Neural network reconstructs human thoughts from brain waves in real time
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Researchers from Russian corporation Neurobotics and the Moscow Institute of Physics and Technology have found a way to visualize a person's brain activity as actual images mimicking what they observe in real time. This will enable new post-stroke rehabilitation devices controlled by brain signals. The team published its research as a preprint on bioRxiv and posted a video online showing their "mind-reading" system at work.

"We're working on the Assistive Technologies project of Neuronet of the National Technology Initiative, which focuses on the brain-computer interface that enables post-stroke patients to control an exoskeleton arm for neurorehabilitation purposes, or paralyzed patients to drive an electric wheelchair, for example. The ultimate goal is to increase the accuracy of neural control for healthy individuals, too," said Vladimir Konyshev, who heads the Neurorobotics Lab at MIPT.

The existing solutions for extracting observed images from brain signals either use functional MRI or analyze the signals picked up via implants directly from neurons. Both methods have fairly limited applications in clinical practice and everyday life.

The brain-computer interface developed by MIPT and Neurobotics relies on artificial neural networks and electroencephalography, or EEG, a technique for recording brain waves via electrodes placed noninvasively on the scalp. By analyzing brain activity, the system reconstructs the images seen by a person undergoing EEG in real time.

Figure 1. Each pair presents a frame from a video watched by a test subject and the corresponding image generated by the neural network based on brain activity. Credit: Grigory Rashkov/Neurobotics

Figure 2. Operation algorithm of the brain-computer interface (BCI) system. Credit: Anatoly Bobe/Neurobotics, and @tsarycyanide/MIPT
In the first part of the experiment, the neurobiologists asked healthy subjects to watch 20 minutes of 10-second YouTube video fragments. The team selected five arbitrary video categories: abstract shapes, waterfalls, human faces, moving mechanisms and motor sports. The latter category featured first-person recordings of snowmobile, water scooter, motorcycle and car races.

By analyzing the EEG data, the researchers showed that the brain wave patterns are distinct for each category of videos. This enabled the team to analyze the brain's response to videos in real time.

In the second phase of the experiment, three random categories were selected from the original five. The researchers developed two neural networks: one for generating random category-specific images from "noise," and another for generating similar "noise" from EEG. The team then trained the networks to operate together in a way that turns the EEG signal into actual images similar to those the test subjects were observing (fig. 2).

"The electroencephalogram is a collection of brain signals recorded from scalp. Researchers used to think that studying brain processes via EEG is like figuring out the internal structure of a steam engine by analyzing the smoke left behind by a steam train," explained paper co-author Grigory Rashkov, a junior researcher at MIPT and a programmer at Neurobotics. "We did not expect that it contains sufficient information to even partially reconstruct an image observed by a person. Yet it turned out to be quite possible."

"What's more, we can use this as the basis for a brain-computer interface operating in real time. It's fairly reassuring. With present-day technology, the invasive neural interfaces envisioned by Elon Musk face the challenges of complex surgery and rapid deterioration due to natural processes — they oxidize and fail within several months. We hope we can eventually design more affordable neural interfaces that do not require implantation," the researcher added.


Provided by Moscow Institute of Physics and Technology

To test the system's ability to visualize brain activity, the subjects were shown previously unseen videos from the same categories. As they watched, EEGs were recorded and fed to the neural networks. The system passed the test, generating convincing images that could be easily categorized in 90 percent of the cases (fig. 1).