

disease who were taking statins. These observations may indicate that diabetic dyslipidemia is a potential residual cardiovascular risk.

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A Novel Deep Learning Approach for Automated Diagnosis of Acute Ischemic Infarction on Computed Tomography



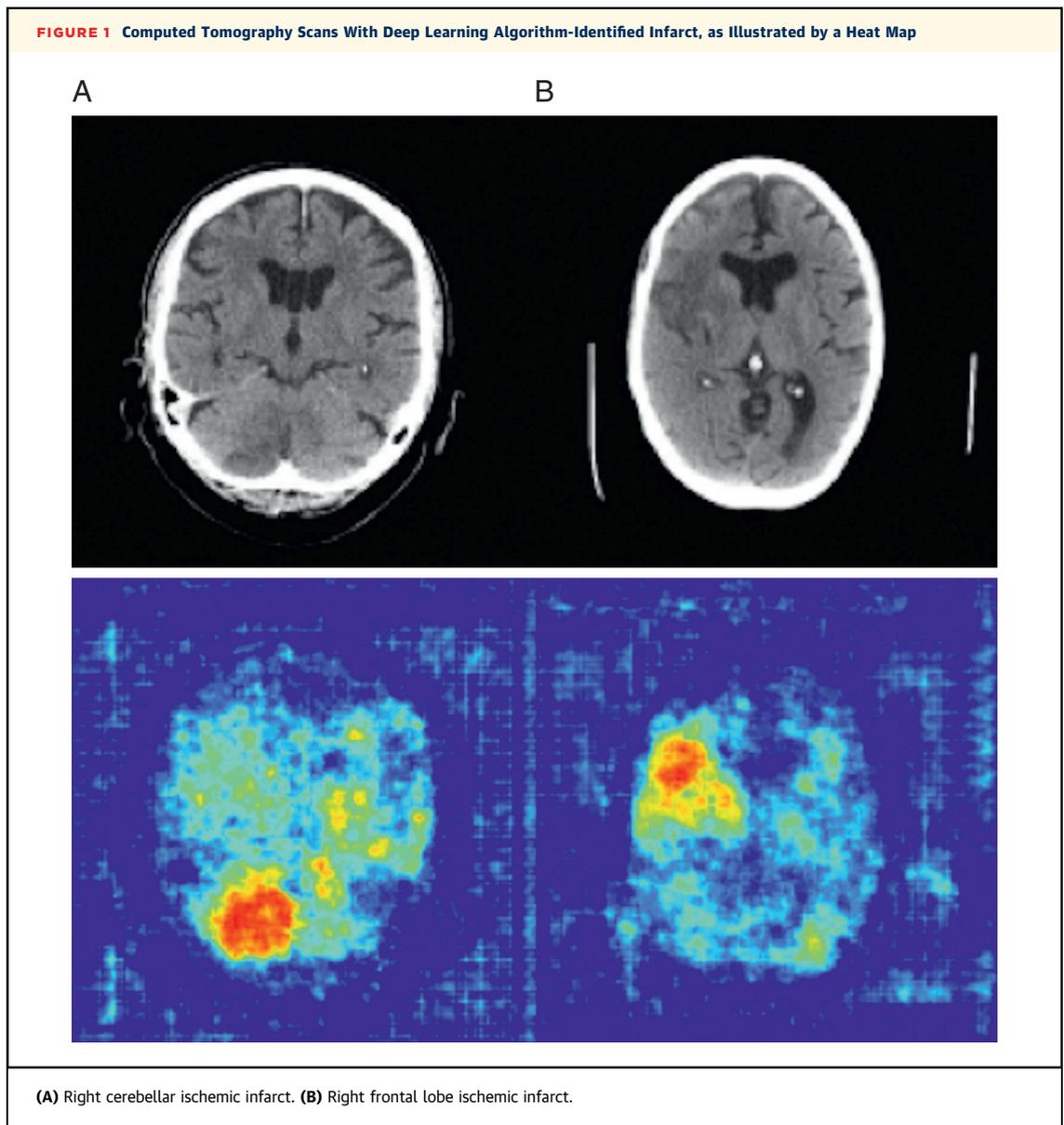
Stroke is the fourth leading cause of death and the leading cause of morbidity and long-term disability in adults (1). Early diagnosis of acute ischemic stroke

(AIS) in patients is difficult and the need for this early diagnosis will increase as the population ages and acute therapies evolve. Deep learning (DL) is a novel machine learning approach that enables automated extraction and classification of imaging features. This study aims to use DL to enhance our ability to evaluate for AIS while offering both automation and confirmation of a diagnosis.

All patients with AIS admitted to the New York-Presbyterian Hospital/Weill Cornell Medical Center between 2011 and 2014 were prospectively registered in the Cornell Acute Stroke Academic Registry. A total of 114 patients with noncontrast-enhanced computed tomography (CT) scan evidence of acute brain infarction were randomly selected. Board-certified neuroradiologists, blinded to the derivation of the model, annotated the images by marking infarct area to obtain an expert consensus interpretation. Digital Imaging and Communication in Medicine data was split randomly into a training set and test set (80:20). A 3-dimensional multiscale, fully convolutional deep learning neural network was developed and trained on the training set of CT images (2). Neural networks are mathematical models built to recognize patterns in images and predict outcomes based on a predefined ground truth (3). The performance of this model was independently tested using the test set and compared with the expert consensus interpretation. Diagnostic accuracy, sensitivity, specificity, and area under the receiver-operating characteristic curve (AUC) for the DL algorithm were calculated at a voxel and image level on the test set. Computer-generated heat maps were created to denote the possibility of infarct (Figure 1).

The mean age of the study sample was 76 ± 13 years, and 62 (55%) patients were female. Imaging datasets were split for the 114 patients into a training set ($n = 92$) and a testing set ($n = 22$). In the training set of 5,888 images, infarction was present in 602 (10.2%) images. In the testing set of 920 images, infarction was present in 130 (14.1%) images. A total of 1.5 billion voxels were used to train the model.

The AUC for the DL algorithm for voxel accuracy was 0.973 (95% confidence interval: 0.972 to 0.974). Diagnostic accuracy, sensitivity, and specificity were 92%, 93%, and 92%, respectively. Positive predictive value (PPV) and negative predictive value (NPV) were 86% and 92%, respectively. The AUC for the DL model for automated diagnosis of infarction at an image level was 0.91 (95% confidence interval: 0.90 to 0.94). The corresponding diagnostic accuracy, sensitivity, and specificity were 88%, 65%, and 91%, respectively. PPV and NPV were 49% and 95%.



These results demonstrate that DL-based neural networks can be trained to identify acute brain infarction on CT scan. Our present results reveal a diagnostic accuracy of 93% compared with expert interpretation by blinded neuroradiologists. Of importance, at a cutpoint AUC of 0.91, we observed a NPV of 95% with a lower PPV of 49%. These findings suggest that the DL algorithms may allow for better ruling out of AIS, and that future methods are required for verification of acute brain infarction by machine learning.

To our knowledge, this study represents the largest to date to evaluate deep learning methods for auto-diagnosis of acute brain infarct on CT scan, with levels of overall diagnostic accuracy comparable to

board-certified specialists. Our results support the use of machine-learning methods for automated diagnosis of stroke to improve the diagnostic efficiency of radiologists and health systems. Future studies may also allow classification of any CT-visualized feature, as well as other types of pathologies.

The study is not without limitations. This study was performed at a single tertiary care academic center, and it is possible that the study results would prove different if CT scans were judged by less well-trained radiologists, a possibility that can be mitigated by the DL solutions. Also, the DL model in this study was trained by supervised learning methods on a generally small patient dataset and lacked a control group for comparison. It is important for future

prospective analysis to be performed to validate our study findings.

In conclusion, machine-learned models using novel DL techniques enable highly accurate automated diagnosis of acute brain infarction. These algorithms have the potential to assist radiologists while improving patient outcomes.

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CMR-Based and Time-Shift Corrected Pressure Gradients Provide Good Agreement to Invasive Measurements in Aortic Coarctation



In aortic coarctation (CoA), clinical guidelines recommend treatment in the presence of a relevant pressure gradient (1). Although reliable noninvasive

measurement approaches would be crucial, the accuracy of currently available methods is limited and cardiac catheterization has remained a clinical reference standard. Four-dimensional velocity-encoded cardiac magnetic resonance (CMR) was shown to be able to map relative pressures in a vessel (2,3). In contrast to invasive peak-to-peak measurements from heart catheterization, however, pressure mapping neglects the time-phase shift and thus adds bias. We aimed to validate a novel CMR-based time-shift corrected pressure mapping (TCPM) approach against catheterization and to compare this method against current noninvasive diagnostic standards, Doppler echocardiography, and cuff-based pressure differences.

We prospectively enrolled 21 patients with CoA (11 men, 10 women, mean age 23 ± 15 years) with clinical indication for cardiac catheterization due to relevant pressure gradients based on echocardiography and/or arterial hypertension. Our local ethical committee approved the study, and written informed consent was obtained from the participants or their guardians. CMR was performed before the catheterization procedure with a 5-element cardiac phased-array coil on a Philips Achieva R3.2.2 1.5-T scanner (Philips Medical System, Best, the Netherlands). During routine catheterization, pressure curve and resulting peak-to-peak gradients were recorded on a Schwarzer Haemodynamic Analyzing System (Heilbronn, Germany) in 2 predefined locations of the aorta. Previously described and clinically established 3- and 4-dimensional velocity encoded CMR sequences were used (3).

Figure 1A illustrates the post-processing and computation workflow to assess CMR-based TCPM. Semiautomatic aortic segmentation was performed with ZIBAmira (Zuse Institute, Berlin, Germany). Pressure mapping including antialiasing, registration, and flow analysis was done with MevisFlow (Fraunhofer MEVIS, Bremen, Germany). To generate relative pressure maps, pressure-Poisson equation was solved as detailed previously (2). CoA pressure differences were assessed at peak flow in the ascending and descending aorta and corrected to take into the account time-shift between pressure and volume flow in both vessels. Data were analyzed using SPSS version 21 (IBM Corporation, Armonk, New York). Bland-Altman plot and mean-equivalence paired *t*-test were performed (4). The null hypothesis in this test assumes measurements between methods differ considering a set clinical threshold of 5.0 mm Hg, corresponding to the variability of current clinical reference standards (invasive heart catheterization).