

Prediction Score for Anticoagulation Control Quality Among Older Adults

Kueiyu Joshua Lin, MD, ScD, MPH; Daniel E. Singer, MD; Robert J. Glynn, PhD, ScD; Suzanne Blackley, MA; Li Zhou, MD, PhD; Jun Liu, MD; Gina Dube, PharmD, CACP, RPh; Lynn B. Oertel, MS, ANP-BC, CACP; Sebastian Schneeweiss, MD, ScD

Background—Time in the therapeutic range (TTR) is associated with the effectiveness and safety of vitamin K antagonist (VKA) therapy. To optimize prescribing of VKA, we aimed to develop and validate a prediction model for TTR in older adults taking VKA for nonvalvular atrial fibrillation and venous thromboembolism.

Methods and Results—The study cohort comprised patients aged ≥ 65 years who were taking VKA for atrial fibrillation or venous thromboembolism and who were identified in the 2 US electronic health record databases linked with Medicare claims data from 2007 through 2014. With the predictors identified from a systematic review and clinical knowledge, we built a prediction model for TTR, using one electronic health record system as the training set and the other as the validation set. We compared the performance of the new models to that of a published prediction score for TTR, SAME-TT₂R₂. Based on 1663 patients in the training set and 1181 in the validation set, our optimized score included 42 variables and the simplified model included 7 variables, abbreviated as PROSPER (Pneumonia, Renal dysfunction, Oozing blood [prior bleeding], Staying in hospital ≥ 7 days, Pain medication use, no Enhanced [structured] anticoagulation services, Rx for antibiotics). The PROSPER score outperformed SAME-TT₂R₂ when predicting both TTR $\geq 70\%$ (area under the receiver operating characteristic curve 0.67 versus 0.55) and the thromboembolic and bleeding outcomes (area under the receiver operating characteristic curve 0.62 versus 0.52).

Conclusions—Our geriatric TTR score can be used as a clinical decision aid to select appropriate candidates to receive VKA therapy and as a research tool to address confounding and treatment effect heterogeneity by anticoagulation quality. (*J Am Heart Assoc.* 2017;6:e006814. DOI: 10.1161/JAHA.117.006814.)

Key Words: anticoagulant • atrial fibrillation • quality control • stroke • venous thromboembolism

Vitamin K antagonist (VKA; eg, warfarin) therapy is an effective anticoagulation option for stroke prevention in patients with nonvalvular atrial fibrillation (AF) and for treatment and secondary prevention of venous thromboembolism (VTE; including deep vein thrombosis and pulmonary embolism).^{1–3} The safety and effectiveness of VKAs, however, depends on regular international normalized ratio (INR) monitoring and anticoagulation control quality, often

measured by the time in therapeutic range (TTR), for which INR 2.0 to 3.0 is the standard therapeutic range for AF and VTE.^{4–6} Patients on VKA with poor anticoagulation quality (ie, low TTR) have been shown to have a higher risk of thromboembolic and bleeding complications and thus a worse risk–benefit ratio.^{4,7,8}

Although clinical trials have shown that direct-acting oral anticoagulants (DOACs) are therapeutically advantageous over

From the Division of Pharmacoepidemiology and Pharmacoeconomics, Department of Medicine, Brigham and Women's Hospital, Harvard Medical School, Boston, MA (K.J.L., R.J.G., J.L., S.S.); Department of Medicine, Massachusetts General Hospital, Harvard Medical School, Boston, MA (K.J.L., D.E.S.); Department of Epidemiology, Harvard T.H. Chan School of Public Health, Boston, MA (K.J.L., D.E.S., R.J.G., S.S.); Department of Medicine, Brigham and Women's Hospital, Harvard Medical School, Boston, MA (L.Z.); Clinical and Quality Analysis, Information Systems, Partners HealthCare System, Boston, MA (S.B.); Clinical Informatics, Partners eCare, Partners HealthCare System, Boston, MA (L.Z.); Department of Pharmacy, Brigham and Women's Hospital, Harvard Medical School, Boston, MA (G.D.); Anticoagulation Management Service, Department of Nursing, Massachusetts General Hospital, Boston, MA (L.B.O.).

Accompanying Data S1, Tables S1 through S7 and Figures S1, S2 are available at <http://jaha.ahajournals.org/content/6/10/e006814/DC1/embed/inline-supplementary-material-1.pdf>

Correspondence to: Kueiyu Joshua Lin, MD, ScD, MPH, Division of Pharmacoepidemiology and Pharmacoeconomics, Department of Medicine, Brigham and Women's Hospital, Harvard Medical School, 1620 Tremont St. Suite 3030, Boston, MA 02120. E-mail: jklin@bwh.harvard.edu

Received May 31, 2017; accepted August 23, 2017.

© 2017 The Authors. Published on behalf of the American Heart Association, Inc., by Wiley. This is an open access article under the terms of the Creative Commons Attribution-NonCommercial-NoDerivs License, which permits use and distribution in any medium, provided the original work is properly cited, the use is non-commercial and no modifications or adaptations are made.

Clinical Perspective

What Is New?

- In patients aged ≥ 65 years, our prediction model for anticoagulation control quality outperformed the published score, SAME-TT₂R₂.
- Time in the therapeutic range was $\geq 70\%$ (area under receiver operating characteristic curve 0.71 versus 0.57, a significant difference.).

What Are the Clinical Implications?

- Our prediction score for anticoagulation quality can help clinicians select the appropriate older adult candidates to receive vitamin K antagonist therapy and can provide researchers with a tool to adjust for confounding and to investigate treatment effect heterogeneity due to predicted anticoagulation quality.

or at least noninferior to VKAs,^{9–11} clinical equipoise still exists when patients are likely to have good anticoagulation control based on pretreatment characteristics.^{7,12} This choice is particularly difficult to make in older adults because DOACs have been associated with a higher risk of major gastrointestinal bleeding than VKAs in the older population.^{13–15} Moreover, chronic kidney disease is highly prevalent in older adults,¹⁶ which makes lack of routine monitoring tests for DOACs a challenge rather than an advantage because some DOACs are substantially renally excreted (eg, 80% for dabigatran). Consequently, it is critical to understand how patient characteristics are associated with anticoagulation quality so we can identify the ideal candidates for VKA therapy.

In the existing literature, there is only 1 published prediction score for anticoagulation quality: the SAME-TT₂R₂ score.¹⁷ It did not consider some clinically important predictors for TTR (eg, polypharmacy, hospitalizations, antibiotic use)^{18–22} and was found to have suboptimal performance in external validation populations (area under receiver operating characteristic curve [AUC] for relevant clinical end points < 0.6).^{23–25} In addition, although the majority of oral anticoagulant users are older adults,^{3,18} SAME-TT₂R₂ was developed with 52.7% of the population aged < 70 years. Because comorbidity profiles vary substantially by age, the generalizability and applicability of SAME-TT₂R₂ in the older population is unclear.

We aimed to develop and validate a new prediction model for TTR, particularly in patients aged ≥ 65 years taking VKA for nonvalvular AF or VTE. Because prior studies found that the TTR predictors identified in AF patients were similar to those in VTE patients,¹⁸ for clinical simplicity we developed 1 score for both indications but validated the performance in patients with nonvalvular AF and VTE separately.

Methods

Data Source

We linked electronic health record (EHR) data from 2 large US academic provider networks with Medicare claims data. The first network consists of 1 tertiary hospital, 2 community hospitals, 17 primary care centers, and 1 anticoagulation clinic that manages VKA-related care for all patients within the network. The second network includes 1 tertiary hospital, 1 community hospital, 16 primary care centers, and an anticoagulation clinic. Patients in network 1 were used as the training set for the prediction model derivation, and those in network 2 were used as the validation set. The EHR database contains information on patient demographics, diagnosis and procedure codes, medications, lifestyle factors, laboratory data, and various clinical notes. Both inpatient and outpatient EHR data were used in this study. The Medicare claims data contain information on demographics, inpatient and outpatient diagnosis and procedure codes, and dispensed medications.²⁶ This study was approved by Partners Health-Care Institutional Review Board (IRB).

Study Population

In the linked Medicare claims–EHR data, we identified all patients aged ≥ 65 years with nonvalvular AF or VTE initiating a VKA from January 1, 2007, to December 31, 2014, with no use of any oral anticoagulants (VKAs or DOACs) in the prior 90 days (new user design²⁷). The VKA initiation date was the index (cohort entry) date. The study cohort was required to have at least 180 days of continuous enrollment in Medicare inpatient, outpatient, and prescription benefits with at least 1 EHR encounter with date of service after January 1, 2007, and before the index date. To ensure our ability to assess the primary outcome reliably, patients were required to have at least 5 INR values recorded in the system. To assess whether this requirement would select an unrepresentative cohort, we compared the distributions of the combined comorbidity score²⁸ in those with versus without at least 5 INRs. We computed standardized differences between proportions of each combined comorbidity score category in those with versus without 5 INRs. A standardized difference of < 0.1 was used to indicate an acceptable discrepancy.²⁹

Outcome Definition of the Anticoagulation Control

We calculated TTR using Rosendaal's method,³⁰ which assigns an INR value to each day by linear interpolation of successive observed INR values with gaps < 56 days. After interpolation, we computed the proportions of time that fell

within the therapeutic range (INR 2.0–3.0). We ascertained TTR starting the 29th day after the index date until the earliest of the following: 12 months after the index date, lack of INRs with a gap ≤ 56 days, death, discontinuation of VKA, or study end (December 31, 2014). We did not assess TTR for the first month because variability of INR values in the first month generally reflects expected fluctuations in INRs during the titration phase of VKA therapy. VKA discontinuation was defined based on an algorithm validated in a prior study in which high agreement with actual VKA use was demonstrated by chart review ($\kappa=0.84$).³¹

Candidate Predictors and Building of the TTR Prediction Model

We conducted a systematic review to identify original articles reporting predictors of anticoagulation control quality (assessed by TTR or INR variability) in users of VKAs, after

multivariate adjustment. Figure 1 summarizes the search terms and the selection process. The significant predictors reported by the selected articles, along with variables deemed clinically important to predict anticoagulation quality, were used as the candidate predictors to build our prediction model. Based on these variables, we built a model predicting continuous TTR by Lasso regression with 5-fold cross-validation, using the data in the training set.³² We referred to the predicted TTR derived from this model as the *geriatric TTR score*. To build a simplified model for clinical use, we excluded biophysiologic variables requiring additional testing and used Lasso regression with a Bayesian information criterion, which tends to generate a more parsimonious model than do other criteria.³³ The points of the scoring system were the nearest integer proportional to the unstandardized coefficient in this simplified model. All predictors were assessed in the 180 days before (and including) the initiation of a VKA, with the exception of receiving structured anticoagulation management service, which was assessed until

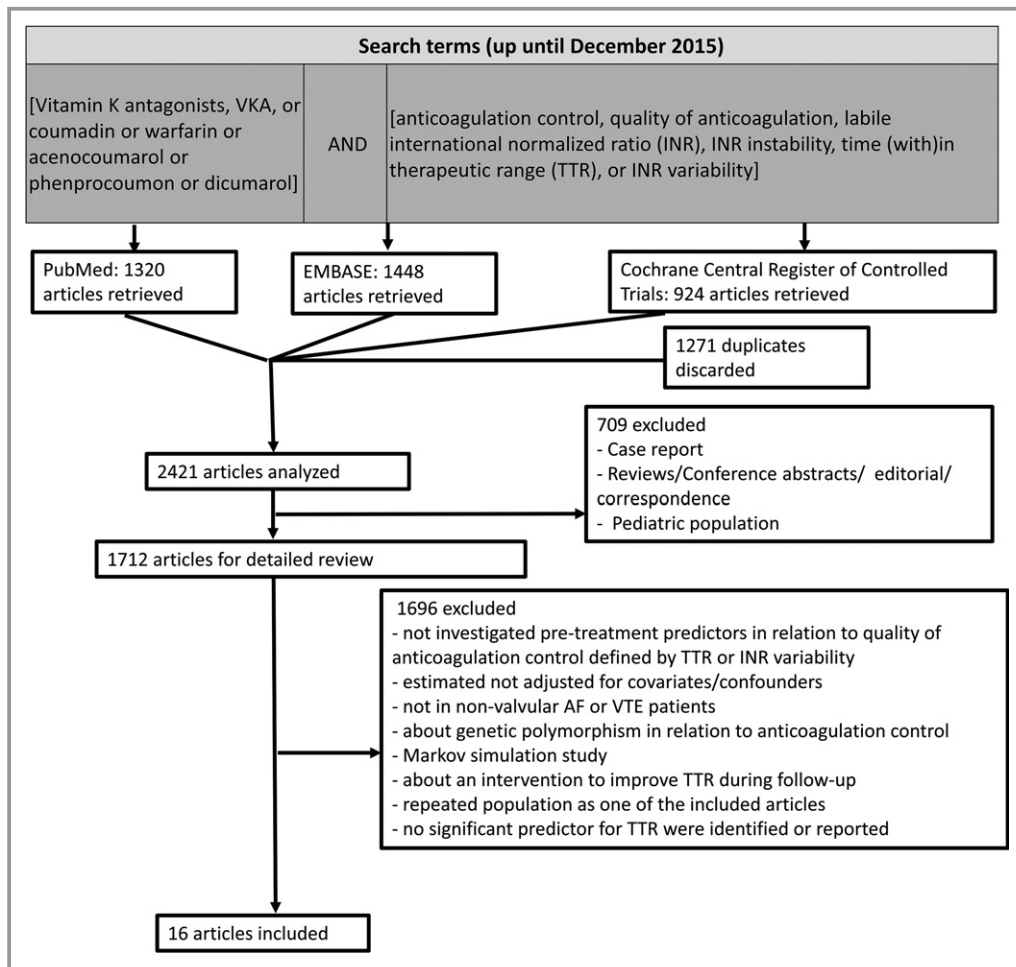


Figure 1. Systematic review on significant predictors for anticoagulation quality (TTR). AF indicates atrial fibrillation; INR, international normalized ratio; TTR, time in therapeutic range; VKA, vitamin K antagonist; VTE, venous thromboembolism.

Table 1. Significant Predictors for Anticoagulation Control Quality From a Systematic Review

First Author	Study Period	Sample Size	Region	Mean Age, y	Female, %	Indication	Significant Positive Predictors	Significant Negative Predictors
Boulanger ³⁹	1998–2003	13709	USA	67	43	NVAF	Male sex	CHF, DM, residing in Northeast vs Midwest or West
Apostolakis ¹⁷	1995–2001	1061	North America	69	41	NVAF	Use of β blocker, verapamil vs amiodarone, age >50 y, male sex	Ethnic minority, smoking (within 2 y), comorbidities defined as >2 of the following: hypertension, DM, coronary artery disease/myocardial infarction, PVD, CHF, previous stroke, pulmonary disease, and hepatic or renal disease
Tomita ⁴⁰	2011–2012	163	Japan	74.4	38	NVAF	Male sex	CHF
Dlott ²⁰	2007–2008	138319	USA	74	49	NVAF	Male sex, age 55–84.9 vs ≥ 85 y	Length of INR testing period, age <55 vs ≥ 85 y, physicians with lower case load, lower median income range, geographic region in the United States
Kim ²²	2006–2008	62156	USA	72.2–73.4	...	NVAF	...	CHF, AST >80 U/L, Alkaline phosphatase >150 U/L, sodium <140 mEq/L use of metolazone, and hospitalization for CHF
Kose ⁴¹	2011–2013	55	Japan	67.8	25	NVAF	...	CHF
Macedo ¹⁸	2000–2013	140078	UK	73.5	44	NVAF	Lipid-lowering drugs, older age	Low/normal BMI (<25), smoking, having acute respiratory infections, chronic lung disease (COPD, asthma), DM, epilepsy; use of pain medications, and number of hospitalizations
Nelson ⁴²	2006–2010	9433	USA	72.6	...	NVAF	Male sex, age >75 y, hypertension	CHF and DM
Pignatelli ⁴³	2008–2013	553	Italy	72.9	40	NVAF	Use of ACEI/ARB	DM
White ⁴⁴	2009–2013	290	USA	70–72	44	NVAF	Older age, male sex	...
Yong ⁴⁵	2003–2012	184161	USA	...	100	NVAF	...	Nonwhite race
Kooistra ⁴⁸	2007–2011	3825	Multination	...	46	VTE	...	Underweight, active cancer at baseline, secondary VTE, INR <2.0 at stop of bridging therapy

Continued

Table 1. Continued

First Author	Study Period	Sample Size	Region	Mean Age, y	Female, %	Indication	Significant Positive Predictors	Significant Negative Predictors
Macedo ¹⁸	2000–2013	70371	UK	65.4	52	VTE	Older age, male sex	Lower BMI, smoking, cancer, chronic use of pain medication, chronic lung disease (COPD or asthma), dementia, DM, epilepsy
Rose ⁴⁶	2006–2008	124619	USA	...	3 (VA population)	Mixed	For TTR in the first 6 mo: hyperlipidemia, older age; for TTR after the first 6 mo: hyperlipidemia, hypertension, older age, male sex	For TTR in the first 6 mo: race, living in a poor area, driving distance (weak association), cancer, DM, CKD, CAD, alcohol abuse, bipolar disorder, substance abuse, dementia, polypharmacy, number of hospitalizations; for TTR after the first 6 mo: nonwhite race, living in a poor area, duration on VKA, cancer, liver disease, epilepsy, CKD, DM, chronic lung disease, CAD, PVD, and heart failure, alcohol abuse, dementia, substance abuse, major depression, and bipolar disorder, number of concomitant medication use, number of hospitalizations
Efird ²¹	2007–2008	1763	USA	65.7–70.7	1.6 (VA population)	Mixed	...	Chronic liver disease, increased levels of AST and creatinine, lower levels of albumin
Nilsson ⁴⁷	1996–2012	2068	Denmark	Female (49.4); male (55.2)	34	Mixed	Male sex	
Paradise ¹⁹	2007–2008	28216	USA	...	3 (VA population)	Mixed	...	Bipolar, depression, psychotic disorders

ACEI indicates angiotensin-converting enzyme inhibitor; ARB, angiotensin II receptor blocker; AST, aspartate aminotransferase; BMI, body mass index; CAD, coronary artery disease; CHF, congestive heart failure; CKD, chronic kidney disease; COPD, chronic obstructive pulmonary disease; DM, diabetes mellitus; INR, international normalized ratio; NVAf, nonvalvular atrial fibrillation; PVD, peripheral vascular disease; TTR, time in therapeutic range; VA, Veterans Administration; VKA, vitamin K antagonist; VTE, venous thromboembolism.

Table 2. Patient Characteristics of the Study Populations

Continuous variables	Training Set (n=1663), mean (SD)	AF Validation Set (n=694), mean (SD)	VTE Validation Set (n=487), mean (SD)
Age, y	77.0 (7.5)	76.5 (7.2)	75.6 (7.1)
Anticoagulation management service participation, %*	0.41 (0.49)	0.55 (0.50)	0.48 (0.50)
TTR [†] , INR 2–3	0.56 (0.25)	0.53 (0.27)	0.47 (0.28)
Percentage of time below range	0.32 (0.26)	0.38 (0.30)	0.43 (0.32)
Percentage of time above range	0.12 (0.14)	0.09 (0.12)	0.09 (0.13)
Mean follow-up time to assess INR, d	179.3 (121.7)	174.5 (131.0)	136.3 (117.9)
Categorical variables	Training Set (n=1663), n (%)	AF Validation Set (n=694), n (%)	VTE Validation Set (n=487), n (%)
Female sex	829 (49.9)	322 (46.4)	282 (57.9)
Race			
Black	45 (2.7)	44 (6.3)	54 (11.1)
White	1519 (91.3)	608 (87.6)	398 (81.7)
Other	99 (6.0)	42 (6.1)	35 (7.2)
Limited English proficiency	157 (9.4)	80 (11.5)	58 (11.9)
Patients with higher education[‡]			
Above median	758 (45.6)	341 (49.1)	231 (47.4)
Below median	879 (52.9)	341 (49.1)	248 (50.9)
Missing	26 (1.6)	12 (1.7)	8 (1.6)
Income level[‡]			
Above median	940 (56.5)	389 (56.1)	262 (53.8)
Below median	697 (41.9)	293 (42.2)	217 (44.6)
Missing	26 (1.6)	12 (1.7)	8 (1.6)
Distance from the nearest provider facility, miles[†]			
<5	610 (36.7)	351 (50.6)	234 (48.1)
5–10	455 (27.4)	102 (14.7)	64 (13.1)
10–20	295 (17.7)	78 (11.2)	58 (11.9)
>20	303 (18.2)	163 (23.5)	131 (26.9)
BMI			
<18.5	24 (1.4)	5 (0.7)	4 (0.8)
18.5–24.9	219 (13.2)	114 (16.4)	74 (15.2)
25–29.9	379 (22.8)	180 (25.9)	118 (24.2)
30–34.9	277 (16.7)	110 (15.9)	77 (15.8)
35–39.9	102 (6.1)	54 (7.8)	44 (9.0)
≥40	96 (5.8)	36 (5.2)	16 (3.3)
Missing	566 (34.0)	195 (28.1)	154 (31.6)
Smoking status			
Current	247 (14.9)	79 (11.4)	58 (11.9)
Not current	1291 (77.6)	555 (80.0)	394 (80.9)
Missing	125 (7.5)	60 (8.7)	35 (7.2)
CHF	427 (25.7)	224 (32.3)	155 (31.8)
Epilepsy	84 (5.1)	26 (3.7)	31 (6.4)
Cancer	628 (37.8)	281 (40.5)	241 (49.5)

Continued

Table 2. Continued

Categorical variables	Training Set (n=1663), n (%)	AF Validation Set (n=694), n (%)	VTE Validation Set (n=487), n (%)
Renal dysfunction	579 (34.8)	263 (37.9)	227 (46.6)
Prior bleeding [§]	364 (21.9)	166 (23.9)	148 (30.4)
Pneumonia	365 (21.9)	125 (18.0)	135 (27.7)
Drug abuse	18 (1.1)	4 (0.6)	3 (0.6)
Chronic liver disease	143 (8.6)	49 (7.1)	65 (13.3)
Psychosis	104 (6.3)	37 (5.3)	39 (8.0)
Hyperlipidemia	1131 (68.0)	488 (70.3)	320 (65.7)
Peripheral vascular disease	283 (17.0)	109 (15.7)	77 (15.8)
Use of β blocker	1071 (64.4)	555 (80.0)	318 (65.3)
Use of ACEI	619 (37.2)	248 (35.7)	168 (34.5)
Use of metolazone	7 (0.4)	25 (3.6)	19 (3.9)
Use of opioids	665 (40.0)	307 (44.2)	298 (61.2)
Use of statins	1019 (61.3)	466 (67.1)	289 (59.3)
Use of acetaminophen	658 (39.6)	209 (30.1)	225 (46.2)
Use of antibiotics	915 (55.0)	402 (57.9)	345 (70.8)
Use of antiplatelet agents	506 (30.4)	251 (36.2)	186 (38.2)
Use of oral steroids	265 (15.9)	104 (15.0)	113 (23.2)
Influenza vaccine	534 (32.1)	227 (32.7)	154 (31.6)
PSA test	256 (15.4)	121 (17.4)	68 (14.0)
Mammography	136 (8.2)	44 (6.3)	43 (8.8)
Pap smear	41 (2.5)	17 (2.4)	16 (3.3)
Falls	171 (10.3)	53 (7.6)	63 (12.9)
Fractures	185 (11.1)	52 (7.5)	71 (14.6)
Parkinson disease	30 (1.8)	18 (2.6)	9 (1.8)
Albumin level, g/dL			
≥ 3.5	775 (46.6)	303 (43.7)	201 (41.3)
2.5–3.49	334 (20.1)	111 (16.0)	123 (25.3)
<2.5	53 (3.2)	17 (2.5)	25 (5.1)
Missing	501 (30.1)	263 (37.9)	138 (28.3)
ALP level, U/L			
≤ 150	1064 (64.0)	402 (57.9)	315 (64.7)
>150	73 (4.4)	25 (3.6)	26 (5.3)
Missing	526 (31.6)	267 (38.5)	146 (30.0)
Sodium level, mmol/L			
>130	1387 (83.4)	529 (76.2)	391 (80.3)
≤ 130	18 (1.1)	12 (1.7)	6 (1.2)
Missing	258 (15.5)	153 (22.1)	90 (18.5)
eGFR, mL/min/1.73m ²			
≥ 60	736 (44.3)	285 (41.1)	225 (46.2)
30–59.9	447 (26.9)	165 (23.8)	105 (21.6)
15–29.9	49 (3.0)	32 (4.6)	18 (3.7)

Continued

Table 2. Continued

Categorical variables	Training Set (n=1663), n (%)	AF Validation Set (n=694), n (%)	VTE Validation Set (n=487), n (%)
<15	92 (5.5)	51 (7.4)	44 (9.0)
Missing	339 (20.4)	161 (23.2)	95 (19.5)
Hospitalization length of stay			
None	571 (34.3)	245 (35.3)	100 (20.5)
1–6 d	497 (29.9)	235 (33.9)	152 (31.2)
≥7 d	595 (35.8)	214 (30.8)	235 (48.3)
Number of hospitalizations			
0	571 (34.3)	245 (35.3)	100 (20.5)
1	621 (37.3)	261 (37.6)	179 (36.8)
≥2	471 (28.3)	188 (27.1)	208 (42.7)
Number of medications [¶]			
<5	529 (31.8)	222 (32.0)	154 (31.6)
5–9	856 (51.5)	338 (48.7)	212 (43.5)
≥10	278 (16.7)	134 (19.3)	121 (24.9)

ACEI indicates angiotensin-converting enzyme inhibitor; AF, atrial fibrillation; ALP, alkaline phosphatase; BMI, body mass index; CHF, congestive heart failure; eGFR, estimated glomerular filtration rate; INR, international normalized ratio; PSA, prostate-specific antigen; TTR, time in therapeutic range; VKA, vitamin K antagonist; VTE, venous thromboembolism.

*Percentage of patients participating in a dedicated anticoagulation management service.

†See the distribution of TTR in training and validation sets in Figure S2.

‡Based on ZIP codes of residence.

§Including all the major upper and lower gastrointestinal, and other extracranial bleeding events.

||In the 180 days before VKA initiation.

¶At the time of VKA initiation.

28 days after VKA initiation (immediately before the start of the follow-up).

Performance of the Geriatric TTR Score Versus SAME-TT₂R₂

We calculated a coefficient-based and simplified version of the SAME-TT₂R₂ score for each patient (see Table S1 for details).¹⁷ Model performance was compared (1) between the geriatric TTR score and the coefficient-based SAME-TT₂R₂ score and (2) between the simplified point system of the geriatric TTR score and the simple SAME-TT₂R₂ score. The validation set was subdivided into AF and VTE populations. We computed the AUC when predicting TTR >70%, a cutoff to indicate good anticoagulation quality in the literature.^{34,35} We then computed the AUC when predicting incidence of a composite clinical outcome of stroke, systemic arterial embolism, VTE, and major bleeding (see detailed definitions in Table S2) that occurred between the 29th and 365th days following the index date (ie, the same ascertainment period as TTR). We also evaluated thromboembolic and bleeding events separately. In addition, we computed Hosmer–Lemeshow goodness-of-fit statistics to assess calibration of the models. The hypothesis testing for AUC comparison was done with methods proposed by DeLong et al.³⁶

Missing Data

The information on smoking and body mass index was recorded in the study EHRs as both structured data and text in the clinical notes. To reduce missing data, we used natural language processing³⁷ to extract information on these 2 variables from the clinical notes; this approach reduced the proportion of patients missing smoking data from 54.4% to 7.8% and of those missing body mass index from 38.5% to 32.2% (see Data S1 for details). For those still missing smoking and body mass index information after natural language processing and with other variables with missing data, we used the missing indicator method in the analysis.

Statistical Analyses

First, we tested the sensitivity of our results to the length of the baseline assessment period (365 instead of 180 days) and the definition of the new initiator of VKA (no use of VKA in the 180 days instead of 90 days before the index date). We calculated the Spearman correlation coefficient between the new scores based on revised strategies and the original score to quantify discrepancies. Second, to evaluate whether our results were sensitive to outliers or

Table 3. The Geriatric TTR Prediction Model for Anticoagulation Control Quality*

Predictor	Coefficient (SE) [†]
Intercept	0.583 (0.030)
AF vs VTE	0.010 (0.013)
Dedicated anticoagulation management service: yes vs no	0.105 (0.014)
Sex, female vs male	-0.016 (0.014)
Black race	-0.046 (0.036)
Nonblack, nonwhite race	0.036 (0.026)
White	Ref
Limited English proficiency	-0.033 (0.021)
Income: below median [‡]	-0.009 (0.014)
Income: missing [‡]	0.018 (0.047)
Income: median or higher [‡]	Ref
Living 10–20 miles from facility [§]	0.016 (0.018)
Living 5–10 miles from facility [§]	0.026 (0.016)
Living >20 miles from facility [§]	-0.041 (0.017)
Living <5 miles from facility [§]	Ref
BMI 25–29.9	0.059 (0.020)
BMI 30–34.9	0.027 (0.021)
BMI 35–39.9	0.065 (0.028)
BMI <18.5	-0.057 (0.050)
BMI ≥40	0.049 (0.029)
BMI missing	0.035 (0.019)
BMI 18.5–24.9	Ref
CHF	-0.019 (0.015)
Epilepsy	-0.016 (0.027)
Cancer	-0.026 (0.012)
Renal dysfunction	-0.043 (0.015)
Prior bleeding	-0.021 (0.015)
Pneumonia	-0.016 (0.016)
Drug abuse	-0.073 (0.056)
Chronic liver disease	-0.025 (0.021)
Psychosis	-0.022 (0.024)
Hyperlipidemia	0.025 (0.014)
No. of regular medications, 5–9	-0.029 (0.014)
No. of regular medications, ≥10	-0.036 (0.020)
No. of regular medications, <5	Ref
Hospitalization d ≥7 d: yes vs no	-0.001 (0.019)
No. of hospitalizations ≥2: yes vs no	-0.001 (0.018)
Albumin 2.5–3.49 g/dL	-0.014 (0.017)
Albumin <2.5 g/dL	-0.075 (0.035)
Albumin missing	0.020 (0.038)

Continued

Table 3. Continued

Predictor	Coefficient (SE) [†]
Albumin ≥3.5 g/dL	Ref
ALP >150 U/L	-0.068 (0.029)
ALP missing	-0.014 (0.037)
ALP ≤150 U/L	Ref
Sodium ≤130 mmol/L	0.017 (0.056)
Sodium missing	-0.078 (0.027)
Sodium >130 mmol/L	Ref
eGFR 15–29.9 mL/min/1.73m ²	-0.038 (0.036)
eGFR 30–59.9 mL/min/1.73m ²	-0.003 (0.015)
eGFR <15 mL/min/1.73m ²	-0.070 (0.029)
eGFR missing	-0.017 (0.023)
eGFR ≥60 mL/min/1.73m ²	Ref
Peripheral vascular disease	-0.016 (0.017)
Use of β blocker	0.028 (0.013)
Use of ACEI	0.018 (0.012)
Use of metolazone	-0.075 (0.089)
Use of opioids	-0.013 (0.016)
Use of statins	-0.027 (0.014)
Use of acetaminophen	-0.024 (0.016)
Use of antibiotics	-0.021 (0.013)
Use of antiplatelet agents	-0.027 (0.015)
Use of oral steroids	-0.012 (0.017)
Influenza vaccine	0.020 (0.012)
PSA test	0.037 (0.018)
Mammography	0.061 (0.022)
Pap smear	-0.048 (0.037)
Falls	-0.021 (0.021)
Fractures	-0.013 (0.021)
Parkinson disease	0.097 (0.043)

ACEI indicates angiotensin-converting enzyme inhibitors; AF, atrial fibrillation; ALP, alkaline phosphatase; BMI, body mass index; CHF, congestive heart failure; eGFR, estimated glomerular filtration rate; PSA, prostate-specific antigen; TTR, time in therapeutic range; VTE, venous thromboembolism.

*Quantified by international normalized ratio (INR) time in therapeutic range (TTR), see the distribution of TTR in training and validation sets in Figure S2.

[†]Unstandardized coefficients based on a lasso regression model including all the variables listed in this table.

[‡]Based on the mean income level of the ZIP code the patients resided in.

[§]Average distance based on ZIP codes from the nearest facility in the network.

^{||}Including all the major upper and lower gastrointestinal, and other extracranial bleeding events.

skewed the distribution of TTR, we repeated our analyses after (1) Box–Cox transformation of TTR³⁸ and (2) exclusion of those with extreme outcomes (TTR=0) from the analysis. The statistical analyses were conducted with SAS 9.4 (SAS Institute Inc).

Table 4. Simplified Geriatric Prediction Score for Anticoagulation Control Quality*: PROSPER

Predictor	Coefficient (SE) [†]	Point
Intercept	0.719 (0.012)	...
Pneumonia	−0.030 (0.015)	1
Renal dysfunction [‡]	−0.068 (0.013)	2
Oozing blood (bleeding history)	−0.026 (0.015)	1
Staying in hospital ≥7 d	−0.029 (0.015)	1
Pain medications	−0.037 (0.013)	1
No Enhanced anticoagulation care [§]	−0.122 (0.012)	4
Rx for antibiotics	−0.030 (0.013)	1

All variables should be assessed in the 6 mo before initiating a VKA, except for no enhanced anticoagulation care, which was assessed at the time of initiation. PROSPER indicates Pneumonia, Renal dysfunction, Oozing blood [prior bleeding], Staying in hospital ≥7 days, use of Pain medications, lack of Enhanced [dedicated and structured] anticoagulation care, Rx for antibiotics; VKA, vitamin K antagonist.

*Quantified by international normalized ratio time in therapeutic range (TTR).

[†]Unstandardized coefficients based on a model selected based on a Bayesian information criterion.

[‡]Renal dysfunction was defined as having records for acute kidney injury, chronic kidney disease, or end-stage kidney disease in the prior 180 days.

[§]Lack of participation (no access or plan) in a dedicated anticoagulation management service when initiating a VKA.

Results

Systematic Review

From a total of 3692 studies, we selected 16 articles (Figure 1 summarizes the search and selection process). Among them, 11 articles investigated patients taking VKA for nonvalvular AF,^{17,18,20,22,39–45} 4 for mixed indications^{19,21,46,47} and 2 for VTE.^{18,48} Based on these selected articles, we identified 8 positive and 42 negative predictors (Table 1). To enrich the candidate predictor pool, we added 23 variables based on clinical knowledge (see Table S3 for the full list of candidate predictors and their definitions).

Patient Characteristics of the Study Population

Among 14250 VKA new initiators with at least 180 days of Medicare enrollment and 1 EHR encounter in the study system, we selected the study cohort with at least 5 INRs recorded in our database, including 1663 patients in the training set, 694 in the AF validation set, and 487 in the VTE validation set. The distribution of combined comorbidity score was similar in patients with versus without 5 INRs with a mean standardized difference of 0.02 between proportions of all combined comorbidity score categories in those included versus excluded (Figure S1). The mean TTR was 0.47 to 0.56 in our training and validation sets (Table 2 and Figure S2). We observed modest differences in some predictors in AF versus VTE validation populations (eg, higher prevalence of cancer and more use of antibiotics in the VTE; Table 2).

Prediction Models for TTR

From 50 predictors identified in the systematic review and 23 additional variables, we built the new geriatric TTR score with 42 predictors through lasso regression ($R^2=0.19$; Table 3). The simplified model included a total of 7 variables ($R^2=0.14$). We summarized these variables using the acronym PROSPER (Pneumonia, Renal dysfunction, Oozing blood [prior bleeding], Staying in hospital ≥7 days, use of Pain medications, lack of Enhanced [dedicated and structured] anticoagulation care, Rx for antibiotics; see Table 4 and Table S3 for detailed definitions). There was no significant difference between the AUCs of PROSPER versus the full geriatric TTR model predicting TTR >70% in the validation set (AUC 0.678 versus 0.680, $P=0.86$ for difference). The 2 most influential predictors of TTR were lack of participation in a dedicated anticoagulation management service (assigned 4 points) and renal dysfunction (assigned 2 points). The rest of the variables were assigned 1 point each.

Comparison of Performance: SAME-TT₂R₂ Versus Geriatric TTR Score

In the training set, the AUC for the geriatric TTR score predicting TTR >70% (AUC=0.71) was substantially larger than that for coefficient-based SAME-TT₂R₂ (AUC=0.57, $P<0.001$ for difference); the AUC for the geriatric TTR score predicting the primary clinical outcome (AUC=0.65) was significantly larger than that for SAME-TT₂R₂ (AUC=0.53, $P<0.001$ for difference). The results were similar in the validation set (Figure 2). This pattern was consistent when the validation set was subdivided into AF and VTE validation sets (Table 5). We also found similar findings when the composite clinical outcome was subdivided into thromboembolic versus bleeding outcomes (Table S4). The Hosmer–Lemeshow goodness-of-fit test for predicting TTR >70% confirmed good calibration for the both the full geriatric model and coefficient-based SAME-TT₂R₂ in the training and validation sets (Table S5).

Comparison of Performance: SAME-TT₂R₂ Simple Scoring System Versus PROSPER

In the training set, the AUC for PROSPER predicting TTR >70% (AUC=0.67) was substantially larger than that for SAME-TT₂R₂ (AUC=0.55, $P<0.001$ for difference); the AUC for PROSPER predicting the primary clinical outcome (AUC=0.62) was significantly larger than that for SAME-TT₂R₂ (AUC=0.52, $P<0.001$ for difference). A similar pattern was observed in the AF and VTE validation sets when predicting both types of outcomes (Table 5). Patients stratified by PROSPER had a clear decreasing trend of mean TTR, ranging from 0.71 to 0.30, in both the training and validation sets (Table 6).

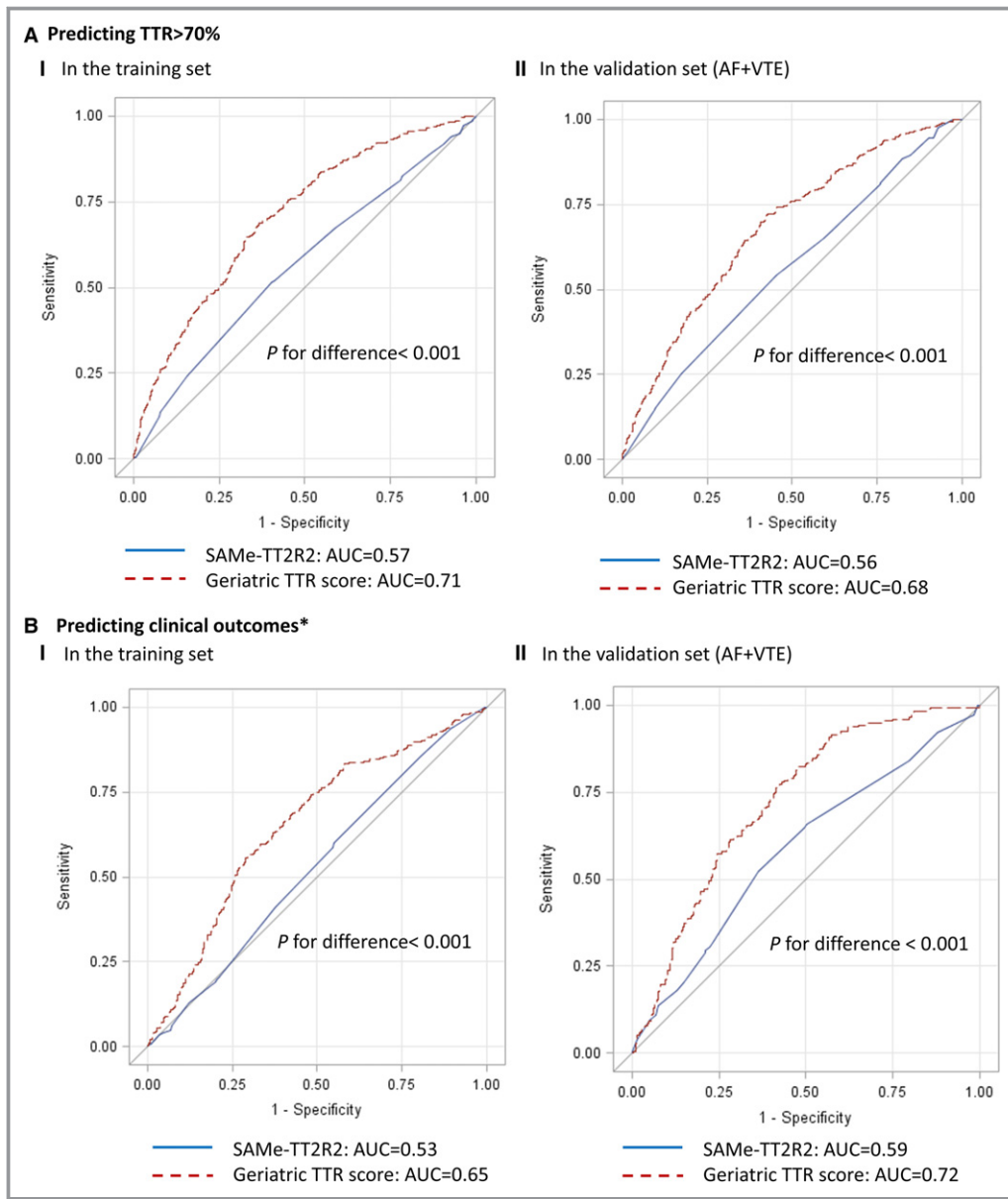


Figure 2. Comparison of AUC: new geriatric score superior to SAME-TT₂R₂. A, Predicting TTR >70%. B, Predicting clinical outcomes. *Composite outcomes of incident stroke, systemic embolism, VTE, and major bleeding events. AF indicates atrial fibrillation; AUC, area under the receiver operating characteristic curve; INR, international normalized ratio; TTR, time in therapeutic range; VTE, venous thromboembolism.

Sensitivity Analyses

After changing the length of baseline assessment period from 180 to 365 days, the revised prediction score was highly correlated with the original one (Spearman coefficient=0.89). After defining new initiation of VKA as no use in the 180 days rather than 90 days before the index date, the revised prediction score was highly correlated with the original one (Spearman coefficient=0.99). The performance of these revised models was similar to that of the original model (Table S6). After Box–Cox transformation, the distribution of TTR became more symmetric (Fisher-Pearson

skewness coefficient⁴⁹ reduced by 46%), resulting in a prediction score highly correlated with the predicted value generated by the original model (Spearman coefficient=0.99). Similar patterns were found when excluding those with TTR 0 (Table S6).

Discussion

We developed and validated a new prediction score in the older adult population. Our geriatric TTR score included 42 predictors, and the simplified clinical scoring system, PROSPER, had 7 variables. The geriatric TTR score and

Table 5. Comparison of Model Performance of Original SAME-TT₂R₂ and Geriatric TTR Score

	Optimized Prediction Models			Simplified Prediction Models		
	SAME-TT ₂ R ₂ * AUC (95% CI)	Geriatric TTR Score AUC (95% CI)	P for Difference	SAME-TT ₂ R ₂ † AUC (95% CI)	PROSPER‡ AUC (95% CI)	P for Difference
Prediction TTR >70%, training set	0.57 (0.54–0.59)	0.71 (0.68–0.73)	<0.001	0.55 (0.52–0.58)	0.67 (0.64–0.69)	<0.0001
Prediction TTR >70%, AF validation set	0.57 (0.52–0.61)	0.66 (0.62–0.70)	0.0011	0.58 (0.53–0.62)	0.67 (0.62–0.71)	0.0016
Prediction TTR >70%, VTE validation set	0.57 (0.51–0.63)	0.74 (0.69–0.79)	<0.001	0.59 (0.54–0.65)	0.71 (0.66–0.77)	0.0003
Prediction clinical outcomes,§ training set	0.53 (0.49–0.56)	0.65 (0.62–0.69)	<0.001	0.52 (0.49–0.56)	0.62 (0.58–0.66)	<0.0001
Prediction clinical outcomes,§ AF validation set	0.60 (0.54–0.66)	0.74 (0.69–0.79)	<0.001	0.60 (0.55–0.66)	0.73 (0.68–0.77)	<0.0001
Prediction clinical outcomes,§ VTE validation set	0.57 (0.51–0.63)	0.67 (0.61–0.72)	0.01	0.59 (0.53–0.65)	0.65 (0.60–0.71)	0.098

AF indicates atrial fibrillation; AUC, area under receiver operating characteristic curve; CI, confidence interval; PROSPER, Pneumonia, Renal dysfunction, Oozing blood [prior bleeding], Staying in hospital ≥7 days, use of Pain medications, lack of Enhanced [dedicated and structured] anticoagulation care, Rx for antibiotics; TTR, time in therapeutic range; VTE, venous thromboembolism.

*Based on original coefficients.

†Simple scoring system of SAME-TT₂R₂.

‡Simplified geriatric TTR scoring system, see details in Table 4.

§Composite outcomes of incident stroke, systemic embolism, VTE, and major bleeding events.

PROSPER outperformed the corresponding coefficient-based and simple version of SAME-TT₂R₂, available for the past 4 years, when predicting TTR ≥70% and thromboembolic and bleeding outcomes for those aged ≥65 years. The performance of PROSPER was not significantly worse than that of the full model in the validation set.

Physicians can use geriatric TTR scores to identify patients with good predicted TTR (>70%) as good candidates for VKA therapy for nonvalvular AF or VTE; otherwise, a DOAC may be preferred unless contraindicated. It is feasible to develop an automated program in an EHR system for computing the

predicted TTR based on the full model as a clinical decision support tool; otherwise, PROSPER can be readily calculated without an aid. Our findings suggest that a PROSPER score >2 is predictive of having poor TTR; therefore, initiating a VKA may not be ideal. This cut point is associated with reasonable specificity (75%) for TTR >70% and sensitivity (85%; Table 7) for TTR <50% (another cut point suggested in the literature to indicate poor anticoagulation quality³⁴). Alternatively, the

Table 7. Sensitivity and Specificity in the Validation Set (AF and VTE)

Cutoff of Simplified Geriatric Score (PROSPER)*	TTR >70%		TTR <50%	
	Sensitivity (%)	Specificity (%)	Sensitivity (%)	Specificity (%)
0	19.5	93.4	97.7	15.9
1	31.4	84.7	92.6	28.7
2	47.1	74.6	85.2	43.5
3	56.0	67.6	79.4	52.3
4	68.9	57.6	70.5	64.6
5	79.5	48.0	59.2	73.4
6	89.1	36.2	46.6	83.6
7	91.8	26.9	35.5	88.4
8	93.5	17.6	22.7	91.5
9	95.2	10.4	13.0	94.2
10	99.7	4.3	5.8	98.7
11	100.0	0	0	100.0

AF indicates atrial fibrillation; PROSPER, Pneumonia, Renal dysfunction, Oozing blood [prior bleeding], Staying in hospital ≥7 days, use of Pain medications, lack of Enhanced [dedicated and structured] anticoagulation care, Rx for antibiotics; TTR, time in therapeutic range; VTE, venous thromboembolism.

*Defining patients with scores lower than or equal to this cutoff as having TTR >70% and those with scores higher than the cut point as having TTR <50%.

Table 6. Mean TTR by Simplified New Geriatric Score

Simplified Geriatric Score (PROSPER)*	Training Set (n=1663)		Validation Set (n=1033)	
	n (%)	Mean TTR (SD)	n (%)	Mean TTR (SD)
0	154 (9.3)	0.71 (0.17)	106 (10.3)	0.70 (0.18)
1	148 (8.9)	0.67 (0.20)	99 (9.6)	0.63 (0.19)
2	118 (7.1)	0.67 (0.18)	121 (11.7)	0.61 (0.21)
3	79 (4.8)	0.64 (0.18)	78 (7.6)	0.58 (0.23)
4	225 (13.5)	0.59 (0.25)	112 (10.8)	0.56 (0.24)
5	217 (13.0)	0.55 (0.25)	102 (9.9)	0.49 (0.31)
6	202 (12.1)	0.55 (0.25)	115 (11.1)	0.49 (0.28)
7	162 (9.7)	0.52 (0.25)	77 (7.5)	0.34 (0.30)
8	128 (7.7)	0.45 (0.28)	74 (7.2)	0.32 (0.26)
9	104 (6.3)	0.43 (0.27)	58 (5.6)	0.31 (0.28)
10	91 (5.5)	0.41 (0.26)	58 (5.6)	0.43 (0.31)
11	35 (2.1)	0.35 (0.25)	33 (3.2)	0.30 (0.23)

PROSPER indicates Pneumonia, Renal dysfunction, Oozing blood [prior bleeding], Staying in hospital ≥7 days, use of Pain medications, lack of Enhanced [dedicated and structured] anticoagulation care, Rx for antibiotics; TTR, time in therapeutic range.

*See details in Table 4.

Table 8. TTR by the Simplified Geriatric Score Categories

Simplified Geriatric Score (PROSPER)	Training Set		Validation Set (AF and VTE)	
	n (%)	Mean (SD)	n (%)	Mean (SD)
0–2	420 (25.3)	0.69 (0.18)	326 (31.6)	0.64 (0.20)
3–6	723 (43.5)	0.57 (0.25)	407 (39.4)	0.52 (0.27)
≥7	520 (31.3)	0.45 (0.27)	300 (29.0)	0.34 (0.28)
Total	1663	0.56 (0.25)	1033	0.51 (0.27)

AF indicates atrial fibrillation; PROSPER, Pneumonia, Renal dysfunction, Oozing blood [prior bleeding], Staying in hospital ≥7 days, use of Pain medications, lack of Enhanced [dedicated and structured] anticoagulation care, Rx for antibiotics; TTR, time in therapeutic range; VTE, venous thromboembolism.

categorization of PROSPER as 0 to 2, 3 to 6, and ≥7 approximately subdivided the population into tertiles that correlated well with TTR. These 3 categories may be used to indicate low, moderate, and high risk of having poor TTR (Table 8). Our work highlights the importance of a structured approach to warfarin management; lack of a dedicated anticoagulation management service was found to be the strongest predictor of poor TTR. This finding is in line with several prior studies in which structured anticoagulation care was shown to improve TTR and to reduce risk of complications.^{50–53} In the current era when DOACs are available, unstructured warfarin management is a particularly unattractive treatment option. If DOAC treatment is not possible and a patient has a PROSPER score >2, providers should encourage patient participation in a dedicated anticoagulation management service or some equivalently well-organized warfarin treatment setting (eg, a practice with a nurse dedicated to managing warfarin). Renal dysfunction, defined as the presence of acute kidney injury, chronic kidney disease, or end-stage kidney disease in the prior 180 days, was also found to be an important predictor of poor TTR. The anticoagulation decision is particularly difficult in AF patients with renal dysfunction, for whom there is uncertainty as to the net benefit of warfarin or DOACs with poor renal function. One approach can be to favor use of a DOAC that is less renally excreted (eg, apixaban) with necessary dose adjustment.

Our score can also be helpful in a research context. First, researchers can evaluate the potential treatment-effect heterogeneity by levels of predicted anticoagulation quality based on the full model with a computer program. This assessment can provide direct evidence for choosing the ideal candidates for VKA versus DOACs based on the pretreatment characteristics predictive of anticoagulation quality. Next, our score can be used as a proxy adjustment tool for confounding by anticoagulation control quality. This adjustment is otherwise difficult because calculating TTR requires intensive INR recording in the study databases, which are often incomplete or nonexistent. Because some researchers do not have information on

biophysiologic variables, we also presented an alternative model excluding these variables that requires additional testing (Table S7). This alternative model had performance similar to the geriatric TTR score (data not shown).

We have demonstrated that the performance of the new geriatric TTR score was clearly superior to that of SAME-TT₂R₂ in the older adult population. The authors of SAME-TT₂R₂ demonstrated good discrimination performance only when predicting those with TTR <5th percentile but not for those with TTR <25th percentile (AUC=0.58).¹⁷ However, the latter is closer to the clinical relevant cut points (eg, TTR >70% usually composes about 30% to 40% of the population^{17,18,34}). Because the majority of VKA users are older adults^{3,18} who are also more vulnerable to developing bleeding complications,⁵⁴ we developed an alternative score dedicated to patients aged ≥65 years. We built a prediction model for both nonvalvular AF and VTE indications because prior studies found that AF and VTE patients share many risk factors for TTR¹⁸ and because including interaction terms with the VKA indications did not materially improve the model in our analysis. We validated the performance of our model in AF and VTE populations separately and found consistent results.

There are some important limitations. We used linked claims EHR data for higher data quality: The claims data provided comprehensive data across care settings, and EHRs provided necessary clinical information to ascertain TTR and important predictors. However, requiring overlap of the 2 databases reduces our sample size substantially and limits our ability to investigate each clinical outcome individually.⁵⁵ Besides, for biophysiologic variables (eg, body mass index, albumin levels), we had 29% to 34% people with missing data in the relevant period. We handled it by the missing indicator method because not having certain tests done could, by itself, be informative of the general health state, and this approach allows use of these scores even if some variables are not available. As a sensitivity analysis, the scores not including these variables with missing data were highly correlated with the original one and had similar model performance (data not shown). In addition, some of the possible determinants of TTR are not available in our data sets, such as diet information and genetic profiles associated with warfarin pharmacokinetics, which may limit our model performance. Consequently, our prediction scores should be used as an aid, not as the only ground for decision-making. Next, we chose to build one prediction model for both nonvalvular AF and VTE indications because prior studies found that AF and VTE patients have many common risk factors for poor anticoagulation quality.¹⁸ We validated the performance of this score in AF and VTE populations separately and found consistent results. Nonetheless, we acknowledge the alternative approach to build separate models for different indications, which may increase specificity at the cost of simplicity and applicability. Last, the

predictors identified in our study should not be interpreted as having causal effect on anticoagulation quality because they could be merely the markers or proxies of the real causal factors.

In conclusion, we developed and validated a prediction score for anticoagulation control quality quantified by TTR in the older adult population. It outperformed the published score, SAME-TT₂R₂, in patients aged ≥65 years when predicting TTR as well as thromboembolic and bleeding events. The full model of the geriatric TTR score can be used as an embedded algorithm within an EHR or for a research study. The simplified scoring system, PROSPER, had comparable performance and can be used in daily practice to help choose the best candidates to receive VKA therapy.

Sources of Funding

Lin received a stipend from the Pharmacoepidemiology program in the Department of Epidemiology, Harvard T.H. Chan School of Public Health and Department of Medicine, Brigham and Women's Hospital, Harvard Medical School. Singer was supported by the Eliot B. and Edith C. Shoolman Fund of the Massachusetts General Hospital (Boston, MA). Drs. Lin and Schneeweiss were supported by PCORI grant 282364.5077585.0007 (ARCH/SCILHS).

Disclosures

Oertel has occasionally participated on Advisory Boards for Roche Diagnostics and Alere. Dr. Schneeweiss is consultant to WHISCON, LLC and to Aetion, Inc., a software manufacturer of which he also owns equity. He is principal investigator of investigator-initiated grants to the Brigham and Women's Hospital from Bayer, Genentech, and Boehringer Ingelheim unrelated to the topic of this study. The remaining authors have no disclosures to report.

References

1. Ansell J, Hirsh J, Hylek E, Jacobson A, Crowther M, Palareti G; American College of Chest P. Pharmacology and management of the vitamin K antagonists: American College of Chest Physicians Evidence-Based Clinical Practice Guidelines (8th Edition). *Chest*. 2008;133:160S–198S.
2. Hart RG, Pearce LA, Aguilar MI. Meta-analysis: antithrombotic therapy to prevent stroke in patients who have nonvalvular atrial fibrillation. *Ann Intern Med*. 2007;146:857–867.
3. Huisman MV, Rothman KJ, Paquette M, Teutsch C, Diener HC, Dubner SJ, Halperin JL, Ma CS, Zint K, Elsaesser A, Bartels DB, Lip GY; Investigators G-A. The changing landscape for stroke prevention in AF: findings from the GLORIA-AF Registry Phase 2. *J Am Coll Cardiol*. 2017;69:777–785.
4. Cancino RS, Hylek EM, Reisman JI, Rose AJ. Comparing patient-level and site-level anticoagulation control as predictors of adverse events. *Thromb Res*. 2014;133:652–656.
5. Rose AJ, Delate T, Ozonoff A, Witt DM. Comparison of the abilities of summary measures of international normalized ratio control to predict clinically relevant bleeding. *Circ Cardiovasc Qual Outcomes*. 2015;8:524–531.
6. Connolly SJ, Pogue J, Eikelboom J, Flaker G, Commerford P, Franzosi MG, Healey JS, Yusuf S; Investigators AW. Benefit of oral anticoagulant over antiplatelet therapy in atrial fibrillation depends on the quality of international normalized ratio control achieved by centers and countries as measured by time in therapeutic range. *Circulation*. 2008;118:2029–2037.
7. Ho CW, Ho MH, Chan PH, Hai JJ, Cheung E, Yeung CY, Lau KK, Chan KH, Lau CP, Lip GY, Leung GK, Tse HF, Siu CW. Ischemic stroke and intracranial hemorrhage with aspirin, dabigatran, and warfarin: impact of quality of anticoagulation control. *Stroke*. 2015;46:23–30.
8. Singer DE, Chang Y, Fang MC, Borowsky LH, Pomernacki NK, Udaltsova N, Go AS. Should patient characteristics influence target anticoagulation intensity for stroke prevention in nonvalvular atrial fibrillation? The ATRIA study. *Circ Cardiovasc Qual Outcomes*. 2009;2:297–304.
9. Granger CB, Alexander JH, McMurray JJ, Lopes RD, Hylek EM, Hanna M, Al-Khalidi HR, Ansell J, Atar D, Avezum A, Bahit MC, Diaz R, Easton JD, Ezekowitz JA, Flaker G, Garcia D, Ghalibaf M, Gersh BJ, Golitsyn S, Goto S, Hermosillo AG, Hohnloser SH, Horowitz J, Mohan P, Jansky P, Lewis BS, Lopez-Sendon JL, Pais P, Parkhomenko A, Verheugt FW, Zhu J, Wallentin L; Committees A and Investigators. Apixaban versus warfarin in patients with atrial fibrillation. *N Engl J Med*. 2011;365:981–992.
10. Connolly SJ, Ezekowitz MD, Yusuf S, Eikelboom J, Oldgren J, Parekh A, Pogue J, Reilly PA, Themeles E, Varrone J, Wang S, Alings M, Xavier D, Zhu J, Diaz R, Lewis BS, Darius H, Diener HC, Joyner CD, Wallentin L; Committee R-LS and Investigators. Dabigatran versus warfarin in patients with atrial fibrillation. *N Engl J Med*. 2009;361:1139–1151.
11. Patel MR, Mahaffey KW, Garg J, Pan G, Singer DE, Hacke W, Breithardt G, Halperin JL, Hankey GJ, Piccini JP, Becker RC, Nessel CC, Paolini JF, Berkowitz SD, Fox KA, Califf RM; Investigators RA. Rivaroxaban versus warfarin in nonvalvular atrial fibrillation. *N Engl J Med*. 2011;365:883–891.
12. Wallentin L, Yusuf S, Ezekowitz MD, Alings M, Flather M, Franzosi MG, Pais P, Dans A, Eikelboom J, Oldgren J, Pogue J, Reilly PA, Yang S, Connolly SJ; Investigators R-L. Efficacy and safety of dabigatran compared with warfarin at different levels of international normalized ratio control for stroke prevention in atrial fibrillation: an analysis of the RE-LY trial. *Lancet*. 2010;376:975–983.
13. Hernandez I, Baik SH, Pintera A, Zhang Y. Risk of bleeding with dabigatran in atrial fibrillation. *JAMA Intern Med*. 2015;175:18–24.
14. Graham DJ, Reichman ME, Wernecke M, Zhang R, Southworth MR, Levenson M, Sheu TC, Mott K, Goulding MR, Houstoun M, MaCurdy TE, Worrall C, Kelman JA. Cardiovascular, bleeding, and mortality risks in elderly Medicare patients treated with dabigatran or warfarin for nonvalvular atrial fibrillation. *Circulation*. 2015;131:157–164.
15. Lauw MN, Eikelboom JW, Coppens M, Wallentin L, Yusuf S, Ezekowitz M, Oldgren J, Nakamya J, Wang J, Connolly SJ. Effects of dabigatran according to age in atrial fibrillation. *Heart*. 2017;103:1015–1023.
16. Stevens LA, Viswanathan G, Weiner DE. Chronic kidney disease and end-stage renal disease in the elderly population: current prevalence, future projections, and clinical significance. *Adv Chronic Kidney Dis*. 2010;17:293–301.
17. Apostolakis S, Sullivan RM, Olshansky B, Lip GY. Factors affecting quality of anticoagulation control among patients with atrial fibrillation on warfarin: the SAME-TT(2)R(2) score. *Chest*. 2013;144:1555–1563.
18. Macedo AF, Bell J, McCarron C, Conroy R, Richardson J, Scowcroft A, Sunderland T, Rotherham N. Determinants of oral anticoagulation control in new warfarin patients: analysis using data from Clinical Practice Research Datalink. *Thromb Res*. 2015;136:250–260.
19. Paradise HT, Berlowitz DR, Ozonoff A, Miller DR, Hylek EM, Ash AS, Jasuja GK, Zhao S, Reisman JI, Rose AJ. Outcomes of anticoagulation therapy in patients with mental health conditions. *J Gen Intern Med*. 2014;29:855–861.
20. Dlott JS, George RA, Huang X, Odeh M, Kaufman HW, Ansell J, Hylek EM. National assessment of warfarin anticoagulation therapy for stroke prevention in atrial fibrillation. *Circulation*. 2014;129:1407–1414.
21. Efirid LM, Mishkin DS, Berlowitz DR, Ash AS, Hylek EM, Ozonoff A, Reisman JI, Zhao S, Jasuja GK, Rose AJ. Stratifying the risks of oral anticoagulation in patients with liver disease. *Circ Cardiovasc Qual Outcomes*. 2014;7:461–467.
22. Kim EJ, Ozonoff A, Hylek EM, Berlowitz DR, Ash AS, Miller DR, Zhao S, Reisman JI, Jasuja GK, Rose AJ. Predicting outcomes among patients with atrial fibrillation and heart failure receiving anticoagulation with warfarin. *Thromb Haemost*. 2015;114:70–77.
23. Lip GY, Haguenoer K, Saint-Etienne C, Fauchier L. Relationship of the SAME-TT(2) R(2) score to poor-quality anticoagulation, stroke, clinically relevant bleeding, and mortality in patients with atrial fibrillation. *Chest*. 2014;146:719–726.
24. Gallego P, Roldan V, Marin F, Galvez J, Valdes M, Vicente V, Lip GY. SAME-TT2R2 score, time in therapeutic range, and outcomes in anticoagulated patients with atrial fibrillation. *Am J Med*. 2014;127:1083–1088.
25. Zhang H, Yang Y, Zhu J. The SAME-TT(2)R(2) score: far from clinical application. *Chest*. 2014;145:418–419.
26. Hennessy S. Use of health care databases in pharmacoepidemiology. *Basic Clin Pharmacol Toxicol*. 2006;98:311–313.

27. Johnson ES, Bartman BA, Briesacher BA, Fleming NS, Gerhard T, Kornegay CJ, Nourjah P, Sauer B, Schumock GT, Sedrakyan A, Sturmer T, West SL, Schneeweiss S. The incident user design in comparative effectiveness research. *Pharmacoepidemiol Drug Saf.* 2013;22:1–6.
28. Gagne JJ, Glynn RJ, Avorn J, Levin R, Schneeweiss S. A combined comorbidity score predicted mortality in elderly patients better than existing scores. *J Clin Epidemiol.* 2011;64:749–759.
29. Austin PC. Balance diagnostics for comparing the distribution of baseline covariates between treatment groups in propensity-score matched samples. *Stat Med.* 2009;28:3083–3107.
30. Rosendaal FR, Cannegieter SC, van der Meer FJ, Briet E. A method to determine the optimal intensity of oral anticoagulant therapy. *Thromb Haemost.* 1993;69:236–239.
31. Go AS, Hylek EM, Chang Y, Phillips KA, Henault LE, Capra AM, Jensvold NG, Selby JV, Singer DE. Anticoagulation therapy for stroke prevention in atrial fibrillation: how well do randomized trials translate into clinical practice? *JAMA.* 2003;290:2685–2692.
32. Tibshirani R. Regression shrinkage and selection via the lasso. *J R Stat Soc Series B.* 1996;58:267–288.
33. Jones RH. Bayesian information criterion for longitudinal and clustered data. *Stat Med.* 2011;30:3050–3056.
34. Razouki Z, Ozonoff A, Zhao S, Jasuja GK, Rose AJ. Improving quality measurement for anticoagulation: adding international normalized ratio variability to percent time in therapeutic range. *Circ Cardiovasc Qual Outcomes.* 2014;7:664–669.
35. Wan Y, Heneghan C, Perera R, Roberts N, Hollowell J, Glasziou P, Bankhead C, Xu Y. Anticoagulation control and prediction of adverse events in patients with atrial fibrillation: a systematic review. *Circ Cardiovasc Qual Outcomes.* 2008;1:84–91.
36. DeLong ER, DeLong DM, Clarke-Pearson DL. Comparing the areas under two or more correlated receiver operating characteristic curves: a nonparametric approach. *Biometrics.* 1988;44:837–845.
37. Friedman C, Hripcsak G. Natural language processing and its future in medicine. *Acad Med.* 1999;74:890–895.
38. Box GEP, Cox DR. An analysis of transformations revisited. *J R Stat Soc.* 1964;26:211–252.
39. Boulanger L, Kim J, Friedman M, Hauch O, Foster T, Menzin J. Patterns of use of antithrombotic therapy and quality of anticoagulation among patients with non-valvular atrial fibrillation in clinical practice. *Int J Clin Pract.* 2006;60:258–264.
40. Tomita H, Kadokami T, Momii H, Kawamura N, Yoshida M, Inou T, Fukuizumi Y, Usui M, Funakoshi K, Yamada S, Aomori T, Yamamoto K, Uno T, Ando S; Group A-Wr. Patient factors against stable control of warfarin therapy for Japanese non-valvular atrial fibrillation patients. *Thromb Res.* 2013;132:537–542.
41. Kose E, Arai S, An T, Kikkawa A, Aoyama T, Matsumoto Y, Hayashi H. Analysis of factors affecting time in therapeutic range control after warfarin administration. *Pharmazie.* 2015;70:494–498.
42. Nelson WW, Desai S, Damaraju CV, Lu L, Fields LE, Wildgoose P, Schein JR. International normalized ratio stability in warfarin-experienced patients with nonvalvular atrial fibrillation. *Am J Cardiovasc Drugs.* 2015;15:205–211.
43. Pignatelli P, Pastori D, Vicario T, Bucci T, Del Ben M, Russo R, Tanzilli A, Nardoni ML, Bartimoccia S, Nocella C, Ferro D, Saiola M, Cangemi R, Lip GY, Violi F. Relationship between Mediterranean diet and time in therapeutic range in atrial fibrillation patients taking vitamin K antagonists. *Europace.* 2015;17:1223–1228.
44. White RD, Riggs KW, Ege EJ, Petroski GF, Koerber SM, Flaker G. The effect of the amiodarone-warfarin interaction on anticoagulation quality in a single, high-quality anticoagulation center. *Blood Coagul Fibrinolysis.* 2016;27:147–150.
45. Yong C, Azarbal F, Abnoui F, Heidenreich PA, Schmitt S, Fan J, Than CT, Ullal AJ, Yang F, Phibbs CS, Frayne SM, Ho PM, Shore S, Mahaffey KW, Turakhia MP. Racial differences in quality of anticoagulation therapy for atrial fibrillation (from the TREAT-AF study). *Am J Cardiol.* 2016;117:61–68.
46. Rose AJ, Hylek EM, Ozonoff A, Ash AS, Reisman JI, Berlowitz DR. Patient characteristics associated with oral anticoagulation control: results of the Veterans Affairs Study to Improve Anticoagulation (VARIA). *J Thromb Haemost.* 2010;8:2182–2191.
47. Nilsson H, Grove EL, Larsen TB, Nielsen PB, Skjoth F, Maegaard M, Christensen TD. Sex differences in treatment quality of self-managed oral anticoagulant therapy: 6,900 patient-years of follow-up. *PLoS One.* 2014;9:e113627.
48. Kooistra HA, Gebel M, Sahin K, Lensing AW, Meijer K. Independent predictors of poor vitamin K antagonist control in venous thromboembolism patients. Data from the EINSTEIN-DVT and PE studies. *Thromb Haemost.* 2015;114:1136–1143.
49. Newell KM, Hancock PA. Forgotten moments: a note on skewness and kurtosis as influential factors in inferences extrapolated from response distributions. *J Mot Behav.* 1984;16:320–335.
50. Wurster M, Doran T. Anticoagulation management: a new approach. *Dis Manag.* 2006;9:201–209.
51. Rudd KM, Dier JG. Comparison of two different models of anticoagulation management services with usual medical care. *Pharmacotherapy.* 2010;30:330–338.
52. Ansell JE. Optimizing the efficacy and safety of oral anticoagulant therapy: high-quality dose management, anticoagulation clinics, and patient self-management. *Semin Vasc Med.* 2003;3:261–270.
53. Chamberlain MA, Sageser NA, Ruiz D. Comparison of anticoagulation clinic patient outcomes with outcomes from traditional care in a family medicine clinic. *J Am Board Fam Pract.* 2001;14:16–21.
54. Pisters R, Lane DA, Nieuwlaar R, de Vos CB, Crijns HJ, Lip GY. A novel user-friendly score (HAS-BLED) to assess 1-year risk of major bleeding in patients with atrial fibrillation: the Euro Heart Survey. *Chest.* 2010;138:1093–1100.
55. Lin KJ, Schneeweiss S. Considerations for the analysis of longitudinal electronic health records linked to claims data to study the effectiveness and safety of drugs. *Clin Pharmacol Ther.* 2016;100:147–159.

Supplemental Material

Data S1.

Natural language processing to improve missing data

We extracted smoking and BMI information from free text clinical notes using MTERMS (Medical Text Extraction, Reasoning and Extraction System)¹, a modular, pipeline-based NLP system for identifying clinical terms in unstructured medical record text. MTERMS takes as input a text file containing a single clinical note which it decomposes into sections (e.g., “History of Present Illness”, “Impression and Plan”), when applicable. Sections are further decomposed to sentences, which are then tokenized and processed through a series of lexical lookup- and regular expression-based modules. MTERMS applies and supports both machine learning and rule-based algorithms for classification tasks. In this project, we used a rule-based approach.

To extract smoking status, we used a lexicon of terms found to be indicative of or associated with smoking status. This set of terms was constructed and used by MTERMS to extract smoking status as part of a study to predict hospital readmission². Terms were compiled from various clinical terminologies and guidelines (e.g., SNOMED-CT, Public Health Service Tobacco Use and Dependence Guidelines). Synonyms, abbreviations, and other lexical variations were manually added to the list by a team of clinical informaticians. Following term extraction, MTERMS identifies and processes certain contextual information associated with the extracted terms, including negation (e.g., “non-smoker”, “quit smoking”) and family history information (e.g., “patient’s father smokes”). A regular expression-based classifier was applied to the extracted output to differentiate between patients who currently smoke, patients who previously smoked but do not currently smoke, and patients who have never smoked.

BMI information was extracted with a similar approach. First, we used MTERMS to perform a lexical search for the terms “BMI”, “body mass index”, and a small set of lexical variations thereof. We also used MTERMS to search for height and weight information using a lexicon compiled from clinical terminologies (e.g., LOINC®)³ and manually supplemented with synonyms, abbreviations, and other lexical variations. Once terms were identified, MTERMS applied a set of regular expressions to locate and extract the actual values associated with the extracted terms, as well as the units in which the values were reported, if applicable. If both height and weight values (and their corresponding units) were identified, an additional post-processing step was applied to calculate the BMI.

Table S1. SAMe-TT2R2 original model⁴	
Variable	Unstandardized coefficient
Female sex	-0.04
Age<50 years old	-0.14
Age 50-60 years old	-0.04
Non-white race	-0.09
Tobacco smoking within 2 years	-0.04
More than two comorbidities*	-0.04
B-blocker use	0.03
Verapamil use	0.05
Amiodarone use	-0.03
* more than two of the following: hypertension, diabetes, coronary artery disease/myocardial infarction, peripheral arterial disease, congestive heart failure, previous stroke, pulmonary disease, and hepatic or renal disease.	

SAMe-TT2R2 simple scoring system⁴		
Acronym	Definition	Points
S	Female sex	1
A	Age< 60 years old	1
Me	More than two comorbidities*	1
T	Treatment with interacting drug, e.g. amiodarone	1
T	Tobacco smoking within 2 years	2
R	Non-white race	2
* more than two of the following: hypertension, diabetes, coronary artery disease/myocardial infarction, peripheral arterial disease, congestive heart failure, previous stroke, pulmonary disease, and hepatic or renal disease.		

Table S2. Definitions of clinical outcomes

Outcome	Hospital Discharge Code(s)	Comments
Primary Outcome: a composite outcome of stroke, systemic embolism, deep vein thrombosis, pulmonary Embolism, and major extracranial bleeding listed below		
Stroke	As primary ICD-9 discharge diagnosis (Dx): 430.x Subarachnoid hemorrhage (SAH) 431.x Intracranial hemorrhage (ICH) 433.x1 Occlusion and stenosis of precerebral arteries with cerebral infarction 434.x1 Occlusion and stenosis of cerebral arteries with cerebral infarction 436.x Acute, but ill-defined cerebrovascular events	In Tennessee Medicaid enrollees aged 50-84, the algorithm had PPV of 97% for primary discharge diagnoses and 89% for primary and secondary ⁵ See Mini-Sentinel report for PPV for individual codes or other algorithms ⁶
Systemic embolism	ICD-9 Diagnoses: 444.x Arterial embolism 362.3x Retinal vascular occlusion 996.7x Other complications of internal (biologic or synthetic) prosthetic device, implant, and graft (including but not limited to embolism or thrombus) 997.7x Vascular complications of other vessels Mesenteric artery (997.71) Renal artery (997.72) Other vessels (997.79) 415.1x Pulmonary embolism and infarction	No validation studies available. Codes are thought to be specific because of the substantial clinical symptomatic and required therapy.
Deep vein thrombosis (DVT)	<u>Validated algorithm:</u> ICD-9 451.1x (Phlebitis and thrombophlebitis of deep vessels of lower extremities) ICD-9 451.2x (of lower extremities, unspecified) ICD-9 451.81 (of Iliac vein) ICD-9 451.9x (of unspecified site) ICD-9 453.1x (thrombophlebitis migrans) ICD-9 453.2x (venous embolism and thrombosis of vena cava) ICD-9 453.8x (venous embolism and thrombosis of other specified veins)	Algorithm for Deep vein thrombosis (DVT): ICD-9 codes of 451.1, 451.2, 451.81, 451.9, 453.1, 453.2, 453.8, 453.9 [hospital discharge, any position] had PPV of 0.72 and specificity > 0.99 in Medicare population ⁷

Outcome	Hospital Discharge Code(s)	Comments
	ICD-9 453.9x (venous embolism and thrombosis of unspecified site) <u>Not in the validated algorithm but will be included following Mini-Sentinel recommendation for VTE outcome:</u> ICD-9 453.40 (Venous embolism and thrombosis of unspecified deep vessels of lower extremity (includes DVT)) ICD-9 453.41 (Venous embolism and thrombosis of deep vessels of proximal lower extremity (includes femoral, iliac, popliteal, thigh, and upper leg)) ICD-9 453.42 (Venous embolism and thrombosis of deep vessels of distal lower extremity (includes calf, lower leg, peroneal, and tibia)) ICD-9 453.0 (Hepatic vein thrombosis)	Mini-Sentinel: while using ICD-9 codes 415.x (PE), 451.x and 453.x (DVT) as a VTE event yielded the highest PPV, for a specific event (DVT or PE) PPV was lower; therefore, the performance of algorithms depends on a population studied ⁸
Pulmonary Embolism (PE)	ICD-9 415.1x (pulmonary embolism and infarction)	PPV of 72% in a community sample (45 YO and older) ⁹
Major extracranial bleeding	Major upper GI bleed, major lower and unspecified GI bleed, major urogenital bleed, major other bleed (for codes see component outcomes below)	No validation studies available for this outcome
Major GI bleeding	Major upper GI bleeding, major lower/unspecified GI bleeding (for codes see component outcomes below)	No validation studies available for this outcome
Major upper GI bleed	ICD-9 diagnoses: 531.0x (acute gastric ulcer with hemorrhage with/without obstruction) 531.2x (with hemorrhage and perforation with/without obstruction) 531.4x (chronic or unspecified gastric ulcer with hemorrhage with/without obstruction) 531.6x (with hemorrhage and perforation with/without obstruction) 532.0x (acute duodenal ulcer with hemorrhage with/without obstruction) 532.2x (with hemorrhage and perforation with/without obstruction) 532.4x (chronic or unspecified duodenal ulcer with hemorrhage with/without obstruction) 532.6x (with hemorrhage and perforation with/without obstruction) 533.0x (acute peptic ulcer of unspecified site with hemorrhage with/without obstruction)	PPV of 87.8% in commercially-insured population ¹⁰

Outcome	Hospital Discharge Code(s)	Comments
	533.2x (with hemorrhage and perforation with/without obstruction) 533.4x (chronic or unspecified peptic ulcer of unspecified site with hemorrhage with/without obstruction) 533.6x (with hemorrhage and perforation with/without obstruction), 534.0x (acute gastrojejunal ulcer with hemorrhage with/without obstruction) 534.2x (with hemorrhage and perforation with/without obstruction) 534.4x (chronic or unspecified gastrojejunal ulcer with hemorrhage with/without obstruction) 534.6x (with hemorrhage and perforation with/without obstruction) 578.0 (hematemesis) OR ICD-9 procedure code 44.43 (endoscopic control of gastric or duodenal bleeding) OR CPT code 43255 (upper gastrointestinal endoscopy including esophagus, stomach, and either the duodenum and/or jejunum as appropriate with control of bleeding, any method)	
Major lower GI bleeding	<u>Lower GI/unspecified GI site bleed¹¹:</u> Diverticulosis of small intestine with hemorrhage: 562.02 Diverticulitis of small intestine with hemorrhage: 562.03 Diverticulosis of colon with hemorrhage: 562.12 Diverticulitis of colon with hemorrhage: 562.13 Hemorrhage of rectum and anus: 569.3x Angiodysplasia of intestine with hemorrhage: 569.85 Blood in stool: 578.1x Hemorrhage of GI tract, unspecified: 578.9	PPVs for individual codes ¹¹ : 562.12 – 91.7% 562.13 – 66.7% 569.3 - 71.4% 569.85 – 100% 578.1 - 81.8% 578.9 -88.2%
Major urogenital bleed	ICD-9 diagnoses: Hematuria: ICD-9 Dx: 599.7 Excessive/frequent menstruation: ICD-9 Dx 626.2x <u>and</u> secondary diagnosis indicating acute bleeding: anemia (280.0, 285.1, 285.9), orthostasis (458.0), syncope (780.2) ¹¹	PPVs for individual codes ¹¹ : 599.7 - 75.0% 626.2 – 100% (2 cases) ¹¹
Other major bleeds	<u>Other major bleeds¹¹:</u> Hemathrosis: 719.1x Hemopericardium: 423.0x Hemoptysis: 786.3x Epistaxis: 784.7x Hemorrhage not specified 459.0x	PPVs for individual codes ¹¹ : 719.1x – 100% 786.3x – 80% 784.7x – 100% 459.0x – 100%

Outcome	Hospital Discharge Code(s)	Comments
	Acute posthemorrhagic anemia 285.1x	285.1x – 100%

Table S3. Definitions of predictors for anticoagulation control quality

Predictors	Definition*	Comments
Age	By category (65-74.9, 75-84.9, ≥85)	
Sex	Female, Male	
Use of Metolazone	Use of Metolazone (generic name of a diuretic)	
Marital_Status	use the levels as recorded plus a missing category (available in the EHR raw demographic table with the variable name "Marital_Status")	
Language	use the levels as recorded plus a missing category (available in the EHR raw demographic table with the variable name "Language")	
Median household income based on patient's zipcode	Quartiles of mean income based on zipcode using 2010 Census for ZIP code level variables and cross walk between zip codes and ZIP Code Tabulation Areas (ZCTAs) file that was obtained at: http://mcdc.missouri.edu/allabout/zipcodes_2010supplement.shtml for data after 2010	
Distance to MGH sites	Classify zip codes based on the 4 groups in the sheet: <5, 5-9.9, 10-19.9, and >20 miles	
Distance to BWH sites	Classify zip codes based on the 4 groups in the sheet: <5, 5-9.9, 10-19.9, and >20 miles	
Anemia	280.xx Iron deficiency anemia 281.xx Other deficiency anemias 282.xx Hereditary hemolytic anemias 283.xx Acquired hemolytic anemias 284.xx Aplastic anemia and other bone marrow failure syndromes 285.xx Other and unspecified anemias	
Prior valve surgery	ICD-9 Dx code of V42.2 (heart valve replaced by transplant), V43.3 (heart valve replaced by a mechanical device/prosthesis) OR ICD-9 procedure code 35.1x (open heart valvuloplasty without replacement), 35.2x (replacement of heart valve), OR one of the following CPT codes: 33660-33665 (atrioventricular valve repair) 33400-33403 (aortic valve valvuloplasty) 33420-33430 (mitral valve repair/valvuloplasty/replacement) 33463-33468 (tricuspid valve repair/valvuloplasty/replacement) 33496 (prosthetic valve dysfunction repair)	
Tobacco smoking	As recorded in the EHR, categorized as current, non-current, or missing	

Predictors	Definition*	Comments
Race	As the original categorization in the claims data, categorized as white, black, and others	
Hypertension	At least 1 Dx of ICD-9 codes 401.x – 405.x OR at least 1 dispensing of a CCB, ACEI, ARB, BB, a thiazide diuretic or a direct antihypertensive agent	
Diabetes	At least 2 outpatient diagnoses of DM (ICD-9 250.X (diabetes)) OR 1 hospital discharge Dx of DM OR 1 diagnosis of DM plus an insulin or oral antidiabetic dispensing	
Heart failure (CHF)	1 inpatient or 2 outpatient claims with any of ICD-9 codes: 428.x, 398.91, 402.01, 402.11, 402.91, 404.01, 404.11, 404.91, 404.03, 404.13, 404.93	Validated in Medicare algorithm: Hospital discharge ICD-9 codes: 428.x, 398.91, 402.01, 402.11, 402.91, 404.01, 404.11, 404.91, 404.03, 404.13, 404.93 PPV 0.97, Spec. 0.97, Sens. 0.76 ⁷
Coronary Artery Disease (CAD)/ Previous MI	ICD-9 Dx: 411.x-414.x, 429.2 or V45.81	
Prior stroke	ICD-9 codes: 433.xx, 434.x, 435.x, 436.x, 437.x	
Peptic Ulcer Disease	Diseases of esophagus: 530.1x – 530.4x, 530.8x, 530.9x Gastric ulcer: 531.x Duodenal ulcer: 532.x Peptic ulcer: 533.x Acute gastritis: 535.0x Other specified gastritis: 535.4x Unspecified gastritis and gastroduodenitis: 535.5x Duodenitis: 535.6x	

Predictors	Definition*	Comments
Prior bleeding: any of the following bleeding events		
Upper GI bleed	ICD-9 diagnosis: 531.0x, 531.2x, 531.4x, 531.6x 532.0x, 532.2x, 532.4x, 532.6x, 533.0x, 533.2x, 533.4x, 533.6x, 534.0x, 534.2x, 534.4x, 534.6x, 578.0 OR ICD-9 procedure code 44.43 OR CPT code 43255	
Lower/ unspecified GI bleed	ICD-9 diagnoses: 562.02, 562.03, 562.12, 562.13, 568.81, 569.3, 569.83, 569.85, 569.86 578.1x, 578.9	
Urogenital bleed	ICD-9 diagnoses: Hematuria: ICD-9 Dx: 599.7 Excessive/frequent menstruation: ICD-9 Dx 626.2x AND secondary diagnosis indicating acute bleeding: anemia (280.0, 285.1, 285.9), orthostasis (458.0), syncope (780.2)	
Other bleeds	Hemathrosis: 719.1x Hemopericardium: 423.0x Hemoptysis: 786.3x Epistaxis: 784.7x Hemorrhage not specified: 459.0x Acute posthemorrhagic anemia: 285.1	
Peripheral Vascular disease or PVD surgery	1 inpatient or 2 outpatient claims with any of the following codes: ICD9 diagnosis: 440.20 - 440.24, 440.29 – 440.32, 440.3, 443.9 ICD9 procedure: 38.08, 38.09, 38.18, 38.48, 38.49, 39.25, 39.5, 39.9, 84.10 - 84.17 HCPCs: 35256, 35286, 35351, 35355, 35361, 35363, 35371, 35372, 35381, 35454, 35456, 35459, 35470, 35473, 35474, 35482, 35483, 35485, 35492, 35493, 35495, 35521, 35533, 35541, 35546, 35548, 35549, 35551, 35556, 35558, 35563, 35565, 35566, 35571, 35621, 35623, 35641, 35646, 35647, 35650, 35651, 35654, 35656, 35661, 35663, 35666, 35671, 27590, 27591, 27592, 27594, 27596, 27880, 27881, 27882, 27884, 27886, 27888	
Prior liver disease	ICD-9 diagnosis: 070.x viral hepatitis 571.x chronic liver disease and cirrhosis	

Predictors	Definition*	Comments
	572.x liver abscess and sequelae of chronic liver disease 573.x other disorders of liver 456.0 – 456.2x esophageal varices 155.0 primary cancer of liver 155.1 cancer of intrahepatic bile ducts 155.2 cancer of liver not specified as primary or secondary 576.8 cholestasis ICD-9 procedure codes: 39.1 intra-abdominal venous shunt 42.91 ligation of esophageal varices	
Renal Dysfunction	Acute renal disease (see below) Chronic renal disease (see below) Diabetic nephropathy (see below) Hypertensive nephropathy (see below) Miscellaneous Renal Insufficiency (see below) or ESRD (see below)	This definition of renal dysfunction is based on prior work ^{12, 13} and will be used for covariate adjustment. It is a combination of several diagnoses that are not well distinguished in claims databases.
Acute Renal Disease	580.0, 580.4, 580.8, 580.9, 581.0, 581.1, 581.2, 581.2, 581.3, 581.8, 581.9, 584.6, 584.7, 584.8, 584.9	
Chronic Renal Insufficiency	582.x, 583.x, 585.x, 586.x, 587.x	
Diabetic Nephropathy	250.4, 250.40, 250.41, 250.42, 250.43	

Predictors	Definition*	Comments
Hypertensive Nephropathy	403.xx, 404.xx	
Miscellaneous Renal Insufficiency	274.10, 440.1, 442.1, 453.3, 581.xx, 593.xx, 753.0, 753.3, 866.00 866.01, 866.1	
ESRD (with and without dialysis)	<p>DIALYSIS</p> <p>ICD-9 procedure: 39.95 hemodialysis 54.98 peritoneal dialysis 38.95 Venous catheterization for renal dialysis 39.27 Arteriovenostomy for renal dialysis 39.42 Revision of arteriovenous shunt for renal dialysis 39.43 Removal of arteriovenous shunt for renal dialysis</p> <p>ICD-9 diagnoses: V45.1 renal dialysis status V56.0 extracorporeal dialysis V56.8 peritoneal dialysis</p> <p>CPT4:</p> <p>90935 HEMODIALYSIS PROC W/SINGLE PHYSICIAN EVALUATION</p> <p>90937 HEMODIALYSIS, REPEATED EVAL, W/WO REVISION DIALYSIS PRESCRIPTION</p> <p>90940 HEMODIALYSIS ACCESS FLOW STUDY, BY INDICATOR DILUTION METHOD, HOOK UP; MEASUREMENT & DISCONNECTION</p> <p>90945 DIALYSIS, OTHER THAN HEMODIALYSIS, SINGLE PHYSICIAN EVAL</p> <p>90947 DIALYSIS PROCEDURE, OTHER THAN HEMODIALYSIS, REPEATED PHYSICIAN EVAL</p> <p>90989 DIALYSIS TRAINING, PATIENT, W/HELPER WHERE APPLICABLE, ANY MODE, COMPLETED COURSE</p>	

Predictors	Definition*	Comments
	<p>90993 DIALYSIS TRAINING, PATIENT, W/HELPER WHERE APPLICABLE, ANY MODE, COURSE INCOMPLETE, PER SESSION</p> <p>99512 HOME VISIT, HEMODIALYSIS</p> <p>99559 HOME INFUSION, PERITONEAL DIALYSIS, PER VISIT</p> <p>OR</p> <p>RENAL TRANSPLANT</p> <p>V42.0 Kidney transplant 55.6x Kidney transplant 996.81 Complication of transplanted kidney</p> <p>CPT4:</p> <p>50360 RENAL ALLOTRANSPLANTATION, IMPLANTATION, GRAFT; W/O DONOR & RECIPIENT NEPHRECTOMY 50365 RENAL ALLOTRANSPLANTATION, IMPLANTATION, GRAFT; W/RECIPIENT NEPHRECTOMY 50380 RENAL AUTOTRANSPLANTATION, REIMPLANTATION, KIDNEY</p> <p>OR</p> <p>RENAL ICD9-defined ESRD</p> <p>585.5 ESKD with no mention of dialysis 585.6 ESKD on dialysis</p>	
COPD	491.xx, 492.xx, or 496.xx	
Asthma	493.xx	
Pneumonia	480.xx – 486.xx, 487.0x, 507.xx	
Alcohol abuse	Alcohol abuse ICD-9 codes: 94.61 – 94.63 – alcohol rehabilitation and detoxification 94.67-94.69 – combined alcohol/drug rehabilitaion and detoxification 303.0x – 303.9x – alcoholism	

Predictors	Definition*	Comments
	291.xx – alcohol-induced mental disorders 357.5x – alcoholic polyneuropathy 425.5x – alcoholic cardiomyopathy 571.1x – acute alcoholic hepatitis 571.2x – alcoholic cirrhosis of liver 571.3x – alcoholic liver damage, unspecified 305.0x -alcohol abuse	
Drug abuse or dependence	292.xx, 304.xx, 305.2x-305.9x	
Epilepsy or convulsions	345.xx, 780.3x	
Cancer	140.x-195.x, 196.x-198.x, 199.x, 200.x-208.x, 230.x-234.x, 235.x-238.x, 239.x, excluding non-melanoma skin cancer (= 173.xx), v10.xx	
Albumin level	albumin <2.5, 2.5-3.49, ≥3.5 g/dL, vs missing indicator	
Serum creatinine	As continuous variable and as a categorical variable (creatinine ≤1, 1.01-1.99 and ≥2 mg/dL vs missing indicator	
Estimated glomerular filtration rate (eGFR)	calculated by serum creatinine by CKD-EPI EQUATION (http://nephron.com/epi_equation) in and classified as (GFR≥60, 45 to <60, 30 to <45, 15 to <30, ESRD if GFR<15 or classified as ESRD based on the criteria above.	
aspartate aminotransferase (AST) level	≤60, >60 to 100, >100 IU/L vs missing indicator	
Serum Alkaline phosphatase	Serum Alkaline phosphatase level > 150, ≤ 150 U/L vs missing indicator	
Serum Sodium	Serum Sodium level < 130, ≥130 mEq vs missing indicator	
BMI	as a categorical variable (BMI≥ 40, 35-39.9, 30-34.9, 18.5-24.9, <18.5 vs missing indicator)	
Underweight	Weight <50 vs ≥50 kg vs missing indicator	
Dementia	290.xx, 294.xx, 330.xx, 331.xx	
Depression	293.83, 296.2x, 296.3x, 298.0x, 300.4x, 309.0x, 309.1x, 309.28, 311.xx	
Bipolar disorder	296.0x, 296.1x, 296.4x, 296.5x, 296.6x, 296.7x, 296.8x, 296.99	

Predictors	Definition*	Comments
Psychotic disorders	290.8x, 290.9x, 295.xx, 297.xx, 298.xx, 299.xx, 780.1x	
*all the codes are assessed in the baseline period: 180 days prior to the index date		

Medication use predictors	Generic drug names
Antihypertensives	Any of the following classes of antihypertensives
ACE-Inhibitors	Benazepril, captopril, enalapril, fosinopril, lisinopril, moexipril, perindopril, quinapril, ramipril, trandolapril
ARBs	Azilsartan, candesartan, eprosartan, irbesartan, losartan, olmesartan, telmisartan, valsartan
Beta-blockers	Acebutolol, atenolol, betaxolol, bisoprolol, carteolol, carvedilol, esmolol, labetalol, metoprolol tartrate, metoprolol succinate, nadolol, nebivolol, penbutolol, pindolol, propranolol, sotalol, timolol
Oral anticoagulants	VKA, dabigatran, rivaroxaban, apixaban
Antiplatelets	Aspirin, lopidogrel, prasugrel, ticlopidine, aspirin–dipyridamole, aspirin, dipyridamole alone, cilostazol, ticagrelor
Antidiabetics	Insulin or acarbose, acetahexamide, albiglutide, alogliptin, canagliflozin, chlorpropamide, dapagliflozin, exenatide, glimepiride, glipizide, glyburide, linagliptin, liraglutide, metformin, miglitol, nateglinide, pioglitazone, pramlintide, repaglinide, rosiglitazone, saxagliptin, sitagliptin, tolazamide, tolbutamide, troglitazone
Statins	Atorvastatin, Fluvastatin, Lovastatin, Pitavastatin, Pravastatin , Rosuvastatin, Simvastatin
Oral corticosteroids	Cortisone, hydrocortisone, prednisone, prednisolone, methylprednisolone, triamcinolone, dexamethasone, bethamethasone
Nonselective nonsteroidal anti-inflammatory drugs	Diclofenac, etodolac, flurbiprofen, ketorolac, ibuprofen, indomethacin, meloxicam, naproxen, piroxicam, sulindac, Ketoprofen, Oxaprozin, Nabumetone, Mefanamic acid, Meclofenamate, Fenoprofen, Diflunisal, Tolmetin
Selective COX-2 inhibitors	Celecoxib, rofecoxib, valdecoxib
Opioids	Codeine, fentanyl, hydrocodone, hydromorphone, meperidine, morphine, oxycodone, propoxyphene, oxymorphone, tramadol

Antibiotics	<p>daptomycin, gemifloxacin, telavancin, ceftaroline, fidaxomicin, amoxicillin, ampicillin, bacampicillin, carbenicillin, cloxacillin, dicloxacillin, flucloxacillin, mezlocillin, nafcillin, oxacillin, penicillin g, penicillin v, piperacillin, pivampicillin, pivmecillinam, ticarcillin, cefacetrile, cephalothin, cefadroxil, cefadroxyl, cefalexin, cephalixin, cefaloglycin, cephaloglycin, cefalonium, cephalonium, cefaloridine, cephaloradine, cefalotin, cephalothin, cefapirin, cephalirin, cefatrizine, cefazaflur, cefazedone, cefazolin, cephalolin, cefradine, cephradine, cefroxadine, ceftazidime, cefaclor, cefamandole, cefmetazole, cefonicid, cefotetan, cefoxitin, cefprozil, cefproxil, cefuroxime, cefuzonam, cefcapene, cefdaloxime, cefdinir, cefditoren, cefetamet, cefixime, cefmenoxime, cefodizime, cefotaxime, cefpimizole, cefpodoxime, cefteteram, ceftibuten, ceftiofur, ceftiolene, ceftizoxime, ceftriaxone, cefoperazone, ceftazidime, ceftidine, cefepime, cefluprenam, cefoselis, ceftazidime, cefpirome, ceftiofur, cefquinome, ceftobiprole, ceftazidime, ertapenem, aztreonam, ceftazidime, avibactam, imipenem, imipenem, cilastatin, doripenem, meropenem, cefaclomezine, cefaloram, cefaparole, cefcanel, cefedrolor, cefempidone, cefetizole, cefivitril, cefmatilen, cefmepidium, ceftiofur, cefsumide, cefuracetimide, ceftiofur, erythromycin, clarithromycin, dirithromycin, roxithromycin, telithromycin, clindamycin, lincomycin, pristinamycin, quinupristin, medication, dalfopristin, amikacin, gentamicin, kanamycin, neomycin, netilmicin, paromomycin, streptomycin, tobramycin, flumequine, nalidixic acid, oxolinic acid, piromidic acid, pipemidic acid, rosoxacin, ciprofloxacin, enoxacin, lomefloxacin, nadifloxacin, norfloxacin, ofloxacin, pefloxacin, rufloxacin, balofloxacin, gatifloxacin, grepafloxacin, levofloxacin, vancomycin, teicoplanin, telavancin, linezolid, cycloserine 2, rifampin, rifabutin, rifapentine, rifalazil, bacitracin, polymyxin b, viomycin, oxifloxacin, pazufloxacin, sparfloxacin, temafloxacin, tosufloxacin, besifloxacin, clinafloxacin, gemifloxacin, sitafloxacin, trovafloxacin, prulifloxacin, sulfamethizole, sulfamethoxazole, sulfisoxazole, trimethoprim-sulfamethoxazole, demeclocycline, doxycycline, minocycline, oxytetracycline, tetracycline, tigecycline, chloramphenicol, metronidazole, tinidazole, nitrofurantoin</p>
ACEI= angiotensin converting enzyme inhibitors, ARB= angiotensin II receptor blockers	

Table S4. Sensitivity analysis: Similar performance with composite vs. specific outcomes				
	SAmE-TT2R2 Score		New geriatric score	
	AUC in the training set	AUC in the validation set (AF+VTE)	AUC in the training set	AUC in the validation set (AF+VTE)
Composite outcomes*	0.53	0.59	0.65	0.72
Thrombotic outcomes**	0.56	0.60	0.65	0.72
Bleeding outcomes***	0.49	0.57	0.64	0.71
AF= atrial fibrillation, VTE=venous thromboembolism, AUC= Area Under Receiver Operating Characteristic curve *Including the thrombotic outcomes and bleeding outcomes listed below ** Including ischemic stroke, arterial thromboembolism, deep vein thrombosis, pulmonary embolism per definitions in Table S2 *** Including intracranial and extracranial major bleeding per definitions in Table S2				

Table S5. Hosmer-Lemeshow Goodness-of-fit table of geriatric TTR model in the validation set

Deciles by predicted TTR	N of patients	TTR>70%		TTR≤70%	
		Observed N of events	Expected N of events	Observed N of events	Expected N of events
1	103	10	10.06	93	92.94
2	103	13	15.18	90	87.82
3	103	20	18.49	83	84.51
4	103	24	21.74	79	81.26
5	103	20	25.62	83	77.38
6	103	33	29.58	70	73.42
7	103	36	33.39	67	69.61
8	103	38	38.35	65	64.65
9	103	44	44.89	59	58.11
10	106	55	55.69	51	50.31

Hosmer and Lemeshow goodness-of-fit Test: Chi-squared=3.4, Degree of freedom=8, P value=0.91
TTR =time in therapeutic range

Table S6. Sensitivity analysis: Similar performance with different analysis strategies			
Version of scores		AUC in the training set	AUC in the validation set (AF+VTE)
Original	TTR>70%	0.71	0.68
	Clinical outcome	0.65	0.72
Defining BAP as 365 days	TTR>70%	0.72	0.67
	Clinical outcome	0.65	0.70
Defining new initiation as no use in the 180 days prior to the index date	TTR>70%	0.71	0.69
	Clinical outcome	0.66	0.72
After Box-Cox transformation of TTR	TTR>70%	0.71	0.68
	Clinical outcome	0.65	0.72
After excluding patients with zero TTRs	TTR>70%	0.70	0.66
	Clinical outcome	0.66	0.72
AF= atrial fibrillation, VTE=venous thromboembolism, BAP=baseline assessment period, AUC= Area Under Receiver Operating Characteristic curve			

Table S7. New geriatric prediction model for anticoagulation control quality* based on only variables available in an insurance claims database

Predictors	Coefficients**	s.e.	Predictors	Coefficients**	s.e.
Intercept	0.655	0.021	# of regular medications 5-9	-0.037	0.014
AF vs. VTE	0.002	0.013	# of regular medications ≥10	-0.050	0.020
Sex female vs male	-0.014	0.014	# of regular medications <5	ref	.
Black race	-0.052	0.037	Hospitalization days≥7 days: yes vs no	-0.018	0.016
Non-black non-white races	0.041	0.025	Use of BB	0.026	0.014
White	ref		Use of ACEI	0.027	0.013
CHF	-0.022	0.016	Use of ARB	0.022	0.017
DM	-0.011	0.014	Use of metolazone	-0.115	0.092
Epilepsy	-0.020	0.029	Use of opioids	-0.022	0.016
Cancer	-0.027	0.012	Use of statins	-0.033	0.015
CR_Dementia	-0.008	0.019	Use of acetaminophen	-0.006	0.016
Renal dysfunction	-0.060	0.014	Use of antibiotics	-0.022	0.013
Prior bleeding‡	-0.029	0.015	Use of oral steroids	-0.018	0.017
COPD or asthma	-0.004	0.014	Influenza vaccine	0.021	0.013
Pneumonia	-0.027	0.016	PSA Test	0.043	0.018
Prior stroke	-0.016	0.015	Colonoscopy	-0.020	0.023
Drug abuse	-0.057	0.058	Mammography	0.068	0.023
Chronic liver disease	-0.039	0.022	Pap smear	-0.073	0.039
Psychosis	-0.019	0.026	Falls	-0.011	0.022
Peripheral vascular disease	-0.034	0.017	Fractures	-0.029	0.021
Hyperlipidemia	0.032	0.015	Parkinson's disease	0.093	0.045

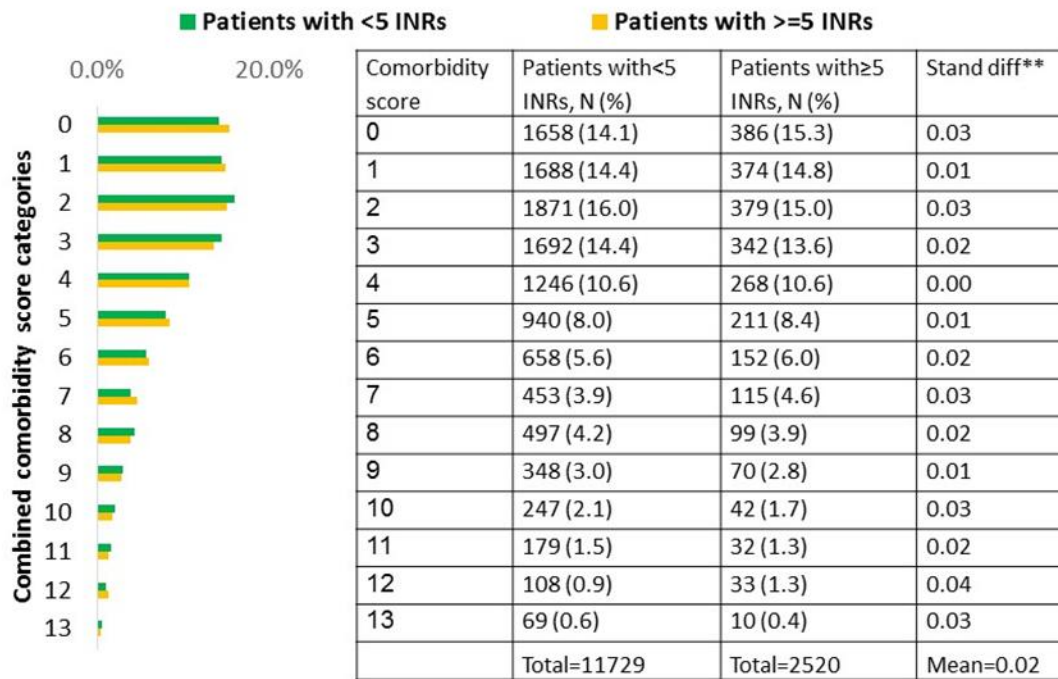
* Quantified by International Normalized Ratio (INR) time in therapeutic range (TTR).

** unstandardized coefficients based on a Lasso regression model including all the variables listed in this table.

***Including all the major upper and lower gastrointestinal and other extracranial bleeding events.

AF= atrial fibrillation, VTE=venous thromboembolism, CHF= congestive heart failure, DM= diabetes mellitus, COPD=chronic obstructive pulmonary disease, BB=beta-blocker, ACEI= angiotensin converting enzyme inhibitors, ARB= Angiotensin II receptor blockers, PSA= Prostate-specific antigen

Figure S1. Similar distribution of combined comorbidity* score in those with vs. without 5 INRs in the study EHR system



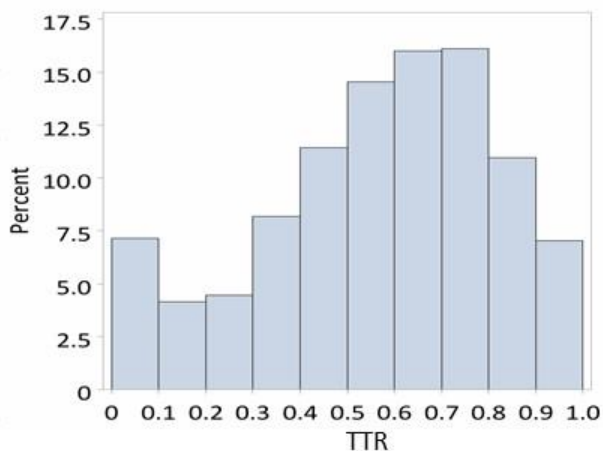
INR= International Normalized Ratio

* J Clin Epidemiol. 2011 Jul;64(7):749-59. **Stand diff= Standardized difference. Comorbidity score categories with cell size <5 were not presented here.

Figure S2. The distribution of TTR in the training & validation sets

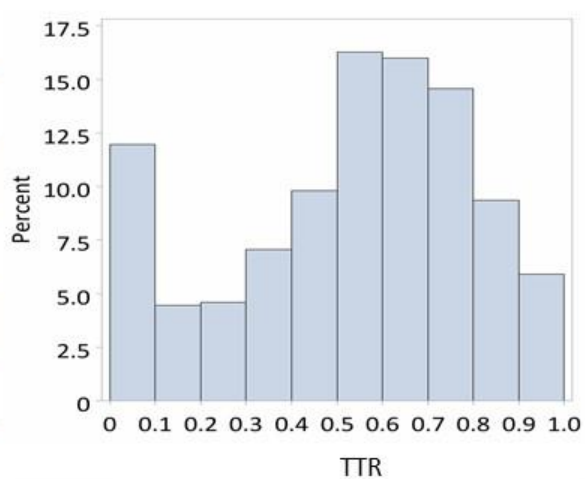
A. Training set

Mean	0.56
SD	0.25
25 percentile	0.41
Median	0.60
75 percentile	0.75



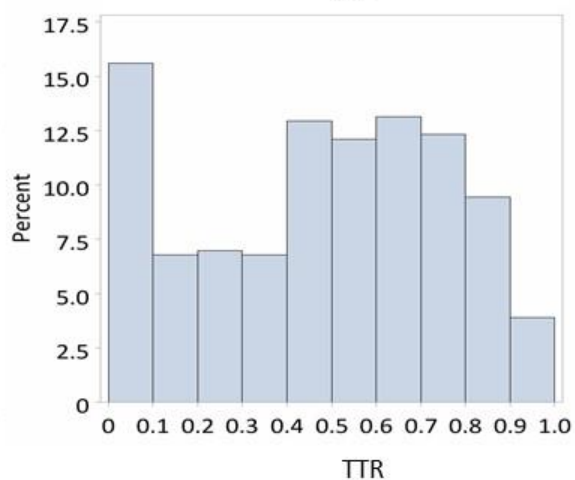
B. AF validation set

Mean	0.53
SD	0.27
25 percentile	0.36
Median	0.57
75 percentile	0.73



C. VTE validation set

Mean	0.47
SD	0.28
25 percentile	0.24
Median	0.50
75 percentile	0.70



TTR= time in therapeutic range, SD=standard deviation ,
 AF= atrial fibrillation, VTE=venous thromboembolism

Supplemental References:

1. Zhou L, Plasek JM, Mahoney LM, Karipineni N, Chang F, Yan X, Chang F, Dimaggio D, Goldman DS, Rocha RA. Using Medical Text Extraction, Reasoning and Mapping System (MTERMS) to process medication information in outpatient clinical notes. *AMIA Annu Symp Proc.* 2011;2011:1639-48.
2. Navathe AS, Zhong F, Lei VJ, Chang FY, Sordo M, Topaz M, Navathe SB, Rocha RA, Zhou L. Hospital Readmission and Social Risk Factors Identified from Physician Notes. *Health Serv Res.* 2017. Mar 13. [Epub ahead of print]
3. McDonald CJ, Huff SM, Suico JG, Hill G, Leavelle D, Aller R, Forrey A, Mercer K, DeMoor G, Hook J, Williams W, Case J, Maloney P. LOINC, a universal standard for identifying laboratory observations: a 5-year update. *Clin Chem.* 2003;49:624-33.
4. Apostolakis S, Sullivan RM, Olshansky B, Lip GY. Factors affecting quality of anticoagulation control among patients with atrial fibrillation on warfarin: the SAME-TT(2)R(2) score. *Chest.* 2013;144:1555-63.
5. Roumie CL, Mitchel E, Gideon PS, Varas-Lorenzo C, Castellsague J, Griffin MR. Validation of ICD-9 codes with a high positive predictive value for incident strokes resulting in hospitalization using Medicaid health data. *Pharmacoepidemiol Drug Saf.* 2008;17:20-6.
6. Andrade SE, Harrold LR, Tjia J, Cutrona SL, Saczynski JS, Dodd KS, Goldberg RJ, Gurwitz JH. A systematic review of validated methods for identifying cerebrovascular accident or transient ischemic attack using administrative data. *Pharmacoepidemiol Drug Saf.* 2012;21 Suppl 1:100-28.
7. Birman-Deych E, Waterman AD, Yan Y, Nilasena DS, Radford MJ, Gage BF. Accuracy of ICD-9-CM codes for identifying cardiovascular and stroke risk factors. *Med Care.* 2005;43:480-5.
8. Tamariz L, Harkins T, Nair V. A systematic review of validated methods for identifying venous thromboembolism using administrative and claims data. *Pharmacoepidemiol Drug Saf.* 2012;21 Suppl 1:154-62.
9. Cushman M, Tsai AW, White RH, Heckbert SR, Rosamond WD, Enright P, Folsom AR. Deep vein thrombosis and pulmonary embolism in two cohorts: the longitudinal investigation of thromboembolism etiology. *Am J Med.* 2004;117:19-25.
10. Wahl PM, Rodgers K, Schneeweiss S, Gage BF, Butler J, Wilmer C, Nash M, Esper G, Gitlin N, Osborn N, Short LJ, Bohn RL. Validation of claims-based diagnostic and procedure codes for cardiovascular and gastrointestinal serious adverse events in a commercially-insured population. *Pharmacoepidemiol Drug Saf.* 2010;19:596-603.

11. Cunningham A, Stein CM, Chung CP, Daugherty JR, Smalley WE, Ray WA. An automated database case definition for serious bleeding related to oral anticoagulant use. *Pharmacoepidemiol Drug Saf.* 2011;20:560-6.
12. Molnar AO, Coca SG, Devereaux PJ, Jain AK, Kitchlu A, Luo J, Parikh CR, Paterson JM, Siddiqui N, Wald R, Walsh M, Garg AX. Statin use associates with a lower incidence of acute kidney injury after major elective surgery. *J Am Soc Nephrol.* 2011;22:939-46.
13. Winkelmayr WC, Schneeweiss S, Mogun H, Patrick AR, Avorn J, Solomon DH. Identification of individuals with CKD from Medicare claims data: a validation study. *Am J Kidney Dis.* 2005;46:225-32.