

# **Predictive value of the QFR in detecting vulnerable plaques in non-flow limiting lesions: a combined analysis of the PROSPECT and IBIS 4 study**

## **Introduction**

Fractional flow reserve (FFR) is the gold standard for assessing the physiologic significance of a coronary artery lesion and planning treatment. This modality, which also appears able to provide useful prognostic information even in non-flow limiting lesions [1], involves advancement of a pressure wire across the lesion and adenosine administration to achieve hyperaemia and thus it comes with additional procedure-related risks of adverse events, extended procedure time and increased cost [2]. Quantitative flow ratio (QFR), a wire-free, hyperaemia-free method has been recently introduced to overcome these limitations and derive FFR from models reconstructed from 3-dimensional quantitative coronary angiography (3D-QCA) [3]. Recent clinical studies have supported the potential clinical value of QFR showing that it enables not only accurate detection of flow limiting lesions but also identification of patients at risk for cardiovascular events [4–7].

Cumulative data has shown that imaging-derived variables combined with computationally-derived physiology indices enable more accurate detection of lesions that are likely to progress and cause events. In the EMERLAD study plaque characteristics and the FFR estimated by computed tomography allowed detection of lesions that caused myocardial infarction with a higher accuracy than plaque morphology (c-index: 0.789 vs. 0.747;  $P=0.014$ ) [8]. No study today however has assessed whether computationally-derived FFR – using the QFR software – has additive predictive value compared with intravascular imaging alone.

## **Methods**

### *Study design*

The present study is a post hoc analysis of data obtained from the IBIS 4 (Intergraded Biomarkers Imaging Study 4) and the PROSPECT (Providing Regional Observations to Study Predictors of Events in the Coronary Tree) clinical studies. The study design of these trials, the definitions and endpoints as well as the inclusion and exclusion criteria have been described in detail elsewhere [9, 10]. In summary,

the IBIS 4 trial was a prospective multi-modality multicenter study which aimed to investigate the effect of aggressive statin therapy on plaque composition and burden in patients admitted with a ST-elevation myocardial infarction (STEMI). All included patients (n=103) had successful revascularization and 3-vessel virtual histology intravascular ultrasound (VH-IVUS) and optical coherence tomography (OCT) imaging at baseline and 13-month follow-up.

The PROSPECT study was a prospective large-scale invasive imaging study that aimed to examine the ability of VH-IVUS in detecting vulnerable plaques that will evolve and cause events. The study included 697 patients admitted with an acute coronary syndrome that had 3-vessel VH-IVUS imaging of the proximal 60-80mm segment and successful percutaneous coronary intervention (PCI) in all the flow limiting lesions.

The patients recruited in the IBIS 4 study were follow-up for a median of 5 years whereas those in the PROSPECT trial for 3.4 years. The clinical, angiographic and intravascular imaging data in both studies allowed detection of untreated lesions that were assessed by VH-IVUS at baseline and caused cardiovascular events (defined as cardiac death, myocardial infarction, or revascularization because of progressive or unstable angina). In addition, in the IBIS 4 study lesions that exhibited disease progression and a significant stenosis (% diameter stenosis, DS  $\geq 70\%$  on QCA) or an intermediate stenosis ( $70\% \geq \%DS \geq 50\%$ ) associated with recurrent angina or evidence of ischemia during exercise test at 13 months follow-up angiography were considered culprit and underwent revascularization. The primary endpoint of the study was lesions assessed by VH-IVUS at baseline that either progressed and required revascularization on repeat angiography, or they caused cardiac death, myocardial infarction, or revascularization because of progressive or unstable angina (MACE).

In the IBIS 4 study all the lesions associated with MACE had either a thin (TCFA, n=13) or a thick-cap fibroatheroma (ThCFA, n=2) phenotype, while in the PROSPECT study 43 of the lesions that were studied by VH-IVUS at baseline and caused events had a fibroatheroma phenotype (TCFA, n=25; ThCFA, n=18), 7 had a non-fibroatheroma phenotype and 4 were unclassified lesions. [9, 10]. From these 43 lesions we analysed only data from the lesions that were included in the PROSPECT endothelial shear stress (ESS) sub-study a study that aimed to investigate the prognostic value of ESS in predicting events [11]. The PROSPECT ESS sub-study included lesions where matching of VH-

IVUS and X-ray angiography was feasible and had a length  $\geq 9$ mm and a vulnerable phenotype (TCFA, ThCFA). In addition, we included a control group of lesions with a vulnerable phenotype that remained quiescent from: 1) the PROSPECT ESS sub-study and from 2) the patients recruited in the IBIS 4 study in Bern University Hospital. Lesions with suboptimal angiographic images (for instance vessel overlapping or foreshortening or poor opacification of the lumen silhouette) and cases where the DICOM file did not include all the required information for QFR analysis were excluded from the study.

### *VH-IVUS analysis*

VH-IVUS analysis was carried out by independent core-laboratories (Cardialysis B.V., Rotterdam, The Netherlands in IBIS 4 and Cardiovascular Research Foundation, New York, NY in PROSPECT) using dedicated software (QIvus, Medis, Leiden, The Netherlands for the IBIS 4 study and QCU-CMS, Medis, Leiden, the Netherlands for the PROSPECT study). In the IBIS 4 trial analysis was performed only for the segment that was evaluated by VH-IVUS at baseline and follow-up whereas in the PROSPECT trial the analysis was carried out for the entire imaged coronary artery.

VH-IVUS segmentation was carried out in every end-diastolic frame; in each frame the lumen and external elastic membrane (EEM) area were annotated and their dimensions the plaque area and burden (PB) and its composition (fibrotic, fibrofatty, calcific and necrotic core area and burden) were estimated. In the studied segments lesions were characterised as  $\geq 3$  consecutive VH-IVUS frames with PB  $\geq 40\%$ ; in each lesion the plaque composition was used to classify it to one of the following phenotypes: pathologic intimal thickening, fibrotic, fibrocalcific, TCFA and ThCFA.[12].

### *2.3. 3D-QCA reconstruction*

Anatomical landmarks such as side branches observed in X-ray angiography and VH-IVUS were utilised to determine the distal and proximal end of the segment that was evaluated by VH-IVUS on coronary angiography. Two end-diastolic angiographic projections that were more than 25° apart, where there was no foreshortening or overlapping of the segment of interest, that permitted accurate delineation of the lumen silhouette were chosen to reconstruct its anatomy utilising an established and well-validated software (QAngio XA 3D RE, Medis, Leiden, the Netherlands) [13]. In the obtained geometries the lesion length, DS and minimum lumen diameter (MLD) were estimated.

#### *1.4. QFR analysis*

QFR was computed using the Medis Suite XA/QAngio XA 3D/QFR software (Medis, Leiden, the Netherlands). QFR was computed using two approaches: 1) assuming a fixed blood flow (fQFR) and 2) taking into account the flow velocity estimated by the time needed for the contrast agent to fill the segment of interest which is known as contrast-flow QFR (cQFR) [3].

#### *1.5. Statistical Methods*

Numerical variables are presented as median and interquartile range while categorical variables as absolute values and percentages. Comparison between numerical variables were performed using the Mann-Whitney U test while categorical variables were compared using the chi-square test. Cox regression analysis was used to identify VH-IVUS, 3D-QCA-derived and QFR predictors associated with MACE. For the 3D-QCA and QFR variables associated with MACE ( $P < 0.05$ ) receiver-operating characteristics curve analysis was performed to identify the best cut-off that predicted MACE. The variables with the highest area under the curve (AUC) that were not co-linear ( $r > 0.5$ ) were entered into a multivariable model to identify independent predictors of MACE.

VH-IVUS and QFR variables were used to classify the lesions in groups. Kaplan-Meier plots were used to display time to event; comparison of MACE rate between groups with different plaque characteristics and QFR values was performed using the log-rank test. Analyses were performed in Stata (version 15.1, StataCorp LP, College Station, TX) and SPSS (version 23; IBM Corp, Armonk, NY, USA). A P-value  $< 0.05$  was considered statistically significant.

## **2. Results**

### *3.1. Patient demographics*

QFR analysis was possible in 17 MACE lesions with a thin or thick cap fibroatheroma phenotype (5 from the PROSPECT ESS and 12 from the IBIS 4 study); (Figure 1). From the 122 lesions that were included in the PROSPECT ESS study and did not cause events, 8 were excluded from the present analysis because of suboptimal angiographic views and 102 were excluded because of insufficient information in the DICOM file for QFR analysis. From the 78 thin or thick cap fibroatheromas that

were recruited in the IBIS 4 study from Bern University Hospital, 12 lesions were excluded from the present analysis because of suboptimal angiographic views. Therefore the final analysis included 17 MACE-lesions and a control group of 78 lesions that remained quiescent.

The baseline characteristics of the patients (n=60) included in the present analysis are shown in Table 1. There were no differences in the baseline demographics between patients who had an MACE and those that did not have an event.

### *3.2. Lesions characteristics*

Table 2 shows the morphological and angiographic variables as well as the QFR values in the studied lesions. All the studied lesions were non-flow limiting by QFR (range: 0.81-1.00).

MACE lesions had a smaller MLA and greater PB on VH-IVUS comparing to non-MACE lesions. On 3D-QCA analysis MACE lesions were longer than the non-MACE lesions and had increased %DS but there were no differences in the MLD between groups. Both cQFR and fQFR were lower in the lesions that caused MACE.

### *3.3. Predictors of MACE lesions*

Two VH-IVUS-derived variables (MLA and PB), two 3D-QCA-derived variables (lesion length, and MLD) and the fQFR and cQFR indices were predictors of MACE lesions in univariate analysis (Table 3). In ROC curve analysis PB (AUC: 0.751, P=0.001), MLA (AUC: 0.734, P=0.003) and cQFR (cutoff 0.97; AUC: 0.733, P=0.003) were the three variables that had the highest AUC for predicting MACE (AUC for fQFR: 0.716, P=0.005). In the multivariable analysis that included cQFR and the presence of lesions with PB>70% and small MLA<4.0mm<sup>2</sup>, a small cQFR (<0.97) was independently associated with future events (HR: 3.53, 95%CI: 1.16-10.75; P=0.027); in this model the presence of lesions with increased plaque burden and small MLA was not an independent predictor of MACE (HR: 1.87, 95%CI: 0.64-5.43; P=0.252; Figure 2).

Lesions were classified in 4 groups according to the presence of  $\geq 2$  out of the 3 high-risk plaque characteristics that in the PROSPECT study were associated with MACE (MLA  $\leq 4$ mm<sup>2</sup> plaque burden  $\geq 70\%$ , TCFA phenotype) and the cQFR value (<0.97). Lesions with high-risk plaque characteristics

and low cQFR had worse prognosis than the other lesions. Similar results were obtained when a high-risk plaque phenotype was defined as the presence of MLA  $\leq 4$  mm<sup>2</sup> and PB  $\geq 70\%$  (Figure 3A, 3B).

### **3. Discussion**

In this study for the first time we examined the value of the computationally derived FFR estimated from QFR software in identifying those non-flow limiting lesions with a vulnerable-lipid-rich phenotype that are likely to progress and cause MACE within 5-year follow-up. We analysed data from two large-scale intravascular imaging studies and found that the cQFR software provides additional prognostic information and that together with the plaque characteristics derived by VH-IVUS enables more accurate detection of the lesions associated with MACE-R.

Large scale prospective intravascular imaging studies of coronary atherosclerosis have demonstrated that an invasive assessment of plaque morphology allows detection of high-risk plaque features and identification of lesions that are likely to progress and cause events [9, 14–16]. However, in these studies the positive predictive value of the studied invasive imaging modalities in predicting MACE was low and therefore their routine use to stratify risk in the clinical setting is currently not recommended [17]. Combination of plaque morphology and physiology and in particular estimation of the endothelial shear stress (ESS) using computational fluid dynamic analysis of intravascular imaging data seems to provide additional prognostic information and detection of vulnerable plaques with a positive predictive value that exceeds  $>50\%$  [11, 18]. However, ESS computation is a time consuming process that requires dedicated software and expertise facts that limit its broad use in the clinical practice.

FFR was introduced to assess in real time the physiologic implications of coronary lesions and detect those that cause flow obstruction and ischemia. Reports have shown an inverse association between baseline FFR and the incidence of future events highlighting the prognostic value of this metric, even in non-flow limiting lesions [19, 20]. Studies also support a high agreement between FFR and QFR in patients with an acute coronary syndrome and underscore the value of QFR in detecting lesions that are likely to progress and cause events [21, 22]. The prognostic value of QFR/FFR is also supported by reports comparing their estimations with IVUS imaging variables showing a weak but statistical

significant correlation between MLA or PB and the FFR values. Moreover, the COMPETE-OCT sub-study and a recent report that used multimodality intravascular imaging to assess plaque morphology demonstrated that significant stenoses are more likely to have a vulnerable phenotype [23, 24]. These findings indicate that there is an association between low FFR values and high-risk plaque features (i.e., MLA, PB and a vulnerable phenotype) that are predictors of future events.

The present study examined for the first time the combined value of computationally-derived FFR and plaque morphology in predicting events. We analysed data from the PROSPECT and IBIS 4 studies and used an established software that is capable of retrospectively processing angiographic imaging data to derive FFR [7]. We found that QFR provided additional prognostic information to plaque morphology and enabled more accurate prediction of MACE. In particular 40.4% of the lesions with high risk plaque characteristics and low QFR values caused MACE while the event rate was lower in lesions with  $QFR \geq 0.96$  or a non-high risk plaque morphology. These findings highlight the potential importance of the combined assessment of plaque morphology and physiology for more accurate characterisation of plaque vulnerability. Future prospective studies or retrospective analyses of currently ongoing intravascular imaging studies of atherosclerosis should be performed to validate the findings from the present report. Such studies may provide insight about the predictive efficacy of this approach in detecting vulnerable plaques and high-risk patients who might benefit from an aggressive treatment of atherosclerosis [25, 26].

Advances in image processing and computational methods are anticipated to facilitate research towards this direction. Several solutions have been proposed in the literature for intravascular imaging analysis and fast computation of the pressure gradient across a lesion at rest or during hyperemia [27–29]. The OFR software is the first user-friendly tool that enables real-time segmentation of optical coherence tomography data and evaluation of lesion severity using computational modelling. Validation of this approach has shown a high agreement between OFR and FFR estimations [30, 31]. Similar software are expected to be designed in the future for the analysis of IVUS imaging data and used to predict more accurate than standalone imaging lesions that are likely to progress and cause events.

#### *4.1. Study Limitations*

The key limitations of the present retrospective study is the fact that a large proportion of lesions were excluded from the analysis as accurate co-registration of IVUS and X-ray imaging was not possible or the DICOM information required to estimate QFR was not available in the angiograms acquired in the PROSPECT study. In addition, QFR analysis was performed in angiographic data that had already been acquired; a fact that may have affected the accuracy of QFR in assessing the hemodynamic severity of a lesion. Moreover, the number of events reported was small and thus in order to avoid model overfitting we included in the multivariate model only 2 non-colinear variables– presence of PB>70% and MLA<4mm<sup>2</sup> and QFR – out of the 6 variables that were associated with MACE in the univariate analysis. Therefore, this analysis should be considered as an exploratory and hypothesis generating analysis requiring validation from larger studies. Moreover, VH-IVUS imaging was carried out using two different imaging systems (s5 in IBIS 4 and Invision Gold in PROSPECT); VH assessment from these two systems could give different estimations for plaque composition but we believe that it will have less effect in the estimation of plaque phenotype. In addition, the analysis of the VH-IVUS data was performed by two different core-labs that however are in close collaboration and use the same classification algorithm to define plaque phenotype [12]. Finally, the follow-up period (median 3.4 years in PROSPECT and 5 years in IBIS 4) and the clinical end-points were different in the two studies as IBIS 4 study also reported revascularizations because of disease progression in coronary angiography and established evidence of ischemia at 13 months follow-up.

#### **4. Conclusions**

In the present post-hoc analysis assessment of plaque physiology using commercially available QFR software appear to provide additional prognostic information and more accurate identification of lesions that are likely to progress and cause MACE at 5-year follow-up compared with plaque morphology assessment derived from VH-IVUS alone. However, the small number of events reported do not allow us to draw firm conclusions; therefore, further confirmatory research is needed in a larger number of patients to quantify the predictive accuracy of combined QFR and intravascular imaging for the identification of vulnerable plaques and vulnerable patients.



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## 5. References

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## Figure legends

**Figure 1.** Flowchart of study design.

**Supplementary 2.** Case examples that underscore the additional value of QFR in predicting lesions associated with MACE-R. Panel (A) shows the angiographic projection of a lesion that progressed and required revascularization at 1-year follow-up. VH-IVUS assessment of this lesion demonstrated a TCFA phenotype with a MLA of 4.11mm<sup>2</sup> and a PB of 71.7% (B). Panel C shows the QCA analysis of the segment assessed by VH-IVUS which indicates a minor stenosis (MLD: 2.0mm, DS: 42.0%), while panel (D) the estimated QFR values (fQFR: 0.94, cQFR: 0.94). Panel (E) portrays the angiographic projection of a lesion that remained quiescent. VH-IVUS examination demonstrated a TCFA phenotype with a MLA of 3.77mm<sup>2</sup> and PB of 60.9% while in QCA analysis the estimated MLD and %DS were similar to the previous lesion (MLD: 2mm, %DS: 34.5). In this occasion however the QFR value was higher (fQFR: 0.98, cQFR: 0.97) than the lesion that progressed and require revascularisation.

**Figure 3.** Kaplan-Meier curves showing the MACE-R rate in lesions classified according to VH-IVUS and QFR metrics. Lesions were classified in groups according to the presence of absence of (A) high-risk plaque morphology (i.e., plaques with  $\geq 2$  high-risk features: defined as MLA<4mm<sup>2</sup>, PB>70% and a TCFA phenotype) and a low cQFR value (<0.97) and of (B) PB>70% and MLA<4mm<sup>2</sup> and low cQFR.