

1 **Biomarker-based prediction of sinus rhythm in atrial**
2 **fibrillation patients: the EAST-AFNET 4 biomolecule study**

3 Larissa Fabritz^{1,2,3,4,5,6}, Christoph Al-Taie^{1,2,3,6}, Katrin Borof², Günter Breithardt^{4,7}, A
4 John Camm⁸, Harry JGM Crijns⁹, Victor Roth Cardoso^{5,10}, Winnie Chua^{5,6}, Silke van
5 Elferen^{1,2,11}, Lars Eckardt^{4,7}, George Gkoutos^{1,5,6,10}, Andreas Goette^{4,6,12,13}, Eduard
6 Guasch¹⁴, Stéphane Hatem^{6,15}, Andreas Metzner², Lluís Mont¹⁴, Vaishnavi Ameya
7 Murukutla^{1,3,6}, Julius Obergassel^{1,2,6}, Andreas Rillig^{2,3}, Moritz F Sinner¹⁶, Renate B
8 Schnabel^{2,3,4}, Ulrich Schotten^{4,6,17}, Laura C Sommerfeld^{1,3,5,6}, Ursula-Henrike
9 Wienhues-Thelen¹⁸, Antonia Zapf^{3,4,19}, Tanja Zeller^{1,2,3}, Paulus Kirchhof^{2,3,4,5,6}

10

11 ¹University Center of Cardiovascular Science, University Heart and Vascular Center
12 Hamburg, University Medical Center Hamburg Eppendorf, Hamburg, Germany

13 ²Department of Cardiology, University Heart and Vascular Center Hamburg, University
14 Medical Center Hamburg Eppendorf, Hamburg, Germany

15 ³German Center for Cardiovascular Research (DZHK), partner site Hamburg/Kiel/Lübeck,
16 Germany

17 ⁴AFNET, Münster, Germany

18 ⁵Institute of Cardiovascular Sciences, University of Birmingham, UK

19 ⁶MAESTRIA Consortium, European Union´s Horizon 2020 research and innovation
20 programme, agreement number 965386

21 ⁷Department of Cardiology II (Electrophysiology), University Hospital Münster, Germany

22 ⁸Clinical Sciences, St George´s University, London, UK

23 ⁹Department of Cardiology, University Hospital Maastricht, Maastricht, The Netherlands

1 ¹⁰MRC Health Data Research UK (HDR), Midlands Site, UK and Institute of Cancer and
2 Genomic Sciences, University of Birmingham, Birmingham, UK.

3 ¹¹Computational and Systems Biology at Hamburg University, Germany

4 ¹²Department of Cardiology and Intensive Care Medicine, St. Vincenz Hospital, Paderborn,
5 Germany

6 ¹³ Otto-von-Guericke University, Medical Faculty, Magdeburg, Germany

7 ¹⁴Hospital Clinic de Barcelona, Institute of Biomedical Research August Pi Sunyer
8 (IDIBAPS), Hospital Clinic, CIBERCV, University of Barcelona, Catalonia, Spain

9 ¹⁵Department of Cardiology, Sorbonne Universités, Faculté de médecine UPMC, Assistance
10 Publique-Hôpitaux de Paris, Pitié-Salpêtrière Hospital, Paris, France.

11 ¹⁶Department of Medicine I, University Hospital Munich, Ludwig Maximilian University of
12 Munich (LMU), Munich, Germany and German Centre for Cardiovascular Research (DZHK),
13 partner site Munich Heart Alliance, Munich, Germany.

14 ¹⁷Department of Physiology, Maastricht University, Maastricht, The Netherlands

15 ¹⁸Roche Diagnostics, Penzberg, Germany

16 ¹⁹Institute of Medical Biometry and Epidemiology, University Medical Center Hamburg
17 Eppendorf, Hamburg, Germany

18

19 Short title: Biomolecules predict sinus rhythm: EAST-AFNET 4

20

21 Corresponding author:

22 Prof. Dr. Paulus Kirchhof

23 Department of Cardiology

24 University Heart and Vascular Centre Hamburg Eppendorf

25 University Medical Center Hamburg Eppendorf

26 Martinist. 52

27 20246 Hamburg

28 Phone: +49 (0)40 7410-52438

29 Fax: +49 (0)40 7410-55862

30 Email: p.kirchhof@uke.de

31 www.uhz.de

1 **Abstract**

2 **Background and Aims.** In patients with atrial fibrillation (AF), recurrent AF and sinus
3 rhythm during follow-up are determined by interactions between cardiovascular disease
4 processes and rhythm-control therapy. Predictors of attaining sinus rhythm at follow-up are
5 not well known.

6 **Methods:** To quantify the interaction between cardiovascular disease processes and
7 rhythm outcomes, 14 biomarkers reflecting AF-related cardiovascular disease processes in
8 1586 patients in the EAST-AFNET 4 biomolecule study (71 years old, 46% women) were
9 quantified at baseline. Mixed logistic regression models including clinical features were
10 constructed for each biomarker. Biomarkers were interrogated for interaction with early
11 rhythm control. Outcome was sinus rhythm at 12 months. Results were validated at 24
12 months and in external datasets.

13 **Results:** Higher baseline concentrations of three biomarkers were independently associated
14 with a lower chance of sinus rhythm at 12 months: angiotensin 2 (ANGPT2) (odds ratio
15 [OR] 0.76 [95% confidence interval 0.65-0.89], $p=0.001$), bone morphogenetic protein 10
16 (BMP10) (OR 0.83 [0.71-0.97], $p=0.017$) and N-terminal pro-B-type natriuretic peptide (NT-
17 proBNP) (OR 0.73 [0.60-0.88], $p=0.001$). Analysis of rhythm at 24 months confirmed the
18 results. Early rhythm control interacted with the predictive potential of NT-proBNP
19 ($p_{\text{interaction}}=0.033$). The predictive effect of NT-proBNP was reduced in patients randomized
20 to early rhythm control (usual care: OR 0.64 [0.51-0.80], $p<0.001$; early rhythm control: OR
21 0.90 [0.69-1.18], $p=0.453$). External validation confirmed that low concentrations of
22 ANGPT2, BMP10 and NT-proBNP predict sinus rhythm during follow-up.

23 **Conclusions:** Low concentrations of ANGPT2, BMP10 and NT-proBNP identify patients
24 with AF who are likely to attain sinus rhythm during follow-up. The predictive ability of NT-
25 proBNP is attenuated in patients receiving rhythm control.

26 Key words: atrial fibrillation; blood biomarker; sinus rhythm; rhythm control; natriuretic
27 peptides; bone morphogenetic protein 10; angiotensin 2; risk prediction; risk score

1
2
3
4
5
6
7
8
9
10
11
12
13
14
15
16
17
18
19
20
21
22
23
24

Introduction

In addition to improving atrial fibrillation (AF)-related symptoms¹, rhythm control therapy² can prevent AF-related cardiovascular events such as stroke, heart failure hospitalizations, and cardiovascular death³. The cardiovascular complication-reducing effect of early rhythm control therapy shown in the EAST-AFNET 4 study is mainly mediated by attaining sinus rhythm at 12-month follow-up⁴. This potentially reflects a reduced AF burden⁵ and lack of progression to non-paroxysmal patterns of AF^{6,7}. Predicting sinus rhythm at 12 months could therefore help to identify patients requiring intensive rhythm control, e.g. with AF ablation^{3,8}. Knowledge of treatable processes contributing to AF at 12-month follow-up can help to develop adjunct therapies aimed at maintaining sinus rhythm and preventing AF progression⁶. Several chronic, interdependent disease processes^{9,10} contribute to AF. Such processes can be aggravated by presence of AF, attenuated by rhythm control, or exist independent of AF^{1,11}. Circulating biomarkers provide quantitative proxies for cardiomyocyte death or injury (troponin [TnT]); atrial metabolic dysfunction and stress (bone morphogenetic protein 10 [BMP10], fatty acid binding protein 3 [FABP3] and insulin-like growth factor binding protein 7 [IGFBP7])^{12,13}; thrombo-inflammation (D-dimer, C-reactive protein [CRP], interleukin-6 [IL-6])^{14,15}; vascular and endothelial dysfunction (angiopoietin 2 [ANGPT2], endothelial specific molecule 1 [ESM1])^{14,15}; frailty (growth differentiation factor 15 [GDF-15]); and cardiac load estimated (natriuretic peptides like N-terminal pro-B-type natriuretic peptide [NT-proBNP])¹⁶. Quantification of biomarkers selected to reflect these disease processes in a single blood draw identifies patient clusters with different risk of cardiovascular events¹⁷. Whether the disease processes reflected by the molecules modify future rhythm in patients with AF has not been investigated.

This analysis of the EAST-AFNET 4 biomolecule study embedded into the Early treatment of Atrial fibrillation for **ST**roke prevention (EAST-AFNET 4) trial² quantified 14 biomarkers reflecting different disease processes in AF that were defined *a priori*⁹. The

1 ability of each biomarker to predict sinus rhythm at 12-month follow-up in patients with and
2 without early rhythm control therapy was evaluated.

3 Validation was performed internally at 24 months, by comparing biomarker-based
4 clusters at baseline by association with sinus rhythm at 12-and 24-month follow-up and by
5 machine learning integrating biomarkers and clinical parameters. Clinical utility was
6 assessed by defining and testing threshold values and by comparison with a clinical score.
7 External validation was performed in two independent datasets of patients with AF.

8

9 **Methods**

10 Details of the prespecified analysis plan of the EAST-AFNET 4 biomolecule study can be
11 found in a separate supplementary material file (**Supplementary file Statistical**
12 **analysis plan SAP**). Post-hoc exploratory analyses were added to gain more insight into
13 the main findings.

14 Derivation dataset (EAST-AFNET 4). EAST-AFNET 4 randomized patients with recently
15 diagnosed AF and stroke risk factors to systematic early rhythm control or usual care
16 including symptom-based rhythm control². All patients were followed-up for a median of 5.1
17 years. The EAST-AFNET 4 biomolecule study collected a baseline blood sample in 1586
18 patients enrolled in the EAST-AFNET 4 trial^{17,18}. In brief, all consenting patients provided a
19 blood sample at baseline. Samples were shipped to the core biostorage facility at UKE
20 Hamburg, spun, shock-frozen and stored at -80°C. EAST-AFNET 4 and its biomolecule
21 study were approved at all participating study sites. Written informed consent was obtained
22 from all patients.

23 Validation datasets.

24 *AXAFA-AFNET 5*. The Anticoagulation using the direct factor Xa inhibitor apixaban during
25 Atrial Fibrillation catheter Ablation: Comparison to vitamin K antagonist therapy (AXAFA-
26 AFNET 5¹⁹) trial was a randomized, investigator-initiated trial comparing continuous

1 vitamin K antagonist therapy to apixaban in 633 patients undergoing a first AF ablation in
2 49 European and US American study sites. The same 14 biomarkers quantified in the
3 derivation dataset were quantified in the AXAFA-AFNET 5 blood samples using the same
4 assays.²⁰ The outcome of interest was rhythm at the final follow-up visit, 120 days after
5 enrolment.¹⁹

6 *BBC-AF atrial fibrillation snapshot.* Details of the BBC-AF cohort have been described
7 before ²¹. In brief, consecutive patients eligible for recruitment had ECG-diagnosed AF or
8 presented with at least two cardiovascular conditions (congestive heart failure, hypertension,
9 diabetes, prior stroke, or vascular disease) to a large teaching hospital (Sandwell and West
10 Birmingham NHS Trust). Patients who did not have a diagnosis of AF underwent 7-day
11 ambulatory ECG monitoring to rule out undiagnosed ECG-documented AF. For this analysis,
12 only patients with ECG-documented AF were included. Follow-up data were collected by
13 assessing local hospital records corroborated against Hospital Episode Statistics data,
14 general practitioner (GP) records, and mortality data from NHS Digital, up to 2.5 years after
15 the final patient was recruited²². This study complied with the Declaration of Helsinki, was
16 approved by the National Research Ethics Service Committee (IRAS ID 97753) and was
17 sponsored by the University of Birmingham. All patients provided written informed consent.

18 *TRUST snapshot.* A snapshot of patients enrolled in the Long-term Outcome and Predictors
19 for Recurrence after Medical and Interventional Treatment of Arrhythmias study (TRUST;
20 NCT05521451), with biomarker concentrations and 12-month rhythm status was created. All
21 patients provided written informed consent. A snapshot of all patients with biomarker
22 concentrations and ECG follow-up at 12-18 months was obtained in June 2024 for
23 validation.

24

25

1 Selection of biomarkers and their quantification. Circulating biomarkers were selected by
2 scientists from the EU-funded CATCH-ME consortium based on relevant disease processes
3 and available high precision high throughput assays⁹. Biomarkers were selected in four
4 steps: 1) Members of the consortium identified candidate biomarkers reflecting disease
5 processes known to contribute to AF and its complications, 2) Deep literature and patent
6 searches for candidate biomarkers and additional novel biomarkers were performed, 3)
7 Expert discussion and Delphi-like votes by the consortium defined most promising
8 candidates, and 4) Availability and feasibility checks to perform measurements of thousands
9 of samples with high precision.

10 Fourteen biomarkers were selected (**Table 1** following clinical characteristics); ANGPT2,
11 BMP10, cancer antigen 125 (CA125), CRP, D-dimer, ESM1, FABP3, fibroblast growth factor
12 23 (FGF23), GDF15, IGFBP7, IL-6, NT-proBNP, TnT and serum creatinine (sCr).

13 Blood samples were collected at all participating sites and shipped to the core lab at
14 University Heart and Vascular Center (UHZ) Hamburg by courier at ambient temperatures
15 (24-48 hours transport time). Upon arrival at UHZ, samples were spun, shock-frozen and
16 stored at -80C for analysis. Biomarkers were centrally quantified using pre-commercial and
17 commercial high-throughput, high-precision platforms (Roche, Penzberg, Germany) in
18 EDTA plasma. The biomarker quantification was provided as an in-kind contribution of
19 Roche to the CATCH ME consortium. Blood samples were shipped to, and quantifications
20 were conducted at the Roche biomarker research facility in Penzberg, Germany.

21 Statistical methods. As this is a secondary outcome analysis of the EAST-AFNET 4 trial, all
22 results are exploratory. Biomarker concentrations were one-percent winsorized²³ from above
23 and logarithmically-transformed (log base e) to normalize skewed concentration ranges for
24 all datasets. Concentrations below the detection limit for CA-125 and D-dimer were replaced
25 with the lowest available value. For the initial testing of prespecified hypotheses, all fourteen
26 biomarkers were used. Validations were done with predictive biomarkers. This analysis does
27 not take into account the probability of chance findings because of performance of multiple

1 comparisons with 14 biomarkers. As a consequence, results should be interpreted as
2 explorative/hypothesis generating and call for further validation.

3 Patients in AF at the time of blood sampling showed higher concentrations in most
4 biomarkers (**Supplementary Table S1**). Rhythm at time of blood sampling was included
5 as a confounder in all subsequent analyses in addition to the features predicting rhythm at 12
6 months in the main EAST-AFNET 4 data set.⁴

7 Mixed logistic regression models were used to assess the predictive value of the 14
8 biomarkers on rhythm at 12 months, with study center as a random intercept. The lme4 R
9 package²⁴ was used. Each biomarker was assessed in a separate model adjusted for sex, age,
10 body mass index, diastolic blood pressure, AF pattern (first-episode, paroxysmal, persistent),
11 left ventricular ejection fraction, rhythm at baseline, and randomized group (usual care or
12 early rhythm control). Those features are associated with rhythm at 12 months in the EAST-
13 AFNET 4 trial⁴. Nested models with additional interaction terms between treatment type and
14 the biomarker of interest were constructed. To obtain p-values for the interaction, each
15 nested model pair was compared by ANOVA for their goodness of fit. Odds ratios and p-
16 values for the biomarker effects under different treatment types were calculated by reference
17 cell coding²⁵. Missing values in heart rhythm and left ventricular ejection fraction were
18 imputed in a 60-times multiple imputed dataset as described earlier², following the
19 recommendations of White, Royston and Wood^{26,27}. A sensitivity analysis constructed
20 prediction models for recurrent AF at 12 and 24 month follow-up without imputation.
21 To further explore the effect of rhythm on the biomarkers, mixed regression models were
22 repeated in subgroups split by baseline rhythm (sinus rhythm or AF) and by rhythm control
23 therapy (early rhythm control or usual care) and odds ratios (OR) for the outcome sinus
24 rhythm at 12 months were calculated using the methods described above.

25 As internal validation, analysis was repeated for sinus rhythm at 24-month follow-up.
26 As sensitivity analysis, the analysis was repeated for recurrent AF up to 24 months.

1 As additional internal validation, patient clusters formed using all biomarker
2 concentrations agnostic to clinical features¹⁷ were tested for prediction of presence of sinus
3 rhythm at 12 and 24-month follow-up. The lowest-risk cluster was used as a reference.

4 As another means of internal validation, we applied a random forest machine
5 learning model (ML) and made use of a mixed effect random forest (MERF) wrapper to
6 account for the center as a random effect. The ML model was fitted with the features used for
7 confounding the generalized linear model as well as of all 14 biomarkers at once. To assess
8 the variable importance we used the models' inherent Gini-based feature importance as well
9 as the model agnostic SHapley Additive exPlanations (SHAP) values.

10 Clinical utility. Cut-off values for clinically useful probabilities of sinus rhythm at 12
11 months (80%) and for AF at 12 months (40%) were determined for all biomarkers that
12 predicted the main outcome. A clinical risk score was developed based on a recent meta-
13 analysis²⁸: Three accepted clinical features predicting recurrent AF, namely left atrial size,
14 AF pattern, and age, were dichotomized with a point scored for persistent AF yes, anterior-
15 posterior left atrial diameter > 50 mm, age >75 years (**Supplementary Table S2**). As
16 many patients with one of these three features attain sinus rhythm at 12 months, the score
17 was considered positively predictive of high risk of AF at 12 months if at least two of the
18 three factors were present. Each of the biomarkers that were independently associated with
19 sinus rhythm at 12 months were added to this clinical score separately, as well as in
20 combination. If at least one biomarker was above the cut-off value, the patient was regarded
21 as high risk of not attaining sinus rhythm. The confusion matrices for correctly and
22 incorrectly classified patients at high-risk-classified of not attaining sinus rhythm were
23 calculated for the reference clinical score alone and all additional, biomarker-enriched
24 scores.

25 Biomarkers' predictive values were tested in the validation datasets using univariate and
26 multivariate models restricted to the features that predicted sinus rhythm at 12 months in
27 the derivation dataset. Python version 3.8.13 was employed for data preprocessing and

1 visualization, R version 4.2.2 for statistical computations²⁹. Relevant code will be made
2 publicly available (<https://github.com/UCCSHH>).

3

4 **Results**

5 Derivation analysis dataset. The 1586 patients with a recent history of AF and stroke risk
6 factors (age 71 years, 45% women) with clinical features, biomarker concentrations and
7 cardiovascular outcomes were equally assigned to both randomized treatment groups.
8 **(Table 1, Supplementary Figure S1).**

9 Association of biomarker concentrations with attaining sinus rhythm at 12 months. Three
10 biomarkers (ANGPT2, BMP10 and NT-proBNP) showed lower concentrations at baseline in
11 patients who were in sinus rhythm at the 12-month follow-up (**Figure 1A**). These three
12 biomarkers were independently associated with sinus rhythm at the 12-month follow-up
13 after multiple corrections for clinical features, early rhythm control, and baseline rhythm
14 (**Figure 1A**). NT-proBNP interacted with early rhythm control therapy at 12-month follow-
15 up ($p=0.033$) and low NT-proBNP concentrations only predicted sinus rhythm at 12 months
16 in patients randomized to usual care (**Figure 1B**). Early rhythm control impacted on the
17 rhythm-predicting effect of NT-proBNP and dampened its predictive value in this group.
18 There was no significant interaction detected between early rhythm control and any of the
19 other 13 biomarkers in this dataset (**Figure 1B**).

20 Biomarker concentrations distributions depicted in violin plots after log transformation
21 (**Figure 2**) show lower concentrations in sinus rhythm versus AF at 12 months. Numbers of
22 mean biomarker concentrations by rhythm at 12-month follow-up and by randomized
23 treatment group are given (**Table 2**).

24 Baseline biomarker concentrations depending on baseline rhythm in the derivation dataset
25 and clinical features are shown in **Table 3**, extended information shown in
26 **Supplementary Table S1**. Post-hoc subgroup analyses by rhythm at the time of baseline

1 assessment (sinus rhythm or AF) and by randomized group (early rhythm control or usual
2 care) find NT-proBNP mainly associated with sinus rhythm at 12 months in patients under
3 usual care. BMP10 and ANGPT2 retained their predictive ability shown in the joint group of
4 all patients also if only the subgroup patients in AF at the time of blood sampling were
5 analysed (**Figure 3**).

6 **Internal validations.** As a first internal validation, the same analysis was performed for
7 the 24-month follow-up. The same biomarkers, ANGPT2, BMP10, and NT-proBNP, were
8 consistently associated with sinus rhythm at 24-month follow-up (**Figure 4**).

9 Repeating the analysis for recurrent AF up to 24 months showed similar results
10 (**Supplementary Table S3**). As further internal validation analysis, unsupervised
11 biomarker-based clustering of EAST patients previously performed was applied to sinus
12 rhythm at 12-month follow-up. Clusters separated by risk of cardiovascular complications,
13 with patients assigned to the high-risk cardiovascular outcome cluster showing a lower
14 likelihood of sinus rhythm at 12 months, patients in the two intermediate cardiovascular risk
15 biomarker clusters showing an intermediate likelihood of sinus rhythm, all tested against the
16 low cardiovascular risk cluster, with the low-risk outcome patient cluster showing the
17 highest likelihood of sinus rhythm at 12 months (**Figure 5A**). These findings were
18 consistent for the high-risk biomarker-based clusters at 24-month follow-up (**Figure 5B**).

19 As further internal validation, a random forest classifier was trained on the EAST-AFNET 4
20 dataset. Its feature performance evaluation confirmed the importance of the three
21 biomarkers alongside AF pattern, rhythm at baseline, and early rhythm control for the
22 outcome of sinus rhythm (**Figure 6**).

23 **Clinical utility.** Thresholds to predict a high probability of attaining sinus rhythm (>80%,
24 low risk of AF) or a high probability of recurrent AF at follow-up (>40%, high risk of AF)
25 were determined for each biomarker (**Table 4, Supplementary Figures S2, S3, S4**). To
26 compare them to clinical features predicting sinus rhythm, a score combining clinical
27 features predicting recurrent AF was created²⁸ (**Supplementary Table S2**). Adding

1 biomarkers using these thresholds improved identification of patients at risk of not attaining
2 sinus rhythm at 12-month follow-up (**Table 5, Supplementary Table S4**).

3 **External validation.** Several separate validation datasets (AXAFA-AFNET 5 trial, BBC-AF
4 and TRUST cohort snapshot **Supplementary Tables S5, S6, S7**) were used. The
5 biomarkers NT-proBNP, BMP10, and ANGPT2 were confirmed as predictive of sinus rhythm
6 in the final follow-up in AXAFA-AFNET 5 (**Figure 7**). The clinical utility of adding the
7 biomolecules to clinical predictors was validated in both cohorts using the thresholds derived
8 in EAST-AFNET 4 (**Supplementary Table S8 and S9**).

9

10

11 **Discussion**

12 Main findings. Three out of fourteen candidate biomarkers, BMP10, ANGPT2 and NT-
13 proBNP, are associated with sinus rhythm at 12-month and 24-month follow-up after
14 correcting for clinical features. Low NT-proBNP, low ANGPT2 and low BMP10
15 concentrations independently predict sinus rhythm in patients at follow-up. NT-proBNP is
16 less predictive of rhythm in patients receiving rhythm control therapy. Adding these
17 biomarkers to a clinical score identifying patients with a low probability of sinus rhythm at
18 12 months (positive with two out of three features: left atrial size >50 mm, persistent AF, or
19 age >75 years) refined risk prediction (Structured Graphical Abstract).

20 Relevance for clinical care and research. In view of the growing choice of medical^{2,30},
21 interventional^{2,31}, and surgical³² treatment options for patients with AF, selecting the best
22 strategy and the patients most benefitting from rhythm control therapy gains importance.
23 Biomarker-based risk estimators have so far mainly been developed to refine anticoagulation
24 decisions in patients with AF³³⁻³⁵. Actionable biomarkers to guide rhythm control therapy are
25 lacking. Similar to stroke prevention estimators, rhythm estimators face the challenge of
26 random factors determining a binary outcome (AF or sinus rhythm). The present results
27 suggest that NT-proBNP, BMP10, and ANGPT2 can stratify patients at high and low risk of

1 attaining sinus rhythm alone and in combination. These biomarkers reflect and identify
2 diseases processes that promote future AF, pointing to potential therapeutic targets for
3 adjunct therapy supporting rhythm control. While a simple clinical score combining
4 enlarged left atrial size, persistent AF, and older age predicted future sinus rhythm
5 reasonably well, adding biomarkers reclassifies a clinically relevant number of patients at
6 high risk of not attaining sinus rhythm at the price of also classifying more patients in sinus
7 rhythm as high-risk.

8 Effect of baseline rhythm on biomarker concentrations. This study shows that
9 ANGPT2 and BMP10 provide additional information on future sinus rhythm when combined
10 with NT-proBNP, especially in patients who are in AF at the time of blood sampling. Most
11 biomarkers studied were elevated when the blood sample was taken in AF. Furthermore, NT-
12 proBNP lost its ability to predict sinus rhythm in patients on rhythm control therapy, and the
13 predictive ability of BMP10 decreased in the subgroup of patients who were in sinus rhythm,
14 but not in the subgroup of patients who were in AF, similar to rhythm-dependent decrease of
15 predictive ability of NT-proBNP^{20,36}. The effects of baseline rhythm on the concentrations
16 and predictive ability of biomarkers should be further investigated in patients with AF
17 undergoing rhythm control therapy.

18 Interpretation of NT-proBNP. NT-proBNP is released by atrial cardiomyocytes in response
19 to stretch and strain, thereby acutely regulating fluid balance in the body, resulting in high
20 concentrations during AF³⁷. In heart failure, NT-proBNP is also released by ventricular
21 cardiomyocytes, further enhancing its concentrations. Atrial stretch has proarrhythmic
22 effects including shortening of the atrial effective refractory period³⁸ and conduction
23 slowing³⁹, partially explaining its prediction of sinus rhythm in this study. NT-proBNP
24 reflects short- and mid-term processes in patients with AF, probably explaining its
25 interaction with rhythm. The possibility that elevated NT-proBNP concentrations predict
26 rhythm during follow-up have been reported before.⁴⁰⁻⁴⁷ NT-proBNP is also associated with
27 incident AF⁴⁸⁻⁵¹ and with cardiovascular events in patients with and without AF and heart

1 failure.²² This analysis demonstrates that the rhythm-predicting ability of NT-proBNP is
2 reduced in patients treated with rhythm control therapy.

3 The NT-proBNP thresholds associated with a high risk of AF at 12 months in this
4 study (>1500 pg/ml) are comparable to the thresholds associated with cardiovascular events,
5 but higher than currently used thresholds e.g. for AF screening⁵² or for diagnosing heart
6 failure with AF and heart failure with preserved ejection fraction⁵³. Based on the present
7 analysis, higher thresholds may have better clinical utility. This warrants further analysis.

8 Interpretation of BMP10 and ANGPT2. BMP10 and ANGPT2 are tightly regulated
9 circulating biomarkers, illustrating their signaling roles in regulating disease processes
10 contributing to AF³. Mechanistic studies of their role in AF are needed to define more precise
11 clinical use cases for these biomarkers in patients with AF.

12 ANGPT2 is a vascular growth factor required for angiogenic remodeling⁵⁴.
13 Overexpression of ANGPT2 in murine models promotes perivascular cardiac inflammation
14 and fibrosis⁵⁵. Pro-inflammatory molecules such as thrombin increase ANGPT2 expression
15 *in vitro*⁵⁶ and inhibition of thrombin in animals with persistent AF improves atrial
16 cardiomyopathy¹⁵. Thus, ANGPT2 mediates the inflammatory communication between
17 endothelial cells and myocardium in AF. Low ANGPT2 might reflect preserved vascular
18 integrity, reducing the inflammatory burden in atrial vascular beds and thereby slowing AF
19 progression.

20 ANGPT2 is associated with recurrent AF in patients after AF ablation²⁰ and with
21 prevalent AF in unselected hospitalized patients⁵⁷. ANGPT2 is elevated in patients with
22 kidney disease⁵⁸, acute lung injury⁵⁹ and sepsis⁶⁰, conditions associated with AF. ANGPT2
23 can also predict heart failure hospitalization in patients with AF⁶¹, similar to NT-proBNP.²²
24 This study is the first to suggest that ANGPT2 can predict sinus rhythm in patients with AF
25 with and without rhythm control therapy. Further research into treatable atrial disease
26 processes regulated by ANGPT2 is warranted.

1 BMP10 is selectively expressed in and released by atrial cardiomyocytes^{16, 62}. BMP10
2 is part of the TGF β growth factor family and regulates vascular smooth muscle cell tone⁶³. Its
3 function in the atria is not well known. BMP10 concentrations are reduced in hereditary
4 forms of pulmonary arterial hypertension⁶⁴, possibly reflecting reduced left atrial
5 metabolism. Its inverse correlation and possible repression by PITX2 in atrial
6 cardiomyocytes^{16, 65} may suggest that elevated BMP10 concentrations could identify a
7 reversible atrial metabolic defect^{13, 17} that may be aggravated by the genomic basis of AF on
8 chromosome 4q25¹³.

9 High concentrations of BMP10 are associated with recurrent AF^{57, 66}, and with
10 cardiovascular events^{17, 67} and stroke in patients with AF. BMP10 may also be associated with
11 atrial fibrosis⁶⁸. Lower BMP10 concentrations in patients in sinus rhythm²⁰, combined with
12 its prediction of future sinus rhythm (**Figure 1**) suggest that a possible BMP10-mediated
13 metabolic defect could partially be secondary to the metabolic demands of AF. Taken
14 together, these results suggest that BMP10 is a potentially actionable biomarker indicative of
15 atrial myopathy and atrial metabolic dysfunction. Further research into the atrial effects of
16 BMP10 and its relation to AF burden⁵ are warranted.

17 Biomolecule-based clustering of patients agnostic to clinical features previously
18 identified four subgroups of patients with AF with a gradual increase in cardiovascular
19 events.¹⁷ The three biomarkers associated with sinus rhythm at 12 months in this study are
20 among the six dominant biomarkers previously defining these patient clusters.¹⁷ The
21 biomarker-based clusters show a certain risk gradient for sinus rhythm at 12 months
22 (**Figure 5**). At difference to the prior study that defined patient clusters based on 14
23 biomarker concentrations agnostic to clinical information, this analysis shows that the three
24 biomarkers NT-proBNP, ANGPT2 and BMP10 predict sinus rhythm in context with clinical
25 parameters. Of note, a simple clinical score was already quite useful in identifying patients
26 who will attain sinus rhythm. This information can help clinicians to select different
27 intensities of rhythm control therapy depending on the likelihood of attaining sinus rhythm.

1 NT-proBNP, ANGPT2 and BMP10 can refine that selection. The present result and the
2 biomarker-clustering also identify potentially treatable drivers of recurrent AF and or
3 cardiovascular events in patients with AF. Based on the known atrial effects of BMP10 and
4 ANGPT2, antihypertensive therapy and metabolic interventions such as SGLT2 inhibitor
5 therapy⁶⁹ could have beneficial effects in patients with elevated BMP10 and ANGPT2
6 concentrations.^{16, 67, 70} The underlying disease processes suggest that the same biomarkers
7 could also be useful to identify patients at risk of AF. The present analysis identifies
8 potentially actionable biomarkers suitable to select the intensity of rhythm control therapy.
9 Further research into the mechanistic links between these biomarkers with baseline and
10 future rhythm, and further evaluations of their clinical utility in different scenarios are
11 warranted.

12 Strengths and limitations. Central quantification of the biomarkers using high-precision
13 assays combined with the rigorous, near-complete follow-up at 12 and 24 months in a
14 controlled clinical trial is a strength of this analysis. The consistent findings at both time
15 points may suggest that the effects can be extrapolated to even longer follow-up, but this
16 would require validation. Another strength of the analysis is the collection of samples in a
17 broad range of care settings in adequately treated patients with AF, and external validation
18 both in a controlled clinical trial and in cohorts of patients with AF enrolled in routine care
19 settings. Validation of the findings using the same assays in different clinical datasets is a
20 strength, but also limits the findings to the assays provided for this study.

21 The study has important limitations. Although the statistical analysis plan was prespecified
22 and validation was possible in different datasets, all results are explorative. This study is
23 limited to 14 preselected biomarkers. Selected biomarkers intentionally reflect overlapping
24 disease processes, creating redundancy that enables robust definition of disease pathways.
25 Collinearity of biomarkers was more deeply investigated in a previous study defining
26 biomarker-based patient clusters agnostic to clinical features¹⁷.

1 Additional biomarkers in AF may emerge from hypothesis-free quantification of many
2 molecules at once e.g. by RNA-sequencing of cardiac tissue⁷¹, quantification of circulating
3 RNAs, and by proteomics^{72,73}. Repeat blood samples were not obtained and no information
4 on changes over time is available. Some data on the changes of BMP10 and NT-proBNP over
5 time have been published^{20,36}.

6 While NT-proBNP can be measured in clinical routine as *in-vitro* diagnostic devices with
7 regulatory approval, the assays for ANGPT2 and BMP10 are not approved for clinical use,
8 restricting them to research settings. Only the consenting portion of the total EAST-AFNET
9 study participants was included in the biomarker study (two thirds), hence there could be a
10 considerable selection bias. Due to time required to setup the biobank, the first 400 patients
11 were not invited to participate in the biomarker study.

12 The present study used serum creatinine rather than estimated glomerular filtration rate in
13 the analyses as the formulas used to estimate kidney function rely on clinical parameters that
14 are used in the regression model, including age, sex, and body mass index. Serum creatinine
15 was not a major predictor of sinus rhythm. Whether estimated kidney function is a better
16 predictor of sinus rhythm was not studied.

17 Validation datasets were smaller than the derivation dataset and therefore did not allow for
18 multiple confounding. Post-hoc subgroup analysis by baseline rhythm in EAST-AFNET 4
19 may have underestimated effects due to smaller group sizes. Almost all patients received
20 guideline-recommended anticoagulation, rate and rhythm control, and often effective
21 treatment of concomitant conditions. 24-hour blood pressure may provide more granular
22 prognostic information than office-based blood pressure, but 24-hour blood pressure
23 readings were not available for this analysis.

24 Left atrial size was used in the clinical score rather than left atrial volume. Indexed left atrial
25 volume can provide more detailed information on left atrial size compared to left atrial
26 diameter, but the predictive value of left atrial volume for recurrent AF is less well
27 established than left atrial size²⁸. Indexed left atrial volume was not available in sufficient

1 patients to be assessed in this study. The predictive ability of the different biomarker-based
2 models is only valid for the specific AF prevalences in the cohorts studied. Further research
3 into the clinical utility of the biomarkers identified here is warranted.

4 The blood samples studied here stem from patients with predominantly Caucasian ethnicity,
5 which may limit the generalizability of the findings to other ethnic groups. Validation in
6 other ethnicities is therefore needed.

7 Testing the relationship between specific blood biomarker levels and a remote outcome
8 observed 12 months later is challenging. In order to limit acute effects of the specific
9 biomarker levels at baseline, we corrected for the acute rhythm at baseline, among other
10 clinical parameters. Prediction of future rhythm by biomarkers depends on several factors,
11 including the underlying biology of each biomarker, spontaneous variations in
12 concentrations, and assay quality. Lack of predictive ability in this study does not rule out
13 relevant biological function of a given molecule. The proposed interventions countering the
14 disease processes associated with biomarkers require further testing.

15 **Conclusion**

16 In conclusion, these findings suggest that NT-proBNP, ANGPT2 and BMP10 can be
17 combined to identify patients with AF at high risk of not attaining sinus rhythm. The disease
18 processes related to ANGPT2 and BMP10 emerge as likely contributors to future rhythm in
19 patients with and without rhythm control therapy. NT-proBNP elevations interact with early
20 rhythm control, potentially suggesting repeat assessment of NT-proBNP to monitor the
21 effectiveness of rhythm control.

22
23 **Acknowledgement:** We thank AFNET staff.

24

25

References

1. Hindricks G, Potpara T, Dagres N, Arbelo E, Bax JJ, Blomstrom-Lundqvist C, *et al.* 2020 ESC Guidelines for the diagnosis and management of atrial fibrillation developed in collaboration with the European Association for Cardio-Thoracic Surgery (EACTS): The Task Force for the diagnosis and management of atrial fibrillation of the European Society of Cardiology (ESC) Developed with the special contribution of the European Heart Rhythm Association (EHRA) of the ESC. *Eur Heart J* 2021;**42**(5):373-498.
2. Kirchhof P, Camm AJ, Goette A, Brandes A, Eckardt L, Elvan A, *et al.* Early Rhythm-Control Therapy in Patients with Atrial Fibrillation. *N Engl J Med* 2020;**383**(14):1305-1316.
3. Linz D, Andrade JG, Arbelo E, Boriani G, Breithardt G, Camm AJ, *et al.* Longer and better lives for patients with atrial fibrillation: the 9th AFNET/EHRA consensus conference. *Europace* 2024;**26**(4).
4. Eckardt L, Sehner S, Suling A, Borof K, Breithardt G, Crijns H, *et al.* Attaining sinus rhythm mediates improved outcome with early rhythm control therapy of atrial fibrillation: the EAST-AFNET 4 trial. *Eur Heart J* 2022;**43**(40):4127-4144.
5. Becher N, Metzner A, Toennis T, Kirchhof P, Schnabel RB. Atrial fibrillation burden: a new outcome predictor and therapeutic target. *Eur Heart J* 2024.
6. Andrade JG, Deyell MW, Khairy P, Champagne J, Leong-Sit P, Novak P, *et al.* Atrial fibrillation progression after cryoablation versus radiofrequency ablation: the CIRCA-DOSE trial. *Eur Heart J* 2023.
7. Kuck KH, Lebedev DS, Mikhaylov EN, Romanov A, Geller L, Kalejs O, *et al.* Catheter ablation or medical therapy to delay progression of atrial fibrillation: the randomized controlled atrial fibrillation progression trial (ATTEST). *Europace* 2021;**23**(3):362-369.
8. Andrade JG, Deyell MW, Macle L, Wells GA, Bennett M, Essebag V, *et al.* Progression of Atrial Fibrillation after Cryoablation or Drug Therapy. *N Engl J Med* 2022.
9. Fabritz L, Guasch E, Antoniades C, Bardinnet I, Benninger G, Betts TR, *et al.* Expert consensus document: Defining the major health modifiers causing atrial fibrillation: a roadmap to underpin personalized prevention and treatment. *Nat Rev Cardiol* 2016;**13**(4):230-7.
10. Nielsen JB, Thorolfsdottir RB, Fritsche LG, Zhou W, Skov MW, Graham SE, *et al.* Biobank-driven genomic discovery yields new insight into atrial fibrillation biology. *Nat Genet* 2018;**50**(9):1234-1239.
11. Schnabel RB, Yin X, Gona P, Larson MG, Beiser AS, McManus DD, *et al.* 50 year trends in atrial fibrillation prevalence, incidence, risk factors, and mortality in the Framingham Heart Study: a cohort study. *Lancet* 2015;**386**(9989):154-62.
12. Billing AM, Kim YC, Gullaksen S, Schrage B, Raabe J, Hutzfeldt A, *et al.* Metabolic Communication by SGLT2 Inhibition. *Circulation* 2023.
13. Reyat J, Sommerfeld L, O'Reilly M, Roth Cardoso V, Thiemann E, Khan A, *et al.* PITX2-deficiency leads to atrial mitochondrial dysfunction. *Cardiovasc Res* 2024;**in press, doi 10.1093/cvr/cvae169**.
14. Hu YF, Chen YJ, Lin YJ, Chen SA. Inflammation and the pathogenesis of atrial fibrillation. *Nat Rev Cardiol* 2015;**12**(4):230-43.
15. Spronk HM, De Jong AM, Verheule S, De Boer HC, Maass AH, Lau DH, *et al.* Hypercoagulability causes atrial fibrosis and promotes atrial fibrillation. *Eur Heart J* 2017;**38**(1):38-50.
16. Reyat JS, Chua W, Cardoso VR, Witten A, Kastner PM, Kabir SN, *et al.* Reduced left atrial cardiomyocyte PITX2 and elevated circulating BMP10 predict atrial fibrillation after ablation. *JCI Insight* 2020;**5**(16).
17. Fabritz L, Chua W, Cardoso VR, Al-Taie C, Borof K, Suling A, *et al.* Blood-based cardiometabolic phenotypes in atrial fibrillation and their associated risk: EAST-AFNET 4 biomolecule study. *Cardiovasc Res* 2024.

- 1 18. Kany S, Al-Taie C, Roselli C, Pirruccello JP, Borof K, Reinbold C, *et al.* Association of genetic
2 risk and outcomes in patients with atrial fibrillation: interactions with early rhythm control in the
3 EAST-AFNET4 trial. *Cardiovasc Res* 2023;**119**(9):1799-1810.
- 4 19. Kirchhof P, Haeusler KG, Blank B, De Bono J, Callans D, Elvan A, *et al.* Apixaban in patients at
5 risk of stroke undergoing atrial fibrillation ablation. *Eur Heart J* 2018;**39**(32):2942-2955.
- 6 20. Chua W, Khashaba A, Canagarajah H, Nielsen JC, di Biase L, Haeusler KG, *et al.* Disturbed
7 atrial metabolism, shear stress, and cardiac load contribute to atrial fibrillation after ablation: AXAFA
8 biomolecule study. *Europace* 2024;**26**(2).
- 9 21. Chua W, Purmah Y, Cardoso VR, Gkoutos GV, Tull SP, Neculau G, *et al.* Data-driven discovery
10 and validation of circulating blood-based biomarkers associated with prevalent atrial fibrillation. *Eur*
11 *Heart J* 2019;**40**(16):1268-1276.
- 12 22. Brady PF, Chua W, Nehaj F, Connolly DL, Khashaba A, Purmah YJV, *et al.* Interactions
13 Between Atrial Fibrillation and Natriuretic Peptide in Predicting Heart Failure Hospitalization or
14 Cardiovascular Death. *J Am Heart Assoc* 2022;**11**(4):e022833.
- 15 23. Dixon WJ, Yuen KK. Trimming and winsorization: A review. *Statistische Hefte* 1974;**15**(2):157-
16 170.
- 17 24. Bates D MM, Bolker B, Walker S. Fitting Linear Mixed-Effects Models Using lme4. *Journal of*
18 *Statistical Software* 2015;**67**(1):1–48.
- 19 25. Figueiras A, Domenech-Massons JM, Cadarso C. Regression models: calculating the
20 confidence interval of effects in the presence of interactions. *Stat Med* 1998;**17**(18):2099-105.
- 21 26. White IR, Royston P, Wood AM. Multiple imputation using chained equations: Issues and
22 guidance for practice. *Stat Med* 2011;**30**(4):377-99.
- 23 27. van Buuren S G-OK. mice: Multivariate Imputation by Chained Equations in R. *Journal of*
24 *Statistical Software* 2011;**45**(3):1-67.
- 25 28. Dretzke J, Chuchu N, Agarwal R, Herd C, Chua W, Fabritz L, *et al.* Predicting recurrent atrial
26 fibrillation after catheter ablation: a systematic review of prognostic models. *Europace*
27 2020;**22**(5):748-760.
- 28 29. Team RC. R: A Language and Environment for Statistical Computin. In: R Foundation for
29 Statistical Computing; 2022.
- 30 30. Ruff CT, Giugliano RP, Braunwald E, Hoffman EB, Deenadayalu N, Ezekowitz MD, *et al.*
31 Comparison of the efficacy and safety of new oral anticoagulants with warfarin in patients with atrial
32 fibrillation: a meta-analysis of randomised trials. *Lancet* 2014;**383**(9921):955-62.
- 33 31. Sohns C, Fox H, Marrouche NF, Crijns H, Costard-Jaeckle A, Bergau L, *et al.* Catheter Ablation
34 in End-Stage Heart Failure with Atrial Fibrillation. *N Engl J Med* 2023;**389**(15):1380-1389.
- 35 32. Whitlock RP, Belley-Cote EP, Paparella D, Healey JS, Brady K, Sharma M, *et al.* Left Atrial
36 Appendage Occlusion during Cardiac Surgery to Prevent Stroke. *N Engl J Med* 2021;**384**(22):2081-
37 2091.
- 38 33. Hijazi Z, Lindback J, Alexander JH, Hanna M, Held C, Hylek EM, *et al.* The ABC (age,
39 biomarkers, clinical history) stroke risk score: a biomarker-based risk score for predicting stroke in
40 atrial fibrillation. *Eur Heart J* 2016;**37**(20):1582-90.
- 41 34. Hijazi Z, Oldgren J, Lindback J, Alexander JH, Connolly SJ, Eikelboom JW, *et al.* The novel
42 biomarker-based ABC (age, biomarkers, clinical history)-bleeding risk score for patients with atrial
43 fibrillation: a derivation and validation study. *Lancet* 2016;**387**(10035):2302-11.
- 44 35. Pol T, Hijazi Z, Lindback J, Oldgren J, Alexander JH, Connolly SJ, *et al.* Using multimarker
45 screening to identify biomarkers associated with cardiovascular death in patients with atrial
46 fibrillation. *Cardiovasc Res* 2022;**118**(9):2112-2123.
- 47 36. Gkarmiris KI, Lindback J, Alexander JH, Granger CB, Kastner P, Lopes RD, *et al.* Repeated
48 Measurement of the Novel Atrial Biomarker BMP10 (Bone Morphogenetic Protein 10) Refines Risk
49 Stratification in Anticoagulated Patients With Atrial Fibrillation: Insights From the ARISTOTLE Trial. *J*
50 *Am Heart Assoc* 2024;**13**(7):e033720.

- 1 37. Yamaji T, Ishibashi M, Nakaoka H, Imataka K, Amano M, Fujii J. Possible role for atrial
2 natriuretic peptide in polyuria associated with paroxysmal atrial arrhythmias. *Lancet*
3 1985;**1**(8439):1211.
- 4 38. Ravelli F, Alessie M. Effects of atrial dilatation on refractory period and vulnerability to atrial
5 fibrillation in the isolated Langendorff-perfused rabbit heart. *Circulation* 1997;**96**(5):1686-95.
- 6 39. Walters TE, Lee G, Spence S, Larobina M, Atkinson V, Antippa P, *et al.* Acute atrial stretch
7 results in conduction slowing and complex signals at the pulmonary vein to left atrial junction:
8 insights into the mechanism of pulmonary vein arrhythmogenesis. *Circ Arrhythm Electrophysiol*
9 2014;**7**(6):1189-97.
- 10 40. Wachter R, Lahno R, Haase B, Weber-Kruger M, Seegers J, Edelmann F, *et al.* Natriuretic
11 peptides for the detection of paroxysmal atrial fibrillation in patients with cerebral ischemia--the
12 Find-AF study. *PLoS One* 2012;**7**(4):e34351.
- 13 41. Zografos T, Maniotis C, Katsivas A, Katritsis D. Relationship between brain natriuretic
14 peptides and recurrence of atrial fibrillation after successful direct current cardioversion: a meta-
15 analysis. *Pacing Clin Electrophysiol* 2014;**37**(11):1530-7.
- 16 42. Asselbergs FW, van den Berg MP, Bakker SJ, Signorovitch JE, Hillege HL, van Gilst WH, van
17 Veldhuisen DJ. N-terminal pro B-type natriuretic peptide levels predict newly detected atrial
18 fibrillation in a population-based cohort. *Neth Heart J* 2008;**16**(3):73-8.
- 19 43. Hwang HJ, Son JW, Nam BH, Joung B, Lee B, Kim JB, *et al.* Incremental predictive value of
20 pre-procedural N-terminal pro-B-type natriuretic peptide for short-term recurrence in atrial
21 fibrillation ablation. *Clin Res Cardiol* 2009.
- 22 44. Freestone B, Gustafsson F, Chong AY, Corell P, Kistorp C, Hildebrandt P, Lip GY. Influence of
23 atrial fibrillation on plasma von willebrand factor, soluble E-selectin, and N-terminal pro B-type
24 natriuretic peptide levels in systolic heart failure. *Chest* 2008;**133**(5):1203-8.
- 25 45. Xu X, Tang Y. Relationship between Brain Natriuretic Peptide and Recurrence of Atrial
26 Fibrillation after Successful Electrical Cardioversion: an Updated Meta-Analysis. *Braz J Cardiovasc*
27 *Surg* 2017;**32**(6):530-535.
- 28 46. Darkner S, Goetze JP, Chen X, Henningsen K, Pehrson S, Svendsen JH. Natriuretic Propeptides
29 as Markers of Atrial Fibrillation Burden and Recurrence (from the AMIO-CAT Trial). *Am J Cardiol*
30 2017;**120**(8):1309-1315.
- 31 47. den Uijl DW, Delgado V, Tops LF, Ng AC, Boersma E, Trines SA, *et al.* Natriuretic peptide
32 levels predict recurrence of atrial fibrillation after radiofrequency catheter ablation. *Am Heart J*
33 2011;**161**(1):197-203.
- 34 48. Cushman M, Judd SE, Howard VJ, Kissela B, Gutierrez OM, Jenny NS, *et al.* N-terminal pro-B-
35 type natriuretic peptide and stroke risk: the reasons for geographic and racial differences in stroke
36 cohort. *Stroke* 2014;**45**(6):1646-50.
- 37 49. Schrage B, Geelhoed B, Niiranen TJ, Gianfagna F, Vishram-Nielsen JKK, Costanzo S, *et al.*
38 Comparison of Cardiovascular Risk Factors in European Population Cohorts for Predicting Atrial
39 Fibrillation and Heart Failure, Their Subsequent Onset, and Death. *J Am Heart Assoc*
40 2020;**9**(9):e015218.
- 41 50. Svennberg E, Henriksson P, Engdahl J, Hijazi Z, Al-Khalili F, Friberg L, Frykman V. N-terminal
42 pro B-type natriuretic peptide in systematic screening for atrial fibrillation. *Heart* 2017;**103**(16):1271-
43 1277.
- 44 51. Patton KK, Ellinor PT, Heckbert SR, Christenson RH, DeFilippi C, Gottdiener JS, Kronmal RA.
45 N-terminal pro-B-type natriuretic peptide is a major predictor of the development of atrial
46 fibrillation: the Cardiovascular Health Study. *Circulation* 2009;**120**(18):1768-74.
- 47 52. Engdahl J, Svennberg E, Friberg L, Al-Khalili F, Frykman V, Kemp Gudmundsdottir K, *et al.*
48 Stepwise mass screening for atrial fibrillation using N-terminal pro b-type natriuretic peptide: the
49 STROKESTOP II study design. *Europace* 2017;**19**(2):297-302.

- 1 53. Parwani AS, Kaab S, Friede T, Tilz RR, Bauersachs J, Frey N, *et al.* Catheter-based ablation to
2 improve outcomes in patients with atrial fibrillation and heart failure with preserved ejection
3 fraction: Rationale and design of the CABA-HFPEF-DZHK27 trial. *Eur J Heart Fail* 2024.
- 4 54. Gale NW, Thurston G, Hackett SF, Renard R, Wang Q, McClain J, *et al.* Angiopoietin-2 is
5 required for postnatal angiogenesis and lymphatic patterning, and only the latter role is rescued by
6 Angiopoietin-1. *Dev Cell* 2002;**3**(3):411-23.
- 7 55. Chen JX, Zeng H, Reese J, Aschner JL, Meyrick B. Overexpression of angiopoietin-2 impairs
8 myocardial angiogenesis and exacerbates cardiac fibrosis in the diabetic db/db mouse model. *Am J*
9 *Physiol Heart Circ Physiol* 2012;**302**(4):H1003-12.
- 10 56. Huang YQ, Li JJ, Hu L, Lee M, Karparkin S. Thrombin induces increased expression and
11 secretion of angiopoietin-2 from human umbilical vein endothelial cells. *Blood* 2002;**99**(5):1646-50.
- 12 57. Chua W, Cardoso VR, Guasch E, Sinner MF, Al-Taie C, Brady P, *et al.* An angiopoietin 2,
13 FGF23, and BMP10 biomarker signature differentiates atrial fibrillation from other concomitant
14 cardiovascular conditions. *Sci Rep* 2023;**13**(1):16743.
- 15 58. Bontekoe J, Lee J, Bansal V, Syed M, Hoppensteadt D, Maia P, *et al.* Biomarker Profiling in
16 Stage 5 Chronic Kidney Disease Identifies the Relationship between Angiopoietin-2 and Atrial
17 Fibrillation. *Clin Appl Thromb Hemost* 2018;**24**(9_suppl):269S-276S.
- 18 59. Bhandari V, Choo-Wing R, Lee CG, Zhu Z, Nedrelo JH, Chupp GL, *et al.* Hyperoxia causes
19 angiopoietin 2-mediated acute lung injury and necrotic cell death. *Nat Med* 2006;**12**(11):1286-93.
- 20 60. Kumpers P, Lukasz A, David S, Horn R, Hafer C, Faulhaber-Walter R, *et al.* Excess circulating
21 angiopoietin-2 is a strong predictor of mortality in critically ill medical patients. *Crit Care*
22 2008;**12**(6):R147.
- 23 61. Benz AP, Hijazi Z, Lindback J, Connolly SJ, Eikelboom JW, Kastner P, *et al.* Plasma
24 angiopoietin-2 and its association with heart failure in patients with atrial fibrillation. *Europace*
25 2023;**25**(7).
- 26 62. Kahr PC, Piccini I, Fabritz L, Greber B, Scholer H, Scheld HH, *et al.* Systematic Analysis of Gene
27 Expression Differences between Left and Right Atria in Different Mouse Strains and in Human Atrial
28 Tissue. *PLoS ONE* 2011;**6**(10):e26389.
- 29 63. Wang L, Rice M, Swist S, Kubin T, Wu F, Wang S, *et al.* BMP9 and BMP10 Act Directly on
30 Vascular Smooth Muscle Cells for Generation and Maintenance of the Contractile State. *Circulation*
31 2021;**143**(14):1394-1410.
- 32 64. Hodgson J, Swietlik EM, Salmon RM, Hadinnapola C, Nikolic I, Wharton J, *et al.*
33 Characterization of GDF2 Mutations and Levels of BMP9 and BMP10 in Pulmonary Arterial
34 Hypertension. *Am J Respir Crit Care Med* 2020;**201**(5):575-585.
- 35 65. Steimle JD, Grisanti Canozo FJ, Park M, Kadow ZA, Samee MAH, Martin JF. Decoding the
36 PITX2-controlled genetic network in atrial fibrillation. *JCI Insight* 2022;**7**(11).
- 37 66. Hennings E, Aeschbacher S, Coslovsky M, Paladini RE, Meyre PB, Voellmin G, *et al.*
38 Association of bone morphogenetic protein 10 and recurrent atrial fibrillation after catheter
39 ablation. *Europace* 2023;**25**(6).
- 40 67. Hennings E, Blum S, Aeschbacher S, Coslovsky M, Knecht S, Eken C, *et al.* Bone
41 Morphogenetic Protein 10-A Novel Biomarker to Predict Adverse Outcomes in Patients With Atrial
42 Fibrillation. *J Am Heart Assoc* 2023;**12**(6):e028255.
- 43 68. Winters J, Kawczynski MJ, Gilbers MD, Isaacs A, Zeemering S, Bidar E, *et al.* Circulating
44 BMP10 Levels Associate With Late Postoperative Atrial Fibrillation and Left Atrial Endomysial
45 Fibrosis. *JACC Clin Electrophysiol* 2024.
- 46 69. Billing AM, Kim YC, Gullaksen S, Schrage B, Raabe J, Hutzfeldt A, *et al.* Metabolic
47 Communication by SGLT2 Inhibition. *Circulation* 2024;**149**(11):860-884.
- 48 70. Hijazi Z, Benz AP, Lindback J, Alexander JH, Connolly SJ, Eikelboom JW, *et al.* Bone
49 morphogenetic protein 10: a novel risk marker of ischaemic stroke in patients with atrial fibrillation.
50 *Eur Heart J* 2023;**44**(3):208-218.

- 1 71. Winters J, Isaacs A, Zeemering S, Kawczynski M, Maesen B, Maessen J, *et al.* Heart Failure,
2 Female Sex, and Atrial Fibrillation Are the Main Drivers of Human Atrial Cardiomyopathy: Results
3 From the CATCH ME Consortium. *J Am Heart Assoc* 2023;**12**(22):e031220.
- 4 72. Barallobre-Barreiro J, Gupta SK, Zoccarato A, Kitazume-Taneike R, Fava M, Yin X, *et al.*
5 Glycoproteomics Reveals Decorin Peptides With Anti-Myostatin Activity in Human Atrial Fibrillation.
6 *Circulation* 2016;**134**(11):817-32.
- 7 73. Ko D, Benson MD, Ngo D, Yang Q, Larson MG, Wang TJ, *et al.* Proteomics Profiling and Risk of
8 New-Onset Atrial Fibrillation: Framingham Heart Study. *J Am Heart Assoc* 2019;**8**(6):e010976.
- 9

ACCEPTED MANUSCRIPT

1 **Figure legends**

2

3 **Graphical Abstract.** In patients diagnosed with atrial fibrillation, low concentrations of
4 NT-proBNP, BMP10 and ANGPT2 at baseline predict sinus rhythm at 12-month follow-up in
5 context with clinical features. This was validated in additional datasets, of which AXAFA-
6 AFNET 5 is depicted here. A treatment interaction shows that NT-proBNP's predictive value
7 is impacted by early rhythm control treatment.

8 *AF, atrial fibrillation; ANGPT2, angiotensin 2; BMP10, bone morphogenetic protein 10; NT-*
9 *proBNP, N-terminal pro-B-type natriuretic peptide*

10

11 **Figure 1.** Low concentrations of the biomarkers NT-proBNP, Angiotensin 2 and Bone
12 morphogenetic protein 10 predict sinus rhythm at 12-month follow-up in the derivation
13 dataset (EAST-AFNET 4). Odds ratios for sinus rhythm at 12-month follow-up (A) and odds
14 ratios by randomized treatment group (B). Forest plot showing odds ratios for each
15 biomarker for the outcome sinus rhythm at 12-month follow-up and 95% confidence
16 intervals. The odds ratio for NT-proBNP shows an interaction between NT-proBNP
17 concentrations and randomized treatment group (early rhythm control or usual care). All
18 odds ratios are corrected for clinical features, age, sex, EAST study center, rhythm at
19 baseline, atrial fibrillation type, randomized treatment group, body mass index, diastolic
20 blood pressure, and left ventricular ejection fraction. Even after multiple confounding, high
21 biomarker concentrations indicate lower odds of sinus rhythm at 12-month follow-up. Low
22 concentrations of NT-proBNP predict sinus rhythm at 12-month follow-up in patients with
23 usual care (only symptomatic rhythm control). High concentrations of NT-proBNP do not
24 necessarily predict lack of sinus rhythm at 12 months if patients receive early rhythm control.

25 *Angiotensin 2 (ANGPT2), bone morphogenetic protein 10 (BMP10), cancer antigen 125 (CA125), C-*
26 *reactive protein (CRP), D-dimer, endothelial specific molecule 1 (ESM1), fatty acid binding protein 3*
27 *(FABP3), fibroblast growth factor 23 (FGF23), growth differentiation factor 15 (GDF15), insulin-like*

1 *growth factor binding protein 7 (IGFBP7), interleukin-6 (IL-6), N-terminal pro-B-type natriuretic*
2 *peptide (NT-proBNP), cardiac troponin (TnT) and serum creatinine (sCr).*

3
4 **Figure 2.** Biomarker concentration distributions at baseline in patients with sinus rhythm
5 (teal) or atrial fibrillation (orange) at 12-month follow-up. Violin plot of the distribution of
6 log-transformed biomarker concentrations for each of 14 biomarkers at baseline, split by the
7 outcome of rhythm at 12-month follow-up. Log-transformed biomarker concentrations are
8 shown on the y-axis and the kernel estimated frequency on the x-axis. Central thick
9 horizontal lines are the median and the thinner lines represent interquartile range.

10 N-terminal pro B-type Natriuretic Peptide, Angiotensin 2 and Bone Morphogenetic Protein
11 10 show an association with sinus rhythm at 12-month follow-up based on the acceptance of
12 a Type 1 error of 5%. P-values were calculated using mixed logistic regression model with site
13 as random effect, adjusted for age, sex, study site, rhythm at baseline, randomized group
14 (early rhythm control or usual care), body mass index, diastolic blood pressure, and left
15 ventricular ejection fraction, those clinical features that were associated with outcomes
16 including sinus rhythm in the main EAST-AFNET 4 trial.

17 *Angiotensin 2 (ANGPT2), bone morphogenetic protein 10 (BMP10), cancer antigen 125 (CA125), C-*
18 *reactive protein (CRP), D-dimer, endothelial specific molecule 1 (ESM1), fatty acid binding protein 3*
19 *(FABP3), fibroblast growth factor 23 (FGF23), growth differentiation factor 15 (GDF15), insulin-like*
20 *growth factor binding protein 7 (IGFBP7), interleukin-6 (IL-6), N-terminal pro-B-type natriuretic*
21 *peptide (NT-proBNP), cardiac troponin (TnT) and serum creatinine (sCr).*

22
23 **Figure 3.** Biomarkers measured at baseline predicting sinus rhythm at 12-month follow-up
24 in all participants of the biomarker study, separately analysed by rhythm at baseline (atrial
25 fibrillation at baseline or sinus rhythm at baseline) and randomized treatment group (early
26 rhythm control or usual care), respectively, in a post-hoc analysis.

1 Of the three biomarkers identified to be predictive of sinus rhythm in the whole cohort, NT-
2 proBNP, ANGPT2 and BMP10, all three biomarkers retained their predictive value in the
3 subgroup of patients randomized to usual care. All three biomarkers also retained their
4 predictive value in the subgroup of patients in atrial fibrillation during blood draw at
5 baseline.

6 *Angiotensin 2 (ANGPT2), bone morphogenetic protein 10 (BMP10), N-terminal pro-B-type*
7 *natriuretic peptide (NT-proBNP)*

8

9 **Figure 4.** Internal validation: Angiotensin 2, Bone morphogenetic protein 10 and NT-
10 proBNP biomarkers at baseline predict Sinus Rhythm at 24-month follow-up even after
11 correction for multiple confounders. Odds ratios for sinus rhythm at 24-month follow-up.
12 This analysis provides an internal validation of the biomarkers predicting sinus rhythm at
13 12-month follow-up (**Figure 1**). All Odds ratios are corrected for clinical age, sex, study site,
14 rhythm at baseline, randomized treatment group (early rhythm control or usual care), body
15 mass index, diastolic blood pressure, and left ventricular ejection fraction, those clinical
16 features that were associated with outcomes including sinus rhythm in the main EAST-
17 AFNET 4 trial⁴. Low concentrations of NT-proBNP, ANGPT2 and BMP10 predict sinus
18 rhythm at 24-month follow-up in patients. Accordingly, high concentrations predict lack of
19 sinus rhythm at 24-month follow-up.

20 *Angiotensin 2 (ANGPT2), bone morphogenetic protein 10 (BMP10), cancer antigen 125 (CA125), C-*
21 *reactive protein (CRP), D-dimer, endothelial specific molecule 1 (ESM1), fatty acid binding protein 3*
22 *(FABP3), fibroblast growth factor 23 (FGF23), growth differentiation factor 15 (GDF15), insulin-like*
23 *growth factor binding protein 7 (IGFBP7), interleukin-6 (IL-6), N-terminal pro-B-type natriuretic*
24 *peptide (NT-proBNP), cardiac troponin (TnT) and serum creatinine (sCr).*

25

26 **Figure 5.** Validation applying biomarker based clusters indicating cardiovascular outcome
27 risk: Patients at high risk of cardiovascular complications as estimated by biomarker-based

1 clusters have reduced odds of sinus rhythm at 12-month and 24-month follow-up. Odds ratio
2 for the high cardiovascular outcome risk (red) and intermediate cardiovascular outcome risk
3 biomarker clusters (orange and green) for sinus rhythm at 12-months follow-up (A above)
4 and at 24-month follow-up (B bottom) tested against the low cardiovascular risk cluster (not
5 depicted as used as reference). All odds ratios are corrected for age, sex, study center, rhythm
6 at baseline, atrial fibrillation type (depicted) randomized treatment group (early rhythm
7 control or usual care), as well as body mass index, diastolic blood pressure, and left
8 ventricular ejection fraction, the clinical features that were associated with outcomes
9 including sinus rhythm in the main EAST-FNET 4 trial⁴.

10 *AF (atrial fibrillation)*

11

12 **Figure 6.** Validation by Random forest analyses identified highest importance for similar
13 biomarkers, alongside rhythm at baseline and AF pattern, as predictors of sinus rhythm at
14 12-month follow-up (Figure 6A – importance, Figure 6B – SHAP value).

15 *Atrial fibrillation (AF), Angiotensin 2 (ANGPT2), baseline (BL), body mass index (BMI), bone*
16 *morphogenetic protein 10 (BMP10), cancer antigen 125 (CA125), C-reactive protein (CRP), D-dimer,*
17 *endothelial specific molecule 1 (ESM1), fatty acid binding protein 3 (FABP3), fibroblast growth*
18 *factor 23 (FGF23), growth differentiation factor 15 (GDF15), insulin-like growth factor binding*
19 *protein 7 (IGFBP7), interleukin-6 (IL-6), mixed-effects random forest (MERF), N-terminal pro-B-*
20 *type natriuretic peptide (NT-proBNP), cardiac troponin (TnT), serum creatinine (sCr) and SHapley*
21 *Additive exPlanations (SHAP)*

22

23

1 **Figure 7. External validation of the prediction of sinus rhythm at the end of**
2 **follow-up by baseline biomarkers in AXAFA-AFNET 5.** AXAFA –AFNET 5 enrolled
3 674 patients undergoing a first AF ablation with at least one stroke risk factor. Patients were
4 randomized to apixaban or vitamin K antagonist therapy without affecting rhythm.
5 Individual models with rhythm at baseline, age, and sex were constructed to determine
6 whether each biomarker predicts sinus rhythm at the end of follow-up 120 days after
7 randomization, 549 patients with sinus rhythm, 71 patients with atrial fibrillation.

8 * *p-values were calculated using logistic regression, adjusted for sex, age, rhythm at baseline and*
9 *treatment group. Abbreviations: Angiopoietin 2 (ANGPT2), bone morphogenetic protein 10*
10 *(BMP10), N-terminal pro-B-type natriuretic peptide (NT-proBNP).*

11

1 **Table 1:** Baseline characteristics and biomarkers in the EAST-AFNET 4 biomolecule study.

Treatment group		Early rhythm control	Usual care	p-value*
n		800	786	
sex: Female		355 (44%)	358 (46%)	0.639
Age (years)		71 [66, 75]	71 [66, 76]	0.711
BMI		28.7 [25.6, 32.1]	29.0 [25.6, 32.5]	0.699
Blood pressure (systolic) (mmHg)		135 [123, 150]	135 [125, 148]	0.730
Blood pressure (diastolic) (mmHg)		80 [74, 90]	80 [74, 90]	0.716
LVEF (%)		60 [55, 65]	60 [55, 65]	0.873
AF type (First episode)		290 (36%)	270 (34%)	
AF type (Paroxysmal)		302 (38%)	288 (37%)	0.839
AF type (Persistent)		208 (26%)	228 (29%)	0.202
Other Clinical Characteristics				
Diabetes		207 (26%)	189 (24%)	0.400
Hypertension		494 (62%)	512 (65%)	0.170
Chronic kidney disease		98 (12%)	97 (12%)	0.956
Estimated Glomerular Filtration Rate (ml/min1.73 m ²)		75 [63 - 87]	76 [64 - 87]	0.734
Previous stroke or transient ischemic attack		114 (14%)	81 (10%)	0.017
Chronic obstructive pulmonary disease		63 (8%)	61 (8%)	0.991
Diastolic LA diameter (mm)		42 [38, 47]	43 [39, 47]	0.730
NYHA class				
No heart failure		523 (65%)	509 (65%)	
I		82 (10%)	88 (11%)	0.555
II		164 (21%)	160 (20%)	0.985
III		31 (4%)	29 (4%)	0.882
EHRA-score				
I		232 (29%)	236 (30%)	
II		386 (48%)	374 (48%)	0.679
III		122 (15%)	122 (15%)	0.914
IV		8 (1%)	9 (1%)	0.839
Missing		52 (7%)	45 (6%)	
Biomarker (unit)				
	Coefficient of variation			
NT-proBNP (pg/ml)	1.51	441 [175 - 966]	467 [187 - 1036]	0.537
ANGPT2 (ng/ml)	0.70	2.53 [1.87 - 3.65]	2.53 [1.87 - 3.75]	0.456
BMP10 (ng/ml)	0.24	2.10 [1.82 - 2.41]	2.11 [1.83 - 2.45]	0.507
FGF23 (pg/ml)	1.27	155 [115 - 218]	153 [115 - 211]	0.244
ESM1 (ng/ml)	0.76	2.04 [1.64 - 2.59]	2.05 [1.63 - 2.63]	0.818
GDF15 (pg/ml)	0.80	1333 [990 - 2000]	1359 [971 - 2005]	0.078
IGFBP7 (ng/ml)	0.26	102 [90.7 - 117]	102 [90.1 - 117]	0.457
IL-6 (pg/ml)	6.62	2.56 [1.64 - 4.04]	2.68 [1.67 - 4.18]	0.479
FABP3 (ng/ml)	0.50	32.0 [26.3 - 39.6]	31.9 [26.4 - 39.6]	0.837
D-dimer (µg/ml)	1.74	0.17 [0.09 - 0.34]	0.16 [0.08 - 0.36]	0.506

TnT (ng/l)	2.26	11.1 [8.02 – 16.6]	11.4 [8.21 – 16.7]	0.337
CRP (mg/l)	3.28	2.02 [0.96 – 4.99]	2.38 [1.04 – 4.75]	0.392
sCr (μmol/l)	0.29	81.7 [70.7 – 95.5]	80.4 [70.0 - 94.5]	0.771
CA125 (U/ml)	1.51	11.5 [8.08 - 15.9]	11.1 [7.93 - 16.1]	0.433

1 * *p*-values were calculated on the unimputed dataset using mixed logistic regression model with site
 2 as random effect, for biomarkers additionally adjusted for sex, age, body mass index, diastolic blood
 3 pressure, left ventricular ejection fraction and AF-type. Distributions are shown as mean and SD for
 4 normally distributed values, as median and IQR for non-normally distributed values and
 5 biomarkers, and as frequency (percentage) for nominal features.

6 Abbreviations: Atrial Fibrillation (AF), sinus rhythm (SR), early rhythm control (ERC), usual care
 7 (UC), body mass index (BMI), left ventricular ejection fraction (LVEF), left atrium (LA), New York
 8 Heart Association Functional Classification of heart failure (NYHA), European Heart Rhythm
 9 Association score (EHRA), angiotensin 2 (ANGPT2), bone morphogenetic protein 10 (BMP10),
 10 cancer antigen 125 (CA125), C-reactive protein (CRP), D-dimer, endothelial specific molecule 1
 11 (ESM1), fatty acid binding protein 3 (FABP3), fibroblast growth factor 23 (FGF23), growth
 12 differentiation factor 15 (GDF15), insulin-like growth factor binding protein 7 (IGFBP7), interleukin-
 13 6 (IL-6), N-terminal pro-B-type natriuretic peptide (NT-proBNP), cardiac troponin (TnT) and
 14 serum creatinine (sCr). Estimated Glomerular Filtration (eGFR) rate was calculated as CKD EPI,
 15 Chronic Kidney Disease Epidemiology Collaboration

16

17

1 **Table 2.** Baseline biomarker concentrations are shown split by patient rhythm at 12-month
 2 follow-up (sinus rhythm or atrial fibrillation) and by randomized group (early rhythm
 3 control or usual care).

Randomisation group	Early rhythm control		Usual care		SR vs. AF 12 month
	Sinus rhythm 12-month FU	AF 12-month FU	Sinus rhythm 12-month FU	AF 12-month FU	p-value*
NT-proBNP (pg/ml)	377 [164 - 859]	750 [376 - 1351]	294 [127 - 700]	782 [437 - 1454]	0.001
ANGPT2 (ng/ml)	2.34 [1.78 - 3.41]	3.45 [2.43 - 5.62]	2.24 [1.7 - 3.09]	3.31 [2.15 - 4.62]	0.001
BMP10 (ng/ml)	2.08 [1.8 - 2.39]	2.21 [1.96 - 2.58]	2.04 [1.79 - 2.34]	2.24 [1.94 - 2.66]	0.010
FGF23 (pg/ml)	151 [112 - 209]	179 [125 - 238]	141 [108 - 197]	168 [129 - 226]	0.429
ESM1 (ng/ml)	2.01 [1.61 - 2.52]	2.17 [1.75 - 2.88]	1.98 [1.57 - 2.52]	2.09 [1.73 - 2.66]	0.218
GDF15 (pg/ml)	1304 [958 - 1934]	1441 [997 - 2008]	1254 [911 - 1782]	1589 [1071 - 2347]	0.461
IGFBP7 (ng/ml)	100 [89 - 114]	108 [93 - 126]	98.8 [88.5 - 110]	104 [94.7 - 119]	0.487
IL-6 (pg/ml)	2.47 [1.57 - 3.88]	2.6 [1.76 - 4.62]	2.37 [1.56 - 3.6]	3.02 [1.98 - 4.65]	0.417
FABP3 (ng/ml)	31.4 [25.6 - 39]	35.3 [28.3 - 43.4]	30.4 [25.7 - 37.8]	33.5 [28.0 - 42.0]	0.151
D-dimer (µg/ml)	0.17 [0.08 - 0.33]	0.19 [0.1 - 0.36]	0.16 [0.08 - 0.32]	0.16 [0.08 - 0.32]	0.638
TnT (ng/l)	10.6 [7.81 - 15.7]	13.0 [9 - 17.6]	10.3 [7.53 - 15.5]	12.5 [8.68 - 17.7]	0.415
CRP (mg/l)	2 [0.95 - 4.65]	1.97 [0.9 - 4.63]	2.07 [0.93 - 4.37]	2.52 [1.12 - 4.87]	0.910
sCr (µmol/l)	81.3 [70 - 95]	83 [72.7 - 94.8]	79.5 [68.0 - 91.9]	84.4 [72 - 97.2]	0.541
CA125 (U/ml)	11.4 [8.0 - 15.8]	12.3 [8.3 - 17.1]	10.8 [7.8 - 15.7]	11.4 [7.96 - 15.9]	0.779

4 * p-values were calculated using mixed logistic regression model with site as random effect, adjusted
 5 for sex, age, body mass index, diastolic blood pressure, left ventricular ejection fraction and
 6 randomization group. Values are shown as median [IQR].

7
 8 Abbreviations: Atrial fibrillation (AF), sinus rhythm (SR), follow-up (FU), angiotensin 2 (ANGPT2),
 9 bone morphogenetic protein 10 (BMP10), cancer antigen 125 (CA125), C-reactive protein (CRP), D-
 10 dimer, endothelial specific molecule 1 (ESM1), fatty acid binding protein 3 (FABP3), fibroblast
 11 growth factor 23 (FGF23), growth differentiation factor 15 (GDF15), insulin-like growth factor
 12 binding protein 7 (IGFBP7), interleukin-6 (IL-6), N-terminal pro-B-type natriuretic peptide (NT-
 13 proBNP), cardiac troponin (TnT) and serum creatinine (sCr).

14
 15
 16

1 **Table 3:** Baseline clinical characteristics used as confounders and biomarker concentrations
 2 in the derivation data set (EAST-AFNET 4 biomolecule study) at baseline by randomized
 3 group and by baseline rhythm. Rhythm at time of blood sampling was included as a fix factor
 4 in the analyses of outcome.

Group	Early rhythm control		Usual care		p-value
Baseline rhythm	Sinus rhythm	Atrial fibrillation	Sinus rhythm	Atrial fibrillation	rhythm*
N	452	348	438	348	
Women	220 (49%)	135 (39%)	221 (51%)	137 (39%)	<0.001
Age, years	70 [65 - 75]	71 [67- 76]	71 [66 - 75]	72 [66 - 76]	0.035
BMI	28.7 [25.8 - 31.6]	28.4 [25.5 - 32.9]	28.4 [25.4 - 31.4]	29.4 [25.9 - 33.3]	0.022
Blood pressure (diastolic) (mmHg)	80 [72, 87]	80 [76, 90]	80 [71, 89]	80 [76, 90]	<0.001
LVEF (%)	60 [57, 65]	59 [50, 64]	60 [59, 65]	60 [51, 64]	<0.001
AF type: First episode	172 (38%)	118 (34%)	155 (35%)	115 (33%)	
AF type: Paroxysmal	235 (52%)	67 (19%)	223 (51%)	65 (19%)	<0.001
AF type: Persistent or long-standing persistent	45 (10%)	163 (47%)	60 (14%)	168 (48%)	<0.001
Biomarker concentrations					
NT-proBNP (pg/ml)	228 [121 - 467]	890 [506 - 1496]	253 [124 - 504]	934 [529 - 1603]	<0.001
ANGPT2 (ng/ml)	2.20 [1.65 - 2.76]	3.39 [2.29 - 5.14]	2.12 [1.63 - 3.00]	3.35 [2.31 - 4.81]	<0.001
BMP10 (ng/ml)	2.03 [1.73 - 2.30]	2.22 [1.96 - 2.58]	2.01 [1.76 - 2.29]	2.25 [1.96 - 2.69]	<0.001
FGF23 (pg/ml)	139 [106 - 194]	178 [128 - 247]	140 [110 - 192]	170 [130 - 243]	0.003
ESM1 (ng/ml)	1.97 [1.58 - 2.44]	2.14 [1.74 - 2.84]	1.96 [1.57 - 2.56]	2.15 [1.74 - 2.78]	0.002
GDF15 (pg/ml)	1251 [938 - 1847]	1478 [1058 - 2188]	1259 [914 - 1761]	1585 [1065 - 2272]	<0.001
IGFBP7 (ng/ml)	99.0 [89.3 - 111.2]	106.8 [93.6 - 125]	99.2 [87.9 - 111]	105 [93.8 - 123]	<0.001
IL-6 (pg/ml)	2.22 [1.50 - 3.58]	3.03 [1.99 - 4.88]	2.42 [1.58 - 3.89]	3.02 [1.95 - 4.59]	0.041
FABP3 (ng/ml)	30.2 [25.1 - 38.1]	34.2 [28.2 - 42.1]	30.9 [25.6 - 37.9]	33.3 [27.1 - 42.6]	0.020
D-dimer (µg/ml)	0.17 [0.08 - 0.32]	0.18 [0.09 - 0.36]	0.15 [0.08 - 0.32]	0.18 [0.09 - 0.4]	0.267
TnT (ng/l)	10.1 [7.39 - 14.5]	12.7 [9 - 18.8]	10.7 [7.6 - 15.7]	12.5 [8.73 - 18.3]	0.436
CRP (mg/l)	1.76 [0.87 - 4.29]	2.48 [1.09 - 5.78]	2.08 [0.93 - 4.52]	2.58 [1.26 - 5.03]	0.130

sCr (μmol/l)	80.0 [69.0 - 93.7]	84.0 [71.0 - 97.0]	79.6 [68.1 - 92.0]	83.9 [71.0 - 97.2]	0.296
CA125 (U/ml)	11.1 [8.01 - 14.9]	12.3 [8.4 - 16.9]	10.8 [8.02 - 15.7]	11.4 [7.84 - 16.7]	0.052

1 * p-values were calculated in the unimputed, pooled dataset (ERC and UC combined) using mixed
2 logistic regression model with site as random effect, for the biomarkers additionally adjusted for sex,
3 age, body mass index, diastolic blood pressure, left ventricular ejection fraction and AF-type, the
4 clinical features that were associated with outcomes including sinus rhythm in the main EAST-AFNET
5 4 dataset⁴.

6 Distributions are shown as mean and SD for normally distributed values, as median and IQR for non-
7 normal distributed values and biomarkers, and as frequency (percentage) for nominal features. For
8 biomarker concentrations there were no differences between the randomized groups, but differences
9 between Sinus rhythm or AF during the baseline visit.

10 Abbreviations: Atrial fibrillation (AF), Early rhythm control (ERC), usual care (UC), body mass index
11 (BMI), atrial fibrillation (AF), left ventricular ejection fraction (LVEF), Angiotensin 2 (ANGPT2),
12 bone morphogenetic protein 10 (BMP10), cancer antigen 125 (CA125), C-reactive protein (CRP), D-
13 dimer, endothelial specific molecule 1 (ESM1), fatty acid binding protein 3 (FABP3), fibroblast growth
14 factor 23 (FGF23), growth differentiation factor 15 (GDF15), insulin-like growth factor binding
15 protein 7 (IGFBP7), interleukin-6 (IL-6), N-terminal pro-B-type natriuretic peptide (NT-proBNP),
16 cardiac troponin (TnT) and serum creatinine (sCr).

17

18

1 **Table 4** Threshold concentrations for NT-proBNP, BMP10, and ANGPT2 determined in the
2 derivation dataset (EAST-AFNET 4 biomolecule study). The lower threshold was defined as
3 the nearest round concentration below which 80% of patients attained sinus rhythm at 12
4 months. The higher threshold was defined as the nearest rounded concentration above which
5 40% of patients were in AF at 12 months.

Biomarker	Low threshold (>80% sinus rhythm at 12 months)	High threshold (>40% AF at 12 months)
NT-proBNP (pg/ml)	<1000	>1500
BMP10 (ng/ml)	<2	>3
ANGPT2 (ng/ml)	<3.5	>3.5

6

7 *Abbreviations: Atrial Fibrillation (AF), angiotensin 2 (ANGPT2), bone morphogenetic protein 10*
8 *(BMP10), N-terminal pro-B-type natriuretic peptide (NT-proBNP)*

9

10

1 **Table 5. Estimated clinical utility of adding NT-proBNP, BMP10, and ANGPT2**
 2 **alone or in combination to a clinical risk score to predict sinus rhythm at 12**
 3 **months.** Sinus rhythm at 12 months was initially predicted by a clinical risk score based on
 4 three validated clinical features (LA size >50mm, persistent AF, age >75 years) alone. This
 5 reference score was then combined with one, a combination of two or all three binarized
 6 predictive biomarkers (see Table 3 on biomarker thresholds: NT-proBNP <1000 pg/ml or
 7 >1500 pg/ml, ANGPT2 <3.5 ng/ml or >3.5 ng/ml, BMP10 <2 ng/ml or >3 ng/ml). If either
 8 the clinical risk score is ≥ 2 or any of the biomarkers added to the model surpasses its
 9 threshold, the model predicts failure to attain sinus rhythm at 12-month follow-up and
 10 predicts AF instead. By definition, there was no reclassification into low-risk groups. All
 11 numbers indicate number of patients with percentages of the predicted class in brackets.

12
 13
 14

	Patients reclassified as high risk of not attaining Sinus Rhythm at 12M (N)	Confusion matrix			
		Predicted Sinus Rhythm (Actual patients in SR: N=1081)		Predicted: AF (Actual patients in AF: N=365)	
		Patients in Sinus Rhythm at 12M	Patients in AF at 12M	Patients in AF at 12M	Patients in Sinus Rhythm at 12M
Clinical model*	Reference	813 (77%)	245 (23%)	75 (40%)	112 (60%)
+ NT-proBNP	135	743 (79%)	201 (21%)	129 (40%)	191 (60%)
+ BMP10	240	670 (79%)	175 (21%)	161 (36%)	279 (64%)
+ ANGPT2	301	650 (81%)	145 (19%)	198 (39%)	303 (61%)
+ NT-proBNP and BMP10	298	638 (80%)	158 (20%)	183 (36%)	315 (64%)
+ NT-proBNP and ANGPT2	345	625 (82%)	130 (18%)	215 (39%)	332 (61%)
+ ANGPT2 and BMP10	410	570 (82%)	125 (18%)	223 (36%)	394 (64%)
+ NT-proBNP and BMP10 and ANGPT2	441	551 (83%)	115 (17%)	234 (36%)	416 (64%)

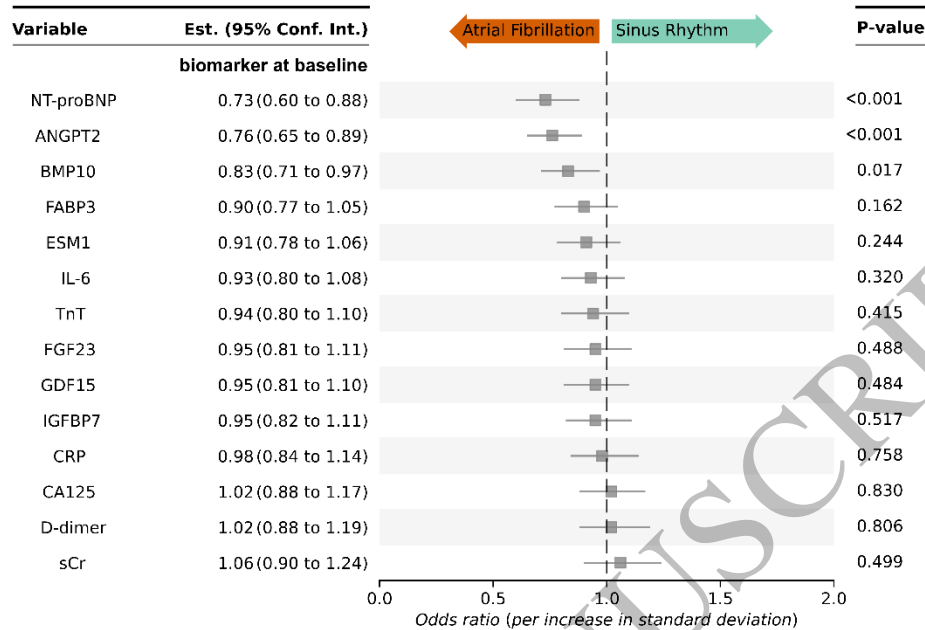
15

16 *Abbreviations: Atrial Fibrillation (AF), angiotensin 2 (ANGPT2), bone morphogenetic protein 10*
 17 *(BMP10), N-terminal pro-B-type natriuretic peptide (NT-proBNP), Sinus Rhythm (SR), 12-month*
 18 *Follow-up (12M)*

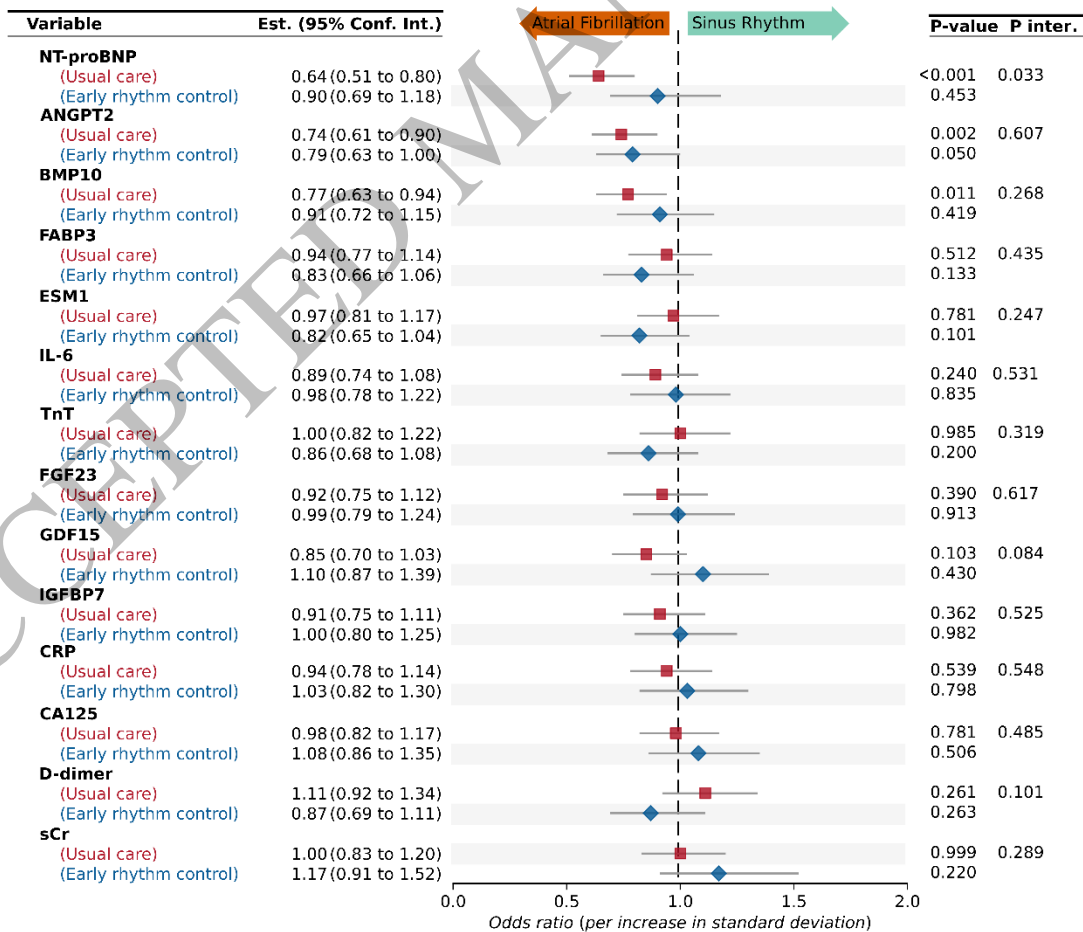
19 There were 140 missing values in outcomes and 225 missing values in LA size. The
 20 additional use of biomarkers for prediction can lead to differing missing values in
 21 predictions made for participants with available outcome data.

22
 23
 24

A Each Biomarker at Baseline and Sinus Rhythm at 12-month Follow-up



B Sinus Rhythm at 12-month Follow-up by Randomized Groups



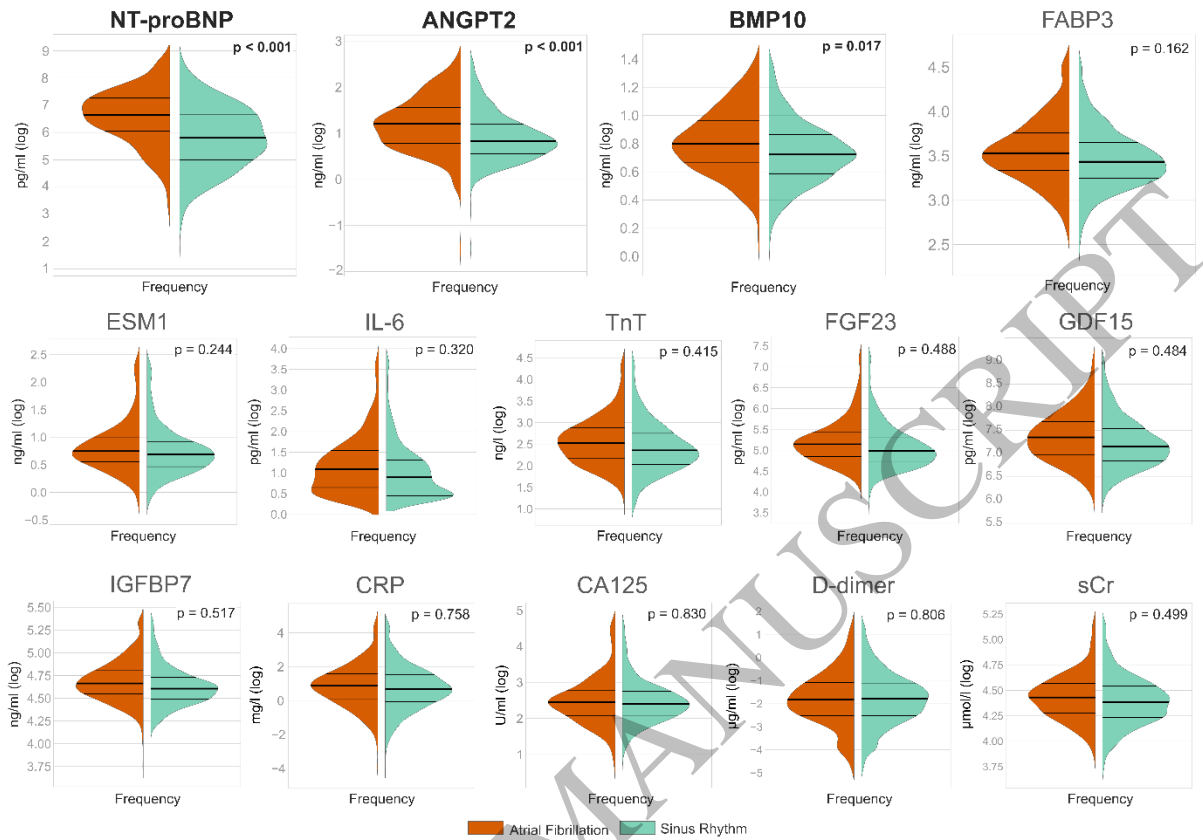
1

2

3

Figure 1
159x227 mm (x DPI)

Biomarker Concentrations at Baseline in Patients with Sinus Rhythm or Atrial Fibrillation at 12-month Follow-up



1
2
3
4

Figure 2
159x121 mm (x DPI)

Each Biomarker at Baseline and Sinus Rhythm at 12-month Follow-up

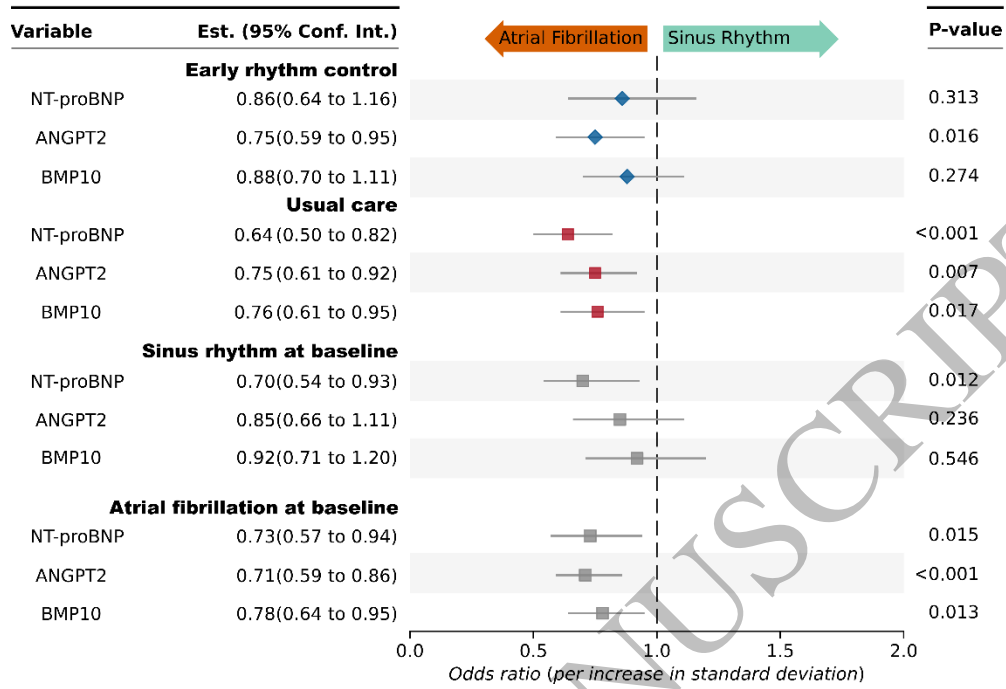


Figure 3
159x102 mm (x DPI)

1
2
3
4

Each Biomarker at Baseline and Sinus Rhythm at 24-month Follow-up

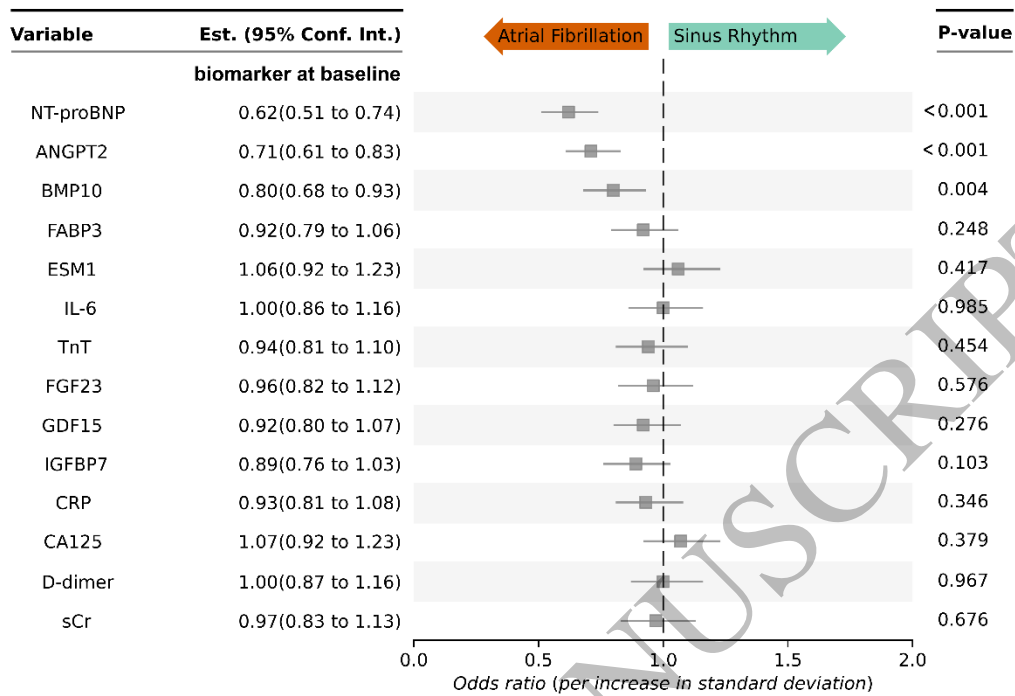
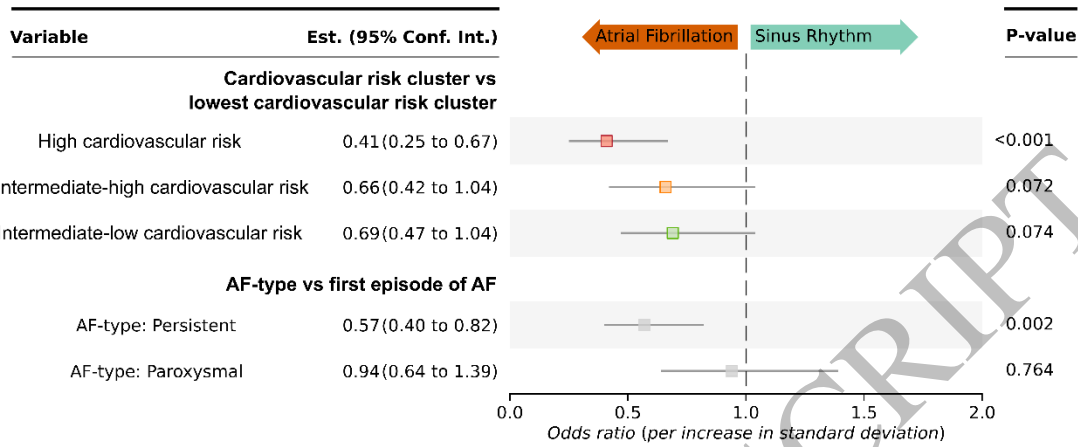


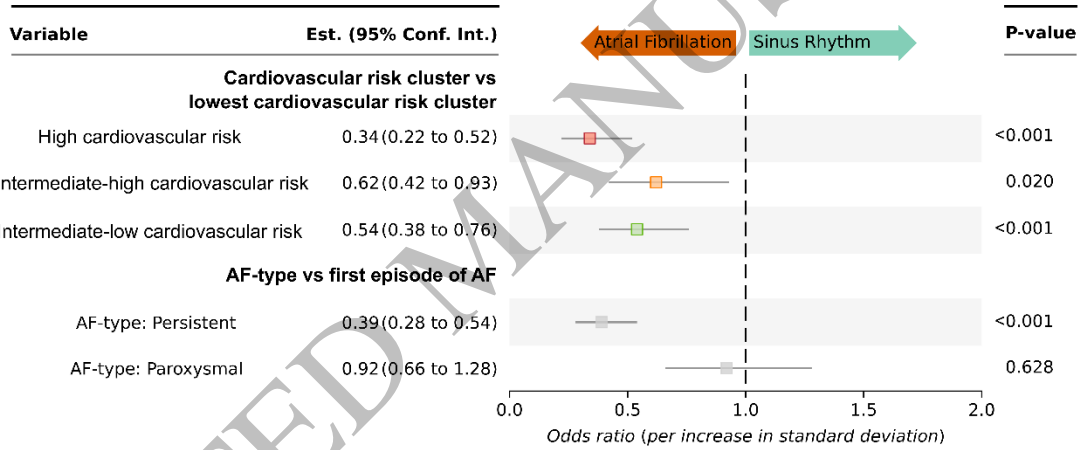
Figure 4
159x102 mm (x DPI)

1
2
3
4

A Cardiovascular Risk Clusters and Sinus Rhythm at 12-month Follow-up

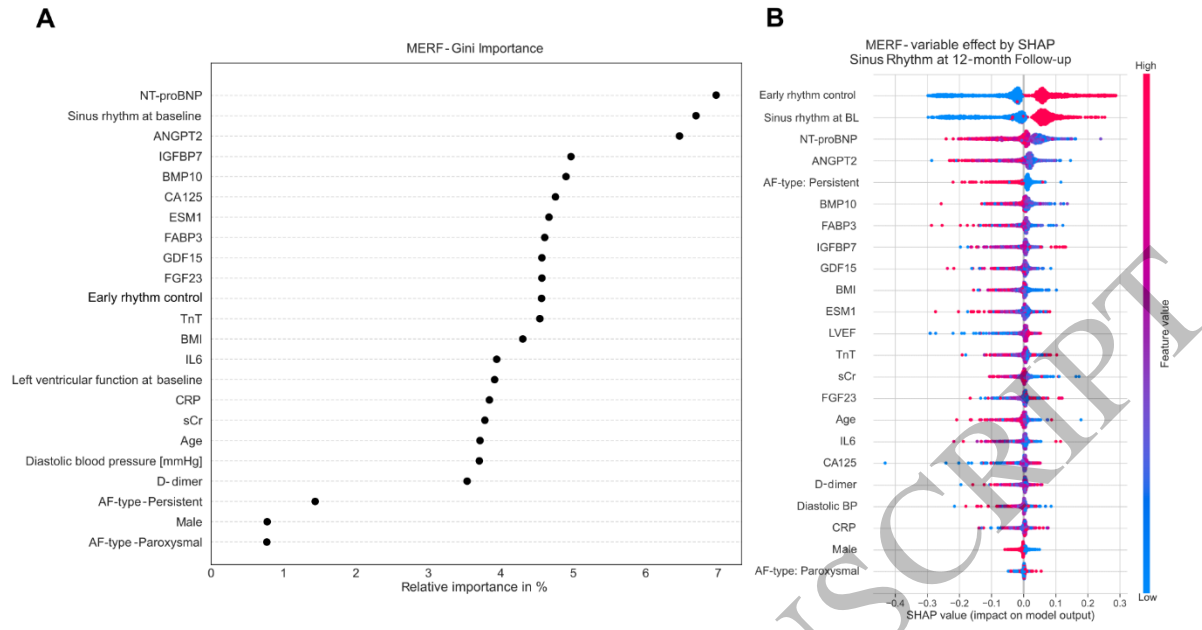


B Cardiovascular Risk Clusters and Sinus Rhythm at 24-month Follow-up



1
2
3
4

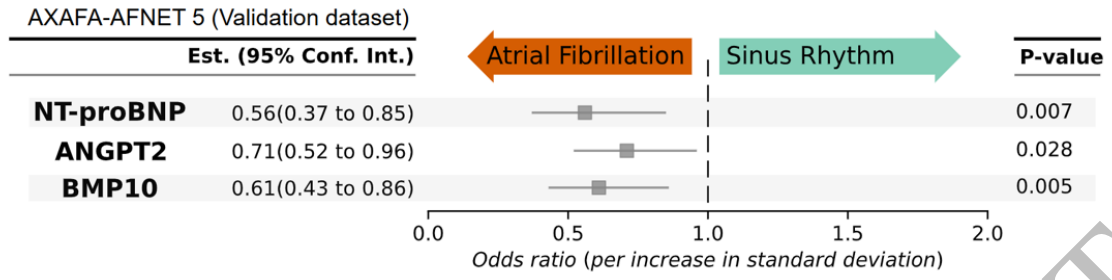
Figure 5
159x147 mm (x DPI)



1
2
3
4

Figure 6
159x86 mm (x DPI)

ACCEPTED MANUSCRIPT



1
2
3
4

Figure 7
159x45 mm (x DPI)

ACCEPTED MANUSCRIPT

1 **Structured Graphical Abstract**

2 Key Question

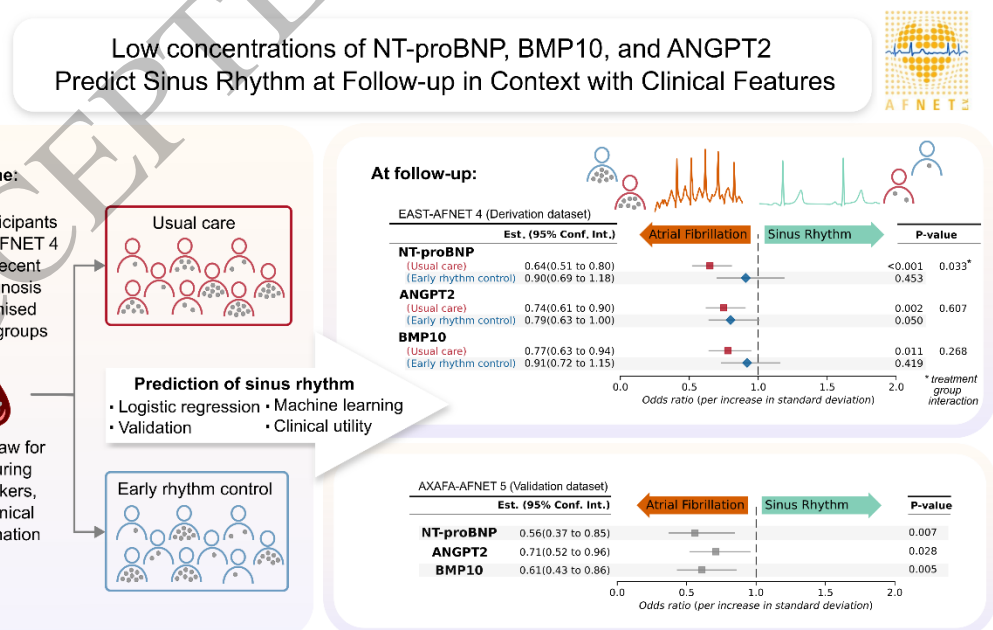
3 Can clinical features or circulating biomarkers measured at baseline predict sinus rhythm
 4 during mid-term follow-up in patients with atrial fibrillation (AF)? Which biomarkers
 5 interact with early rhythm control therapy?

6 Key Finding

7 Low baseline concentrations of angiotensin 2 (ANGPT2), bone morphogenetic protein 10
 8 (BMP10), and N-terminal pro-B-type natriuretic peptide (NT-proBNP) predicted sinus
 9 rhythm at follow-up in the EAST-AFNET 4 trial and in two external validation datasets. NT-
 10 proBNP had reduced predictive value in patients treated with early rhythm control. ANGPT2
 11 and BMP10 added most information in patients who were in AF when blood samples were
 12 taken. The three biomarkers refined prediction of sinus rhythm compared to a clinical risk
 13 score.

14 Take Home Message

15 Low concentrations of NT-proBNP (<1000 pg/ml), ANGPT2 (<3.5 ng/ml) and BMP10 (<2
 16 ng/ml) identified patients with a high chance of attaining sinus rhythm during follow-up
 17 when added to a clinical risk score. Combining NT-proBNP with ANGPT2 and BMP10 is
 18 particularly useful in patients in AF at the time of blood sampling and in patients on rhythm
 19 control.



20

21