Supplementary Appendix

Prediction of short-term atrial fibrillation risk using primary care electronic health records

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Supplementary Introduction

Supplementary Table S1. Algorithms that have been derived and/or validated in community-based EHR for predicting AF

Algorithm	Study	Study	EHR cohort	Age eligibility	Discrimination	Follow-	Variable frequently missing in routinely-collected
8	Aim	÷	(country)	(years)	(c-statistic)	up	primary care EHR
Models origina	ally derive	ed for another pu	rpose but tested for		ident atrial fibrill	ation	
	EV	Chao 2013	NHIRD (TW)	≥18	0.713	10	
	EV	Saliba 2016	ClalitHS (IL)	≥50	0.728	3	
	EV	Li 2019	YMID (CN)	≥18	0.632	11	
CHADS ₂	EV	Li 2019	NHIS-HEALS (KR)	≥18	0.637	11	N/A
	EV	Kim 2020	NHIS-NSC (KR)	≥18	0.652	5	
	EV	Saliba 2016	ClalitHS (IL)	≥50	0.744	3	
	EV	Li 2019	YMID (CN)	≥18	0.687	11	
CHA ₂ DS ₂ -	EV	Li 2019	NHIS-HEALS (KR)	≥18	0.637	11	N/A
VASc	EV	Himmelreich 2020	Nivel-PCD (NL)	≥40	0.669	5	N/A
	EV	Kim 2020	NHIS-NSC (KR)	≥18	0.654	5	
	EV	Suenari 2017	NHIRD (TW)	≥20	0.716	9	
	EV	Li 2019	YMID (CN)	≥18	0.633	11	
НАТСН	EV	Li 2019	NHIS-HEALS (KR)	≥18	0.646	11	N/A
	EV	Kim 2020	NHIS-NSC (KR)	≥18	0.669	5	
	EV	Hu-WS 2020	NHIRD (TW)	≥18	0.771	14	
Machine Lear	ning mod	els					·
	D	Hill 2019	CPRD (UK)	≥30	0.827	11	Height, weight, BMI, SBP, DBP
Pfizer-AI	EV	Sekelj 2020	Discover (UK)	≥30	0.870	8	
NHIRD	D	Hu-WS 2019	NHIRD (TW)	≥18	0.948	14	Follow-up duration (years)
NHIS-NSC	D	Kim 2020	NHIS-NSC (KR)	≥18	0.845	5	BMI, SBP, Triglycerides, total cholesterol, HDL cholesterol, LDL cholesterol, eGFR, GGT, fasting blood glucose, Haemoglobin, AST, Socioeconomic status
Regression Mo	odels deriv	ved in electronic	health records				

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	D	Li 2019	YMID (CN)	>19	0.750	11	
	D	LI 2019		≥18	0.730	11	
	EV	Li 2019	NHIS-HEALS (KR)	≥18	0.654	11	
C ₂ HEST	EV	Hu-WS 2020	NHIRD (TW)	≥18	0.790	14	N/A
			DCRS, DNPR, DPR (DK)	65	0.588		
	EV	EV Lip 2020		70	0.594	5	
				75	0.593		
MHS	D	Aronson 2018	MHS (IL)	≥50	0.743	10	BMI, SBP
					0.857	1	
Taiwan AF	р	D Chao 2021	NHIRD (TW)	≥40	0.825	5	N/A
Taiwan AF	D				0.797	10	IN/A
					0.756	16	
InGef	D	Schnabel 2022	InGef (G)	≥45	0.829	1	N/A
Regression mo	Regression model derived in a prospective cohort design						
CHARGE-AF	EV	Hill 2019	CPRD (UK)	≥30	0.725	11	Height, weight, SBP, DBP

Heart

Algorithm	Study Aim	Study	EHR cohort (country)	Age eligibility (years)	Discrimination (c-statistic)	Follow- up	Variable frequently missing in routinely-collected primary care EHR
Models original	lly derive	ed for another pu	pose but tested for	prediction of inc	ident atrial fibrill	ation	
CHA ₂ DS ₂ - VASc	EV	Himmelreich 2020	Nivel-PCD (NL)	≥40	0.669	5	N/A
Machine Learn	Machine Learning models						
CPRD	D	Hill 2019	CPRD (UK)	≥30	0.827	11	Height, weight, BMI, SBP, DBP
CPRD	EV	Sekelj 2020	Discover (UK)	≥30	0.870	8	
Regression Mo	dels deri	ved in electronic h	ealth records				
		V Lip 2020	DCRS, DNPR,	65	0.588	5	N/A
C ₂ HEST	EV			70	0.594		
			DPR (DK)	75	0.593		
InGef	D	Schnabel 2022	InGef (G)	≥45	0.829	1	N/A
Regression mod	Regression model derived in a prospective cohort design						
CHARGE-AF	EV	Hill 2019	CPRD (UK)	≥30	0.725	11	Height, weight, SBP, DBP

AF, Atrial Fibrillation; CHADS₂, Congestive heart failure, Hypertension, Age >75, Diabetes mellitus, prior Stroke or transient ischemic attack [2 points]; CHA₂DS₂-VASc, Congestive heart failure, Hypertension, Age >75 [2 points], Stroke/transient ischemic attack/thromboembolism [2 points], Vascular disease, Age 65-74, Sex Category; CHARGE-AF, Cohorts for Heart and Aging Research in Genomic Epidemiology; C₂HEST, Coronary artery disease / Chronic obstructive pulmonary disease [1 point each], Hypertension, Elderly (Age \geq 75, 2 points), Systolic heart failure, Thyroid disease (hyperthyroidism); ClalitHS, Clalit Health Services; CPRD, Clinical Practice Research Datalink; D, derivation; DCRS, Danish Civil Registration system; DK, Denmark; DNPR, Danish National Patient Register; DPR, Danish Prescription Regster; EHR, electronic health record; EV, external validation; G, Germany; HATCH, Hypertension, Age, stroke or Transient ischemic attack, Chronic obstructive pulmonary disease, Heart failure; IL, Israel; KR, Republic of Korea; MHS, Maccabi Healthcare Services; NHIRD, National Health Insurance Research Database; NHIS-HEALS, National Health Insurance Service - Health screening Cohort; NHIS-NSC, National Health Insurance Service-based National Sample Cohort; Nivel-PCD, Netherlands Institute for Health Services Research Primary Care Database; NL, Netherlands; TW, Taiwan; UK, United Kingdom; YMID, Yunnan Medical Insurance Database.

Supplementary Methods

Supplementary Table S3. Read codes and ICD-10 codes used to define the outcomes of atrial fibrillation or atrial flutter

Code	Description
Readcodes	
G573200	Paroxysmal atrial fibrillation
G573400	Permanent atrial fibrillation
G573500	Persistent atrial fibrillation
3272	ECG: atrial fibrillation
G573000	Atrial fibrillation
G573300	Non-rheumatic atrial fibrillation
G573.00	Atrial fibrillation and flutter
G573z00	Atrial fibrillation and flutter NOS
3273	ECG: atrial flutter
G573100	Atrial flutter
ICD-10 codes	
I48	Atrial fibrillation and flutter

Training of the Random Forest classifier

Each decision tree used Gini impurity, commonly used in classification and regression tree (CART) algorithms, to measure the split quality.¹ The minimum impurity split threshold for each node, above which a node will split into two or more branches, was set to 10^{-7} . The minimum number of samples required to split a node was set to two. The minimum samples per leaf was set to one. All the algorithm's hyperparameters were tuned using the grid search method, in which all possible combinations were evaluated, resulting in 1000 trees, mtry = 8 (the number of random features to consider in each tree) and nodesize = 12 (number of patients classified at that node).

Supplementary Table S4. Baseline demographic and comorbidity variables used in algorithms tested for predicting incident AF in community-based electronic health records

Algorithm	Demographics	Comorbidities
CHADS ₂	Age	Hypertension, CHF, diabetes mellitus, CVA
CHA ₂ DS ₂ -VASc	Age, sex	Hypertension, CHF, stroke/TIA/thromboembolism, vascular disease
CHARGE-AF	Age, race, smoking status	Anti-hypertensive medication, MI, CHF, DM
C ₂ HEST	Age	Hypertension, ischaemic heart disease, CHF, COPD, thyroid disease
НАТСН	Age	Hypertension, CHF, stroke/TIA, COPD
InGef	Age, sex	Anti-hypertension medication, heart failure medication, chronic kidney disease, disorderd of lipoprotein metabolism and other lipidaemias, pulmonary heart diseases cardiac arrhythmias, other cerebrovascular disease, diverticular disease of intestine, dorsalgia, breathing abnormalities
MHS	Age, sex	Anti-hypertensive medication, MI, CHF, peripheral vascular disease, inflammatory disease in a female, COPD
NHIRD	Age (years), age group, sex	Hypertension, CHF, COPD, rheumatological disease, dyslipidaemia, DM, CVA or TIA, sleep disorder, cancer, hyperthyroidism, vascular disease, gout, CKD or ESRD, anaemia
NHIS-NSC*	Age, sex, smoking (pack- year), alcohol	Hypertension, CHF, MI, vascular disease, stroke/TIA, COPD
Pfizer-AI	Age, sex, race, smoking status	Hypertension, anti-hypertensive medication, CHF, congenital heart disease, MI, LVH, type 1 DM, type 2 DM
Taiwan AF	Age, sex, alcohol excess	Hypertension, CHF, IHD, ESRD

AF, Atrial Fibrillation; CHADS₂, Congestive heart failure, Hypertension, Age >75, Diabetes mellitus, prior Stroke or transient ischemic attack [2 points]; CHA₂DS₂-VASc, Congestive heart failure, Hypertension, Age >75 [2 points], Stroke/transient ischemic attack/thromboembolism [2 points]; CHARGE-AF, Cohorts for Heart and Aging Research in Genomic Epidemiology; C₂HEST, Coronary artery disease / Chronic obstructive pulmonary disease [1 point each], Hypertension, Elderly (Age ≥75, 2 points), Systolic heart failure, Thyroid disease (hyperthyroidism); CHF, chronic heart failure; CKD, chronic kidney disease; COPD, chronic obstructive pulmonary disease; CPRD, Clinical Practice Research Datalink; CVA, cerebrovascular accident; DM, diabetes mellitus; ESRD, end-stage renal disease; HATCH, Hypertension, Age, stroke or Transient ischemic attack, Chronic obstructive pulmonary disease, Heart failure; IHD, ischaemic heart disease; LVH, left ventricular hypertrophy; MHS, Maccabi Healthcare Services; MI, myocardial infarction; NHIRD, National Health Insurance Research Database; NHIS-HEALS, National Health Insurance Service - Health screening Cohort; NHIS-NSC, National Health Insurance Service-based National Sample Cohort; TIA, transient ischaemic attack. * In Kim 2020 prediction model development using machine learning was completed both with and without the predictor $PM_{2.5}$ - which is fine particular matter air pollution. In this analysis we have only included the model without $PM_{2.5}$ as it is judged not to be a predictor that would be routinely available in primary care or population EHR.

Supplementary Table S5. Candidate variables added after literature search with accompanying reference demonstrating association

Comorbidity associated with / predictive of atrial fibrillation	Categorisation	Reference demonstrating association with AF and rationale for categorisation
Cardiac surgery	Valvular,	Greenberg JW, Lancaster TS, Schuessler RB, et al.
	Non-valvular	Postoperative atrial fibrillation following cardiac surgery: a persistent complication. Eur J Cardiothorac Surg 2017;52(4):665-72.
		Within overall cardiac surgical procedures incidence of post-operative AF is 35%, isolated CABG has an incidence of 20—30% and isolated valve surgeries have an incidence of 35-40
Deep venous thrombosis	-	Lutsey P, Norby F, Alonso A, et al. Atrial fibrillation and venous thromboembolism: evidence of bidirectionality in the Atherosclerosis Risk in Communities Study. J Thromb Haemost 2018;16(4):670-79.
Infective Endocarditis	-	Ferrera C, Vilacosta I, Fernández C, et al. Usefulness of new-onset atrial fibrillation, as a strong predictor of heart failure and death in patients with native left- sided infective endocarditis. The American journal of cardiology 2016;117(3):427-33.
Electrophysiology procedure affecting the atria	-	Strickberger SA, Man KC, Daoud EG, et al. Adenosine-induced atrial arrhythmia: a prospective analysis. Ann Intern Med 1997;127(6):417-22. Khachab, H., and B. Brembilla-Perrot. "Prevalence of
		atrial fibrillation in patients with history of paroxysmal supraventricular tachycardia." International journal of cardiology 166.1 (2013): 221- 224.
Hypertrophic cardiomyopathy	-	Siontis KC, Geske JB, Ong K, et al. Atrial fibrillation in hypertrophic cardiomyopathy: prevalence, clinical correlations, and mortality in a large high-risk population. Journal of the American Heart Association 2014;3(3):e001002.
Inflammatory bowel disease	-	Boos CJ. Infection and atrial fibrillation: inflammation begets AF. Eur Heart J 2020
Intensive care unit admission	-	Klein Klouwenberg PM, Frencken JF, Kuipers S, et al. Incidence, predictors, and outcomes of new-onset atrial fibrillation in critically ill patients with sepsis. A cohort study. Am J Respir Crit Care Med 2017;195(2):205-11.
Infection	Gastrointestinal	Gundlund A, Olesen JB, Butt JH, et al. One-year
	Influenza	outcomes in atrial fibrillation presenting during
	Respiratory	infections: a nationwide registry-based study. Eur
	Sepsis	Heart J 2020;41(10):1112-19.
		Chang T-Y, Chao T-F, Liu C-J, et al. The association between influenza infection, vaccination, and atrial fibrillation: A nationwide case-control study. Heart Rhythm 2016;13(6):1189-94.
		Klein Klouwenberg PM, Frencken JF, Kuipers S, et al. Incidence, predictors, and outcomes of new-onset atrial fibrillation in critically ill patients with sepsis. A cohort study. Am J Respir Crit Care Med

		2017;195(2):205-11.
		In a cohort study among infections precipitating AF the order of risk is as follows: Pneumonia > sepsis > urinary tract infection > gastrointestinal infection
	Urinary	
Myocarditis	-	Wang Z, Wang Y, Lin H, et al. Early characteristics of fulminant myocarditis vs non-fulminant myocarditis: a meta-analysis. Medicine 2019;98(8)
Pulmonary embolus	-	Ptaszynska-Kopczynska K, Kiluk I, Sobkowicz B. Atrial fibrillation in patients with acute pulmonary embolism: clinical significance and impact on prognosis. BioMed research international 2019;2019
Pericarditis	-	Imazio M, Lazaros G, Picardi E, et al. Incidence and prognostic significance of new onset atrial fibrillation/flutter in acute pericarditis. Heart 2015;101(18):1463-67.
Pulmonary hypertension	-	Olsson KM, Nickel NP, Tongers J, et al. Atrial flutter and fibrillation in patients with pulmonary hypertension. Int J Cardiol 2013;167(5):2300-05.
Surgery (non-cardiac)	Colorectal	Siu CW, Tung HM, Chu KW, et al. Prevalence and
	Thoracic	predictors of new-onset atrial fibrillation after
	Vascular	elective surgery for colorectal cancer. Pacing Clin Electrophysiol 2005;28:S120-S23.
		Onaitis M, D'Amico T, Zhao Y, et al. Risk factors for
		atrial fibrillation after lung cancer surgery: analysis of
		the Society of Thoracic Surgeons general thoracic
		surgery database. The Annals of thoracic surgery 2010;90(2):368-74.
		Philip I, Berroëta C, Leblanc I. Perioperative challenges of atrial fibrillation. Current Opinion in Anesthesiology 2014;27(3):344-52.
		Thoracic surgery is associated with the greatest risk of post-operative AF amongst non-cardiac surgeries followed by colorectal then vascular surgery
Valvular heart disease	Mitral stenosis /	Iung B, Leenhardt A, Extramiana F. Management of
	rheumatic	atrial fibrillation in patients with rheumatic mitral
	valvular disease	stenosis. Heart 2018;104(13):1062-68.
	Non-mitral valve	
	/ other valves	Levy S. Factors predisposing to the development of
	Mitral regurgitation	atrial fibrillation. Pacing Clin Electrophysiol 1997;20(10):2670-74.
		Grigioni F, Avierinos J-F, Ling LH, et al. Atrial fibrillation complicating the course of degenerative mitral regurgitation: determinants and long-term outcome. J Am Coll Cardiol 2002;40(1):84-92.
		The association of mitral stenosis and rheumatic valve disease with AF is greater than mitral
		regurgitation followed by diseases of other valves
Vascular dementia	-	Ott A, Breteler MM, De Bruyne MC, et al. Atrial
		fibrillation and dementia in a population-based study:
Waight	Ohang	the Rotterdam Study. Stroke 1997;28(2):316-21.
Weight	Obese Overweight	Lavie CJ, Pandey A, Lau DH, et al. Obesity and atrial fibrillation prevalence, pathogenesis, and prognosis:
	over weight	normation provatence, pathogenesis, and prognosis.

Under-weig	ht effects of weight loss and exercise. J Am Coll Cardiol 2017;70(16):2022-35.
	Frost L, Hune LJ, Vestergaard P. Overweight and obesity as risk factors for atrial fibrillation or flutter: the Danish Diet, Cancer, and Health Study. The American journal of medicine 2005;118(5):489-95.
	Lee S-R, Choi E-K, Park CS, et al. Direct oral anticoagulants in patients with nonvalvular atrial fibrillation and low body weight. J Am Coll Cardiol 2019;73(8):919-31.
	Obesity is associated with a greater risk of AF than being overweight. Low body weight is associated with a higher risk of AF than normal weight.

Supplementary Table S6. Variable categorisations with rationale

Comorbidity	Categorisation	References and Rationale for categorisation
associated with /		
predictive of atrial		
fibrillation Demographics		
Age	-	Hindricks G, Potpara T, Dagres N, et al. 2020 ESC
8-		Guidelines for the diagnosis and management of atrial
		fibrillation developed in collaboration with the European
		Association of Cardio-Thoracic Surgery (EACTS). Eur
		Heart J 2020
		Incidence of AF increases with age (therefore included as
		a continuous variable)
Sex	Men	Hindricks G, Potpara T, Dagres N, et al. 2020 ESC
	Women	Guidelines for the diagnosis and management of atrial
		fibrillation developed in collaboration with the European
		Association of Cardio-Thoracic Surgery (EACTS). Eur
		Heart J 2020
		AF is more common in men
Ethnicity	Asian	Shen AY-J, Contreras R, Sobnosky S, et al. Racial/ethnic
	Black	differences in the prevalence of atrial fibrillation among
	Mixed	older adults-a cross-sectional study. J Natl Med Assoc
	Other	2010;102(10):906-14.
	Pacific Asian	
	White	Chiang C-E, Zhang S, Tse HF, et al. Atrial fibrillation
		management in Asia: from the Asian expert forum on atrial fibrillation. Int J Cardiol 2013;164(1):21-32.
		normation. Int J Cardiol 2015,104(1).21-52.
		White, Asian, pacific Asian, and black ethnicities have
		different odds ratios of development of AF
Alcohol use	Ex-	Samokhvalov AV, Irving HM, Rehm J. Alcohol
	Light,	consumption as a risk factor for atrial fibrillation: a
	Moderate	systematic review and meta-analysis. European Journal of Preventive Cardiology 2010;17(6):706-12.
	Excess Unspecified	Fleventive Caldiology 2010,17(0).700-12.
	Unspecified	There is a monotonic dose-response relationship between
		alcohol consumption and AF incidence
Smoking	Current	Heeringa J, Kors JA, Hofman A, et al. Cigarette smoking
	Ex	and risk of atrial fibrillation: the Rotterdam Study. Am Heart
		J 2008;156(6):1163-69.
		Watanabe I. Smoking and risk of atrial fibrillation: Elsevier,
		2018.
		Current and ex-smokers are at increased risk of AF, with
		a higher risk in current smokers.
Weight	Obese	See table S4
	Overweight]
Comarhidition	Under-weight	
Comorbidities Adult congenital	-	
heart disease	-	
Anaemia	-	-
Cancer	Leukaemia	Thompson PA, Lévy V, Tam CS, et al. Atrial fibrillation in
	Lymphoma	CLL patients treated with ibrutinib. An international
	Metastasis	retrospective study. Br J Haematol 2016;175(3):462-66.

	Skin cancers other than	Sorigue M. Guel Conflored E. Correis O. et al. Insider		
	melanoma Solid organ	Sorigue M, Gual-Capllonch F, Garcia O, et al. Incidence, predictive factors, management, and survival impact of atrial fibrillation in non-Hodgkin lymphoma. Ann Hematol 2018;97(9):1633-40.		
		Han H, Chen L, Lin Z, et al. Prevalence, trends, and outcomes of atrial fibrillation in hospitalized patients with metastatic cancer: findings from a national sample. Cancer medicine 2021;10(16):5661-70.		
		AF risk is higher in patients with leukaemia and lymphoma, especially treated with iritunib. Solid organ cancers (such as lung and colorectal cancer) are more likely to undergo surgery. Metastatic disease is associated with higher risk of AF compared to non-metastatic disease. Skin cancers other than melanoma have a lower risk of metastasis and hence AF.		
Cardiac surgery	Valvular,	See table S4		
Chaonia Iridures	Non-valvular Stage 1-2	Alonso A, Lopez FL, Matsushita K, et al. Chronic kidney		
Chronic kidney disease	Stage 3 Stage 4	disease is associated with the incidence of atrial fibrillation: the Atherosclerosis Risk in Communities (ARIC) study. Circulation 2011;123(25):2946-53.		
	Stage 5	Cheduaton 2011,125(25).25 to 55.		
	Unspecified	Risk of AF increases as CKD stage worsens and if there is		
	Other	proteinuria		
COPD	-	-		
Cerebro-vascular	Intracerebral	Hindricks G, Potpara T, Dagres N, et al. 2020 ESC		
accident	haemorrhage	Guidelines for the diagnosis and management of atrial		
	Subarachnoid	fibrillation developed in collaboration with the European		
	haemorrhage	Association of Cardio-Thoracic Surgery (EACTS). Eur		
	Unspecified	Heart J 2020		
		Association with AF is higher for ischaemic strokes than		
Dishatas Mallitas	Cood control	haemorrhagic strokes		
Diabetes Mellitus	Good control Poor control	Dublin S, Glazer NL, Smith NL, et al. Diabetes mellitus, glycemic control, and risk of atrial fibrillation. J Gen Intern		
	Unspecified / secondary	Med 2010;25(8):853-58.		
	Onspectfied / secondary	Neu 2010,25(0).055-50.		
		Poorer glycaemic control is associated with a higher risk of AF compared to better glycaemic control or no diabetes		
Deep venous	-	-		
thrombosis				
Dyslipidaemia	-	-		
Infective Endocarditis	-	-		
Electrophysiology	_			
procedure affecting	-	-		
the atria				
Gout	-	-		
Hypertrophic	-	-		
cardiomyopathy				
Heart failure	-	-		
Hypertension	Poor control	Dzeshka MS, Shantsila A, Shantsila E, et al. Atrial		
	Unspecified / secondary	fibrillation and hypertension. Hypertension 2017;70(5):854-		
	Unspecified / secondary	61.		

		associated with a higher risk of developing AF			
Hyperthyroidism	-	-			
Inflammatory bowel	-	-			
disease					
Intensive care unit	-	-			
admission					
Ischaemic heart	Chronic	Huxley RR, Lopez FL, Folsom AR, et al. Absolute and			
disease		attributable risks of atrial fibrillation in relation to optimal			
	Myocardial infarction	and borderline risk factors: the Atherosclerosis Risk in			
	Percutaneous coronary	Communities (ARIC) study. Circulation 2011;123(14):1501-			
	intervention	08.			
		Pizzetti F, Turazza F, Franzosi M, et al. Incidence and prognostic significance of atrial fibrillation in acute myocardial infarction: the GISSI-3 data. Heart 2001;86(5):527-32.			
		There is a high risk of AF in the acute setting of myocardial infarction as well as evidence in the context of			
		underlying chronic coronary syndromes.			
Infection	Gastrointestinal	See table S4			
	Influenza				
	Respiratory	1			
	Sepsis	1			
	Urinary				
Left ventricular	-	-			
hypertrophy					
Myocarditis	-	-			
Obstructive sleep	-	-			
apnoea					
Pulmonary embolus	-	-			
Pericarditis	-	-			
Pulmonary hypertension	-	-			
Peripheral vascular	-	_			
disease	-	-			
Rheumatological	Autoimmune connective	Lee E, Choi E-K, Jung J-H, et al. Increased risk of atrial			
condition	tissue diseases	fibrillation in patients with Behçet's disease: a nationwide			
condition	Rheumatoid arthritis	population-based study. Int J Cardiol 2019;292:106-11.			
	Spondyloarthropathies				
	Vasculitides	Moon I, Choi E-K, Jung J-H, et al. Ankylosing spondylitis: a novel risk factor for atrial fibrillation—a nationwide population-based study. Int J Cardiol 2019;275:77-82.			
		Melduni RM, Cooper LT, Gersh BJ, et al. Association of Autoimmune Vasculitis and Incident Atrial Fibrillation: A Population-Based Case-Control Study. Journal of the American Heart Association 2020;9(18):e015977.			
		Naaraayan A, Meredith A, Nimkar A, et al. Arrhythmia prevalence among patients with Polymyositis- Dermatomyositis in the United States: an observational study. Heart Rhythm 2021			
		Songnan W, Shengma C. GW24-e2483 Catheter ablation of atrial fibrillation in patients with autoimmune rheumatic diseases. Heart 2013;99(Suppl 3):A197-A97.			
		Giallafos I, Triposkiadis F, Oikonomou E, et al. Incident			

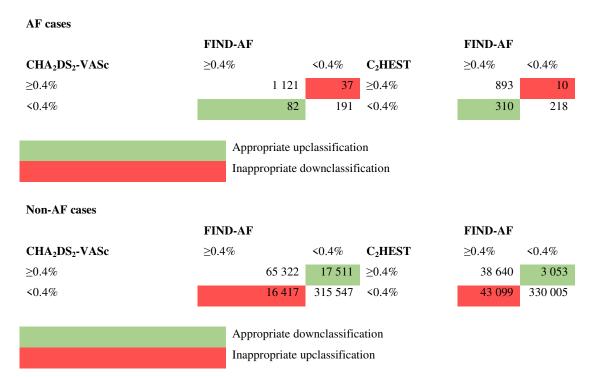
Surgery (non- cardiac) Colorectal See table S4 Thoracic Vascular Vascular - Valvular heart Mitral stenosis / rheumatic valvular disease See table S4 Non-mitral valve / other valves - Mitral regurgitation -	Smoking	Current	 atrial fibrillation in systemic sclerosis: the predictive role of B-type natriuretic peptide. Hellenic J Cardiol 2014;55:313- 21. Pugnet G, Gouya H, Puéchal X, et al. Cardiac involvement in granulomatosis with polyangiitis: a magnetic resonance imaging study of 31 consecutive patients. Rheumatology 2017;56(6):947-56. Lindhardsen J, Ahlehoff O, Gislason GH, et al. Risk of atrial fibrillation and stroke in rheumatoid arthritis: Danish nationwide cohort study. BMJ 2012;344 Each of the subtypes of rheumatological disease are associated with differing risks of development of AF. Here they have been categorised in clinical sub-type. See table S4
Surgery (non-cardiac) Colorectal See table S4 Thoracic Thoracic Vascular Vascular Systemic Embolism - Valvular heart Mitral stenosis / rheumatic valvular disease Non-mitral valve / other valves Non-mitral regurgitation	Shioking		
Vascular Systemic Embolism - Valvular heart Mitral stenosis / rheumatic valvular disease See table S4 Non-mitral valve / other valves Non-mitral valve / other valves			See table S4
Systemic Embolism - Valvular heart Mitral stenosis / rheumatic valvular disease See table S4 Mon-mitral valve / other valves Mitral regurgitation		Thoracic	
Valvular heart disease Mitral stenosis / rheumatic valvular disease See table S4 Non-mitral valve / other valves Non-mitral valve / other valves Mitral regurgitation		Vascular	
disease rheumatic valvular disease Non-mitral valve / other valves Mitral regurgitation	Systemic Embolism	-	-
valves Mitral regurgitation		rheumatic valvular disease	See table S4
		valves	
	Vascular dementia	-	-

Supplementary Results

Supplement Table S7. Baseline characteristics of training and testing datasets

	Training set	Testing set
	n (%)	n (%)
	1 664 911	416 228
Demographics		
Age, years	49.90 (15.43)	49.90 (15.42)
Sex (women)	844 083 (50.7)	211 478 (50.8)
Comorbidities		
Diabetes mellitus	58 513 (3.5)	14 268 (3.4)
Stroke or TIA	30 871 (1.9)	7 794 (1.9)
Ischaemic heart disease	62 980 (3.8)	15 622 (3.8)
Hypertension	200 217 (12.0)	50 106 (12.0)
Heart failure	11 577 (0.7)	2 790 (0.7)
Dyslipidaemia	48 719 (2.9)	12 170 (2.9)
Hyperthyroidism	13 069 (0.8)	3 233 (0.8)
COPD	20 294 (1.2)	5 129 (1.2)
Chronic kidney disease	23 794 (1.4)	6 014 (1.4)
Anaemia	53 962 (3.2)	13 383 (3.2)
Cancer	58 725 (3.5)	14 783 (3.6)
Valvular heart disease	7 946 (0.5)	1 927 (0.5)
Mean CHA ₂ DS ₂ -VASc score	0.98 (1.04)	0.98 (1.04)

Supplementary Table S8. Net reclassification using FIND-AF



Net reclassification indices

Index	CHA ₂ DS ₂ -VASc	C ₂ HEST
Case reclassification (NRI+ [95%	0.031 (0.026-0.048)	0.021 (0.19-0.23)
CI])		
Non-case reclassification (NRI-	0.0026 (0.0015-0.0032)	-0.096 (-0.0980.095)
[95% CI])		
Net reclassification (NRI [95% CI])	0.032 (0.029-0.051)	0.113 (0.098-0.135)

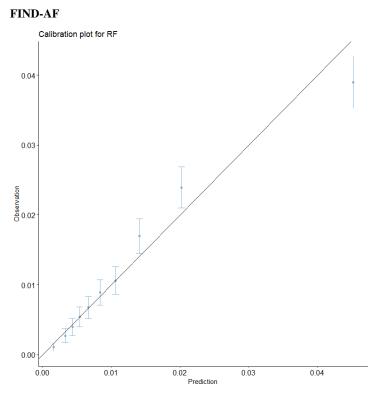
Supplementary Table S9. Baseline characteristics of testing set, stratified by incident AF and predicted AF risk

	Incident atrial fibrillation		FIND-AF predicted risk	
	no AF	AF	Lower risk	Higher risk
	n (%)	n (%)	n (%)	n (%)
	414 676	1 552	333 286	82 942
Demographics				
Age, years	49.82 (15.38)	73.87 (12.47)	44.11 (10.40)	73.24 (8.75)
Sex (women)	210 646 (50.8)	755 (48.6)	170 568 (51.2)	41 210 (49.7)
Ethnicity				
Asian	8 258 (2.0)	21 (1.5)	7 385 (2.2)	894 (1.1)
Black	6 390 (1.5)	9 (0.6)	5 786 (1.7)	613 (0.7)
Other	27 805 (6.7)	106 (7.4)	22 033 (6.6)	5 878 (7.1)
Unknown	93 630 (22.6)	36 (2.5)	91 505 (27.5)	2 161 (2.6)
White	278 714 (67.2)	1 259 (88.0)	206 577 (62.0)	73 396 (88.5)
Comorbidities				
Diabetes mellitus	14 649 (3.5)	171 (11.0)	6 328 (1.9)	8 072 (9.7)
Stroke or TIA	7 467 (1.8)	189 (12.2)	1 376 (0.4)	6 375 (7.7)
Ischaemic heart disease	15 483 (3.7)	314 (20.2)	3 299 (1.0)	12 486 (15.1)
Hypertension	49 494 (11.9)	621 (40.0)	20 139 (6.0)	29 594 (35.7)
Heart failure	2 745 (0.7)	132 (8.5)	163 (0.0)	2 748 (3.3)
Dyslipidaemia	12 122 (2.9)	121 (7.8)	6 095 (1.8)	5 984 (7.2)
Hyperthyroidism	3 203 (0.8)	44 (2.8)	1 883 (0.6)	1 370 (1.7)
COPD	4 987 (1.2)	106 (6.8)	1 111 (0.3)	4 019 (4.8)
Chronic kidney disease	5 839 (1.4)	99 (6.4)	2 938 (0.9)	2 990 (3.6)

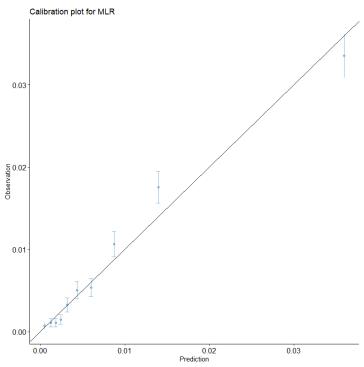
Anaemia	13 165 (3.2)	106 (6.8)	9118 (2.7)	4251 (5.1)
Cancer	14 710 (3.5)	186 (12.0)	6120 (1.8)	8303 (10.0)
Valvular heart disease	1 881 (0.5)	84 (5.4)	562 (0.2)	1414 (1.7)
Mean CHA ₂ DS ₂ -VASc score (SD)	0.97 (1.03)	2.74 (1.40)	0.62 (0.62)	2.42 (1.14)

AF, atrial fibrillation; CHA₂DS₂-VASc, Congestive heart failure, Hypertension, Age >75 years [2 points], Stroke/transient ischemic attack/thromboembolism [2 points], Vascular disease, Age 65-74 years, Sex Category; COPD, chronic obstructive pulmonary disease; TIA, transient ischaemic attack

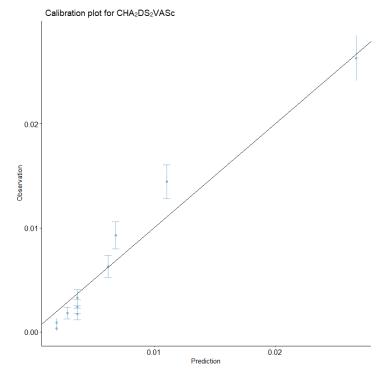
Supplement Figure S1. Calibration plots



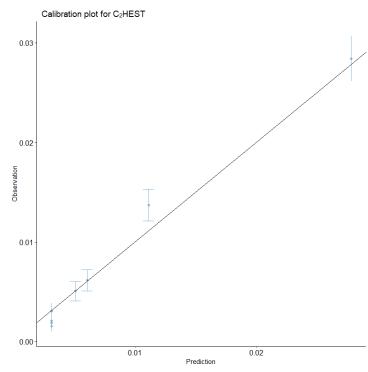
Multivariable logistic regression



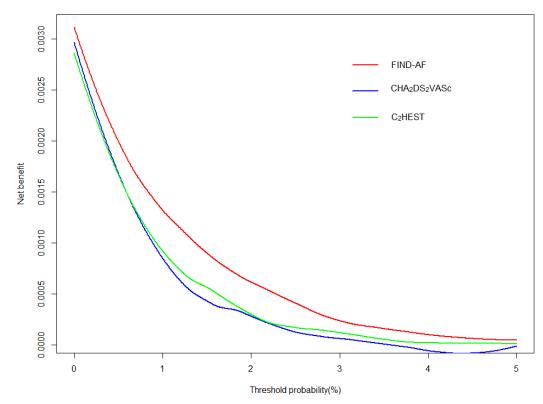




C₂HEST



Supplementary Figure S2. Decision curve analysis for FIND-AF versus CHA₂DS₂-VASc and C₂HEST



Decision curve to predict incident AF

References

1. Raileanu LE, Stoffel K. Theoretical comparison between the gini index and information gain criteria. Annals of Mathematics and Artificial Intelligence 2004;41(1):77-93.