



PERSONALIZED HCC CARE THROUGH AI INNOVATIONS: A MINI REVIEW

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Abstract: Hepatocellular carcinoma (HCC) continues to pose a significant global health challenge, particularly due to the limited treatment options available in its advanced stages. Emerging technologies such as artificial intelligence (AI), including machine learning (ML) and deep learning, have shown considerable promise in transforming the landscape of HCC diagnosis, prognosis, and therapeutic strategies. This review highlights the latest progress in AI applications for HCC, emphasizing personalized treatment, biomarker identification, imaging-assisted diagnostics, prediction of treatment outcomes, and tools for clinical decision-making. AI-based approaches contribute to greater precision and efficiency in managing HCC, ultimately aiming to improve patient care and survival.

keywords: Artificial intelligence, Biomarker discovery, Hepatocellular carcinoma, Patient outcomes, Personalized medicine.

INTRODUCTION

Hepatocellular carcinoma (HCC) is the most prevalent primary liver cancer and the third leading cause of cancer-related mortality worldwide^[1]. Its high lethality is attributed to late-stage diagnosis, underlying liver disease, and resistance to conventional treatments^[2]. The heterogeneity of HCC, arising from diverse etiological factors such as chronic hepatitis B and C infections, alcohol-related liver disease, and metabolic dysfunction-associated steatotic liver disease (MASLD), complicates treatment strategies^[1]. Given this complexity, personalized therapeutic approaches are essential to improving patient outcomes.

In recent years, artificial intelligence (AI) has brought significant advancements to the field of hepatology by enabling the integration and interpretation of large-scale, diverse datasets, including imaging modalities like CT, MRI, and ultrasound, as well as genomic, histopathological, and clinical data^[3]. AI-powered tools support earlier diagnosis by detecting subtle signs of malignancy in imaging^[2], assist in prognosis by utilizing risk stratification models^[4], and enhance treatment planning by customizing interventions based on tumor biology and patient-specific factors^[5]. Machine learning and deep learning models have also contributed to refining liver cancer staging, predicting treatment response, and identifying novel therapeutic targets, paving the way for precision medicine in HCC management^[3].

I. AI IN EARLY DIAGNOSIS AND DETECTION

1.1 AI in Imaging-Based Diagnosis

Imaging modalities such as contrast-enhanced computed tomography (CT) and magnetic resonance imaging (MRI) play a central role in HCC diagnosis. AI-driven radiomics and DL-based algorithms have significantly improved tumor detection rates by identifying subtle imaging patterns that may be overlooked by radiologists^[5].

1.1.1 Deep Learning for Image Interpretation:

Convolutional neural networks (CNNs), a type of deep learning (DL) model, have demonstrated exceptional ability in analyzing liver imaging scans. These networks extract complex tumor characteristics, such as shape, texture, and vascular patterns, which contribute to more precise differentiation between HCC and benign liver lesions. AI-based systems outperform conventional imaging interpretation by reducing observer variability and enhancing diagnostic confidence^[4].

1.1.2 AI-Enhanced Radiomics:

Radiomics involves extracting quantitative imaging features that reveal tumor heterogeneity and progression patterns. AI-powered radiomics models integrate these imaging biomarkers to predict tumor aggressiveness and potential response to therapy. These advanced models assist in early-stage HCC detection by identifying imaging signatures indicative of malignancy before visible tumor progression occurs^[6].

1.1.3 Integration of Imaging and Biomarkers:

In addition to imaging, AI has been applied to serum biomarker analysis to improve diagnostic accuracy. Traditional biomarkers such as alpha-fetoprotein (AFP) have limited sensitivity, especially in early-stage HCC. AI-driven models analyze multiple serum biomarkers, combining them with imaging data to enhance specificity in HCC detection. These hybrid approaches provide a more comprehensive assessment, ensuring early diagnosis and better patient stratification for treatment planning^[7].

1.2 AI in Histopathology and Biomarker Analysis

Histopathological examination of liver tissue remains the gold standard for confirming HCC. However, manual interpretation of biopsy samples can be time-consuming and subject to inter-observer variability. AI-based computational pathology has revolutionized this field by automating the classification and grading of liver tumors, improving diagnostic precision and efficiency^[8].

1.2.1 Automated Tumor Classification:

AI models trained on large datasets of liver biopsy images can accurately distinguish HCC from other liver malignancies, such as cholangiocarcinoma and metastatic tumors. These models analyze cellular morphology, tissue architecture, and staining patterns to provide highly reliable diagnostic classifications^[9].

1.2.2 Assessment of Microvascular Invasion (MVI):

MVI is a key prognostic factor influencing tumor recurrence and patient survival. AI-powered histopathological analysis can detect MVI with greater accuracy than conventional microscopy. By identifying microscopic vascular infiltration patterns, AI assists pathologists in refining HCC grading, which is crucial for surgical decision-making and treatment planning^[2, 5].

1.2.3 AI-Driven Biomarker Discovery:

Beyond histology, AI has facilitated the identification of novel molecular markers associated with HCC progression. Machine learning models analyze genomic, transcriptomic, and proteomic datasets to uncover genetic mutations, gene expression profiles, and protein signatures that correlate with early tumor development. These discoveries have paved the way for precision oncology, where targeted therapies can be tailored to the molecular characteristics of individual tumors^[10].

1.3 AI for Prognostic Modeling and Risk Stratification

Accurately predicting disease progression and recurrence risk is essential for optimizing HCC treatment and long-term patient management. AI-driven prognostic models integrate diverse clinical, imaging, and molecular data to generate personalized risk assessments^[11].

1.3.1 Predicting Patient Outcomes: AI-based nomograms, which combine radiomics and clinical data, have been developed to estimate survival probabilities for HCC patients. These models assess tumor size, vascular invasion, liver function scores, and treatment history to generate individualized prognostic predictions. Clinicians can use these insights to tailor treatment strategies, balancing curative and palliative interventions^[7, 12].

1.3.2 Recurrence Risk Prediction: One of the major challenges in HCC management is the high recurrence rate following surgical resection and liver transplantation. AI-powered machine learning models analyze preoperative imaging, tumor histology, and molecular biomarkers to identify patients at increased risk of recurrence. These models provide actionable insights that guide post-surgical surveillance and adjuvant therapy decisions^[1, 5].

1.3.3 Stratification for Personalized Treatment: AI algorithms can categorize HCC patients into distinct risk groups based on their tumor biology and clinical profile. This risk stratification enables personalized treatment planning, where high-risk patients may receive aggressive interventions such as combination therapy, while low-risk patients may be monitored with less intensive approaches.

II: AI IN TREATMENT DECISION SUPPORT

Artificial intelligence (AI) has significantly transformed the management of hepatocellular carcinoma (HCC) by enabling personalized treatment selection and optimizing therapeutic strategies. Through advanced machine learning (ML) techniques, AI-driven models analyze patient-specific data—including clinical history, imaging, genomics, and biomarker profiles—to guide treatment decisions. These models enhance the efficacy of surgical interventions, transarterial therapies, systemic treatments, and recurrence prediction, ultimately improving patient outcomes and long-term survival rates.

2.1 AI in Surgical and Transarterial Therapies

AI has played a crucial role in selecting optimal treatment strategies for patients undergoing surgical resection or locoregional therapies such as transarterial chemoembolization (TACE) and transarterial radioembolization (TARE). By leveraging real-world clinical data, AI models can predict patient response to these interventions, assisting clinicians in making informed decisions^[13].

2.1.1 Surgical Decision-Making: AI-powered predictive models assess factors such as liver function, tumor burden, and overall patient health to determine the likelihood of successful surgical resection. These models provide risk stratification by evaluating potential postoperative complications, recurrence risks, and long-term survival rates. By integrating imaging and histopathological data, AI can also predict microvascular invasion, a key factor influencing surgical outcomes^[14].

2.1.2 Transarterial Therapy Optimization: TACE and TARE are widely used for intermediate-stage HCC, but their effectiveness varies among patients. AI-driven models analyze imaging patterns, tumor vascularity, and patient characteristics to identify candidates most likely to benefit from these procedures. By assessing radiological features from contrast-enhanced CT and MRI scans, deep learning algorithms improve the selection of embolization techniques and predict

treatment responses. Furthermore, AI-based real-time monitoring of treatment efficacy allows for adjustments in therapy, enhancing precision in patient management^[15, 16]

2.2 Systemic and Immunotherapy Optimization

The advent of AI in systemic and immunotherapy has paved the way for more personalized treatment strategies, ensuring that patients receive the most effective therapeutic regimens based on their tumor biology and genetic makeup^[3, 16].

2.2.1 Targeted Therapy Prediction: Tyrosine kinase inhibitors (TKIs), such as sorafenib and lenvatinib, are commonly used in advanced HCC, but patient response varies significantly. AI models analyze genomic and clinical data to predict which patients will respond favorably to TKIs, allowing for personalized treatment adjustments. By evaluating molecular signatures, AI can help identify resistance mechanisms and suggest alternative therapeutic options^[4, 13].

2.2.2 Immunotherapy Response Prediction: Immune checkpoint inhibitors (ICIs) have revolutionized HCC treatment, but their efficacy is influenced by factors such as tumor microenvironment and genetic alterations. AI-driven biomarker analysis can assess immune-related gene expression, PD-L1 levels, and tumor-infiltrating lymphocytes to determine which patients are most likely to benefit from immunotherapy. These models also aid in predicting potential immune-related adverse events, facilitating proactive management strategies^[2, 11].

2.3 AI-Driven Biomarker Discovery and Multi-Omics Analysis

AI has enabled the integration of multi-omics data—including genomics, transcriptomics, proteomics, and metabolomics—leading to the discovery of novel biomarkers for HCC diagnosis, prognosis, and treatment response prediction^[8].

2.3.1 Genomic and Transcriptomic Biomarkers:

Machine learning models analyze large-scale sequencing data to identify genetic mutations and expression profiles associated with HCC progression. AI-driven analysis of transcriptomic data has revealed molecular subtypes of HCC that respond differently to therapies, enabling more precise treatment selection^[3].

2.3.2 Proteomic and Metabolomic Insights:

AI-powered proteomics studies have identified circulating protein biomarkers that improve early detection and disease monitoring. Similarly, metabolomic analyses supported by AI have uncovered metabolic signatures linked to HCC development, helping differentiate aggressive tumors from indolent ones^[5, 11].

2.3.3 Predictive Biomarker Models:

AI integrates data from multiple biological layers to generate predictive models that estimate treatment response and disease progression. These models enhance precision oncology approaches by stratifying patients based on molecular characteristics, enabling early intervention and more effective therapeutic strategies

2.4 AI for Recurrence Prediction and Long-Term Outcome Assessment

Recurrence after curative treatment remains a major challenge in HCC management. AI-driven prognostic models have shown superior performance in predicting recurrence risks and long-term survival compared to traditional staging systems such as the Barcelona Clinic Liver Cancer (BCLC) classification^[13].

2.4.1 Post-Surgical Recurrence Prediction:

AI algorithms analyze preoperative imaging, histopathological features, and molecular data to predict the likelihood of tumor recurrence following surgical resection. These models help clinicians implement tailored surveillance strategies and adjuvant therapies to reduce recurrence risks^[8].

2.4.2 TACE and Ablation Outcome Prediction:

Machine learning models incorporating radiomics, serum biomarkers, and clinical parameters can forecast HCC recurrence following TACE or radiofrequency ablation (RFA). By identifying high-risk patients, AI assists in optimizing follow-up strategies and recommending combination therapies for improved long-term outcomes^[2].

2.4.3 AI-Based Radiomics for Survival Prediction:

AI-driven radiomic models extract quantitative imaging features that go beyond traditional radiological assessments. These models assess tumor heterogeneity, vascular invasion, and treatment response, providing non-invasive predictions of overall survival and disease progression^[17].

III: CHALLENGES AND FUTURE DIRECTIONS

Despite significant advancements in artificial intelligence (AI) for hepatocellular carcinoma (HCC) management, several obstacles must be overcome before these technologies can be widely implemented in clinical practice. These challenges span data quality, regulatory considerations, clinical adoption, and the need for further validation studies. Addressing these issues will be crucial for ensuring AI's reliability, safety, and effectiveness in real-world applications.

3.1 Data Standardization and Model Generalization

A major limitation in AI-driven HCC management is the variability in data sources, formats, and quality. Clinical datasets, including medical imaging, laboratory results, and patient records, are often collected using different protocols across hospitals and research centers. Imaging data, for instance, can vary in resolution, contrast settings, and scan timing, leading to inconsistencies when training AI models. Additionally, genomic and biomarker datasets may be derived from diverse populations with different underlying risk factors, making it difficult to generalize AI predictions across patient groups. Establishing standardized data collection and annotation frameworks, along with federated learning approaches that enable AI models to train on distributed datasets without compromising privacy, can improve model robustness and reliability^[4, 10].

3.2 Regulatory and Ethical Considerations

AI deployment in clinical settings must adhere to strict regulatory guidelines to ensure patient safety and transparency. One of the primary concerns is the interpretability of AI models, especially deep learning algorithms, which often function as "black boxes" with unclear decision-making processes. Clinicians may be hesitant to trust AI-generated recommendations unless they are explainable and backed by strong clinical evidence. Regulatory agencies, such as the U.S. Food and Drug Administration (FDA) and the European Medicines Agency (EMA), require rigorous validation before approving AI-driven diagnostic or therapeutic tools. Furthermore, ethical considerations such as data privacy, potential biases in AI models, and equitable access to AI-driven healthcare solutions must be addressed to prevent disparities in cancer diagnosis and treatment^[3,11].

3.3 Integration into Clinical Workflows

For AI to have a meaningful impact on HCC management, it must be seamlessly integrated into existing clinical workflows. Many hospitals and healthcare systems rely on electronic health records (EHRs) and radiology information systems that may not be compatible with AI tools. Developing AI models that can work alongside existing clinical decision support systems (CDSS) is essential to ensuring smooth adoption. Additionally, real-time AI predictions need to be presented in a user-friendly format that enables clinicians to make informed decisions quickly. Training healthcare professionals to interpret AI-generated insights and incorporating AI education into medical curricula can further facilitate adoption^[13,17].

3.4 Prospective Validation and Large-Scale Clinical Trials

Most AI applications in HCC have been developed and tested on retrospective datasets, limiting their real-world applicability. To ensure reliability, AI models must undergo large-scale prospective validation in diverse patient populations. Multicenter clinical trials are necessary to evaluate the accuracy, safety, and effectiveness of AI-driven diagnostic and prognostic models. These studies should also assess AI's impact on patient outcomes, such as earlier tumor detection, improved treatment selection, and long-term survival rates. Collaborations between AI researchers, hepatologists, oncologists, and regulatory agencies will be vital in conducting such trials and translating AI advancements into routine clinical care^[5].

3.5 Advancing Explainable AI and Multi-Omics Integration

The future of AI in HCC management lies in developing interpretable models that provide clinicians with clear, actionable insights. Explainable AI (XAI) techniques, such as attention mechanisms in deep learning or feature importance analysis in machine learning, can enhance transparency by showing how AI arrives at specific conclusions. Additionally, integrating multi-omics data—including genomics, proteomics, metabolomics, and transcriptomics—can offer a comprehensive understanding of HCC biology, leading to more precise risk stratification and treatment personalization. AI-driven models that incorporate these diverse data sources could uncover novel therapeutic targets and biomarkers, further advancing precision oncology for liver cancer^[11].

Future research should focus on developing robust, explainable AI models that integrate multi-omics data, ensuring holistic patient assessment. Prospective validation studies with diverse patient cohorts will be crucial for refining AI algorithms and establishing their clinical utility in HCC management^[17].

CONCLUSION:

AI has revolutionized HCC management by enhancing early detection, prognostic assessment, and personalized treatment strategies. Through the integration of advanced imaging analysis, biomarker discovery, and predictive modeling, AI offers a powerful tool for optimizing patient care. As AI methodologies continue to evolve, their incorporation into hepatology holds great promise for improving survival outcomes and refining therapeutic strategies. Addressing current challenges through interdisciplinary collaboration, data harmonization, and regulatory compliance will be essential for the successful clinical adoption of AI-driven innovations in HCC management.

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