1 Title page

Title: Atrial Fibrillation in Embolic Stroke of Undetermined Source: Role of advanced
imaging of left atrial function.

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1 Abstract

Background: Atrial fibrillation (AF) is detected in over 30% of patients following an
embolic stroke of undetermined source (ESUS) when monitored with an implantable loop
recorder (ILR). Identifying AF in ESUS survivors has significant therapeutic implications
and AF risk is essential to guide screening with long-term monitoring. The present study
aimed to establish the role of Left Atrial (LA) function in subsequent AF identification and
develop a risk model for AF in ESUS.

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9 Methods: We conducted a single-centre retrospective case-control study including all 10 patients with ESUS referred to our institution for ILR implantation from December 2009 11 to September 2019. We recorded clinical variables at baseline and analyzed transthoracic 12 echocardiograms in sinus rhythm. Univariate and multivariable analyses were performed 13 to inform variables associated with AF. Lasso regression analysis was used to develop a 14 risk prediction model for AF. The risk model was internally validated using bootstrapping. 15

Results: Three hundred and twenty-three patients with ESUS underwent ILR 16 implantation. In the ESUS population, 293 had a stroke, whereas 30 had suffered a TIA 17 as adjudicated by a senior stroke physician. AF of any duration was detected in 47.1%. 18 Mean follow-up was 710 days. Following lasso regression with backward elimination, we 19 combined increasing lateral PA (the time interval from the beginning of p wave on surface 20 electrocardiogram to the beginning of A' wave on pulsed wave tissue Doppler of the lateral 21 22 mitral annulus) (OR 1.011), increasing Age (OR 1.035), higher diastolic blood pressure (DBP) (OR 1.027) and abnormal LA reservoir Strain (OR 0.973) into a new PADS score. 23

- 1 The probability of identifying AF can be estimated using the formula: Model discrimination
- 2 was good (AUC 0.72). The PADS score was internally validated using bootstrapping with
- 3 1000 samples of 150 patients showing consistent results with an AUC of 0.73.
- 4
- 5 **Conclusions:** The novel PADS score can identify the risk of AF on prolonged monitoring
- 6 with ILR following ESUS and should be considered a dedicated risk-stratification tool for
- 7 decision-making regarding the screening strategy for AF in stroke.
- 8

9 Keywords:

- 10 Atrial fibrillation, embolic stroke of undetermined source, ESUS, transient ischaemic
- 11 attack, prediction model, risk score
- 12

13 Lay Summary

14 One third of patients with a type of stroke called Embolic Stroke of Unknown Source (ESUS) also have a heart condition called Atrial Fibrillation (AF), which increases their 15 risk of having another stroke. However, we don't know why some patients with ESUS 16 develop AF. To figure this out, we studied 323 patients with ESUS and used a special 17 device to monitor their heart rhythm continuously for up to three years, an implantable 18 loop recorder (ILR). We also looked at their medical history, performed a heart 19 ultrasound, and identified some factors that increase the risk of identifying AF in the 20 future. 21

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- Factors associating with future AF include older age, higher diastolic blood
 pressure and problems with the coordination and function of the upper left
 chamber of the heart called left atrium.
- Based on these factors, we created a new scoring system that can identify
 patients who are at higher risk of developing AF better than the current scoring
 systems, the PADS score. This can potentially help doctors provide more
- 7 targeted and effective treatment to these patients, ultimately aiming to reduce
- 8 their risk of having another stroke.
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1 Main Text

2 3

4 Introduction

Stroke is one of the leading causes of morbidity and mortality in the Western world, 5 affording an increasing financial burden to healthcare systems.¹ The global lifetime risk of 6 stroke in individuals over the age of 25 is estimated at 25%.² In approximately one third 7 of patients with ischaemic stroke no immediate cause is identified, classified as Embolic 8 Stroke of Undetermined Source (ESUS).^{3,4} With detailed investigations, a significant 9 proportion of patients with ESUS (> 30%) are subsequently identified as having underlying 10 paroxysmal atrial fibrillation (pAF), which may explain the index event.^{5,6} Correctly 11 identifying AF in ESUS survivors is vital as it guides clinicians toward initiation of 12 anticoagulation, which reduces stroke recurrence by almost 65%.^{7,8} 13

In the absence of AF, recent trials have suggested that anticoagulation offers no clinical benefit and may be of harm in ESUS survivors.^{9,10} However, subgroup analysis of one of these trials has provided evidence that patients with markers for increased risk of AF, may derive benefit from empirical anticoagulation.¹¹ Therefore, the ability to identify individuals at risk for AF is of vital clinical importance.

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Unfortunately, pAF remains challenging to diagnose in practice.^{7,12} Long-term monitoring using an implantable loop recorder (ILR) has proven to be the optimal method for screening of pAF.^{5,6,13,14} The usefulness of ILR in the context of ESUS is recognized by both the recent American Heart Association (AHA)¹⁵ and European Society of Cardiology (ESC) guidelines.¹² Indeed, implantation of an ILR in all ESUS patients would be an ideal method of identifying AF in this cohort, but this practice is resource-intensive, expensive, and not yet widely accepted.¹⁶ The recent ESC guidelines acknowledge this, and
recommend the use of ILR in a targeted group of stroke patients only, yet the guidance
did not provide a method by which suitable individuals should be identified.¹²

4

Individual risk assessment is therefore a potential method by which patients with a high 5 likelihood of subsequent AF could be targeted for ILR implantation. Several risk scores 6 have been developed and existing risk scores have been utilized to predict AF in patients 7 following an ischaemic stroke or transient ischaemic attack (TIA).¹⁷⁻¹⁹ A significant 8 limitation of the studies attempting to develop AF risk prediction models in an ESUS 9 population is the lack of prolonged cardiac rhythm monitoring with an ILR to diagnose AF, 10 which reduces the sensitivity of the scoring system, as lack of long-term monitoring leads 11 to underestimation of AF episodes. Indeed, none of the risk scores perform sufficiently 12 well in patients with ESUS to be incorporated in the guidelines and are not widely used.²⁰⁻ 13 29 14

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Therefore, there is an urgent unmet clinical need for a robust risk-score that can reliably
 predict the development of AF in an ESUS population and potentially help clinicians target
 ILR implants more effectively.

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We hypothesized that imaging parameters of left atrial (LA) function would be associated with subsequent AF, and combined with other imaging and clinical parameters can help build a risk model to predict AF in patients with ESUS. Such a model could help risk stratifying ESUS survivors with regards to the AF future risk and thus tailor utilization of
 ILR monitoring.

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5 Methods

This was a single centre retrospective case- control study. The study was approved by the UK Health Research Authority (16/NW/0527) in 2016 and institutional approval from Cambridge University Hospitals NHS Foundation Trust. The North West-Preston Research Ethics committee waived the need for patient consent for this retrospective study. The study complied with the 1975 Declaration of Helsinki for research and the STROBE guidelines for observational studies were followed.

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13 Study population

14 We included all adults undergoing ILR implant to screen for AF following a cerebrovascular event of unknown aetiology between December 2009 and September 15 2019. All patients were prospectively enrolled in a dedicated clinical database, which was 16 retrospectively interrogated. Cerebrovascular events of unknown cause (ESUS) included 17 ischaemic stroke or Transient Ischaemic Attack (TIA; defined as neurological signs 18 resolving within 24 hours). Prior to referral for ILR, all patients had a 12-lead 19 electrocardiogram (ECG) confirming sinus rhythm and underwent a minimum of 24 hours 20 cardiac rhythm monitoring via inpatient telemetry or Holter monitor, which excluded AF. 21 22 Patients underwent transthoracic, transoesophageal or bubble echocardiography to identify other potential sources of embolism. Patients with patent foramen ovale (PFO), 23

regardless of the presence of atrial septal aneurysm, were included in the study. We 1 elected to include patients with PFO as this is a common finding occurring in over 25% of 2 the population.³⁰ Additionally, although its prevalence it higher amongst patient with 3 ESUS the condition itself has not been shown to increase the risk of ischaemic stroke.^{31,32} 4 All patients underwent either Carotid Doppler, computed tomography angiography (CTA) 5 or magnetic resonance angiography (MRA) to ensure that there was no significant 6 intracranial or extracranial significant vessel stenosis (>50%) or occlusion in the arterial 7 distribution of the index stroke or TIA. Patients with > 50% stenosis that was not in the 8 arterial distribution of the index event were included in the study. All patients had either 9 brain CT or MRI or both. Referral for ILR was at the discretion of the stroke physicians 10 after completion of the investigations and exhaustive exclusion of other explanations for 11 the index event. 12

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14 Study variables

15 <u>Demographic, anthropometric and clinical variables</u>

Demographic and anthropometric data, clinical risk factors, smoking status and alcohol 16 intake were collected from electronic and paper medical records. Additionally, we 17 recorded systolic blood pressure (SBP) and diastolic blood pressure (DBP) at the first 18 19 clinic visit following index stroke. Medications at discharge for patients admitted with an ESUS or following clinic visit for those referred for outpatient review were also recorded. 20 Results of blood biomarkers at the time of admission with a stroke or review at the 21 22 outpatient clinic were collected. A summary of the variables collected is shown in Supplementary Table 1. 23

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8	Echocardiographic variables
9	Echocardiograms performed up to one year prior to ILR implantation were included in the
10	analysis. All the echocardiographic images were digitally stored in an Image Vault (GE
11	Vingmed Ultrasound AS, Cambridge, United Kingdom). Analysis was undertaken offline
12	by British Society of Echocardiography accredited cardiologist (PAC) using
13	EchoPac v203.59 (GE), who was blinded to whether patients had subsequent AF or not.
14	Intra-observer variability was assessed using Bland-Altman plot, which did not show any
15	significant variability (supplementary figure 1a and 1b).
16	
17	Conventional echocardiographic data was obtained in accordance with American Society
18	of Echocardiography and European Association of Cardiovascular Imaging
19	recommendations.35,36,37,38,39 From the parasternal long-axis view the following
20	parameters were recorded: left ventricular (LV) dimensions and mass, aortic root
21	dimensions and LA diameter. LA volume, LV end-systolic and end-diastolic volumes and
22	LV ejection fraction (LVEF%) were determined using Simpson's biplane method from the
23	apical 4- and 2-chamber views. Diastolic function was described with E wave deceleration
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We calculated scores that have previously been used for AF risk prediction including

HAVOC,^{20,21} CHA₂DS₂VASc,^{22,26} HATCH,²⁶ C₂HEST,²³ Brown ESUS-AF,²⁴ NDAF²⁷ as

well as HAS-BLED^{12,33} and ORBIT risk scores³⁴ as shown in **Supplementary Table 2**.

time, E/A and E/E' ratio, based upon the average of the septal and lateral E' values. Atrial 1 electromechanical delay reflecting atrial dyssynchrony was 2 assessed usina electrocardiographic P-wave to lateral tissue Doppler A' wave, which will henceforth be 3 referred to as the lateral PA. This was defined as the time interval from the onset of the 4 p-wave on the surface ECG to the onset of the A' wave obtained using pulsed tissue 5 Doppler imaging of the lateral mitral annulus in the apical 4-chamber window (figure 6 1).^{40,41} A number of studies have assessed atrial electromechanical delay using tissue 7 Doppler imaging rather than electrophysiological studies.^{41–43} 8

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10 11

Figure 1 shows the measurement of lateral PA interval by tissue Doppler imaging. Lateral PA was obtained from the lateral mitral annulus in apical 4-chamber view as the time interval from the beginning of p wave on surface ECG to the beginning of A' wave. In this case lateral PA was measured as 35ms.

- 15 ECG, electrocardiogram
- 16

2 LA strain was determined using speckle tracking technique from standard grayscale images obtained from the apical 4- and 2-chamber windows and semi-automated 3 software (Echopac, GE). The LA endocardial border was manually traced, and the region 4 of interest was adjusted to optimize the inclusion of the atrial myocardium. The onset of 5 the QRS complex was chosen as the zero-reference point. In each view, the LA was 6 automatically divided into six segments giving time-deformation curves for a total of 12 7 segments. The average of all 12 segments was used to define three atrial strain 8 parameters including: LA reservoir strain defined as the peak atrial longitudinal strain; LA 9 contractile strain as the value corresponding to the onset of the p-wave on the surface 10 ECG; and LA conduit strain was as the difference between LA reservoir and contractile 11 strain (figure 2) ^{44,45} More positive LA strain values indicated a more favourable strain. 12

13

14 A summary of the additional parameters and how measurements were obtained is

15 shown in **supplementary Table 3**.



Figure 2 shows an example of LA strain measured using speckle strain analysis. For each apical view the software produces six time-deformation curves corresponding to six atrial segments (coloured traces). The average strain curve is defined for each window (white dotted trace). Three aspects of atrial strain (reservoir, contractile, conduit) are defined and annotated (see main text for details). The average value for reservoir

and contractile strain for all twelve segments is recorded. The conduit strain is calculated as the difference
 between reservoir and contractile strain.

3 LA, left atrium

- 4 5
- 6 ILR implant

ILRs (Medtronic Reveal XT, Reveal DX and SJM Confirm) were implanted 7 subcutaneously in an appropriately mapped left parasternal position. The Medtronic 8 9 Reveal LINQ was inserted at 45 degrees relative to the sternum above the fourth intercostal space in the V2-V3 electrode orientation using dedicated incision and insertion 10 tools. The ILRs were programmed with the AF detection algorithm "on", and tachycardia, 11 bradycardia, and patient activated detection on. The ILRs detect AF either by using 12 specific AF detection algorithm, or by recording episodes of tachycardia, bradycardia or 13 pause, which on further inspection are found to be AF. The Reveal LINQ and XT have 14 specific AF detection algorithms.^{46,47} Whilst the algorithms detect AF of duration greater 15 than 2 minutes, manual inspection of automatic and patient recorded episodes, allowed 16 for detection of shorter durations of AF. The ILRs were interrogated monthly or whenever 17 the patient activated the device. Until 2012 the ILRs were interrogated in the hospital and 18 thereafter remotely via the Medtronic CareLink [™] monitoring network. 19

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- 22 Outcome

The outcome was the detection of any AF or atrial flutter (AFL) of any duration on ILR. There is no of consensus of how much AF is harmful to patients with ESUS. Indeed, even the European Society of Cardiology guidelines are based on expert consensus. As such, we chose any duration of AF as an end-point on the basis that ESUS survivors are a highrisk cohort for further thromboembolic events. Furthermore, AF begets more AF,⁴⁸ and
the minimum duration of AF that increases thromboembolic risk is not known at this time.
We considered AF and AFL as interchangeable, as the risk of thromboembolism and
need for anticoagulation are similar.^{49,50}

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All auto-triggered and patient triggered episodes on ILR were reviewed by a senior cardiac physiologist and two cardiologists specialized in cardiac arrhythmias and accredited by the European Heart Rhythm Association (PAC, PP) to confirm presence of AF or AFL. In case of disagreement, the traces were reviewed by a third cardiologist for final adjudication. Additionally, we recorded time to ILR implantation and time to detection of first AF episode.

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13 Statistical analysis

14 Continuous variables are reported as means (standard deviation [SD]) for parametric data and median (interguartile range [IQR]) for non-parametric data after testing for normality. 15 Categorical variables were reported as proportions. Between groups comparisons were 16 made using independent t- test for parametric data and Mann Whitney U test for non-17 parametric data, after testing for normality. Categorical variables were compared using 18 chi-square test and Fisher's exact test if counts <5. Dichotomous variables with positive 19 events less than 30 were not included in the analysis, due to difficulty in demonstrating 20 homoscedasticity. 21

To investigate the relationship of all variables with the risk of developing AF, univariate and multivariable logistic regression models were fitted on the original data without imputed values using R statistical software. However, univariate and multivariable regression was only used to inform predictive variables. The final prediction model was based on lasso regression.

6

7 Missing Data

⁸ We excluded variables with >35% missing data in line with accepted statistical ⁹ practice.^{51,52}We created and analyzed 100 multiply imputed datasets where the missing ¹⁰ values were <35%. Incomplete variables were imputed under fully conditional ¹¹ specification, using the default settings of the MICE 3.12 package in R.^{53,54} The ¹² parameters of substantive interest were estimated in each imputed dataset separately ¹³ and combined using Rubin's rules. For comparison, we also performed the analysis on ¹⁴ the subset of complete cases.

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17 Model selection

Variable selection for the final model was guided by using a lasso model in each of the imputed datasets (library Glmnet in R).⁵⁵ In each of the 100 imputed datasets we ran a multivariable model with a lasso (L1) penalty to perform variable selection. Variables that were selected in at least 90 of the100 models were then considered for the final lasso model.

2 Results

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A total of 323 patients were included in the study. The mean follow up was 710 days 4 (standard deviation [SD] 442). Of the 323 patients, 152 (47.1%) were found to have 5 6 episodes of AF of any duration. Median time from ILR implant to AF detection was 177 days (interguartile range [IQR] 47, 439) and from stroke onset to AF detection 421 days 7 (IQR 261, 677). See Table 1 and supplementary table 4 for patient demographic data, 8 and clinical and echocardiographic variables both for the entire population and separately 9 for patients with and without post-stroke AF. Table 2 reflects the distribution of the 10 different atrial arrhythmias and presence of symptoms. In short, mean age was 54.7 years 11 (SD 14.8). The AF group was significantly older than the non-AF group (59.3 \pm 13.8 12 versus 50.5 ± 14.4, p < 0.0001). One hundred and twenty-six patients were females (39%). 13 Hypertension was a frequent finding in both AF and non-AF cohorts, but blood pressure 14 control was good. LV mass indexed to body surface area was significantly higher amongst 15 patients with AF (p=0.046) reflecting likely the higher rate of hypertension in the AF arm 16 (p=0.019). Moreover, all three aspects of LA strain were significantly more impaired in the 17 AF cohort (all p values <0.05). Of note, 117 patients had a PFO, of whom 47 (40.2%) 18 went on to develop AF, whereas of the 206 patients without a PFO, 105 (51.0%) 19 developed AF (p=0.06). 20

Among patients with post-stroke AF, 79 (52.0%) had the first episode detected within the first six months of monitoring, 29 (19.1%) at six to 12 months, 30 (19.7%) during the second year of monitoring and 15 (9.9%) after two years of monitoring **(figure 3)**.



1

Figure 3 shows time of AF detection in our population, indicating that 107 (70.4%) were shown to have AF
 within 12 months from implantation.

7 Risk factors for AF and score development

8 Univariate analysis is shown in **table 3**. Only variables with p-value <0.1 are included in

9 this table.

- 10
- Following lasso regression, we combined increasing lateral PA (OR 1.011), increasing
 age (OR 1.035), higher DBP (OR 1.027) and abnormal LA reservoir strain (OR 0.973)
 into the new PADS score (Lateral <u>PA, Age, D</u>iastolic BP, LA reservoir <u>S</u>train) (table 4).
- 15

1 The probability of identifying AF can be estimated using the following formula.

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 $e^{-4.06427051 + \ln(1.011) lateral PA + \ln(1.035) age + \ln(1.027) DBP + \ln(0.973) LA \ reservoir \ strain}$ $Probability of AF = \frac{c}{1 + e^{-4.06427051 + \ln(1.011) lateral PA + \ln(1.035) age + \ln(1.027) DBP + \ln(0.973) LA reservoir strain}}$ 3 4 where age is patient's age, DBP the diastolic blood pressure at first clinic visit following 5 stroke (mmHg), lateral PA the time interval from the beginning of p wave on surface ECG 6 to the beginning of A' wave on pulsed wave Doppler (ms) and LA reservoir strain the left 7 atrial reservoir strain obtained using speckle tracking echocardiography (%). 8 9 Using this score, we can estimate the predicted risk for an individual developing/ 10 identifying AF in the next three years (which is the battery life of the ILR) using the 11 formula shown above, and is shown in supplementary table 5. 12 13 For example, in a patient with ESUS and the following values: Lateral PA 81 ms, Age 14 64 years, DBP 86 mmHg, LA Reservoir strain 17%, the absolute risk of identifying AF 15 in the next three years is 70.0%. Alternatively, in someone with Lateral PA 40 ms, Age 16 17 37 years, DBP 61 mmHg, LA Reservoir strain 45%, the absolute risk of identifying AF in the next three years is 12.3%. 18 19 20 We assessed model discrimination using the area under the curve (AUC) of the Receiver Operating Characteristic (ROC) Curve. The PADS model showed an AUC of 0.72. 21 Furthermore, we internally validated the model using bootstrapping with 1000 samples of 22 23 150 patients showing consistent results with an AUC of 0.73.

PADS outperformed all the other scores known to "predict" AF; HAVOC (AUC 0.56),
CHA2DS2-VASc (AUC 0.58), HATCH (AUC 0.58), C2HEST (0.58), Brown ESUS AF (0.60)
HAS-BLED (0.61) and ORBIT scores (0.55).

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10 **Discussion**

11 PADS score development and validation

Our study was conducted to address the pressing need of identifying an appropriate 12 group of post-ESUS patients that would benefit from ILR monitoring. We investigated 13 clinical and echocardiographic parameters for AF and found that the combination of 14 advanced age, increased DBP, increasing lateral PA and impaired LA reservoir strain 15 associates with AF. Most of these factors have been demonstrated to be associated with 16 an increased risk of AF in stroke survivors in other studies. Indeed, advanced age is one 17 of the strongest predictors of AF and has been incorporated in several risk scores targeted 18 to this population.^{20,22,24,25,27,56–59} Likewise, elevated DBP reflecting elevated LA pressure 19 is also another risk factor for AF.⁶⁰ Additionally, our study showed that increased lateral 20 PA, a marker indicative of atrial electromechanical delay and reflecting LA dyssynchrony 21 is independently associated with AF. This specific relationship has not been reported 22 before amongst ESUS patients. However, increasing lateral PA has been identified as a 23 significant and independent associate of AF amongst 63 patients with pAF and 83 24 controls.⁴¹ Most importantly, similar to several studies, we found impaired LA function 25 assessed by LA strain to be associated with AF.⁶¹ This is in line with current literature 26

where LA reservoir strain has been shown to increase predictive value when added to
existing risk scores.⁶⁰

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Using these variables, we derived and validated the new PADS score, to assess the risk 4 5 of AF in patients with ESUS, a new score that outperformed all the existing scores in this field, when area under the curve is considered as a performance marker. Moreover, with 6 all ESUS patients recommended to undergo transthoracic echocardiography, the PADS 7 8 score is a relatively easy score to calculate, with only 4 variables required. Atrial strain is simple, reproducible and validated to calculate, and using manufacturer's strain analysis 9 modules, can, after atrial contouring, automatically produce mean time-deformation 10 curves. For a detailed review of how this can be undertaken please see the article by 11 Voigt et al.62 12

13

To correctly diagnose the presence of pAF and avoid underestimation of episodes, we used the gold-standard method for AF screening; monitoring with an ILR. We included LA function in our analysis intentionally, as it has been shown in the literature to be a strong and independent predictor of AF, superior to many other variables.^{63,64} To our knowledge this is the first study aimed at developing an AF risk prediction model targeted specifically to ESUS patients using ILR and incorporating advanced imaging parameters of LA function.

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1 Usefulness of PADS score

2 Our risk model provides an estimate of the percentage likelihood of AF within three years of ILR implantation, and individual institutions can tailor this predictive data as they see 3 fit to target their resource most effectively. For example, it can help identify patients at 4 "high", "medium" or "low" risk. Depending on its use, the "high" or "moderate" risk (such 5 as those with an absolute risk of more than 50% according to the authors of the current 6 paper), can be prioritized for an ILR, whilst those with a low risk (e.g. those with <20%) 7 an ILR can be deferred. Using the patient example in **supplementary table 5**, it is clear 8 that the first case with a 70% risk of identifying AF would warrant closer follow up and 9 a low threshold for ILR implantation (if this is not done routinely in the institution the 10 individual presents), whilst the second patient would have a much lower yield in 11 identifying AF had an ILR been implanted. Furthermore, this risk estimation can help 12 inform cost-effectiveness analyses with regards to ILR use, as the use in the moderate 13 14 and high-risk patients will be more cost-effective than the low-risk patients.

15

16 Incidence and duration of Atrial Fibrillation

The incidence of post-stroke AF of any duration in our population is 47.1% and similar to the one reported by Kwong et al, who investigated 9589 patients (age \ge 40) with cryptogenic stroke or TIA (45.3%). Stroke survivors with AF in this study were identified using international classification of disease codes.²⁰ It higher though than previously reported by Asaithambi et al, who looked at the prevalence of AF of any duration with ILR monitoring amongst 234 cryptogenic stroke survivors. They found an AF incidence of 29%, but the follow up was shorter comparing to our study.⁶⁵ The incidence of AF lasting >30s in our study was 31.0% and almost identicial to previously reported by cryptogenic stroke and underlying atrial fibrillation (CRYSTALAF) (30.0%).⁶ Our findings with regards to detection rate for AF lasting \geq 2 minutes (22.6%) are also similar to results published by Ziegler et al. This group examined 1247 patients with cryptogenic stroke and found an incidence of AF lasting \geq 2 minutes (detected by ILR) of 21.5% at 2 years.¹⁴

6

With regards to duration of AF we also feel, similar to Asaithambi et al., that in the context 7 of stroke, AF of any duration is clinically relevant and warrants extensive monitoring to 8 identify longer episodes at the very least, if not consideration of anticoagulation.65 This is 9 supported by the results of a recent Spanish study, which showed that anticoagulating 10 even short episodes of AF results in a decrease of stroke recurrence, although the study 11 did define AF episodes as being a minimum of 1 minute in duration.⁶⁶ In detail, the 12 investigators randomized 191 ESUS patients aged 50-89 years (mean 75.6) to either 13 14 conventional monitoring or ultra-early monitoring using ILR following ESUS. AF lasting >1min was detected in 58.5% of patients in the ILR group versus 21.3% in the usual care 15 group during 30±10 months of follow up. Consequently, anticoagulation therapy was 16 initiated in 65.5% in the ILR arm versus 37.6% of patients in the control arm. This led to 17 a much lower stroke recurrence rate in the ILR arm, 3.3% versus 10.9% in the 18 19 conventional arm, indicating that anticoagulating short AF episodes is beneficial.

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In contrast, the Atrial Fibrillation Detected by Continuous ECG Monitoring Using
 Implantable Loop Recorder to Prevent Stroke in High-risk Individuals (The LOOP Study)
 randomized 6004 individuals aged 70-90 years with at least one risk factors for stroke to

1:3 ratio of ILR monitoring or usual care. Anticoagulation was commenced if AF lasted \geq 1 2 6 min was detected. During a mean follow up of 64.5 months, AF was detected in 31.8% 3 in the ILR group versus 12.2% in the control group. Despite a three-times increase in the anticoagulation therapy in the ILR arm (29.7% versus 13.1%), there was no significant 4 reduction in the risk of stroke or system embolism (p=0.11).⁶⁷ However, the LOOP 5 investigators examined patients with risk factors for stroke, rather than patients with 6 unexplained stroke- a group recognized to be at higher thromboembolic risk. It is likely, 7 that anticoagulating even short episodes of AF is beneficial and reduces stroke 8 recurrence in patients with ESUS although this would need to be identified in prospective 9 randomized studies. 10

11

12 Future directions

Our risk prediction model also has the potential to identify a group of ESUS patients in sinus rhythm that could benefit from anticoagulation. Further studies are needed in this direction to assess the effectiveness of anticoagulating those at the highest risk of AF.

17 Study limitations

This was a retrospective case- control single centre study; however, our institute is the regional center for ILR implantation in post-stroke patients and is receiving referrals across a population of over 2 million people. Referrals for ILR were done at the discretion of the treating stroke physician, when they felt that other causes of stroke were excluded, and that the patient warranted a more prolonged search for AF. Therefore, selection bias could have occurred. TTE analysis was performed retrospectively in scans already

obtained and several measurements could not be performed as images were suboptimal. 1 2 Due to the retrospective nature of the study, where medical records were reviewed and no patient contact was necessary, we have not been able to collect data regarding 3 4 ethnicity. Moreover, parameters where over 35% of the values were missing were 5 excluded. This included parameters that have previously been identified as strong predictors of AF such as NT-pro BNP and troponin. LA reservoir strain and lateral PA 6 were missing at random in 24% and 32% of cases respectively. This was within our a 7 priori cut-off for multiple imputation, but a lower degree of missing data might have 8 provided more accurate results. During the study period, the institution practice was to 9 explant the ILR following AF detection, which precluded accurate analysis of AF burden. 10 Although we have internally validated our risk model, we have not been able to provide 11 external independent validation. Validating the PADS model in an unselected population 12 of ESUS patients would be useful. 13

14

On the other hand, strengths of our study include it being the first study aimed at developing a risk prediction model in patients specifically following ESUS incorporating TTE parameters of LA function. In addition, we used long-term monitoring with an ILR for AF detection, proving to be the best method with the highest diagnostic yield. We also included all adults diagnosed with stroke or TIA referred for an ILR to our institution, having no age limit in the inclusion criteria.

21

1 Conclusion

We have developed and internally validated the PADS risk prediction model to assess the individual risk of AF in post-stroke survivors. We incorporated imaging parameters of LA function and diagnosed AF using ILRs. This score outperformed existing AF prediction risk scores. PADS score can thus be utilized as a risk-stratification tool for decisionmaking in relation targeting ILR implant to identify AF in ESUS survivors. In addition, it may provide the ability to target anticoagulation in a suitable group of stroke patients at high risk of future AF who are currently in sinus rhythm.

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10 Disclosures

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15 Authorship

PAC, JP, PJP and VSV contributed to the conception and design of the work. PAC, RC,
LR, KK, EAW, TM, VT and VSV contributed to the acquisition, analysis, or interpretation
of data for the work. UB and AP did the statistical analysis for the project. PAC, RC and
VT drafted the manuscript. LR, UB, AP, TM, EAW, KK, JP, PJP and VSV critically revised
the manuscript. All gave final approval and agreed to be accountable for all aspects of
work ensuring integrity and accuracy.

22 Data Availability

23 Available from the corresponding author upon request

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Table 1. Baseline characteristics.							
Variable	All patients (n 323)	AF (n 152)	No AF (n 171)	P value**			
Demographic and anthropometric variables							
Age, mean (SD)	54.7 (14.8)	59.4 (13.9)	50.5 (14.4)	<0.001			
Female, n (%)	126 (39.0)	60 (39.5)	66 (38.6)	0.872			
BMI, mean (SD)	27.76 (4.7)	27.44 (4.6)	28.05 (4.8)	0.242			
Clinical variables		<u> </u>					
CCF, n (%)	1 (0.3)	0 (0)	1 (0.6)	0.319			
HTN, n (%)	131 (40.6)	72 (47.4)	59 (34.5)	0.019			
CAD, n (%)	22 (6.8)	9 (5.9)	13 (7.6)	0.548			
Diabetes, n (%)	38 (11.8)	19 (12.5)	19 (11.1)	0.699			
Cancer, n (%)	20 (6.2)	15 (9.8)	5 (2.9)	0.015			
SBP, mean (SD)	129.0 (17.6)	132.1 (16.8)	126.2 (17.9)	0.013			
DBP, mean (SD)	74.7 (10.6)	76.56 (10.7)	73.1 (10.2)	0.004			
>50% stenosis in a major extracranial/ intracranial vessel, n (%) *	16 (5.0)	11.(7.2)	5 (2.9)	0.075			
HTN treatment, n (%)	128 (39.6)	69 (45.4)	59 (34.5)	0.046			
Statins, n (%)	266 (82.3)	132 (86.8)	134 (78.4)	0.046			
Lymphocytes (10 ⁹ cells/l), mean (SD)	2.0 (1.0)	1.8 (0.7)	2.1 (1.2)	0.073			
neutrophil/lymphocyte ratio, median (IQR)	2.5 (1.8, 3.6)	2.7 (1.9, 3.8)	2.3 (1.7, 3.5)	0.035			
Platelet/lymphocyte ratio, median (IQR)	123.1 (95.3, 173.3)	131.7 (101.5, 175.0)	117.6 (92.1, 166.7)	0.046			
eGFR (ml/mln/1.73 m ²), mean (SD)	89.9 (24.5)	85.5 (22.34)	93.7 (25.8)	0.005			
CRP (mg/dL), median (IQR)	2.0 (1.0, 6.0)	2.0 (1.0, 6.0)	2.0 (1.0, 5.2)	0.374			
Alkaline phosphatase	81.0 (67.0, 101.0)	86.0 (71.0, 104.0)	78.0 (65.0, 96.0)	0.033			
Echocardiographic va	ariables						
LV mass indexed (g/m ²), mean (SD)	83.8 (19.0)	86.0 (19.6)	81.3 (18.1)	0.046			
LVEF biplane (%),	61 1 (57 9 65 0)	60 7 (57 9 64 2)	61 9 (57 3 65 2)	0 166			
LV GLS (%), mean	01.1 (57.9, 65.0)	00.7 (57.9, 04.2)	01.9 (07.3, 05.2)	0.100			
(SD) Average S' wave	16.3 (3.4)	16.2 (3.1)	16.4 (3.7)	0.756			
(cm/s), mean SD	8.7 (1.9)	8.5 (2.0)	8.9 (1.8)	0.100			
E wave deceleration time (ms), median (IQR)	217.0 (187.0, 254.0)	222.0 (191.0, 263.0)	210.0 (180.0, 239.0)	0.007			
E/A ratio, median			10(0912)	0.022			
Septal E' wave (m/s), mean (SD)	7.7 (2.5)	7.2 (2.2)	8.2 (2.7)	0.002			

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Lateral E' wave						
(cm/s), mean (SD)	10.3 (3.5)	9.9 (3.3)	10.7 (3.7)	0.073		
Lateral PA (ms),						
mean (SD)	74.7 (19.7)	78.2 (20.4)	71.4 (18.5)	0.011		
LAV maximum						
indexed (ml/m²),						
median (IQR)	25.3 (21.1, 30.8)	26.3 (21.5, 32.2)	24.2 (20.8, 28.9)	0.079		
LAV min indexed						
(ml/m²), median						
(IQR)	10.8 (8.7, 13.4)	11.3 (9.3, 14.0)	10.6 (8.2, 13.0)	0.018		
LA reservoir strain						
(%), mean (SD)	27.5 (9.1)	25.3 (7.3)	29.7 (10.1)	<0.001		
LA contractile strain						
(%), mean (SD)	15.0 (5.9)	13.4 (4.4)	14.9 (5.1)	0.018		
LA conduit strain (%),						
median (IQR)	12.1 (8.8, 17.1)	11.2 (8.3, 15.0)	13.2 (9.5, 19.1)	0.003		
Existing scores						
HAVOC, median						
(IQR)	1 (0,3)	2 (0,3)	1 (1,3)	0.041		
CHA ₂ DS ₂ -VASc,						
median (range)	3 (3,4)	4 (3,5)	3 (3,4)	0.004		
HATCH, median		\sim				
(IQR)	2 (2,3)	3 (2,3)	2 (2,3)	0.003		
C ₂ HEST score,						
median (IQR)	0 (0,1)	1 (0, 1)	0 (0,1)	0.004		
Brown ESUS AF,						
median (IQR)	0 (0,1)	0 (0,1)	0 (0,0)	<0.001		
NDAF, median (IQR)	3 (1,3)	3 (1,3)	3 (1,3)	0.215		
HASBLED, median	- ()-)		- ()-)			
(IQR)	2 (2.3)	3(2,3)	2 (2,3)	<0.001		
()	_ (_,;;)		_ (_,~,			
ORBIT, median (IQR)	1 (1,1)	1 (1,2)	1 (1,1)	0.245		
AF, atrial fibrillation; BMI, body mass index; CAD, coronary artery disease; CCF, congestive cardiac failure; cm,						
centimetre; CRP, C reactive protein; DBP, diastolic blood pressure; dL, decilitre; eGFR, estimated glomerular filtration						

rate; GLS, global longitudinal strain; HTN, hypertension; IQR, interquartile range; kg, kilogram; I, litre; LA, left atriu LAEF, left atrial emptying fraction; LAV, left atrial volume; LVEF, left ventricular ejection fraction; LVIDd, left ventricular internal diameter in end-diastole; LVIDs, left ventricular internal diameter in systole; m, meter; m² squared meter; m milligram; ms, millisecond; s, second; SBP, systolic blood pressure; SD, standard deviation; U, international units * not in the arterial distribution of the index event

**Quoted P value is for the difference between the AF and non AF groups

Table 2. Atrial arrhythmia characteristics						
Rhythm	Number of patients with arrhythmia	Number of episodes	Number of patients with symptomatic episodes			
Atrial fibrillation	114	375	10 (8.8%)			
Atrial flutter	38	188	5 (13.2%)			

Table 3. Univariate analysis.				
Lower Cl	OR	Upper Cl		
1.03	1.04	1.06		
1.09	1.71	2.67		
1.01	1.02	1.03		
1.01	1.03	1.06		
1.01	1.58	2.47		
1.01	1.82	3.30		
0.57	0.77	1.03		
0.98	0.99	1.00		
1.00	1.02	1.05		
1.00	1.01	1.02		
1.00	1.01	1.03		
1.00	1.01	1.01		
0.21	0.42	0.83		
0.76	0.84	0.94		
0.87	0.94	1.01		
0.78	0.90	1.02		
1.00	1.02	1.03		
1.00	1.03	1.06		
1.02	1.08	1.14		
0.92	0.95	0.97		
0.89	0.94	0.99		
0.89	0.92	0.97		
	Lower Cl 1.03 1.09 1.01 1.01 1.01 1.01 1.01 1.01 0.57 0.98 1.00 1.00 1.00 1.00 1.00 0.21 0.76 0.87 0.78 1.00 1.00 1.00 1.00 1.00 0.87 0.78 1.00 1.00 0.89 0.89 0.89	Lower Cl OR 1.03 1.04 1.09 1.71 1.01 1.02 1.01 1.03 1.01 1.03 1.01 1.58 1.01 1.82 0.57 0.77 0.98 0.99 1.00 1.01 1.00 1.01 1.00 1.01 1.00 1.01 1.00 1.01 1.00 1.01 0.21 0.42 0.76 0.84 0.87 0.94 0.78 0.90 1.00 1.02 1.00 1.03 1.02 1.08 0.92 0.95 0.89 0.94 0.89 0.92		

confidence interval; CRP, C reactive protein; DBP, diastolic blood pressure; dL, decilitre; eGFR, estimated glomerular filtration rate; HTN, hypertension; LA, left atrium; LAV, left atrial volume; OR, odds ratio; s, SBP, systolic blood pressure
* not in the arterial distribution of the index event

Table 4. PADS risk prediction model.					
Variable	Low CI	OR	High CI		
Lateral PA	1.00	1.01	1.03		
Age	1.02	1.04	1.05		
DBP	1.00	1.03	1.05		
LA reservoir strain	0.94	0.97	1.00		
CI, confidence interval; DBP, diastolic blood pressure; LA, left atrium; OR, odds					
ratio					

