

# Artificial Intelligence in Decoding ADHD (Attention Deficit Hyperactivity Disorder)

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## Abstract

Attention Deficit Hyperactivity Disorder (ADHD) is a common neurodevelopmental disorder that affects both children and adults, is characterized by symptoms of inattention, hyperactivity, and impulsivity. Globally, ADHD affects approximately 5-7% of children and adolescents, as well as 2.5% of adults. The condition significantly impacts daily functioning, social relationships or interactions, and academic or professional performance. ADHD appears differently in different age groups, with classroom disruptions and difficulty in social interactions being common in children, while adults may struggle with time management, organizing, and maintaining interpersonal connections. The disorder is categorized into three main types: mainly inattentive, mostly hyperactive-impulsive, and combination presentation, which are defined by dominant symptom patterns. Diagnosis involves ongoing symptom assessment, with treatment options including medications, behavioural therapies, and management strategies tailored to individual needs. The growing use of Artificial Intelligence (AI) in healthcare has significantly improved ADHD diagnosis and treatment, offering higher precision, efficiency, and personalization. AI algorithms enhance diagnostic accuracy by analyzing large datasets and identifying complex patterns in medical images, enabling early detection of ADHD and related conditions. Moreover, AI-driven treatment plans personalize therapeutic techniques based on individual patient data, improving outcomes and reducing adverse side effects. Benefits of AI include improved diagnostic accuracy, increased efficiency through automation, development of personalized medicine, and reduced healthcare costs. This review explores the role of AI in ADHD diagnosis and treatment, focusing on its transformative potential in improving patient care and advancing precision medicine. Understanding AI applications in healthcare can lead to way for more effective ADHD therapy management and improved patient quality of life.



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## INTRODUCTION

Attention Deficit Hyperactivity Disorder (ADHD) is a common condition that affects both children and adults and it is a neurodevelopmental disorder characterized by symptoms such as inattention, hyperactivity, and impulsivity. This condition has worldwide prevalence of approximately 5-7% among children and adolescents and 2.5% among adults.<sup>1</sup>

### *Symptoms and Diagnosis*

ADHD, or attention deficit hyperactivity disorder, is diagnosed based on presence of ongoing symptoms that have occurred over some time and are evident throughout various life situations. Some of the symptoms are inattention, hyperactivity, and impulsivity.<sup>1,2</sup>

answers, and struggling with self-control. This leads to challenges in relationships and functioning in everyday work.

ADHD, or attention-deficit/hyperactivity disorder, is seen differently in children and adults. In the case of children, ADHD is typically noticed during school years of children. This can occur due to classroom disruptions, challenges with focusing on schoolwork, and difficulties with social interactions with others. On other hand, in adults, symptoms of ADHD might not be as clear, but they can still have major impact on daily life and relationships. These symptoms could be evident as struggling with time management, organization, maintaining focus, and maintaining interpersonal connections and *etc.*

### Symptoms of Attention-Deficit/Hyperactivity Disorder (ADHD)

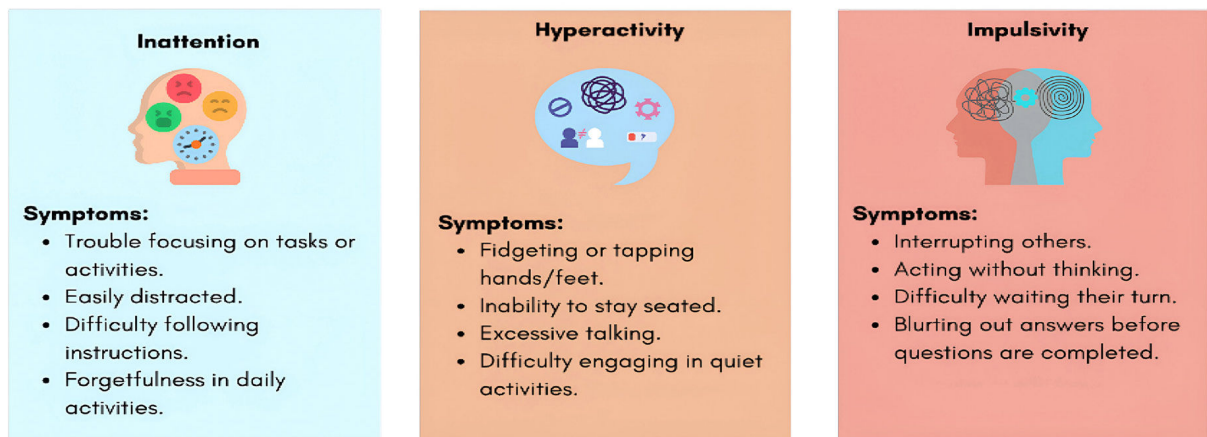


Fig. 1: Symptoms of ADHD (Source: Provided by author)

- Inattention involves difficulty focusing on particular task, following instructions, and completing tasks. Individuals with ADHD make careless mistakes, struggle with time management problems and organisation, and find it difficult to remain on track.
- Hyperactivity is characterized by excessive movement, playing, and restlessness. Individuals with ADHD may have difficulties sitting still, waiting, and taking turns, which can lead to their interactions and behaviour in social and academic settings.
- Impulsivity refers to impulsive actions such as interrupting others, delivering

### *Types of ADHD*

There are three main types of ADHD. 3, each characterized by the primary symptoms:

### *Impact and Treatment*

ADHD can have significant impact on an individual's life, leading to difficulties in social relationships, academic and professional performance, and daily functioning. Treatment options include medication, behavioural management techniques, and other practical strategies. A comprehensive evaluation typically includes review of past and current symptoms, medical exam and history, and use of adult rating scales or checklists.<sup>2,3</sup>

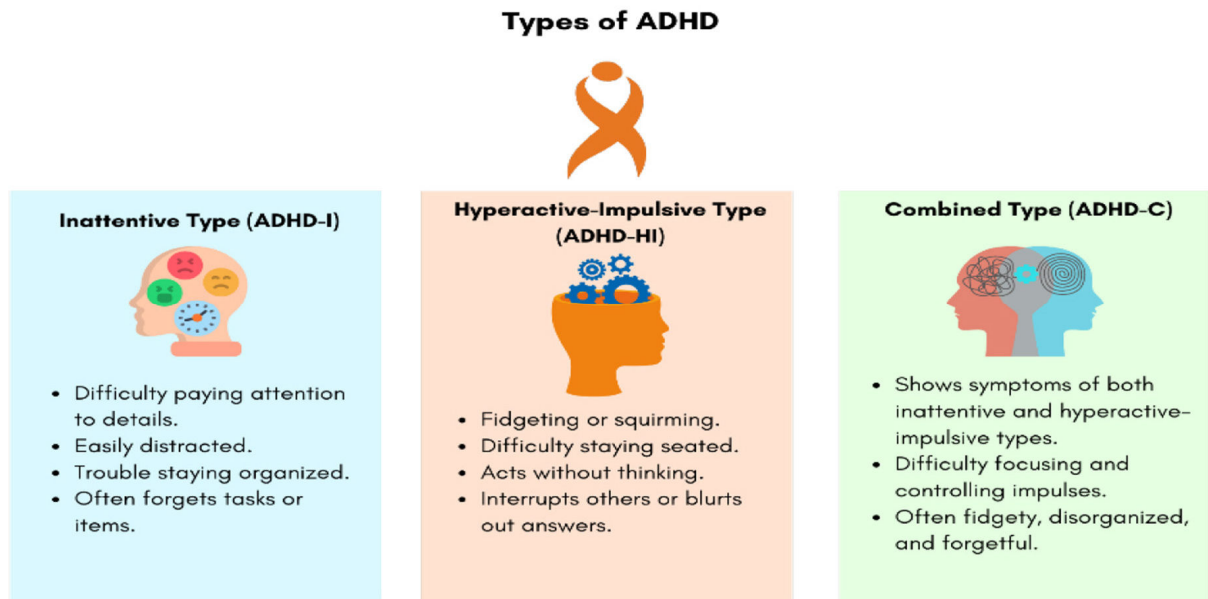


Fig. 2: Types of ADHD (Source: Provided by author)

### ***Importance of accurate diagnosis and effective treatment***

Accurate diagnosis and effective treatment of Attention Deficit Hyperactivity Disorder (ADHD) are essential for improving the quality of life of individuals who are affected by this disorder. Proper diagnosis ensures that individuals acquire appropriate medication to meet their specific needs, while effective treatment can help manage symptoms, improve functioning, and prevent or mitigate the negative consequences associated with ADHD.<sup>4</sup>

#### *Accurate Diagnosis*

Accurate diagnosis of ADHD is essential for several reasons:

1. Differentiating ADHD from other conditions: ADHD can happen at the same time as other mental health issues, like anxiety, depression, or learning problems. Accurate diagnosis helps in distinguishing ADHD from these other problems and makes sure that each issue gets the right treatment.
2. Tailoring interventions: Accurate diagnosis lets healthcare professionals to make personalized treatment plans that address the specific symptoms and needs of each individual patient. This personalized approach works better than using the same approach for everyone.

3. Providing appropriate accommodations: An accurate ADHD diagnosis can help individuals get help and support services in educational and at-work settings, like extended time on tests, reduced distractions, or assistive technologies. These services can really improve academic and professional outcomes.

#### *Effective Treatment*

Effective treatment of ADHD involves combination of medication, behavioural interventions, and practical strategies. The goals of effective treatment are as follows:<sup>5-7</sup>

1. Manage symptoms: Effective treatment can help individuals with ADHD manage their symptoms that they are facing, such as inattention, hyperactivity, and impulsivity, which can improve their ability to function in everyday life.
2. Improve functioning: Effective treatment can help individuals with ADHD improve their academic, professional, and social functioning by providing them with the tools and strategies to overcome the challenges associated with the disorder while working.
3. Prevent or mitigate negative consequences: Untreated ADHD patients can lead to negative consequences, such as poor academic performance, social difficulties,

substance abuse, increased risk of accidents, etc. Effective treatment can help prevent or mitigate these negative outcomes and can bring some positive outcomes.

### Role of AI in Improving Diagnosis and Treatment

Integration of Artificial Intelligence (AI) has brought about a major shift in various fields, including healthcare, where it has significantly improved precision and effectiveness of medical diagnoses and treatments. By increasing medical diagnosis and treatment, AI has potential to transform healthcare delivery, ultimately leading to improved patient outcomes and reducing healthcare expenses and costs. In this section, we will explore the impact of AI on improving diagnosis and treatment, highlighting its different applications, advantages, and challenges.<sup>8,9</sup>

#### *AI in Medical Diagnosis*

Utilization of AI has significantly improved the standards of medical diagnosis by analyzing large datasets and identifying patterns that might not be apparent to human clinicians. AI algorithms can be accurately trained to identify/recognized pathological or diseased patterns in medical images such as X-rays, CT scans, and MRI scans, enabling them to deliver accurate diagnoses. Particularly in identification of diseases like cancer, Alzheimer's, and stroke, where early detection is important for effective treatment, this technology has proven to be immensely effective.<sup>10</sup>

#### *AI in Treatment*

AI can also play important role in treatment by advancing clinicians in developing personalized treatment plans. By analysing patient data, medical history, and treatment outcomes, AI algorithms can provide accurate recommendations for optimal treatment strategies which is patient specific. This personalized approach holds promising improvement in treatment outcomes, minimized side effects, and increased patient satisfaction, representing significant advancement in domain of personalized medicine.<sup>10-12</sup>

#### *Benefits of AI in Diagnosis and Treatment*

Artificial Intelligence (AI) integration in healthcare has gained some attention due to its potential to revolutionize medical system. This integration provides few benefits, including improved accuracy in diagnosis and personalized treatment plans. Some benefits include:

- **Improved Accuracy:** AI is best at analyzing large datasets, identifying complex patterns, and extracting insights that can be overlooked by humans. This capability is advantageous in diagnostics, enabling early detection of diseases and precise treatment planning.
- **Increased Efficiency:** By automating repetitive responsibilities such as administrative work, data analysis, and image recognition, AI releases healthcare professionals to focus on complex patient cases. This increases overall efficiency in healthcare approach and optimizes resource utilization.
- **Personalized Medicine:** AI algorithms can learn complex patient data, including genetic information, medical history, lifestyle factors, and real-time monitoring data. AI facilitates development of personalized treatment plans adapted to individual patient needs, thereby optimizing therapeutic efficacy and minimizing adverse outcomes.
- **Reduced Costs:** the deployment of AI in healthcare leads to cost savings through various approaches. These include reduction of unnecessary diagnostic tests and procedures, avoidance of medication errors, improved patient outcomes, and streamlined operational workflows.

### Current State of AI Applications in ADHD Research Machine Learning Algorithms for Symptom Detection

In recent years, machine learning algorithms have been increasingly utilized within field of Attention Deficit Hyperactivity Disorder (ADHD) research. These algorithms have demonstrated great promise in their ability to analyze extensive datasets and uncover complex patterns that may not be readily detectable to human clinicians. This has led to increased diagnostic accuracy and development of personalized treatment plans for individuals with ADHD.

One of the key applications of machine learning in ADHD research is development of classification models. These models are designed to differentiate between individuals with ADHD and healthy controls, as well as those with other chronic disorders. The main objective is to identify the most significant features that set ADHD apart from other conditions and to create models that can effectively



predict the presence of ADHD symptoms. These efforts can greatly enhance our understanding of ADHD and improve the accuracy of diagnostic procedures.<sup>13</sup>

### Applications of Machine Learning in ADHD

Machine learning algorithms have been applied in various ways to detect ADHD symptoms:<sup>13,14</sup>

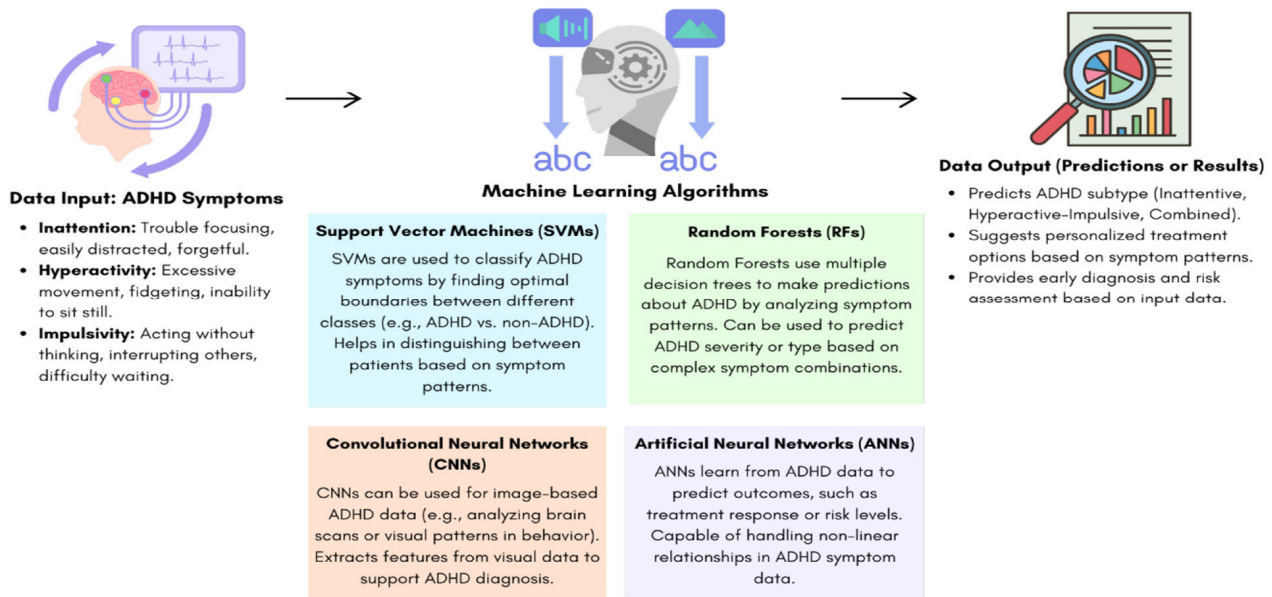


Fig. 3: ML Algorithm can be applied on ADHD symptoms (Source: Provided by author)

### Performance of Machine Learning Algorithms

Machine learning algorithms have been shown to be highly effective in detecting ADHD symptoms. For example, a study using linear support vector machine (SVM) classifier achieved an accuracy of 66.1% in differentiating children and adolescents with ADHD from those without condition.<sup>15</sup> Another study using a convolutional neural network (CNN) achieved an accuracy of 95.83% in detecting ADHD in children.<sup>16</sup>

### Challenges and Future Directions

Despite promising results, there are several challenges and future directions to consider:<sup>17</sup>

- **Data Quality:** The quality of data used to train machine learning models is essential for their performance. The accuracy and completeness of the data is crucial.
- **Feature Selection:** Selecting relevant features for machine learning model is critical for its performance. This can be a challenging task, particularly when dealing with large datasets.
- **Interpretability:** Machine learning models can be difficult to interpret, which can make it challenging to understand basic

mechanisms of ADHD. Developing more understandable models is essential for their adoption in clinical practice.

- **Generalizability:** Machine learning models trained on specific datasets may not be suitable for other datasets. It is essential to develop models that can be categorized into different datasets for their widespread application.

### Neural networks for brain imaging analysis

Recent advances in neuroimaging have led to widespread use of neural networks in analysis of brain imaging data to enhance diagnosis and comprehension of Attention Deficit Hyperactivity Disorder (ADHD). The complex capabilities of these neural networks allow them to effectively differentiate complex patterns within various types of brain imaging data, including functional magnetic resonance imaging (fMRI), diffusion weighted imaging (DWI), and structural magnetic resonance imaging (sMRI), leading to identification of key biomarkers and neural mechanisms associated with ADHD.<sup>18,19</sup>

Neural networks have been used in development of classification models to differentiate between individuals with ADHD and healthy controls, as

well as subjects with other disorders/diseases. By analyzing wide range of datasets, these models can reveal complex patterns that might not be seen by humans, resulting in significantly improvement of diagnostic accuracy and preparation of personalized treatment plans.<sup>20,21</sup>

*Applications of Neural Networks in ADHD*

Neural networks have been applied in various ways to analyze brain imaging data in ADHD research:

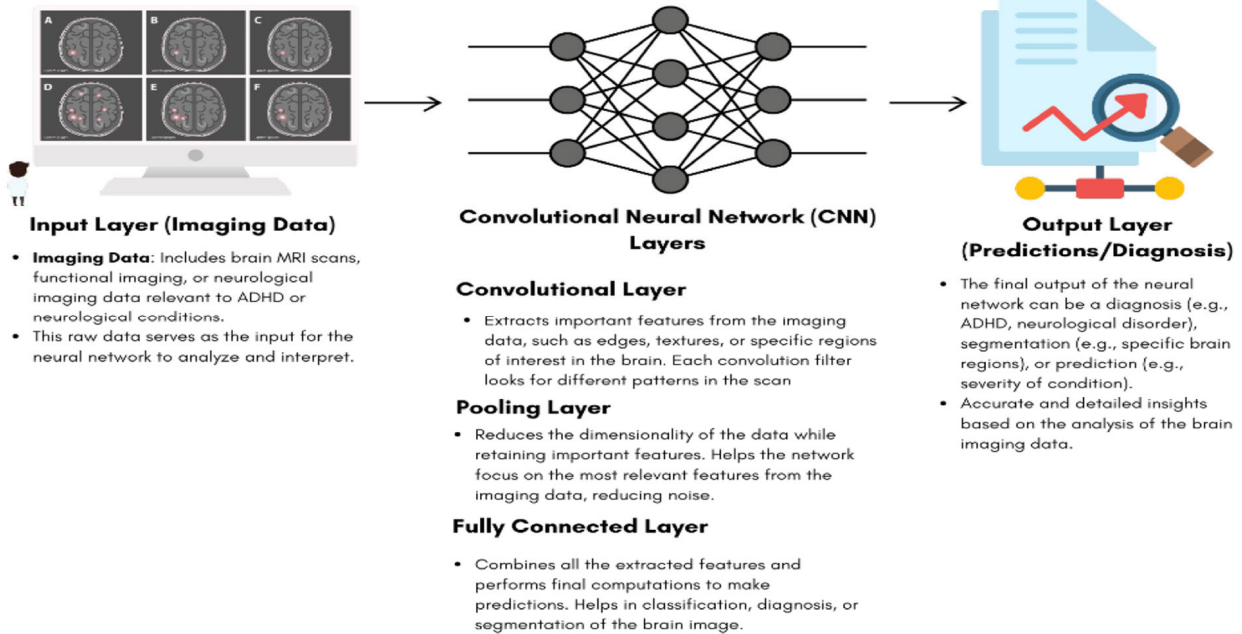


Fig. 4: Neural Network applied on imaging data (Source: Provided by author)

*Performance of Neural Networks*

Neural networks have been shown to be effective in analyzing brain imaging data in ADHD research. For example, a study using a GCN achieved an accuracy of 95.83% in comparing individuals with ADHD from typically developing individuals.<sup>22</sup>

*Challenges and Future Directions*

Despite promising results, there are several challenges and future directions to consider:<sup>17</sup>

- **Data Quality:** The quality of brain imaging data used to train neural networks is crucial for their performance. The accuracy and completeness of the data is crucial.
- **Interpretability:** Neural networks can be difficult to interpret, which can make it challenging to understand basic mechanisms of ADHD. The development of more interpretable models is essential for their implementation in clinical practice.
- **Generalizability:** Neural networks trained on specific datasets may not be suitable for other datasets. It is essential to develop models that can be categorized

into different datasets for their widespread application.

*Future Directions and Potential Benefits*

*Integration of AI with other diagnostic tools*

Integration of Artificial Intelligence (AI) with traditional diagnostic tools represents transformative changes in medical diagnostics, enhancing accuracy, efficiency, and effectiveness of healthcare approach. AI have capabilities that extend to various diagnostic methods, such as medical imaging, laboratory tests, genomics, proteomics, comprehensive patient data analysis and many others.

AI can analyze medical images like X-rays, CT scans, and MRIs to improve diagnostic accuracy and also increase its speed. It is also used in interpreting laboratory test results, identifying patterns in blood tests and biopsies for more accurate diagnoses and treatment. In addition, AI is also utilized in genomics and proteomics facilitates the identification of genetic biomarkers and disease predictions. By integrating and analyzing complex patient data—including medical history, demographics, and

lifestyle factors—AI can uncover patterns and enhance diagnostic accuracy.<sup>23</sup>

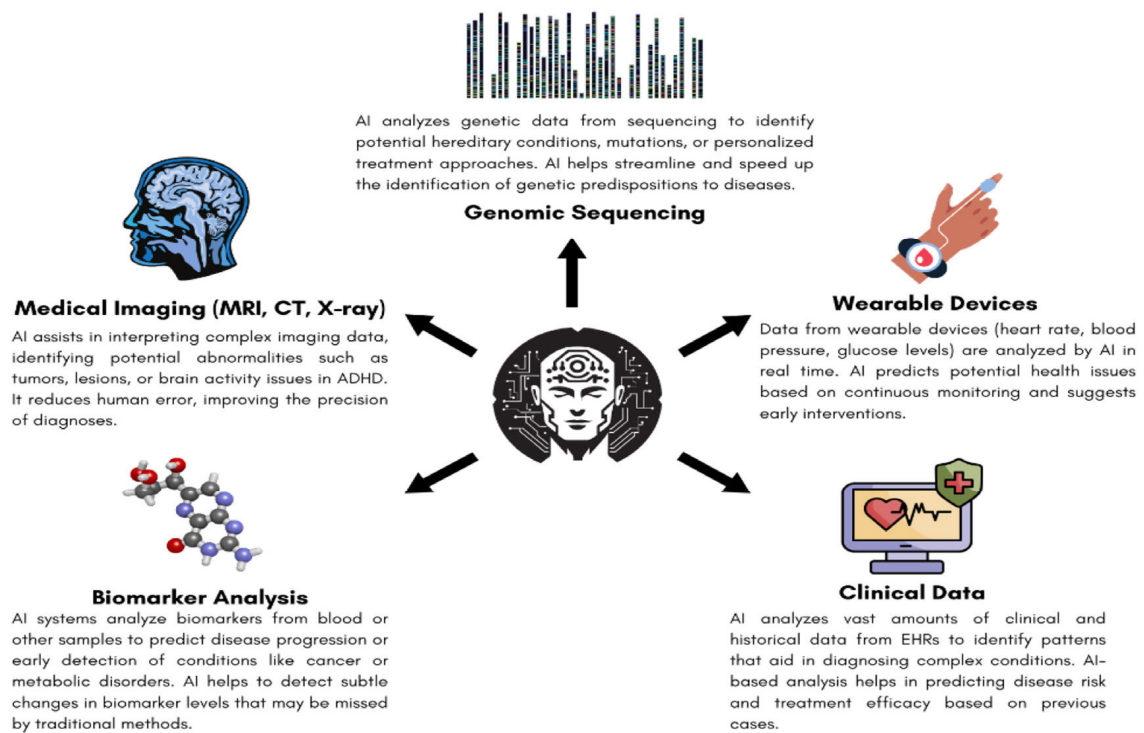


Fig. 5: AI integration with other diagnostic tools (Source: Provided by author)

Future directions for AI integration in diagnostics include multimodal data analysis, where AI examines multiple data sources simultaneously to provide complete understanding of a patient's health condition at once and development of explainable AI is crucial for promoting transparency and trust in AI-assisted diagnostics. AI can provide real-time feedback to healthcare professionals, enabling more informed decision-making and improved patient outcomes. Additionally, AI facilitates personalized medicine by creating treatment plans for an individual's genetic and medical history.<sup>24,25</sup>

While potential benefits of integrating AI with diagnostic tools are extensive. AI improves diagnostic accuracy by processing large datasets and identifying patterns beyond human capabilities of identifying the patterns. It reduces the time and resources required for diagnosis and treatment, reducing the time and resources required for diagnosing and treating. Improved diagnostic accuracy and efficiency lead to better patient results and lower healthcare costs. Furthermore, AI have ability to provide timely and accurate information to patients enhance their engagement in their own care, resulting in improved health outcomes.<sup>26,27</sup>

### *Personalized treatment strategies using AI*

Incorporation of Artificial Intelligence (AI) in field of healthcare has unlocked variety of opportunities for developing personalized treatment strategies. AI can process large datasets and recognize complex signatures and patterns. It allows healthcare professionals to develop treatments for individual patients, utilizing their distinct characteristics, genetic profiles, and disease progressions.<sup>28</sup>

AI algorithms can easily integrate and analyze large amounts of data such as electronic health records, genomic information, imaging scans, and real-time data from wearable devices used for patient monitoring. By analyzing this large amount of information, AI can predict necessary patterns and connections that may be avoided by humans. This deep analysis yields comprehensive understanding of each patient's health profile, equipping healthcare professionals to make precise diagnoses and create personalized treatment plans.<sup>29,30</sup>

AI also allows in real-time patient monitoring, offering continuous feedback on their health status. Wearable devices equipped with biosensors can capture data on fundamental signs, activity levels, sleep patterns, and other relevant metrics/



information, which is then analyzed by AI/ML algorithms in real time. This analysis allows healthcare professionals to detect any changes from normal ranges or potential health risks, thereby allowing healthcare professionals to provide timely treatments and diagnoses. Moreover, AI models can continually learn and adapt to changing patient requirements, ensuring dynamic and responsive personalized treatment approaches.<sup>31</sup>

While potential benefits of personalized AI-driven treatment strategies are significant, there are also some challenges that need to be addressed. That are ensuring quality, completeness, and interoperability of patient data is crucial for effectiveness of these strategies. It is essential to develop ethical guidelines and governance frameworks for effective and equitable use of AI in healthcare. Improving the interpretability of AI algorithms to gain trust and acceptance among healthcare professionals and patients should be a key priority. Additionally, it is important to ensure that AI models developed for personalized treatment strategies are widely accepted and validated across diverse patient populations for widespread adoption. Developing collaboration between AI experts, healthcare providers, and patients while educating stakeholders on both potential benefits and limitations of AI in healthcare is crucial for successful implementation.<sup>32</sup>

## CONCLUSION

Using AI in ADHD research is a promising approach to enhancing diagnosis and treatment strategies for individuals affected by Attention Deficit Hyperactivity Disorder (ADHD) and integrating Artificial Intelligence (AI) tools, particularly machine learning algorithms and neural networks, has shown substantial potential in revolutionizing how ADHD disease is diagnosed and treated. The current state of AI applications in ADHD research highlights effectiveness of machine learning algorithms in symptom detection, classification models, and analysis of brain imaging data. These AI-driven approaches have demonstrated high accuracy rates in differentiating individuals with ADHD from healthy controls and other chronic disorders, leading to improved diagnostic precision and development of personalized treatment plans specific to individual patient needs. Despite outstanding progress, there are numerous challenges and future directions that merit attention. Data quality remains critical aspect, necessitating accurate and comprehensive

datasets for training AI models effectively. Feature selection and interpretability of AI models are ongoing challenges, focusing on need for more understandable and transparent algorithms to facilitate their adoption in clinical practice. The future prospects of AI in ADHD research are promising. Integrating AI with traditional diagnostic tools provides a transformative approach, enhancing diagnostic accuracy, efficiency, and effectiveness across various healthcare sectors. Multimodal data analysis and self-understandable AI methods are key areas for advancement, enabling deeper understanding of patient health conditions and encouraging transparency and trust in AI-assisted diagnostics. Personalized treatment strategies using AI represent significant advancement in healthcare. AI can process large amounts of patient data, analyze complex patterns, and offer real-time monitoring capabilities revolutionize development of patient specific treatment plans. However, ensuring data quality, addressing ethical considerations, enhancing interpretability, and validating AI models across large patient populations are crucial steps in realizing full potential of personalized AI-driven treatment strategies and diagnosis. In conclusion, integration of AI in ADHD research and healthcare holds immense advantages. It provides improved diagnostic accuracy, personalized treatment approaches, improved patient outcomes, and reduced healthcare costs. Collaborative efforts between AI experts, healthcare providers, patients, and stakeholders are essential for overcoming challenges and advancing implementation of AI-based solutions in ADHD diagnosis and treatment.

## REFERENCES

1. da Silva, B.S., Grevet, E.H., Silva, L. *et al.* An overview on neurobiology and therapeutics of attention-deficit/hyperactivity disorder. *Discov Ment Health* 3, 2 (2023). <https://doi.org/10.1007/s44192-022-00030-1>
2. Williams OC, Prasad S, McCrary A, Jordan E, Sachdeva V, Deva S, Kumar H, Mehta J, Neupane P, Gupta A. Adult attention deficit hyperactivity disorder: a comprehensive review. *Ann Med Surg (Lond)*. 2023 Apr 12;85(5):1802-1810. doi: 10.1097/MS9.0000000000000631. PMID: 37228994; PMCID: PMC10205222.
3. Luo, Y., Weibman, D., Halperin, J. M., & Li, X. (2019). A review of heterogeneity in attention deficit/hyperactivity disorder (ADHD). *Frontiers in Human Neuroscience*, 13. <https://doi.org/10.3389/fnhum.2019.00042>



4. Arnold, L. E., Hodgkins, P., Caci, H., Kahle, J., & Young, S. (2015). Effect of treatment modality on long-term outcomes in attention-deficit/hyperactivity disorder: A systematic review. *PLOS ONE*. <https://doi.org/10.1371/journal.pone.0116407>
5. Al Kuwaiti A, Nazer K, Al-Reedy A, Al-Shehri S, Al-Muhanna A, Subbarayalu AV, Al Muhanna D, Al-Muhanna FA. A Review of the Role of Artificial Intelligence in Healthcare. *Journal of Personalized Medicine*. 2023; 13(6):951. <https://doi.org/10.3390/jpm13060951>
6. Ramsay JR. Assessment and monitoring of treatment response in adult ADHD patients: current perspectives. *Neuropsychiatr Dis Treat*. 2017; 13:221-232. <https://doi.org/10.2147/NDT.S104706>
7. Vitiello, B., & Sherrill, J. (2007). School-based interventions for students with attention deficit hyperactivity disorder: Research implications and prospects. *School Psychology Review*, 36(2), 287-290. <https://doi.org/10.1080/02796015.2007.12087946>
8. Shaw, M., Hodgkins, P., Caci, H. *et al*. A systematic review and analysis of long-term outcomes in attention deficit hyperactivity disorder: effects of treatment and non-treatment. *BMC Med* 10, 99 (2012). <https://doi.org/10.1186/1741-7015-10-99>
9. Alowais, S.A., Alghamdi, S.S., Alsuhebany, N. *et al*. Revolutionizing healthcare: the role of artificial intelligence in clinical practice. *BMC Med Educ* 23, 689 (2023). <https://doi.org/10.1186/s12909-023-04698-z>
10. Devi, K. J., Alghamdi, W., N, D., Sathish, T., & Alghamdi, W. (2023). Artificial Intelligence in Healthcare: Diagnosis, Treatment, and Prediction. *E3S Web of Conferences*, 399(3). <https://doi.org/10.1051/e3sconf/202339904043>
11. Khan, A. (2023). Transforming Healthcare through AI: Unleashing the Power of Personalized Medicine. *International Journal of Multidisciplinary Sciences and Arts*, 2(1). <https://doi.org/10.47709/ijmdsa.v2i1.2424>
12. Schork, N.J. (2019). Artificial Intelligence and Personalized Medicine. In: Von Hoff, D., Han, H. (eds) *Precision Medicine in Cancer Therapy . Cancer Treatment and Research*, vol 178. Springer, Cham. [https://doi.org/10.1007/978-3-030-16391-4\\_11](https://doi.org/10.1007/978-3-030-16391-4_11)
13. Ghasemi E, Ebrahimi M, Ebrahimie E. Machine learning models effectively distinguish attention-deficit/hyperactivity disorder using event-related potentials. *Cogn Neurodyn*. 2022 Dec;16(6):1335-1349. doi: 10.1007/s11571-021-09746-2. Epub 2022 Feb 15. PMID: 36408064; PMCID: PMC9666608.
14. Alkahtani, H.; Aldhyani, T.H.H.; Ahmed, Z.A.T.; Alqarni, A.A. Developing System-Based Artificial Intelligence Models for Detecting the Attention Deficit Hyperactivity Disorder. *Mathematics* 2023, 11, 4698. <https://doi.org/10.3390/math11224698>
15. Mikolas, P., Vahid, A., Bernardoni, F. *et al*. Training a machine learning classifier to identify ADHD based on real-world clinical data from medical records. *Sci Rep* 12, 12934 (2022). <https://doi.org/10.1038/s41598-022-17126-x>
16. Taghi Beyglou, B., Shahbazi, A., Bagheri, F., Akbarian, S., & Jahed, M. (2022). Detection of ADHD cases using CNN and classical classifiers of raw EEG. *Computer Methods and Programs in Biomedicine Update*, 2, 100080. <https://doi.org/10.1016/j.cmpbup.2022.100080>
17. Zhang, J., Zhang, Zm. Ethics and governance of trustworthy medical artificial intelligence. *BMC Med Inform Decis Mak* 23, 7 (2023). <https://doi.org/10.1186/s12911-023-02103-9>
18. Wang, XH., Jiao, Y. & Li, L. Identifying individuals with attention deficit hyperactivity disorder based on temporal variability of dynamic functional connectivity. *Sci Rep* 8, 11789 (2018). <https://doi.org/10.1038/s41598-018-30308-w>
19. Eloyan, A., Muschelli, J., Nebel, M. B., Liu, H., Han, F., Zhao, T., Caffo, B. (2012). Automated diagnoses of attention deficit hyperactive disorder using magnetic resonance imaging. *Frontiers in Systems Neuroscience*, 6, Article 61. <https://doi.org/10.3389/fnsys.2012.00061>
20. Chen, M., Li, H., Wang, J., Dillman, J. R., Parikh, N. A., & He, L. (2019). A Multichannel Deep Neural Network Model Analyzing Multiscale Functional Brain Connectome Data for Attention Deficit Hyperactivity Disorder Detection. *Radiology: Artificial Intelligence*, 1(1), e190012. <https://doi.org/10.1148/ryai.2019190012>
21. Lim, L., Marquand, A., Cubillo, A. A., Smith, A. B., Chantiluke, K., Simmons, A., Mehta, M., & Rubia, K. (2013). Disorder-Specific Predictive Classification of Adolescents with Attention Deficit Hyperactivity Disorder (ADHD) Relative to Autism Using Structural Magnetic Resonance Imaging. *PLOS ONE*, 8(5), e63660. <https://doi.org/10.1371/journal.pone.0063660>
22. Hu Y, Ran J, Qiao R, Xu J, Tan C, Hu L, Tian Y. Identifying ADHD-Related Abnormal Functional Connectivity with a Graph Convolutional Neural Network. *Neural Plast*. 2024 Apr 30;2024:8862647. doi: 10.1155/2024/8862647. PMID: 38715980; PMCID: PMC11074862.
23. Alshehri S, Alahmari KA, Alasiry A. A Comprehensive Evaluation of AI-Assisted

- Diagnostic Tools in ENT Medicine: Insights and Perspectives from Healthcare Professionals. *J Pers Med.* 2024 Mar 28;14(4):354. doi: 10.3390/jpm14040354. PMID: 38672981; PMCID: PMC11051468.
24. Dave, M., Patel, N. Artificial intelligence in healthcare and education. *Br Dent J* 234, 761-764 (2023). <https://doi.org/10.1038/s41415-023-5845-2>
  25. Sezgin E. Artificial intelligence in healthcare: Complementing, not replacing, doctors and healthcare providers. *Digit Health.* 2023 Jul 2;9:20552076231186520. doi: 10.1177/20552076231186520. PMID: 37426593; PMCID: PMC10328041.
  26. Khalifa, M., Albadawy, M., & Iqbal, U. (2024). Advancing clinical decision support: The role of artificial intelligence across six domains. *Computer Methods and Programs in Biomedicine Update*, 5, 100142. <https://doi.org/10.1016/j.cmpbup.2024.100142>
  27. Lin SY, Mahoney MR, Sinsky CA. Ten Ways Artificial Intelligence Will Transform Primary Care. *J Gen Intern Med.* 2019 Aug;34(8):1626-1630. doi: 10.1007/s11606-019-05035-1. Epub 2019 May 14. PMID: 31090027; PMCID: PMC6667610.
  28. Davenport, T.H., Glaser, J.P. Factors governing the adoption of artificial intelligence in healthcare providers. *Discov Health Systems* 1, 4 (2022). <https://doi.org/10.1007/s44250-022-00004-8>
  29. Bajwa J, Munir U, Nori A, Williams B. Artificial intelligence in healthcare: transforming the practice of medicine. *Future Healthc J.* 2021 Jul;8(2):e188-e194. doi: 10.7861/fhj.2021-0095. PMID: 34286183; PMCID: PMC8285156.
  30. Yeasmin, S. (2019). Benefits of Artificial Intelligence in Medicine. In 2019 2nd International Conference on Computer Applications & Information Security (ICCAIS) (pp. 1-6). Riyadh, Saudi Arabia. <https://doi.org/10.1109/CAIS.2019.8769557>
  31. He J, Baxter SL, Xu J, Xu J, Zhou X, Zhang K. The practical implementation of artificial intelligence technologies in medicine. *Nat Med.* 2019 Jan;25(1):30-36. doi: 10.1038/s41591-018-0307-0. Epub 2019 Jan 7. PMID: 30617336; PMCID: PMC6995276.
  32. Alshehri S, Alahmari KA, Alasiry A. A Comprehensive Evaluation of AI-Assisted Diagnostic Tools in ENT Medicine: Insights and Perspectives from Healthcare Professionals. *J Pers Med.* 2024 Mar 28;14(4):354. doi: 10.3390/jpm14040354. PMID: 38672981; PMCID: PMC11051468.

