

# Detecting coronary plaque vulnerability using computed tomography radiomics: the one stop shop for plaque vulnerability?

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**This editorial refers to ‘Identification of invasive and radio-nuclide imaging markers of coronary plaque vulnerability using radiomic analysis of coronary computed tomography angiography’, by M. Kolossváry *et al.*, pp. 1250–1258.**

Advanced atherosclerotic plaques prone to cause acute coronary syndromes (ACS) are characterized by large lipid-rich necrotic cores, increased amounts of inflammatory cells, and thin fibrous caps. These atherosclerotic lesions are designated vulnerable plaques and have a high probability of rupture and atherothrombotic-driven ACS.<sup>1</sup>

An important strategy to lower the incidence of ACS is improved cardiovascular risk assessment. One of the most powerful ways to achieve this goal is by imaging the coronary arteries. Nowadays, there are several intravascular and non-invasive imaging modalities which can be used to detect high-risk coronary plaque characteristics. Serial coronary intravascular ultrasonography (IVUS) is capable of detecting thin-cap fibroatheroma which predispose for myocardial infarction.<sup>2</sup> Optical coherence tomography (OCT) provides high spatial-resolution images of superficial atherosclerotic plaque microarchitecture and can be used to measure fibrous cap thickness and ulcerations, calcium deposits, fibrous and lipid-rich plaque components, as well as molecular and cellular constituents of plaques.<sup>3</sup> A non-invasive technique that can identify and localize high-risk coronary plaque is <sup>18</sup>F-sodium fluoride PET (<sup>18</sup>F-NaF PET). Intense uptake of <sup>18</sup>F-NaF has been demonstrated in coronary plaques with high-risk features in patients with stable coronary artery disease.<sup>4</sup> Another non-invasive technique is magnetic resonance imaging. Magnetic resonance imaging can identify endothelial dysfunction, coronary plaque, and coronary intraplaque haemorrhage.<sup>5</sup>

Arguably, the most widely used non-invasive imaging modality to assess coronary plaque in clinical practice today is coronary computed tomography angiography (CCTA). Coronary computed tomography angiography can detect positive remodelling, low-attenuation plaque, and spotty calcification, which are known precursors of ACS. These atherosclerotic plaque characteristics are also

associated with myocardial ischaemia and indicative of more malignant, rapidly progressive coronary artery disease.<sup>6</sup>

It is encouraging to see so many diagnostic options to interrogate the different pathophysiological processes underlying plaque formation in the coronary arteries. Nevertheless, no single test is currently capable of capturing all known harbingers of vulnerability in a cost-effective and patient friendly manner, leaving physicians managing patients with coronary atherosclerosis with an urgent unmet need.

Kolossváry *et al.*<sup>7</sup> present a novel approach to bridge this gap by probing the utility of advanced analysis of the obtained CCTA grey scale values—also known as texture analysis or ‘radiomics’—to indicate the presence of different markers of vulnerability as assessed with multiple invasive and non-invasive tests. Radiomics transcends simple visual analysis of computed tomography (CT) images by conversion of digital medical images into high-dimensional data. These data lend themselves to mathematical extraction of advanced quantitative descriptors that are generally not used in clinical practice. This is supplemented by quantification of so-called semantic features, i.e. the terms imagers use to describe lesions.<sup>8</sup> Radiomics are a powerful set of techniques that have already led to discovery of previously unknown relationships between imaging features, immune response, inflammation, and survival in patients with lung cancer,<sup>9</sup> and Kolossváry *et al.* are to be commended for extending this methodology to imaging of coronary arteries.

In the present, highly interesting study, 44 coronary plaques from 25 patients with suspected coronary artery disease were analysed. All patients underwent CCTA, <sup>18</sup>F-NaF PET, IVUS, and OCT within a 90-day period. For all tests, the presence and extent of conventional markers of plaque vulnerability were established and correlated with the presence of at least two conventional features of plaque vulnerability at CCTA. Subsequently, the investigators assessed the correlation between radiomics features and markers of plaque vulnerability. Overall, 935 radiomics features were calculated and the incremental value of these features over conventional features was investigated. The main finding of the study is that radiomics markers significantly improved diagnostic accuracy for detection of vulnerable

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plaque features from CCTA. The highest gain in accuracy was achieved for the radiomics parameter 'surface of high attenuation voxels', which had an area under the receiver operator curve of 0.87 for identification of  $^{18}\text{F}$ -NaF PET positive lesions vs. 0.65 for conventional high-risk plaque features. Based on these results, the authors correctly conclude that advanced texture analysis of coronary CT images may allow identification of pathophysiological processes from routinely acquired non-invasive coronary CT images that were only thought to be accessible with completely different imaging modalities.

The present study provides us with unique insights. First, it is highly informative to compare different measures of plaque vulnerability in individual patients and to learn to which extent different imaging modalities are complementary. Second, the investigators convincingly show that radiomics analysis greatly increases the amount of information that can be extracted from CCTA, which can provide new insights into its ability to function as a one-stop-shop technique for comprehensive coronary plaque characterization. As such, the investigators have significantly advanced the state-of-the-art. Furthermore, they should be lauded for making available their source code to other interested investigators.<sup>10</sup>

The primary limitations of the study are the small sample size and cross-sectional design. However, it is very difficult to obtain a large cohort of patients willing and able to undergo four imaging tests within a short time. A challenge when dealing with such a small number of patients is to avoid spurious correlations between the large number of radiomics parameters and the small number of conventional vulnerable plaque parameters. In order to deal with this problem, the investigators used stratified cross-validation with a large number of repeats to derive robust estimates of diagnostic performance. This is an accepted way of getting a sense of the generalizability of the diagnostic accuracy in a real-world setting, but—as the investigators note—no substitute for confirmation of the findings in a much larger and independent cohort. In addition, it would be interesting to investigate the evolution of the radiomics parameters over time and in response to initiation of therapy, and to investigate their value for prediction of ACS over established clinical and imaging parameters. Another limitation is the need for manual segmentation of plaques prior to radiomic analysis. This may be overcome by application of deep learning techniques.<sup>11,12</sup>

The results of the present study lay the ground-work for much more detailed analysis of coronary plaque phenotypes. Especially promising in this regard are deep learning techniques that do not rely on hand-crafted features for identification of plaque vulnerability. Preliminary work using such techniques has already demonstrated

that deep learning-based analysis can aid in the identification of flow-limiting coronary lesions,<sup>13</sup> and it is highly likely that identification of vulnerable plaques and patients can be further improved as well.

In conclusion, Kolossváry *et al.* should be congratulated with this important study that ushers in the era of high dimensional, quantitative coronary plaque analysis in routine clinical practice. The methods described in this study complement clinically used analysis and have the potential to greatly improve individual risk prediction.

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