
Denoised Ultra-Low-Dose CT for Pneumonia Detection



Pulmonary infections pose a significant risk to immunocompromised individuals, contributing to high morbidity and mortality rates. Early and accurate detection of pneumonia is critical to guiding appropriate treatment, particularly in cases of invasive fungal pneumonia, which requires distinct therapeutic strategies. Computed tomography (CT) is widely regarded as the reference standard for imaging in immunosuppressed populations due to its superior ability to detect infection-related abnormalities. However, the potential long-term risks of radiation exposure from repeated CT scans are a major concern, particularly for younger patients and those requiring frequent imaging. Even low-dose CT scans contribute significantly to cumulative radiation exposure, prompting the development of ultra-low-dose CT (ULDCT) techniques to mitigate these risks.

ULDCT significantly reduces the effective radiation dose while maintaining diagnostic capability. However, its widespread clinical application has been hindered by increased image noise, which reduces the clarity of anatomical structures and diagnostic confidence. Recent advances in artificial intelligence have introduced deep learning-based noise reduction techniques, enhancing image quality and allowing for further reductions in radiation exposure without compromising diagnostic performance. A recent review published in *Radiology: Cardiothoracic Imaging* evaluates the effectiveness of denoised ULDCT in assessing pneumonia in immunocompromised individuals, comparing its diagnostic accuracy with normal-dose CT.

Diagnostic Accuracy of Denoised ULDCT

This prospective study enrolled 54 immunocompromised adults who presented with fever and were referred for chest CT scans to assess potential pneumonia. Each participant underwent both a normal-dose CT scan and an ULDCT scan, followed by AI-based noise reduction of the ULDCT images. The normal-dose CT served as the reference standard for diagnosis, and two radiologists, blinded to clinical information, reviewed the images for signs of pneumonia and associated pulmonary findings.

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The results demonstrated that denoised ULDCT offered improved diagnostic accuracy compared with raw ULDCT. It effectively identified individuals with and without pneumonia and enhanced the detection of specific infection patterns, including fungal, bacterial, viral and *Pneumocystis jirovecii* pneumonia. The AI-based denoising process helped reduce false-positive results caused by excessive image noise in raw ULDCT, thereby improving specificity and overall diagnostic confidence. Sensitivity and specificity for various infection types were higher with denoised ULDCT than with unprocessed ULDCT, bringing its performance closer to that of normal-dose CT. This suggests that the integration of AI-driven noise reduction could enhance the reliability of ULDCT as a diagnostic tool for pneumonia in immunocompromised patients.

Image Quality Improvement and Feature Detection

A key advantage of denoised ULDCT over standard ULDCT was its ability to improve image quality, particularly in the visualisation of fine pulmonary details. Specific features associated with pneumonia, such as macronodules, small nodules and ground-glass opacities, were more accurately detected following the application of noise reduction. Denoised ULDCT provided clearer differentiation of these abnormalities, allowing for a more precise classification of infection types and severity.

Further improvements were observed in the detection of tree-in-bud opacities and interlobular septal thickening, both of which are important indicators of specific lung infections. The AI-enhanced images demonstrated increased sensitivity for these fine structures, which were often obscured by noise in standard ULDCT. Signal-to-noise ratio measurements confirmed a statistically significant improvement in image clarity, reinforcing the efficacy of denoised ULDCT in enhancing diagnostic precision. Given that these features play a critical role in differentiating

between various pulmonary conditions, the ability to visualise them more clearly strengthens the potential of denoised ULDCT as a viable alternative to normal-dose CT in pneumonia assessment.

Radiation Dose Reduction and Clinical Implications

One of the primary benefits of ULDCT is its substantial reduction in radiation exposure compared with normal-dose CT. In this study, the median radiation dose for ULDCT was only 1.95% of that of normal-dose CT (0.12 mSv vs 6.15 mSv). This drastic decrease in radiation burden is particularly relevant for immunocompromised patients, many of whom require repeated imaging for infection monitoring and disease progression assessment. Reducing radiation exposure in these individuals is crucial for minimising long-term health risks associated with cumulative radiation doses.

Despite its advantages, the use of raw ULDCT has been limited due to the loss of image clarity caused by high noise levels. The findings of this study indicate that AI-based noise reduction can address this limitation effectively, preserving image quality while maintaining the radiation-saving benefits of ULDCT. Denoised ULDCT was particularly beneficial in reducing false-positive diagnoses caused by image noise, ensuring that patients without pneumonia were not mistakenly classified as infected. Given these results, denoised ULDCT could serve as a practical alternative to normal-dose CT in clinical scenarios where repeated imaging is required. Its adoption in routine practice could provide a safer imaging solution without sacrificing diagnostic accuracy, making it an attractive option for managing pneumonia in immunocompromised individuals.

Denoised ULDCT represents a significant advancement in pneumonia detection for immunocompromised patients. By integrating AI-driven noise reduction with ultra-low radiation exposure, it successfully enhances diagnostic accuracy while addressing concerns related to repeated CT imaging. The study findings suggest that denoised ULDCT not only improves pneumonia detection but also enhances the visualisation of critical pulmonary features that aid in distinguishing infection types. Its ability to reduce radiation exposure without compromising diagnostic precision makes it a promising tool for infection monitoring in vulnerable patient populations.

Further research with larger patient cohorts is necessary to validate these findings and expand the clinical applicability of denoised ULDCT. Nonetheless, the study provides compelling evidence that AI-enhanced ULDCT could play a crucial role in optimising pneumonia diagnosis and patient safety. With continued refinement, this approach could contribute to improved imaging protocols for immunocompromised individuals, offering a more effective balance between radiation risk and diagnostic performance.

Source: [Radiology: Cardiothoracic Imaging](#)

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