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Improving the radiological diagnosis of hepatic artery thrombosis after liver transplantation: Current approaches and future challenges.

Future challenges in HAT diagnosis.

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Abstract

Hepatic artery thrombosis (HAT) is a devastating vascular complication following liver transplantation, requiring prompt diagnosis and rapid revascularization treatment to prevent graft loss. At present, imaging modalities such as ultrasound, computed tomography, and magnetic resonance play crucial roles in diagnosing hepatic artery thrombosis (HAT). Although imaging techniques have improved sensitivity and specificity for HAT diagnosis, they have limitations that hinder the timely diagnosis of this complication. In this sense, the emergence of artificial intelligence (AI) presents a transformative opportunity to address these diagnostic limitations. The development of machine learning algorithms and deep neural networks has demonstrated the potential to enhance the precision diagnosis of liver transplant complications, enabling quicker and more accurate detection of HAT. This article examines the current landscape of imaging diagnostic techniques for HAT and explores the emerging role of AI in addressing future challenges in the diagnosis of HAT after liver transplant.

Key Words: liver transplantation; postoperative complications; hepatic artery; thrombosis; radiology; artificial intelligence.

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Core Tip: HAT is a severe vascular complication after liver transplant requiring prompt diagnosis and intervention to prevent graft loss and patient death. However, current imaging methods have limitations. AI, especially deep learning, holds promising potential to enhance precise and accurate HAT diagnosis. This article explores current HAT imaging techniques and highlights the potential role of AI-based methods, aiming to improve diagnostic performance and recipient survival.

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INTRODUCTION

Liver transplantation has emerged as the treatment of choice for patients with end-stage liver diseases (ESLD), including advanced stages of both cholestatic and non-cholestatic cirrhosis, as well as the early stages of hepatocellular carcinoma (HCC)^[1-3].

In recent years, there has been a sustained increase in liver transplant cases, resulting in improved survival prognoses and quality of life for ESLD patients^[4,5].

Continuous progress in the development of surgical techniques and novel immunosuppressive agents has contributed to enhanced survival rates among recipients^[6,7], with a current five-year survival rate of up to 75%^[8-10]. However, strict postoperative multidisciplinary surveillance is imperative to identify and address potential complications that may affect both the graft and the recipient^[11,12].

Despite ongoing advancements in the field of liver transplantation, postoperative vascular complications, particularly those related to the hepatic artery (HA), remain one of the primary causes of graft failure and recipient mortality^[11].

Hepatic artery thrombosis (HAT) is a severe complication after liver transplantation, associated with biliary complications such as ischemic cholangiopathy, which may occur even after a successful revascularization treatment, resulting in late graft loss and therefore having a critical impact on quality of life^[12,13]. Furthermore, HAT is considered as a risk factor for development of biliary stones in liver graft, which is associated with recurring cholangitis, secondary biliary cirrhosis, and graft failure^[14,15].

HAT can be classified according to its temporal onset. Thrombotic occlusion of the HA occurring within the first 30 days following liver transplantation is classified as early HAT (eHAT), which is believed to result from technical problems and perioperative risk factors such as artery kinking, donor arterial anatomic variation, different diameters of the arteries in the anastomosis, or low quality of the donor's or recipient's arteries^[16,17].

On the other hand, the later development of HAT, known as late HAT, is usually related to ischemic or immunologic risk factors such as Cytomegalovirus-positive donors and hepatitis C seropositive recipient^[17-19].

A large study including 4234 cases of adult and pediatric liver transplants reported an overall HAT incidence of 5%, which was higher in pediatric liver transplant recipients than in adults (8% *vs* 3.9% respectively)^[20]. In addition, a systematic review comprising 21822 cases of orthotopic liver transplantation (OLT), reported an overall incidence of eHAT of 4.4% with an overall mortality of 33.3%, which was also significantly higher in children (34.3%) than in adults (25%)^[16].

Strikingly, the cause of this difference remains unknown. Nevertheless, the most likely explanation is the small size of the vessels and the associated technical difficulties of anastomosing^[16,21,22].

The reported incidence of late HAT is highly variable, ranging widely from 1% to 25%, with mortality rates of 50%^[20,23].

In addition, median times reported to diagnosis of eHAT were 6.9 days (range: 1–17.5 days postoperative), while for late HAT, median times were 6 months, ranging from 1.8 to 79 months^[16].

The clinical presentation of HAT widely varies according to the timing of onset and the development of collateral vessels, which could maintain blood flow to the allograft^[17,20]. Clinically, eHAT manifests with fever, abdominal pain, elevated transaminases, and leukocytosis, which can be followed by septic shock. ^[20,24–27].

Late HAT has an insidious course, characterized by progressive abdominal pain, alteration of liver function tests, relapsing fever, recurring cholangitis, and bacteremia^[27,28].

Color-Doppler ultrasound (CDUS) is the modality of choice for the postoperative surveillance of liver graft vasculature during the postoperative period, which could depict hemodynamic changes that require further assessment with second-line diagnostic tools such as computed tomography angiography (CTA) or conventional hepatic arterial angiography^[17,29,30].

Currently, there are three different modalities for HAT treatment: retransplantation, surgical revascularization, and endovascular revascularization^[17,25]. However, the most effective treatment approach remains controversial^[24]. Bekker *et al*^[16], reported that retransplantation was more frequently performed in pediatric liver transplant recipients (61.9%) than in adults (50%), and was the treatment of choice in the overall cases of eHAT.

In another large study, retransplantation was performed in 71% of patients with eHAT and 51% of patients with late HAT^[20].

CTA is the second line of choice when a hemodynamic HA abnormality is suspected on DUS evaluation^[29,31]. The interpretation of CTA still requires a detailed evaluation of all abdominal vascular structures, which is a time- and labor-intensive process that requires high expertise in abdominal imaging.

Although several studies have reported the high sensitivity of CTA for HAT diagnosis, its specificity remains somewhat low (83.5-87.5%)^[32,33].

In this sense, considering the invasiveness and risk of diagnostic angiography, which is the current gold standard for HAT confirmation, it is necessary to improve the diagnostic performance of CTA^[34].

The emergence of artificial intelligence (AI), particularly deep learning (DL) algorithms, is gaining growing attention for its performance in image-recognition tasks, achieving high performance on CTA analysis^[35-37].

Recent studies have developed different DL-based algorithms, which have resulted in shorter time and high diagnostic performance for CTA diagnosis of vessel occlusion at different anatomic sites, thus improving management outcomes of vascular time-dependent pathologies^[37-41].

This article explores the current landscape of multimodality imaging for HAT, highlighting the potential of DL-based algorithms as emerging technologies that could improve HAT diagnosis post-liver transplantation

MULTIMODALITY IMAGING OF HEPATIC ARTERY THROMBOSIS AFTER LIVER TRANSPLANTATION.

Ultrasound (US) evaluation is the modality of choice for assessing liver graft vasculature. It offers a rapid, comprehensive, and accurate grayscale and CDUS evaluation of liver parenchyma at the patient's bedside^[29,42], which allows precise

assessment of the entire graft vasculature, particularly the blood flow in the HA anastomosis^[28].

³ The arterial anastomosis is typically located in the porta hepatis and can be identified by the presence of intense focal color aliasing and elevated velocity on spectral Doppler images surrounding the porta hepatis^[43]. Normal Doppler evaluation of the HA shows a continuous diastolic flow with a rapid systolic upstroke, an acceleration time of less than 80 msec, and a resistive index (RI) that ranges between 0.5 and 0.7^[44]. (Figure 1).

In 1996, Nolten and Sproat^[45] described some qualitative hemodynamic changes in the HA that may anticipate its thrombotic occlusion, including the loss of diastolic flow, dampening of the systolic peak, and finally, the complete loss of arterial flow. The detection of low-velocity and high-resistance flow, nonvisualization, or absence of Doppler color flow in the HA and its intrahepatic branches are findings highly suggestive of HAT^[24,46], requiring prompt assessment using CTA or conventional hepatic angiography^[44,47].

CTA plays a crucial role in the detection of HAT following liver transplantation. Its high-resolution, contrast-enhanced images provide detailed anatomical information, making it a crucial tool for diagnosing HAT, allowing the assessment of vessel patency and identification of thrombus formation, as well as the evaluation of collateral circulation and ischemia-related biliary complications such as biloma and abscess^[48,49].

The lack of opacification of HA and its intrahepatic branches strongly suggest eHAT. However, it should be confirmed in specific detail with maximum intensity projection (MIP) images ^[25,48]. (Figure 2). On the other hand, the development of collaterals, mainly raised from the phrenic arteries, and the temporal onset are crucial signs that suggest the diagnosis of late HAT^[11,16].

Magnetic resonance (MR) offers detailed images of the graft parenchyma and biliary ducts within the postoperative surveillance period (Figure 3). However, it may be less readily available and time-consuming compared to the US and CTA^[50]. In addition, retrospective studies have reported similar diagnostic accuracy to US but with a higher number of false positives and a more demanding examination^[51].

As mentioned above, hepatic arterial angiography is considered the gold standard for the diagnosis of HAT, which can involve diagnostic and therapeutic approaches for the endovascular management of this complication^[52] (Figure 4).

EMERGING ROLE OF DEEP LEARNING IN HEPATIC ARTERY THROMBOSIS DIAGNOSIS.

AI has emerged as a revolutionary technology with a critical impact on the field of medicine. By enhancing diagnostic accuracy, improving efficiency, and enabling early detection of diseases, its continued integration into radiology practices holds the promise of further improving patient care and advancing our understanding of complex diseases^[53,54].

Recent research showed that AI-based technology can significantly support the field of liver transplantation by optimizing organ allocation, donor-recipient matching, survival prediction analysis, and the diagnosis of postoperative complications in liver graft recipients^[55,56].

As mentioned above, AI algorithms have improved the analysis of medical images, detecting subtle abnormalities, quantifying disease progression, and identifying patterns that might be challenging for human radiologists to discern^[57].

DL is a subfield of AI based on neural networks inspired by the human brain structure. It focuses on using artificial neural networks with multiple layers, often referred to as deep neural networks, to model and solve complex tasks and approximate very complex nonlinear relationships^[57,58].

Convolutional Neural Networks (CNNs) are a type of DL artificial neural network specifically designed for processing and analyzing visual data, such as images and videos, for tasks involving visual perception. Therefore, the emerging CNN algorithms have had a profound impact on the field of radiology, revolutionizing the way medical images are interpreted, analyzed, and utilized for diagnosis and treatment planning^[59].

Currently, the increasing use of CNN algorithms in medical image analysis has demonstrated interesting results in improving rapid frontline CTA detection of life-threatening large vessel occlusion, with promising diagnostic performance^[60-64].

Tajbakhsh *et al* investigated the feasibility of a novel CNN algorithm as an emergent mechanism to improve the diagnosis of thromboembolism detection, showing that their DL algorithm outperforms classic machine learning techniques with a sensitivity of 83% for detecting thromboembolism on CTA^[65,66]. Additionally, they also developed a novel computer-aided embolism diagnosis system, providing radiologists with an effective visualization tool to conveniently examine the vessel lumen from multiple perspectives and confidently report filling defects. Their vessel-oriented image representation (VOIR) offers a multi-view representation of the embolus, summarizing the 3D contextual information around it^[67].

Huan *et al*, developed the PENet-3D CNN model to detect thromboembolic occlusion using the entire volumetric CTA imaging data, achieving an AUROC of 0.85^[68]. Later, they optimize their model by integrating clinical data from the electronic medical record to achieve 0.87 [95%CI: 0.871-0.875], 0.87 [95%CI: 0.872-0.877], and 0.947 [95%CI: 0.946-0.948] of sensitivity, specificity, and AUROC respectively, for the task of automatically detecting thromboembolism on volumetric CTA image analysis^[69].

Ma *et al* proposed a new DL model for embolism detection using the CNN-based network Gradient-weighted Class Activation Mapping (Grad-Cam), a localization technique that provides visual explanations on CTA scans. The algorithm achieved a sensitivity of 0.86 with a specificity of 0.85^[70], which is competitive with radiologists' sensitivities ranging from 0.67 to 0.87 and specificities of 0.89–0.99 for embolism detection on CTA^[71].

A recent multicenter study was performed to validate a DL-based application designed with CNN (CINA-PE), to automatically detect embolism on CTA and alert radiologists for urgent interpretation. This algorithm achieved a sensitivity of 91.4% (95%CI: 86.4–95.0%) and specificity of 91.5% (95%CI: 86.8–95.0%), leading to an accuracy of 91.5%^[72]. Additionally, an automated CNN-based algorithm designed by Fu *et al*, that could be trained to complete lumen segmentation automatically reduced the radiologists' report writing time of CTA from 28.8 minutes \pm 5.6 to 12.4 minutes \pm 2.0. Therefore, it offers a time-saving and accurate method to analyze CTA to provide optimized clinical workflow^[39].

These experiments further confirm the potential of DL algorithms for medical imaging applications^[59]. In particular, the implementation of CNN-based algorithms for image analysis in patients with high clinical suspicion of thrombotic occlusion of HA within the perioperative period could improve the diagnostic performance of the radiologist, optimizing its sensitivity, specificity, and report writing time. Thereby leading to an early and efficient multidisciplinary workflow and therapeutic response, ultimately improving patient prognosis.

CONCLUSION

Despite the continuous advances in the field of liver transplantation, HAT remains a significant cause of morbidity and mortality in recipient patients. While there are different imaging studies that allow the assessment of the HA, they have limitations that prevent an early diagnosis of this complication

AI can potentially revolutionize HAT detection by enhancing the interpretation of imaging data and facilitating rapid and precise diagnosis. The integration of AI into existing imaging modalities, such as CTA, holds the potential to streamline clinical workflows, reduce healthcare costs, and ultimately improve patient outcomes.

Future investigations should be focused on improving the diagnostic performance of non-invasive imaging techniques for life-threatening diseases. HAT is a severe complication that significantly increases the risk of graft loss and patient mortality. In this regard, emergent DL-based algorithms have demonstrated high diagnostic performance for arterial occlusion at different anatomical sites. Considering these findings, the development of new DL algorithms focused on the CTA analysis of the liver graft vasculature could assist radiologists in improving sensitivity, specificity, and diagnostic reporting time for HAT, thus enhancing early treatment for this time-dependent complication

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