

# APPLICATIONS OF ARTIFICIAL INTELLIGENCE IN HEARING AIDS AND AUDITORY IMPLANTS: A SHORT REVIEW

## Contributions:

A Study design/planning  
B Data collection/entry  
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D Data interpretation  
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## Abstract

Artificial intelligence (AI) has been broadly used for a long time, but in terms of hearing aids it only came into the limelight in 2004. Although at that time AI was yet to be incorporated into a hearing aid, with improvement in technology, slowly AI began to be introduced. AI included trainable hearing aids, own voice processing, brain-controlled hearing aids, strategies to improve speech perception in noise, and wind noise management in both hearing aids and cochlear implants. This short review discusses these AI technologies and their utility for the user.

**Key words:** speech perception • artificial intelligence (AI) • hearing aid • cochlear implant (CI)

## ZASTOSOWANIA SZTUCZNEJ INTELIGENCJI W APARATACH I IMPLANTACH SŁUCHOWYCH – KRÓTKI PRZEGLĄD

### Streszczenie

Od dłuższego czasu sztuczna inteligencja (SI) jest szeroko stosowana, jednak w odniesieniu do aparatów słuchowych znalazła się w centrum zainteresowania dopiero po roku 2004. Chociaż w tamtym czasie SI nie była jeszcze wbudowywana w aparaty słuchowe, w miarę rozwoju technologii zaczęło się to powoli zmieniać. SI zaczęto stosować w: samouczących się aparatach słuchowych, do przetwarzania własnego głosu, w aparatach słuchowych kontrolowanych mózgiem, do poprawy percepcji mowy w hałasie i kontroli szumu wiatru, zarówno w aparatach słuchowych, jak i implantach ślimakowych. W niniejszym przeglądzie omówione zostały wymienione technologie SI i ich przydatność dla użytkownika.

**Słowa kluczowe:** percepcja mowy • sztuczna inteligencja (SI) • aparat słuchowy • implant ślimakowy (CI)

### Introduction

Artificial intelligence (AI), often synonymous with machine learning, refers to computer expertise in simulating human intelligence as applied to problem-solving, logical reasoning, and tackling complex problems [1]. In other terms, it is the intelligence exhibited by machines, compared with the natural intelligence of humans and animals [2,3]. It is a branch of computer science that involves developing computer programs to devise applications ranging from running shoes, medical imaging, and robotic vacuum cleaners to navigation systems, and so on. These present a picture of 'Science Fiction' scenarios where robots play a significant role [3]. AI can accomplish tasks that demand human intelligence, for example, recognizing speech, seeing, translating, and making decisions. AI is a broad combination of deep learning and machine learning which includes algorithms for self training, feature extraction, and prediction of future outcomes. Nevertheless, all AIs are not equally created. Among various levels, the most common is symbolic AI, where a human task is done by machine learning [4].

In the hearing aid and cochlear implant industry, AI technology has become the focus of current research and a future trend in the industry. Hearing devices have proven they are successful solutions to hearing loss, and the use of AI

to make these devices even better is extraordinary [5]. In meeting an individual's needs, AI can improve the quality of perception and enhance ease of use. AI is a technological breakthrough that has improved hearing capabilities and advanced users' lifestyles [6]. Current AI features are trainable hearing aids and own voice processing which enables automatic programming of a hearing aid based on the present environment. Future technologies include brain controlled hearing aids with the help of AI algorithms like DNN (deep neural network) and CNN (convoluted neural network) [7]. Details of AI or machine learning algorithms in hearing aids and implants are discussed below.

### Material and methods

Articles published from various peer reviewed journals were searched in different databases: Medline/Pubmed Central, J-GATE, Google Scholar, and a manual Google search. Advanced search based on Boolean operators and keywords were used. White papers from various companies were also considered. The latest technology from different companies were searched on company websites and were included. The following MESH terms were used in the databases above: (((((((artificial intelligence) OR (AI)) OR (deep learning)) OR (machine learning)) AND (hearing aids)) OR (cochlear implants)) OR (CI)) OR (hearing

devices)) OR (auditory implants). Inclusion criteria were not kept strict as peer reviewed publications in this area are limited. Inclusion criteria were articles related to AI in hearing aids and cochlear implants, randomized control trials, clinical trials, case reports, and unpublished white papers. Articles related to AI algorithms in hearing devices were also considered. Articles on recent advances in hearing technology but not on AI were excluded. The data extraction was done by two independent reviewers.

## Results and discussion

A significant utility of AI technology is to enable a hearing impaired individual to perceive speech with better quality, especially in different environmental settings. Even though noise reduction algorithms already exist, an AI algorithm can automatically change its program settings based on the environment. It is said that AI technology can overcome the cocktail party effect in hearing-impaired individuals. AI technology also makes use of a hearing device easy. It gives form to the auditory intention of the user enabling them to achieve what they want [8].

Hearing aids with AI algorithms ‘learn’ the user’s hearing preferences, and via a set of observations and questions, it decides how one wishes to hear [9]. It uses complex processing to achieve a high-definition, natural sound which ensures that all the subtleties are heard. It reduces the need for manual adjustments by the wearer as it tailors and orients the gain naturally. This has paved the way for many innovations such as Signia’s application of AI to ‘Own Voice Processing’ (OVP) [10]. This OVP features natural sounding of the wearer’s voice; it ‘learns’ to recognize the voice and processes it separately while external sounds are undisturbed. This is a solution for those who cite the sound of their own voice as a drawback to using a hearing aid, and it only takes a few seconds [11].

There are many applications of machine learning, such as brain-controlled hearing aids that track the listener’s brainwaves and, based on a parallel with the source of sound in the environment, decide which speaker the listener is attending to. The device will then amplify the attended speaker relative to the others to facilitate hearing that speaker in a crowd. This is called auditory attention decoding (AAD) which automatically extracts the speaker from a mixed audio source. However, speaker-independent separation of speech is very challenging, and progress towards a solution is difficult. Frameworks have been proposed for better performance, and one hopes it is just around the corner. Therefore for better speech detection, neural network models of the auditory cortex and auditory attention is the next focus. A critical component of such a system is a real-time, low-latency speech-separation algorithm based on deep neural network models. These models approximate the computation performed by biological neurons, and have proven to be extremely effective in many machine learning tasks. The ongoing research in this area focusing on advancing our understanding of auditory attention and its neural markers in the human auditory cortex. The aim is to remove the technological barriers to AAD for improving speech intelligibility and reducing the listening effort in people with hearing loss [12].

Throwing light on another successful use of AI, machine learning has enabled hearing aids to detect speech and carry out language translation, track physical and mental health, and sense falls. Using recorded data, predictions can be made, making it possible to detect speech and provide an amplified translation to the ear in real time. Motion sensors can detect falls and then convey alerts and GPS locations to people such as caretakers or family members. These hearing aids are even intelligent enough to monitor spatial locations and log them accordingly. Later, it can use the GPS to determine when the user goes back to the same places and can send a notification if they want to adjust or automatically change preferred settings. Automatic preference adjustment in both quiet and noisy situations gives better speech perception based on the spatial sounds present [6a].

Machine learning or integral algorithms predict outcomes with learning experiences resulting from an individual or human input [13]. The best example is the Widex AI based upon the “SoundSense Learn” feature that is personalized to hone the hearing aid settings. The Widex Evoke program, named “SoundSense Adapt”, learns the user’s preferences in a suitable listening environment. It also allows the user to access and control personalized choices by choosing the setting through binary comparisons. It gets smarter by learning from all users by capturing unspecified data preferences and sending them to the Widex Cloud. SoundSense also facilitates high-speed processing and automation in the hearing aid [14].

Starkey Livio hearing aids expand their utility to other domains such as tracking daily footsteps, activity level, social listening, and active engagement while interacting to lessen cognitive decline, as well as many other wellness benefits. These advanced features are successful with hearing technology that uses integrated sensors and AI. Even though other devices have similar characteristics, research indicates that the ear can receive more accurate information. A newer version is the Livio Edge AI that provides superior sound quality in challenging listening situations. It has built-in AI technology activated with a simple tap in its Edge mode [15]. An easy way to control it is by using the Thrive Hearing Control app that provides the user with control options [16].

The new Syncro hearing aid released by Oticon includes voice priority processing (VPP) in which AI parallel processing is employed. Sequential processing may be faster and is able to select the preferred processing option, but it does not compare different outcomes; this can result in a nonoptimal choice of solution due to the unpredictability of communication in noisy surroundings. On the other hand, parallel processing means the ability to process and distinguish different outcomes, providing the best solution. The VPP optimizes speech output and noise reduction by using three mechanisms. Multi-band adaptive directionality provides a polar pattern for each band of frequency; TriState Noise Management categorizes the noise situations into well-defined listening modes; and voice aligned compression provides compression over an expanded bandwidth based on eight independent channels. These multiple channels optimize the signal provided to the user. In essence, VPP aims to provide the best

voice-to-noise ratio using decision-making from the AI parallel processing system [17].

Other AI algorithms used in hearing technology are deep neural networks (DNNs). Deep learning is a sub-field of machine learning that uses algorithms inspired by the brain's biological structure and functioning to provide machines with intelligence. DNNs became popular in 2019 and their utility in hearing devices began soon after. A recent paper on DNN in hearing aids was by Park et al. [6a] where they found that the DNN could cancel noise by classifying the incoming noise based on its type, thereby providing better speech quality compared to normal hearing aid algorithms. Another article by Lee et al. [7] compared two AI technology algorithms in hearing aids, one a DNN and another a convolution neural network (CNN). Their findings were that the CNN outperformed the DNN in terms of its speech enhancement strategy [7]. There have been other works documenting the utility of DNN algorithms in separating speech and noise and allowing speech to be emphasised [18].

Machine learning has also been used in cochlear implants. Some of the machine learning has been used in enhancing speech perception in noise, modelling auditory physiology, ambient noise and music processing, automated evoked potential measurement (Auto NRT, Auto ART), signal artifact filtering, post operative performance, surgical anatomy location prediction, electrode placement, and robotic surgery. Most of the algorithms used for these functions are deep neural networks, artificial neural networks, support vector machines, decision tree classifiers, linear regression, and Gaussian mixture models. The implant undergoes a training mode in all these situations; after these algorithms are fitted performance becomes more stronger [19-22].

AI has been refined to provide quick and efficient results in cochlear implants [19]. Finding the best setting for the cochlear implant's speech processor is a challenge, the goal being to provide intelligible speech and good sound information without relying on visual cues. AI fits the variability with adaptive intelligence. The Med-El cochlear implant with Rondo 3 speech processor also has adaptive intelligence technology. The new version with Automatic Sound Management 3.0 automatically adjusts to the sound in the environment to provide better hearing [23,24].

Other sound processors such as the Cochlear Kanso 2 and Nucleus N7 have novel algorithms such as SmartSound iQ with SCAN technology. It is designed to provide natural

hearing by capturing sound and adjusting it to the environment (speech, noise, speech in noise, wind, quiet, music) [25-27]. To compensate for the cocktail party effect, ForwardFocus helps to focus sound and reduce background noise, especially in face-to-face interactions.

Smart technology can remember settings and programs used in different environments. With the preferred choices, it can scan the environment and provide the best hearing sensation accurately. A variety of sound combinations such as enjoying music, lyrics, speech in quiet, listening effortlessly and comfortably to the conversation in the presence of noise have paved the way for Advanced Bionics Auto-Sound OS with HiResolution Sound that adapts intelligently and automatically for every listening situation [28].

AI is ubiquitous in mobile phones, computers, and tablets. With the wireless connecting link between these appliances and hearing devices, Smartphone compatibility has given rise to innumerable benefits [29]. Using the AI of such devices can bestow quality listening even if it's television or music listening on smartphones. It helps to tailor the auditory listening with personal preferences depending on the use [6].

A notable state-of-the-art technology must be Elon Musk's Neuralink, a neuroprosthesis with a 'Neural Lace' technology. It is a brain-computer interface that allows people to control wired devices for motor activities. This has expanded for auditory capabilities as the chip allows the user to 'hear' similar to an auditory cortical implant. Other benefits such as streaming music effortlessly into the brain are evolving faster and will soon be put into practice [30-32].

## Conclusion

This short review has given an insight into the utility of the various machine learning techniques used in hearing devices. There is a need to incorporate machine learning algorithms into hearing devices as it improves speech perception in noise, enables automation, accelerates the time it takes to process information, eases the job of audiologists, and aids the comfort of individuals using these devices. The utility also extends to surgery and intra-op monitoring, where it reduces errors. The literature has also shown an improvement in performance with the incorporation of these algorithms. Currently it may not be feasible for every individual to afford hearing devices with advanced features, but in the near future these devices will become affordable and improve quality of life.

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