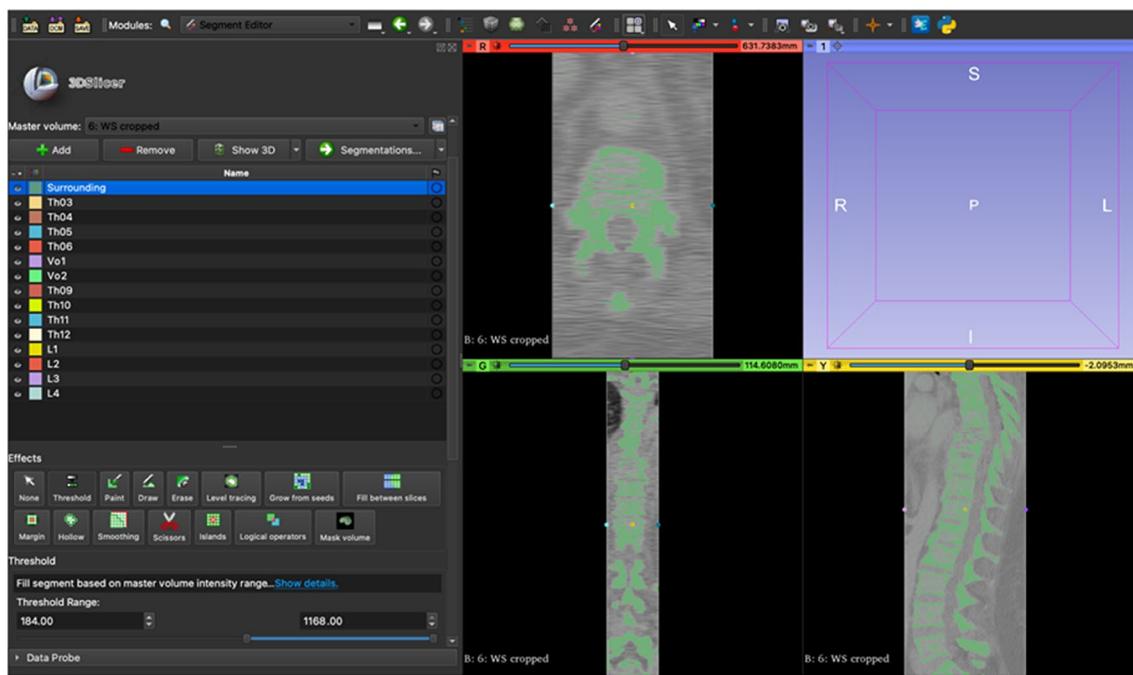


Fig. 1 3D Slicer workflow: Application of Hounsfield Unit thresholding to isolate bone tissue from surrounding soft tissues, with manual adjustments to optimize contrast for each scan

volume. Manual oversight was critical to ensure accurate propagation of the segmentation (Fig. 2).

Manual Refinements: Tools such as “Paint” and “Erase” were used to fine-tune segmentation, specifically excluding posterior elements like pedicles to focus solely on vertebral bodies (Fig. 3). Protruding osteophyte attachments were also removed in order to focus uniformly on the vertebral body as a cylindrical shape (Fig. 3B). Vertebrae with excessive bony overgrowth or spondylophytes that impaired clear boundary delineation were excluded from DQ analysis due to unreliable volumetric estimation.

Estimation of pre-infection volume of infected vertebral bodies

To assess vertebral body (VB) volume changes due to infection, pre-infection volumes (V_{original}) were estimated using linear regression analysis on non-infected VBs, yielding a coefficient of determination (R^2) to confirm linearity. The simple regression model.

$$f(x) = m \cdot x + t$$

was applied to estimate the original volume of the infected VBs. Subsequently, the “Destruction Quotient” (DQ) was calculated to assess volume changes due to infection:

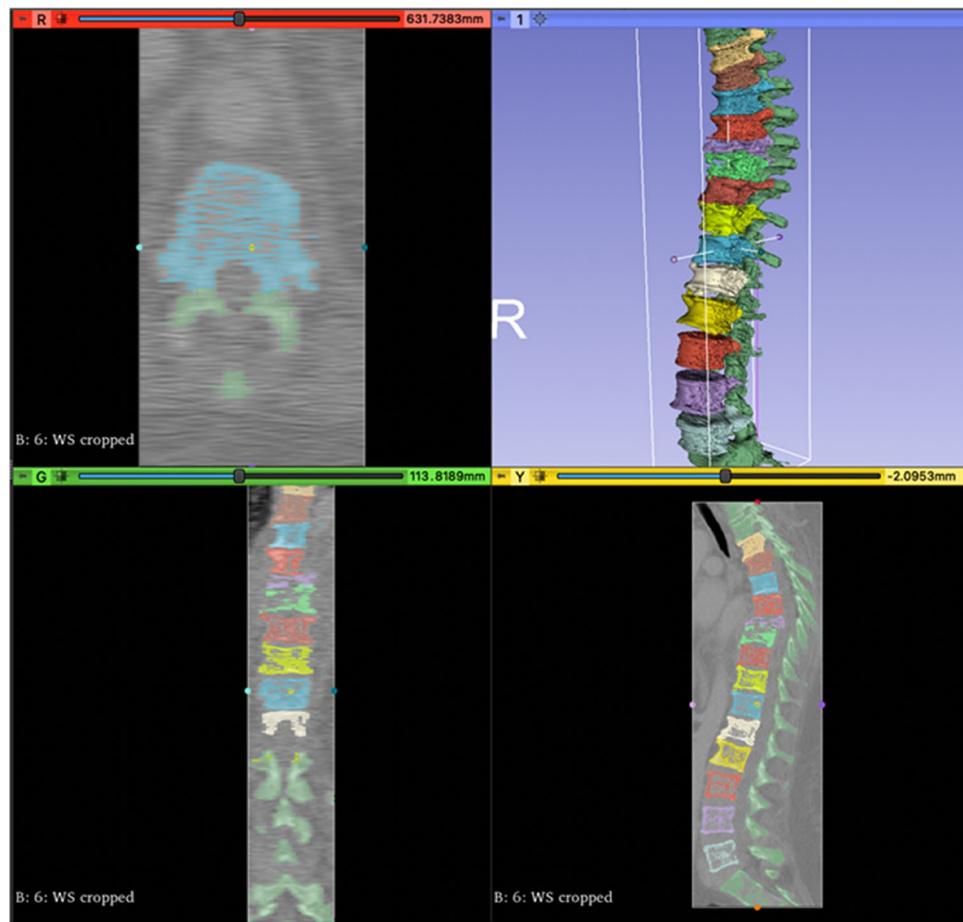
$$DQ = \frac{\text{Measured Volume of Infected VB (}V_{\text{measured}}\text{)}}{\text{Estimated Original Volume of Infected VB (}V_{\text{original}}\text{)}}$$

Values <1 indicate volume loss and >1 suggest volume gain. Patients with preexisting severe degeneration of the infected segment were excluded from the final evaluation of volume changes in vertebral osteomyelitis, as the high difficulty in segmenting vertebral body borders and differentiating between spondylophytes compromised the accuracy of measurements.

Statistical methods

Data analysis was conducted using SPSS Version 28. Parametric data were reported as mean \pm SD, non-parametric data as median (Min–Max). Bootstrapped 95% CIs (2,000 samples, BCa method) were calculated. VB volume progression linearity was assessed via simple linear regression (R^2). Normality was tested with Shapiro-Wilk. Parametric data were analyzed with t-tests, while Mann-Whitney U and Wilcoxon signed-rank tests were used for independent and paired non-parametric comparisons, respectively. Kruskal-Wallis tests compared multiple non-parametric groups, and Spearman’s rho assessed correlations. Statistical significance was set at $p < 0.05$.

Fig. 2 Visualization of the spine in 3D Slicer. Three-dimensional reconstruction of the spine generated using the “Grow from Seeds” function within the 3D Slicer software, illustrating the segmentation process for vertebral bodies



Results

Patient characteristics

Out of $N=255$ eligible patients $n=61$ PVO patients were included. After applying the CT selection criteria, $n=31$ patients were deemed suitable for volumetric analysis (Fig. 4). The cohort comprised $n=15$ women (48.4%) and $n=16$ men (51.6%), with an average age of 67.0 ± 9.2 years. The mean BMI was 32.4 kg/m^2 .

In $n=25$ (80.6%) cases pathogens were identified, with $n=4$ (19.4%) cases of polymicrobial infections. There were $n=6$ culture negative cases. Among the identified pathogens ($n=29$), Staphylococcus species were the most prevalent, accounting for $n=17$ (58.6%) cases, with *Staphylococcus aureus* (STAU) being the most frequent individual pathogen, identified in $n=13$ (44.8%) cases. Enterococci were the second most common Gram-positive group, representing $n=3$ (10.3%) cases. Gram-negative bacteria were identified in $n=6$ (20.7%) cases, with Enterobacteriaceae (including *Escherichia coli*, *Klebsiella pneumoniae*, *Citrobacter freundii* and *Enterobacter cloacae*) being the predominant subgroup ($n=5$). One *Pseudomonas aeruginosa* infection was

documented. Overall, Gram-positive pathogens accounted for $n=23$ (79.3%) of infections, while Gram-negative pathogens represented $n=6$ (20.7%).

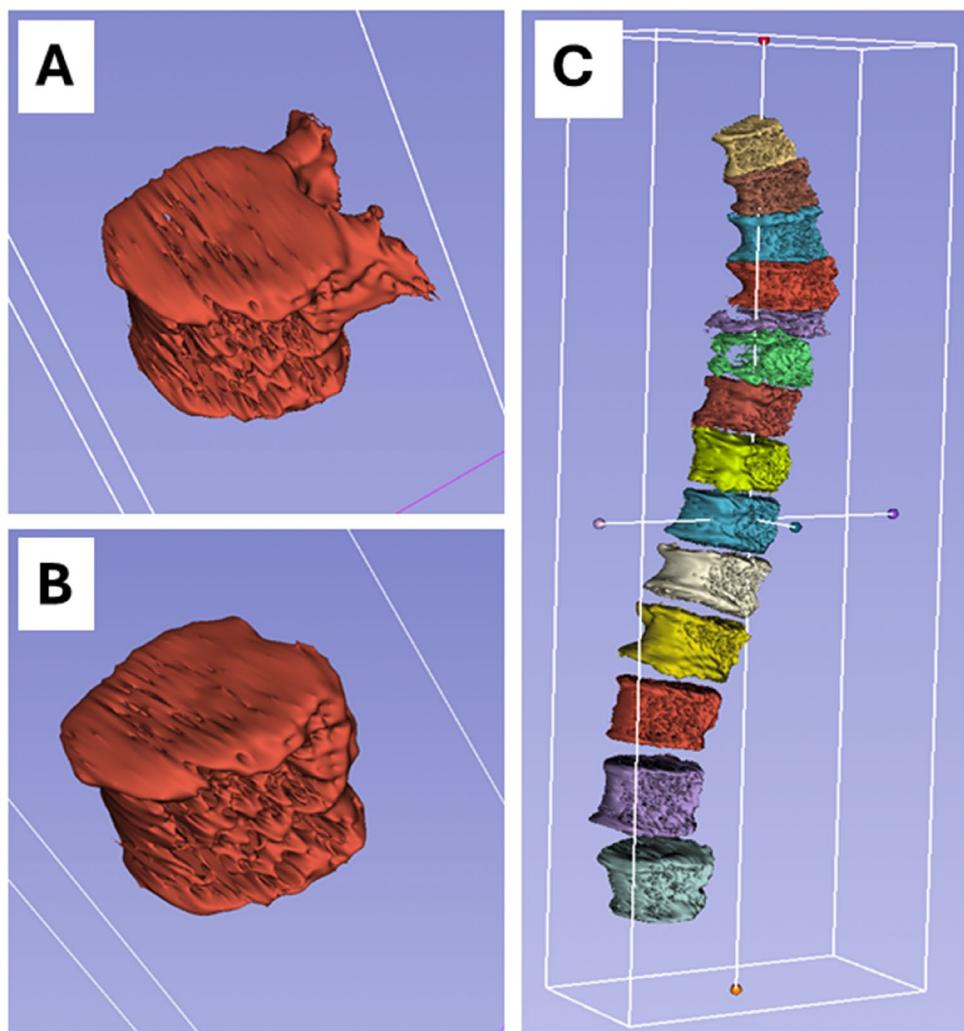
3D volumetry

$N=267$ VBs were measured with 3D volumetry. A total of $n=212$ non-infected VBs were measured, with an average volume of 21.2 cm^3 . The average volume of thoracic VBs was 17.8 cm^3 and 30.7 cm^3 for lumbar VBs. The average volume of each VB is given in Fig. 5.

Linearity assessment of vertebral volume data

The values used to assess linearity were derived from VB volume measurements from the studies by Limthongkul et al. and Molloy et al., as well as the data from this study. For the dataset from Limthongkul et al. [9], an R^2 of 0.96 was obtained for the spinal segment from Th01 to L4. The data from Molloy et al. [19] yielded an R^2 of 0.99 for the segment from Th06 to L4. With an R^2 of 0.95 for Th01–L4 and consistently high values across studies, vertebral body volume growth was assumed linear, justifying the applied regression model.

Fig. 3 Adjustment of a vertebral body (VB) segmentation in 3D Slicer using the “Erase” tool. The image illustrates the refinement process to meet defined criteria. (A) Initial segmentation; (B) Refined segmentation after adjustments. Example of a segmented spine following volume measurement in 3D Slicer. The 5th and 6th thoracic vertebral bodies from the top are identified as the infected segments, highlighted within the segmented model (C)



Linear regression model

The linear regression was conducted case by case and yielded in individual slopes (m). The mean slope for males was 2.3 ± 0.8 (95%CI=1.94–2.75, Fig. 6A), while for females, it was 1.8 ± 0.7 (95%CI=1.46–2.19, Fig. 6B).

Volume of the infected VBs

A methodical example is given in Fig. 7, that shows the individual regression model ($(y)=2.2*x+3.0$; $R^2=0.94$) of a patient with VO of Th9 and Th10.

Vertebral bodies at the lumbar spine (L1–L4) were affected 28 times by PVO, and at the thoracic spine (Th01–Th12) 27 times, giving a mean of $n = 1.8$ affected VBs per included patient. The most frequently affected VBs were L2 (9 cases), followed by L1 and L3 (7 cases each). For the detailed analysis, only the mainly affected VB of an infected segment was considered. Cases with severe degenerative deformation of the infected segment, marked by excess

bone mass in spondylophytes and $DQ > 1$, were excluded from the final analysis as detailed in the methods.

This resulted in a sub-group of $n=20$ PVO patients (female $n=8$ and male $n=12$), aged 67.1 ± 10.0 years and a mean BMI of 31.6 ± 6.2 kg/m 2 . The mean calculated original volume (V_{original}) of these VBs was 24.1 ± 10.5 cm 3 . The mean 3D-volumetric volume of these infected VBs was 17.8 ± 9.3 cm 3 , giving a mean difference of 6.3 ± 5.0 cm 3 and a median DQ 0.81 (0.42–0.93). The average measured volume was significantly smaller than the calculated original volume ($p < 0.001$; Fig. 8).

At the thoracic spine the mean V_{original} was 20.4 ± 8.3 cm 3 and the median loss of vertebral bone was 18% (7–57%). At the lumbar spine the mean calculated initial volume was 30.8 ± 10.3 cm 3 and the median loss of vertebral bone volume was 30% (11–58%). The mean, absolute difference between the calculated original Volume and the measured Volume was significantly higher at the thoracic spine (9.6 ± 6.1 cm 3) compared to the lumbar spine (4.4 ± 2.6 cm 3 ; $p < 0.05$). A moderate, statistically significant positive

Fig. 4 Flowchart illustrating the inclusion and exclusion process for patients with PVO, resulting in 31 patients suitable for 3D volumetric analysis

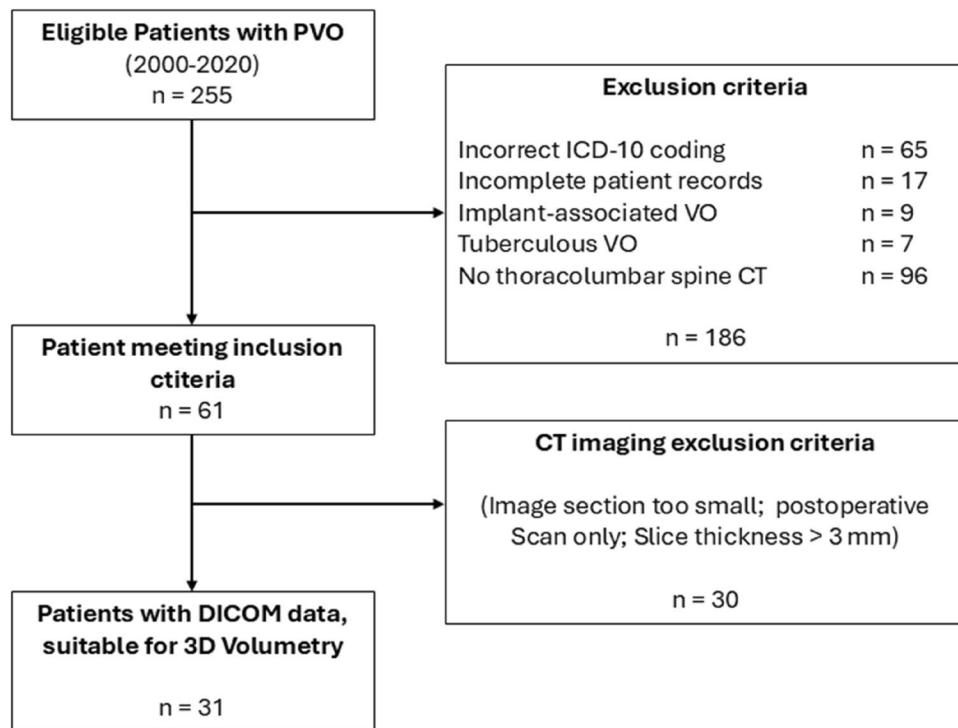
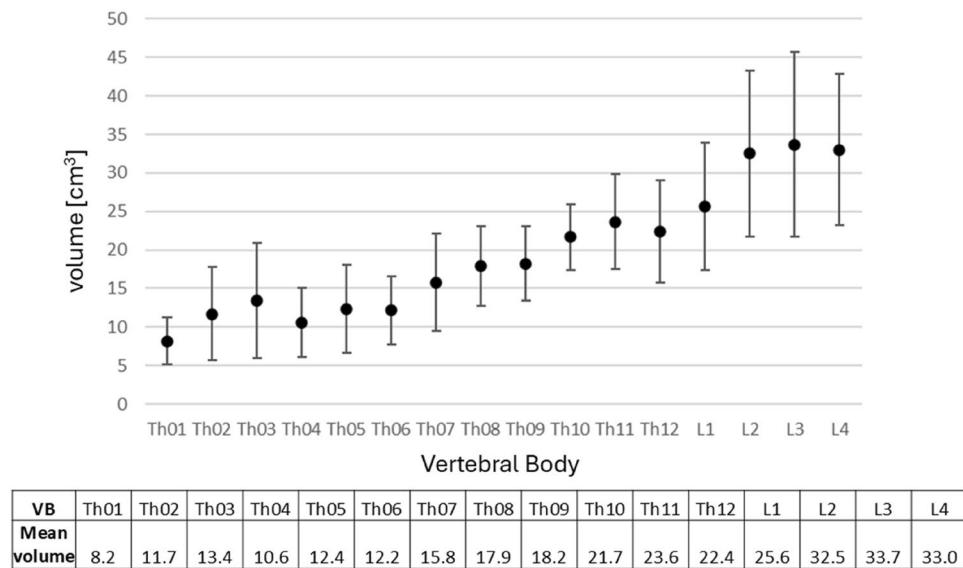


Fig. 5 Mean and standard deviation of vertebral body volumes for all measured vertebrae



correlation of $p=0.427$ between patient sex and the DQ was seen, with female patients loosing 32% (15–57%) of the original bone volume and male patient 12% (7–58%; median DQ male: 0.88 (0.42–0.98); median DQ female: 0.68 (0.43–0.85; $p<0.05$).

While the lowest median DQ of 0.66 (0.65–0.89) was seen in patients with Gram-negative pathogens, the Kruskal-Wallis Test did not show a statistically significant difference between the 4 pathogen groups (STAU; other Gram-positive; Gram-negative; culture-negativew; Fig. 9; $p>0.05$). The median VB volume loss in patients with STAU was

19% ($n=10$; 11–57%), in patients with other Gram-positive pathogens 26% ($n=2$; 11–41%), in patients with Gram-negative pathogens 34% ($n=3$; 11–35%) and in patients with negative cultures 12% ($n=5$; 7–58%). However, only in patients with STAU there was a statistically significant difference between V_{original} and V_{measured} of $6.1 \pm 4.9 \text{ cm}^3$ ($p<0.01$).

Fig. 6 Linear regression slopes for male (A) and female patients (B), presented as functions shown as dotted trendlines. The grey areas represent the 95% Confidence intervals. The central line within the 95%CIs represents the mean slope with x-axis-section ($t=0$)

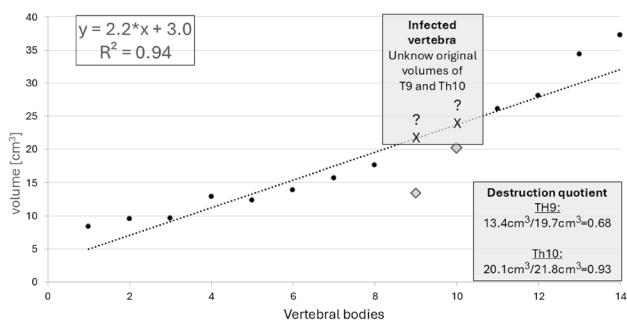
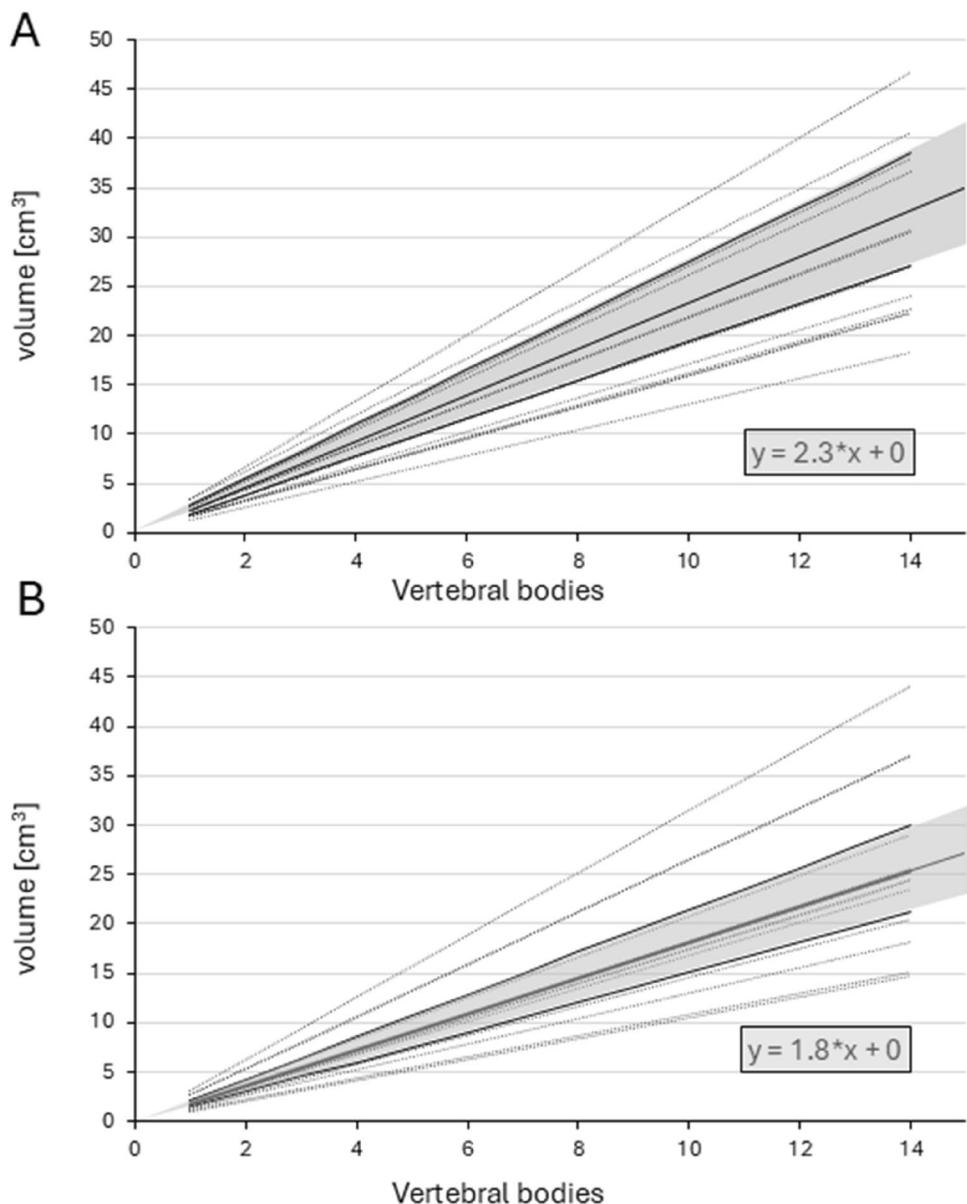


Fig. 7 Example case of a patient with PVO of Th9 and Th10. Although unknown, given the individual equation, the original volume, V_{original} of the two infected VBs was estimated ($V_{\text{original, Th9}} = 19.7\text{cm}^3$ and $V_{\text{original, Th10}} = 21.8\text{cm}^3$) and actual volumes were measured ($V_{\text{measured, Th9}} = 13.4\text{cm}^3$ and $V_{\text{measured, Th10}} = 20.1\text{cm}^3$), enabling calculation of the DQ in the subsequent step

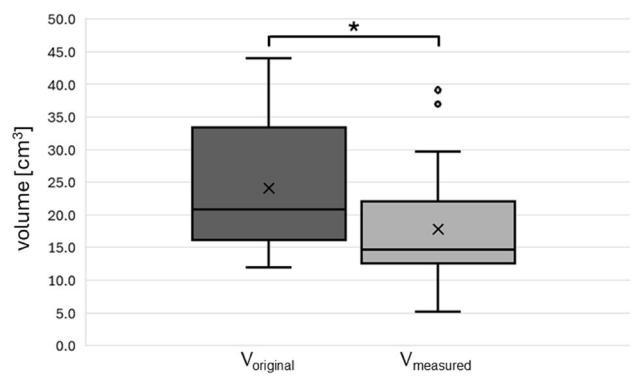


Fig. 8 Boxplots display the median (horizontal line), mean (x), interquartile range (25th–75th percentile), and minimum–maximum values (whiskers). Outliers are shown as individual points. The measured volume (V_{measured}) was significantly smaller than the original calculated volume (V_{original} ; $p < 0.001$, *)

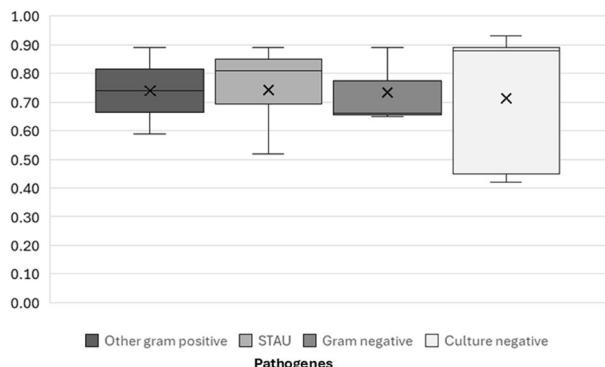


Fig. 9 Boxplots display the median (horizontal line), mean (x), interquartile range (25th–75th percentile), and minimum–maximum values (whiskers) for the DQs of different pathogen groups: Other Gram-positive, STAU, Gram-negative, and Culture-negative

Discussion

This study introduces, for the first time to the authors' knowledge, a patient-specific mathematical approach to estimate the original (pre-infection) vertebral body volume and quantify the extent of osteolytic lesions in PVO. By deriving a linear function based on the cranio-caudal increase in VB volumes and requiring only two accurately measured healthy VBs, the estimation of the “original” volume V_{original} of an infected or otherwise affected vertebra is enabled.

Multiple studies, including those by Limthongkul et al. and Molloy et al., consistently demonstrate a near-linear increase in vertebral body (VB) volume from Th01 to L4, with high coefficients of determination (0.95–0.99) [9, 19]. Our results echo this trend, showing an $R^2=0.95$ for the mean volume of non-infected thoracolumbar VBs, thus reinforcing the linear growth model (Table 1). However, at L5 the literature indicates the linearity breaks off, likely due to anatomical and biomechanical variations; hence, simple linear regression is not reliably applicable at that level. Demographic differences in spinal anatomy must be considered, emphasizing the need for an individualized approach supported by suitable demographic data and highlighting the potential importance of big data integration. It must also

be taken into account that the assumed linearity may be limited in patients with congenital anomalies, postoperative or degenerative changes.

Accurate bone volume estimation is crucial for diagnosis and treatment. In vertebroplasty or kyphoplasty, it helps determine the optimal bone cement volume for structural restoration [19]. This quantification is likewise valuable in the context of osteolytic diseases, including not only PVO but also metastatic tumor infiltration, where volumetric bone destruction necessitates robust classification systems and clearer indications for operative intervention [23]. By integrating an automatically derived bone volume DQ, more nuanced, patient-specific classifications could be developed, and prognostic accuracy could be increased.

Manual segmentation in open-source software, such as 3D Slicer, remains time-consuming and prone to interobserver variability [17, 24]. However, modern segmentation tools leverage artificial intelligence (AI) and advanced imaging algorithms to achieve near real-time volumetric analyses [11]. Chen et al. evaluated the accuracy and efficiency of Brainlab's Elements SmartBrush Spine for auto-segmenting clinical target volumes (CTV) in spine SBRT, showing high consistency and agreement with clinically used CTVs while reducing inter-person variation and contouring time [25]. Integrating the linear regression model into a software pipeline would potentially facilitate automated measurement of an adjacent healthy VB, application of the patient-specific equation, and generation of the DQ. This approach holds promise for integration into AI-based algorithms that can automatically detect vertebral fractures on routine CT scans [26]. Subtle volumetric changes or occult fractures could thus be flagged early, guiding clinicians to intervene in osteoporotic patients and those with multiple myeloma, metastatic lesions, or spinal infections.

Incorporating Houndsfield Units (HU) data could refine the model. Dieckmeyer et al. identified level-specific vBMD thresholds with opportunistic QCT, showing strong links between low vBMD and fracture risk [27]. Similarly, a combined volumetric-HU model could thus capture both the quantity and quality of lost bone mass,

Table 1 Comparison of vertebral volume measurement methods, sample sizes, and results

Study	Molloy et al., 2003 [19]	Odaci et al., 2003 [20]	Komemushi et al., 2009 [21]	Limthongkul et al., 2010 [22]	Caula et al., 2016 [23]	Current Study
Method	Archimedes' Principle	Cavalieri principle	CT -Volumetry	CT -Volumetry BrainLab	Cylinder: $V=\pi R$	CT Volumetry: 3D Slicer
Number of Patients/Vertebrae	10 P/120 VB	2 P/10 VB	8 P/104 VB	40 P/680 VB	129 P/645* VB	31 P/267 VB
Total Spine [cm ³]	(from T6) 29.5	–	(from T5) 26.3	(excl. L5) 20	–	(excl. L5) 21.2
Thoracic Spine [cm ³]	–	–	–	15	–	17.8
Lumbar Spine [cm ³]	(incl. L5) 41	(incl. L5) 30.6	–	(excl. L5) 35.1	(excl. L5) 35.5	(excl. L5) 30.7

improving PVO patient risk stratification for collapse and other complications.

In the current cohort, there was no statistically significant difference in the DQ with respect to pathogen type. These subgroup comparisons should be interpreted with caution due to the small number of patients within each pathogen group, which may limit statistical power and generalizability. Nevertheless, only in patients with STAU infection the differences between V_{original} and V_{measured} were statistically significant. It is plausible that more aggressive organisms like the high-virulent STAU induce earlier or more extensive bone volume loss [28, 29]. Widaa et al. highlighted, that *Staphylococcus aureus* protein A plays a pivotal role in osteomyelitis by inducing bone destruction and bone loss through apoptosis of osteoblasts, inhibition of bone formation, and activation of osteoclasts via soluble RANKL secretion [28]. Coagulase-negative staphylococci (CONS), with *Staphylococcus epidermidis* exhibit lower virulence [3, 30]. They primarily play a role in chronic and implant-associated infections [31–33].

In fact, the lowest DQ of 0.66 and thus the highest amount of bone loss was seen in the group with Gram-negative pathogens. They are known to exhibit a moderate to high virulence and are associated with high mortality rates in PVO patients [5, 34]. Kang et al. found, that pyogenic spondylitis caused by Gram-negative bacteria was strongly associated with genitourinary and intra-abdominal infections, presenting more frequently with severe sepsis, but with similar mortality and clinical outcomes compared to Gram-positive cocci [35].

However, the factors for severe courses of disease in PVO are multifold [4, 5, 36]. The relationship between pathogen virulence, immune status, and bone health in vertebral osteomyelitis is complex [37, 38].

Integrating osteoimmunology and bone turnover differentiation may improve understanding and treatment. This study provides a framework for quantifying bone loss, with future research linking volumetric data to pathogen virulence and patient-specific turnover to elucidate infection-driven bone destruction.

All in all, this study could lay the foundation for assessing the severity of the course of the disease by determining the degree of destruction of an infected vertebral body at the time of diagnosis. From a clinical standpoint, the DQ may support surgical decision-making by identifying patients at risk of instability. It could serve as a longitudinal marker of structural deterioration and be integrated into predictive models to guide stabilization strategies or intensification of antimicrobial therapy. Correlation with bone metabolism biomarkers may improve understanding of infection-induced osteolysis. Beyond PVO, the approach may also be valuable in osteoporotic fractures and metastatic spinal

disease. The volumetry could be a helpful tool in kyphoplasty planning to estimate the optimal volume of bone cement to be applied.

Limitations

This study is limited by its small sample size and retrospective design, which inherently restricts the generalizability of the findings. Additionally, the cases included CT scans taken at varying stages of disease progression and ongoing therapy, introducing heterogeneity in bone destruction and treatment effects. Furthermore, our regression-based model assumes a linear trend in vertebral body volume along the thoracolumbar spine. This assumption may not hold in patients with congenital anomalies, prior trauma, or advanced degenerative disease, potentially limiting the accuracy of the estimated pre-infection volumes. In addition, metal implants from prior spinal surgeries (e.g., pedicle screws) may create artifacts that impair segmentation accuracy and volumetric calculations. The primary aim was not to evaluate clinical outcomes or treatment modalities. Instead, the focus was on validating the feasibility of the linear regression model and its application in a series of 3D-segmented vertebral bodies, providing a foundation for future, standardized quantification of bone volume loss in pyogenic vertebral osteomyelitis. Future multicenter studies are warranted to validate the robustness and applicability of this method in more diverse patient populations.

Conclusion

We introduce patient-specific equations to estimate pre-infection vertebral volumes, enabling objective bone loss assessment via the “Destruction Quotient.” This method enhances clinical decision-making by supplementing imaging with quantitative metrics. AI-driven segmentation and densitometry may further enhance accuracy and prognostic value.

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Author contributions S.L., M.B., M.K., and V.A. conceived the study (Conceptualization). S.L., M.B., M.K., J.S., and M.S. developed the methodology. M.B., M.H., and G.N. implemented the software. S.L., M.B., J.K., and M.L. conducted validation. S.L., M.B., M.K., and J.S. performed formal analysis. S.L., M.B., J.S., J.K., S.B., and M.S. carried out the investigation. V.A., G.N., S.B., and M.L. supplied resources. M.B., J.K., M.H., M.S., G.N., and S.B. curated the data. S.L. and M.B. drafted the original manuscript. All authors (S.L., M.B., J.S., J.K., S.B., M.S., G.N., M.H., M.L., V.A., M.K.) reviewed and edited the manuscript. S.L., M.B., M.K., and M.H. prepared the visualizations. V.A., M.K., and M.L. provided supervision. S.L., M.B., and M.K. managed project administration. S.L. and M.B. contributed

equally to the project and manuscript. All authors approved the final version of the manuscript and agree to be accountable for its content.

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Data availability The datasets generated and analyzed during the current study are available from the corresponding author upon reasonable request.

Declarations

Competing interests The authors declare that there are no conflicts of interest with respect to the authorship or publication of this article.

Ethical Approval and Informed Consent Statements The study was approved by the institutional ethics committee of the University Medical Center Regensburg (Reference number: 12-218_3-101). Due to the retrospective nature of the analysis of fully anonymized data, the requirement for written informed consent was waived by the committee.

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