Prediction of occult atrial fibrillation in patients after cryptogenic stroke and transient ischaemic attack: PROACTIA

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Aims

Studies with implantable cardiac monitors (ICMs) show that one-third of patients with cryptogenic stroke/transient ischaemic attack (TIA) have episodes of subclinical atrial fibrillation (SCAF) and benefit switching from antiplatelet- to anticoagulant therapy. However, ICMs are costly and resource demanding. We aimed to build a score based on participant’s baseline characteristics that could assess individual risk of SCAF.

Methods and results

In a prospective study, 236 eligible patients with a final diagnosis of cryptogenic stroke/TIA had an ICM implanted during the index hospitalization. Pre-specified evaluated variables were: CHA2DS2-VASc, P-wave duration, P-wave morphology, premature atrial beats (PAC)/24 h, supraventricular tachycardia/24 h, left atrial end-systolic volume index (LAVI), Troponin-T, NT-proBNP, and D-dimer. SCAF was detected in 84 patients (36%). All pre-specified variables were significantly associated with SCAF detection in univariate analysis. P-wave duration, followed by PAC/24 h, NT-proBNP, and LAVI, had the largest ratio of SCAF prevalence between its upper and lower quartiles (3.3, vs. 3.2, vs. 3.1 vs. 2.8, respectively). However, in a multivariate analysis, only PAC/24h, P-wave duration, P-wave morphology, and LAVIs remained significant predictors and were included in the PROACTIA score. Subclinical atrial fibrillation prevalence was 75% in the highest vs. 10% in the lowest quartile of the PROACTIA score with a 10-fold higher number of patients with an atrial fibrillation burden >6 h in the highest vs. the lowest quartile.

Conclusion

The PROACTIA score can identify patients with cryptogenic stroke/TIA at risk of subsequent SCAF detection. The large difference in SCAF prevalence between groups may provide a basis for future tailored therapy.

Clinical trial registration

Clinical Trial Registration: ClinicalTrials.gov; NCT02725944.

Keywords

Embolic stroke of undetermined source • Subclinical atrial fibrillation • Implantable loop recorder • Implantable cardiac monitors • Secondary stroke prevention

Introduction

About one-third of all stroke cases are cryptogenic, i.e. strokes that remain without an identifiable cause even after a comprehensive evaluation. However, 20–34% of cryptogenic stroke cases have episodes of subclinical atrial fibrillation [SCAF—i.e. asymptomatic episodes of atrial fibrillation (AF) detected by monitoring devices] during extended monitoring and follow-up, with higher detection rates the longer the monitoring time. Implantable cardiac monitors (ICMs) with validated algorithms for AF detection have demonstrated a superior ability to detect SCAF compared to other rhythm monitoring strategies. Furthermore, modern ICM have home-monitoring capabilities that enable timely initiation of oral
What’s new?

- The most comprehensive prospective evaluation of predictors for device detected subclinical atrial fibrillation (SCAF) in patients with cryptogenic stroke or transient ischaemic attack (TIA).
- In univariate analysis CHA²DS²-VASc, P-wave duration, P-wave morphology, premature atrial beats/24 h, supraventricular tachycardia/24 h, left atrial end-systolic volume index, Troponin-T, NT-proBNP, and D-dimer were significantly associated with SCAF.
- In multivariate analysis, only premature atrial contractions/24 h, P-wave duration, P-wave morphology, and LAVIs remained significant predictors and were included in the PROACTIA score.
- The highest vs. the lowest quartile of PROACTIA score had a seven times higher prevalence of SCAF and a 10-fold higher number of patients with an atrial fibrillation burden >6 h.
- The PROACTIA score based on easily available baseline variables can provide a basis for individually tailored therapy in patients with cryptogenic stroke or TIA.

Methods

Study design and population

The PROACTIA study is an event-driven prospective single centre cohort study.

Patients hospitalized for first-time stroke or TIA, were screened for study inclusion with the registration of medical history, carotid Doppler ultrasound, 12-lead electrocardiogram (ECG), 24 h-Holter ECG registration and verification of ischaemic stroke by cerebral computed tomography, and/or magnetic resonance imaging as part of the routine examinations. Eligible patients with embolic stroke of undetermined source (ESUS, i.e. a non-lacunar stroke without proximal arterial stenosis or known cardio-embolic source) were invited to the study and were examined with echocardiography. After verifying the stroke or TIA as cryptogenic according to the TOAST (Trial of Org 10172 in Acute Stroke Treatment) criteria, the patients were enrolled in the study. All study participants had an ICM (Reveal Linq; Medtronic, Inc., Minneapolis, MN, USA) implanted prior to discharge. Stroke recurrence during follow-up was defined as a new clinical event with a corresponding new area of brain infarct. Patients with known or newly detected AF, non-AF indications for OAC or contraindications for OAC, were excluded. Those unable to sign the informed consent or with a life expectancy of <2 years were also excluded.

The study was conducted according to the Declaration of Helsinki and was approved by the Regional Committees for Medical and Health Research Ethics with reference number 2014/1260. All study participants provided written informed consent before study inclusion. The trial is registered at ClinicalTrials.gov (No. NCT02725944).

Data sources/measurement

The following variables were pre-specified to be evaluated as possible predictors of the individual risk for underlying SCAF:

- CHA²DS²-VASc score.
- Number of supraventricular tachycardia (SVT)/24 h and premature atrial contractions (PAC)/24 h in baseline 24 h Holter ECG.
- P-wave duration and P-wave morphology in baseline 12-lead ECG.
- Left atrial systolic volume index (LAVIs) in baseline echocardiography.
- Biomarkers: D-dimer, high-sensitivity cardiac Troponin-T, NT-proBNP.

12-lead resting ECG was recorded on paper at a speed of 50 mm/s, and P-wave duration and P-wave morphology were measured manually. P-wave duration was measured as the longest P-wave on 12-lead ECG. P-wave morphology was considered positive if the P-wave was biphasic in the inferior leads.

The 24 h Holter ECG was registered using OxyHolter® Recorder (Maynard, MA, USA). Episodes of irregular ventricular rhythm without detectable sinus P-waves lasting more than 30 s were considered as AF episodes, and the patients were treated with oral anticoagulants (OACs), primarily direct oral anticoagulants (DOACs), and not included in the study. Supraventricular tachycardia was defined as three or more consecutive PAC beats. The total number of SVTs and PACs were calculated automatically by OxyHolter® Recorder software and reviewed manually to exclude artefacts. The total numbers were divided by monitoring time and given as SVT/24 h and PAC/24 h.

Transthoracic echocardiographic images were obtained using GE Vivid E9 (GE Healthcare, Horten, Norway), with MSS probe, and stored electronically for later off-line analysis with ComPACS (v.10.6. MedMatic, Genova, Italy). Left atrial volume was determined by the modified Simpson’s biplane model and indexed using body surface area. The analyses were performed according to current guidelines.

The ICM was implanted subcutaneously under local anaesthesia according to the vendor’s recommendation. No prophylactic antibiotics were used. Of the 251 implanted monitors only one adverse event was observed (pocket infection within 1 week after the implantation). The monitor was programmed for AF only and atrial tachycardia with a tachycardia alert just above the estimated maximal heart rate level according to the formula, ‘220 minus age’. Atrial fibrillation was defined as an episode of irregular heart rhythm, without detectable P-waves, lasting more than 30 s, and adjudicated by two cardiologists. Details of the AF detection algorithm have been reported previously. The home-monitoring analyses were performed once weekly. Current medication was registered from the patient’s charts upon discharge.

Antplatelet therapy was switched to OAC (primarily DOACs) once AF was detected.

Missing data

Initial ECG data were lost for two patients and were, therefore, extracted from the ECG registered during the control visit after the 1 year follow-up. The NT-proBNP and D-dimer analyses from the admission blood sampling were not available in 10 and 27 patients, respectively. They were substituted with values obtained from blood sampling upon study inclusion. Testing in additional patients did not reveal any systematic variance between admission and inclusion values (data not shown).
Bias
All ECG and echocardiographic analyses were performed at the time of inclusion without knowledge of later SCAF detection status to eliminate the risk of bias. Furthermore, the principal investigator (LSS) interpreted the ICM recordings without access to the clinical background data or results, and ICM-detected AF episodes were adjudicated by a second cardiologist (HK) blinded to the baseline data. In cases of disagreement, a third blinded cardiologist determined the result. Data analyses were performed according to the original plan.

Ethical approval
Ethical approval was obtained from the Regional Committees for Medical and Health Research Ethics with reference number 2014/1260.

Informed consent
All patients provided informed written consent.

Statistical methods
Continuous values are expressed as mean and standard deviation (SD), or median and interquartile range (IQR) when non-normally distributed. Nominal variables are presented as counts and percentages. Non-detectable values for Troponin-T <0.05 and for D-dimer <0.3 were given as 0.04 and 0.2, respectively. A two-sided t-test was used for comparison of normally distributed variables and the non-parametric Kruskal–Wallis test for non-normally distributed values. Categorical data were analysed using two-tailed χ² statistics with Yates’s correction. Univariate- and multivariate regression analyses were performed to estimate the risk for SCAF. To be considered for inclusion in a multivariate risk model, baseline variables were required to be recorded in ≥90% of patients and found to be significant (P ≤ 0.05) in a univariate logistic regression model. Data for PAC/24 h, SVT/24 h, Tnt, proBNP were log-transformed prior to regression analysis due to non-normal distribution. To avoid non-computable values in patients with 0 PAC/24 h or 0 SVT/24 h the following transformation were used: log(1 + PAC/24 h) and log(1 + SVT/24 h). Receiver operator characteristic (ROC) curves and area under the curve (AUC) were calculated for the estimation of incremental prognostic information on SCAF prediction. Variable selection for the logistic regression model was based on AUC scores calculated using leave-one-out cross-validation (LOOCV). The AUC score is the area under the ROC curve, or equivalently the probability that the model scores a random positive case higher than a random negative one. Leave-one-out cross-validation works by leaving one data point out of the data set at a time and estimating a separate model instance for

![Figure 1](flowchart.png)

*Figure 1* Flowchart describing the screening and enrolment process. ICM, implantable cardiac monitor; TIA, transient ischaemic attack.
The clinical characteristics of the 236 study participants are presented in Table 1. The inclusion process is presented in Figure 1. The demographics and clinical characteristics of the 236 study participants are presented in Table 1.

Results

The inclusion process is presented in Figure 1. The demographics and clinical characteristics of the 236 study participants are presented in Table 1.

Table 1 Demographics and clinical characteristics of study participants

|                        | Total n = 236 | Non-AF n = 152 | SCAF n = 84 | P-value |
|------------------------|---------------|----------------|-------------|---------|
| Female, n (%)          | 90 (38)       | 59 (39)        | 31 (37)     | 0.726   |
| Male, n (%)            | 146 (62)      | 93 (61)        | 53 (63)     |         |
| TIA, n (%)             | 61 (26)       | 44 (29)        | 17 (20)     | 0.151   |
| Stroke, n (%)          | 175 (74)      | 109 (71)       | 67 (80)     |         |
| Age, years, mean (SD)  | 68.6 (12.5)   | 66.7 (13.0)    | 72.1 (10.7) | 0.001   |
| BMI, kg/m^2, mean (SD) | 27.2 (4.5)    | 27.0 (4.3)     | 27.6 (4.8)  | 0.316   |
| Current smoker, n (%)  | 28 (12)       | 18 (12)        | 10 (12)     | 0.854   |
| CHA2DS2-VASc, median (IQR) | 4 (3–5)   | 4 (3–5)        | 5 (4–6)     | 0.014   |
| HT, n (%)              | 150 (64)      | 87 (57)        | 63 (75)     | 0.007   |
| Vascular disease, n (%)| 40 (17)       | 25 (16)        | 15 (18)     | 0.782   |
| CHF, n (%)             | 8 (3)         | 3 (2)          | 5 (6)       | 0.106   |
| DM, n (%)              | 30 (13)       | 17 (11)        | 13 (16)     | 0.343   |
| DVT/PE, n (%)          | 6 (3)         | 2 (1)          | 4 (5)       | 0.105   |
| P-duration, ms, mean (SD) | 109.6 (18.7) | 106.1 (16.7)  | 116 (20.5)  | 0.026   |
| P-morphology bimodal, n (%) | 122 (51.7) | 67 (44)        | 56 (67)     | 0.002   |
| PAC/24-h ECG, median (IQR) | 90 (35–460) | 66 (28–206)    | 347 (59–1917) | <0.001 |
| SVT/24-h ECG, median (IQR) | 1 (0–5)     | 1 (0–3)        | 3 (1–27)    | <0.001 |
| D-dimer, mg/L, median (IQR) | 0.4 (0.2–0.7) | 0.3 (0.2–0.6) | 0.5 (0.3–0.9) | <0.001 |
| TnT, ng/L, median (IQR) | 11.5 (7–18)  | 10 (7–16)      | 14.5 (9.3–26.8) | <0.001 |
| NT-proBNP, ng/L, median (IQR) | 143 (59–387) | 104.5 (51–283) | 245 (102–774) | <0.001 |
| LAVI, mL/m^2, mean (SD) | 37 (11)       | 35 (9)         | 42 (12)     | <0.001 |
| Beta-blockers, n (%)   | 61 (26)       | 34 (22)        | 27 (32)     | 0.095   |
| ACEI/ARB, n (%)        | 104 (44)      | 59 (39)        | 45 (54)     | 0.026   |
| Ca-channel blockers, n (%) | 59 (25)   | 33 (22)        | 26 (31)     | 0.11    |
| Diuretics, n (%)       | 39 (17)       | 23 (27)        | 16 (11)     | 0.001   |
| Statins, n (%)         | 206 (87)      | 134 (88)       | 72 (86)     | 0.297   |
| Ezetimib, n (%)        | 10 (4)        | 8 (5)          | 2 (2)       |         |
| ASA, n (%)             | 27 (11)       | 16 (11)        | 11 (13)     | 0.541   |
| Antiplatelet therapy other than ASA, n (%) | 212 (90) | 136 (90) | 76 (91) | 0.014 |

SCAF, subclinical atrial fibrillation; TIA, transient ischaemic attack; SD, standard deviation; IQR, interquartile range; 25th percentile and 75th percentile values; BMI, body mass index; HT, hypertension; CHF, congestive heart failure; DM, diabetes mellitus; DVT, deep venous embolism; PE, pulmonary embolism; ACEI, angiotensin-converting enzyme inhibitors; ARB, angiotensin receptor blockers (angiotensin subtypes I and II: AT1 and AT2); PAC, atrial premature beat; LAVI, left atrial volume index.
three were gastrointestinal bleedings (one on DOAC treatment, one on antiplatelet therapy and one on DOAC and antiplatelet therapy due to diagnosed acute coronary syndrome during follow-up).

All the pre-specified variables were significantly associated with SCAF detection in univariate analyses. The univariate results are presented together with the prevalence of SCAF detection on follow-up for each quartile of the individual pre-specified variables in Table 2.

Of the pre-specified variables, only PAC/24t, P-wave duration, P-wave morphology, and LAVIs remained as significant predictors in the multivariate analysis, and the predictive power was not improved by adding other variables to the equation. The PROACTIA score was, therefore, based on the formula derived from the multivariate analysis:

\[
0.05472 \times \text{LAVIs mL/m}^2 + 0.95928 \times \log(1 + \text{PAC/24 h}) + 0.03615 \times \text{Pdur ms} + 1.05513 \times \text{Pmorph}.
\]

Receiver operator characteristic-curve characteristics of different prediction models with their respective AUC values and confidence intervals are presented in Figure 2.

Demographics and clinical characteristics of study participants divided by quartiles of PROACTIA score are described in Table 3. The SCAF prevalence in the highest quartile was 7.3 times higher than the SCAF prevalence observed in the lowest quartile (75% vs. 10%) (Figure 3). Furthermore, the AF burden within the lowest PROACTIA score quartile was significantly lower than the corresponding values observed in the higher quartiles (Figure 3).

**Discussion**

The PROACTIA study detected SCAF in 36% of patients which is in line with previous reports using ICM in cryptogenic stroke patients as...
summarized in a recent meta-analysis. All patients with detected SCAF were switched to OAC within days of detection. With this individualized strategy, we observed a very low rate of recurrent stroke (2/100 patient years) combined with a low bleeding incidence (0.9/100 patient year). This is in line with the meta-analysis by Tsivgoulis et al. that demonstrated a decreased risk ratio of recurrent stroke—0.45 (95% CI, 0.21–0.97) in cryptogenic stroke/ESUS patients who underwent prolonged rhythm monitoring with the initiation of OAC upon detection of SCAF episodes >30 s versus standard of care without prolonged rhythm monitoring. In the 2021 Guideline for the prevention of stroke in patients with stroke and TIA long-term rhythm monitoring is recommended for cryptogenic stroke patients with the initiation of OAC regardless of the time spent in AF because of the high risk of recurrent stroke in these patients even with brief subclinical episodes of AF. With the high prevalence of SCAF in cryptogenic stroke/TIA patients, the effect of DOAC treatment in unselected cryptogenic stroke/TIA patients was tested in NAVIGATE ESUS and RESPECT ESUS. Neither showed a benefit of DOAC treatment in unselected cryptogenic stroke/TIA patients. However, a subgroup analysis of the NAVIGATE-ESUS study, selecting patients with increased left atrium diameter (9% of the total study population), demonstrated a marked reduction in yearly stroke recurrence among those treated with rivaroxaban as compared to aspirin (1.7% vs. 6.7%). This result supports the notion that individually tailored therapy, based on baseline characteristics, may be feasible for cryptogenic stroke and TIA patients.

Patients in PROACTIA received individually tailored therapy during follow-up by means of ICMs. However, as stated initially, available resources and costs are currently limiting the use of ICM technology in patients with cryptogenic stroke and TIA. Hence, there is an urgent need for novel measures enabling reliable risk stratification based on baseline data in these patients. Therefore, we aimed to build such a risk stratifying model. We found SCAF to be associated with all the pre-specified risk factors in univariate analysis. However, in multivariate analysis only PAC/24t, P-wave duration of 12-leads ECG; PAC; atrial premature beat; SVT; supraventricular tachycardia; LAVI, left atrial volume index.
Several studies have evaluated scoring systems to predict AF, but most have relied on non-invasive methods for AF detection with a lower AF detection rate than provided by modern ICM technology, or were not restricted to a cryptogenic stroke/TIA population thus complicating comparison with our study. We have, therefore, focused on studies that have utilized ICM for SCAF detection in patients with cryptogenic stroke or TIA. The largest reported study was by Xu et al., who retrospectively reviewed 389 patients followed with an ICM implanted after cryptogenic stroke with a SCAF detection rate of 26%. They found age and left atrial enlargement to be independently predictive of SCAF detection, while CHA2DS2-VASc and co-morbidities such as hypertension, diabetes, obstructive sleep apnoea, and coronary artery disease were not. However, no Holter ECG data, ECG measures, or biomarkers were reported. A history of SVT/PAC was twice as frequent in patients with SCAF, but was not statistically significant in their study.

However, concordant to our findings, other ICM-based studies that did report Holter ECG data found significant associations between frequency of PAC on 24 h Holter and SCAF detection. Furthermore, ECG indices of atrial myopathy such as P-wave duration and morphology, have also been found associated with SCAF detection by ICM.

The CHA2DS2-VASc score has been a cornerstone in the evaluation of AF patients with respect to the risk of stroke and the need for OAC. However, the association between CHA2DS2-VASc score and SCAF detection is not well documented and was not found in the meta-analysis from Tsivgoulis et al.

In a substudy of CRYSTAL AF Zhao et al. evaluated both CHADS2 and the clinically based HAVOC score with points for age (≥75), obesity, congestive heart failure, hypertension, coronary artery disease, peripheral vascular disease, and valve disease, and found higher SCAF prevalence with increasing CHADS2 and HAVOC score with the best prediction achieved by the HAVOC score. However, SCAF prevalence in their high-risk group was only 3 times higher than in their low-risk group (33% vs. 11%) as compared to 7.3 with the PROACTIA score. Furthermore, of the individual HAVOC score components—age was the only significant univariate predictor of SCAF. Concordantly, in PROACTIA, age alone was a more powerful predictor than CHA2DS2-VASc, but neither age nor CHA2DS2-VASc or any of the other clinical variables measured in PROACTIA remained significant as independent predictors in our final multivariate model, nor did they add predictive power to our risk score when LAVIs or PAC/24 h were included in the analysis. Similarly, even though the biomarkers were increased in patients with subsequent SCAF detection, they did not improve the predictive power conferred by the ECG markers and LAVIs.

Limitations of the study

Even though PROACTIA is one of the largest ICM-based studies, the number of events were limited to 8.4 events for each pre-specified predictor. However, the final model contains only four predictors, which gives more than 20 events per predictor. The LOOCV was applied in the model selection process in order to counteract any risk of bias, which can indeed otherwise be a risk for datasets of limited size. The given AUC performance was also calculated using LOOCV, which effectively eliminates the ‘double use of data problem’ that arises when a model is estimated and evaluated on the same data set. However, the results should be confirmed in an independent validation cohort.

Atrial fibrillation is associated with multiple risk factors on the population level. On the individual level, not all risk factors may be present. However, those risk factors that are present may contribute synergistically in a dose-dependent manner. In the PROACTIA score the dose–response relationship of the risk factors is respected, as the score sum is calculated based on the exact values of the individual parameters from each patient, instead of cut-offs. We,
therefore, believe this enables a more robust and precise individual risk prediction of SCAF. The mathematical expression composing the PROACTIA is complex. However, with the aid of an app-based risk calculator, clinicians will easily be able to estimate individual risk in their daily practice.

**Clinical implications**

The PROACTIA score enables the identification of cryptogenic stroke/TIA patients with low, intermediate and a high risk of subsequent SCAF detection. This may provide the basis for a prospective multi-centre study of tailored therapy in which low-risk patients receive neither prolonged rhythm monitoring nor OAC, intermediate-risk patients receive prolonged rhythm monitoring, preferably by ICM, and high-risk patients receive OAC and anti-platelet medication as the primary treatment.

**Conclusions**

The study results support the primary hypothesis that the PROACTIA score built on easily available baseline ECG, 24 h Holter, and echocardiographic left atrium parameters, can identify patients with cryptogenic stroke/TIA at risk of subsequent SCAF detection. The prevalence of SCAF was 7.3 times higher in patients with the highest vs. the lowest quartile of PROACTIA score. The large difference in SCAF prevalence between groups may provide a basis for future tailored therapy in this large and resource-demanding patient population.

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**Conflict of interest:** None declared.

**Data availability**

The data that support the findings of this study are available from the corresponding author upon reasonable request.

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