Absolute risk and predictors of the growth of acute spontaneous intracerebral haemorrhage

Citation for published version:

Digital Object Identifier (DOI):
10.1016/S1474-4422(18)30253-9

Link:
Link to publication record in Edinburgh Research Explorer

Document Version:
Version created as part of publication process; publisher's layout; not normally made publicly available

Published In:
The Lancet Neurology

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Absolute risk and predictors of the growth of acute spontaneous intracerebral haemorrhage: a systematic review and meta-analysis of individual patient data


Summary
Background Intracerebral haemorrhage growth is associated with poor clinical outcome and is a therapeutic target for improving outcome. We aimed to determine the absolute risk and predictors of intracerebral haemorrhage growth, develop and validate prediction models, and evaluate the added value of CT angiography.

Methods In a systematic review of OVID MEDLINE—with additional hand-searching of relevant studies—bibliographies—from Jan 1, 1970, to Dec 31, 2015, we identified observational cohorts and randomised trials with repeat scanning protocols that included at least ten patients with acute intracerebral haemorrhage. We sought individual patient-level data from corresponding authors for patients aged 18 years or older with data available from brain imaging initially done 0·5–24 h and repeated fewer than 6 days after symptom onset, who had baseline intracerebral haemorrhage volume of less than 150 mL, and did not undergo acute treatment that might reduce intracerebral haemorrhage volume. We estimated the absolute risk and predictors of the primary outcome of intracerebral haemorrhage growth (defined as >6 mL increase in intracerebral haemorrhage volume on repeat imaging) using multivariable logistic regression models in development and validation cohorts in four subgroups of patients, using a hierarchical approach: patients not taking antiplatelet therapy at intracerebral haemorrhage onset (who constituted the largest subgroup), patients taking anticoagulant therapy at intracerebral haemorrhage onset, patients from cohorts that included at least some patients taking anticoagulant therapy at intracerebral haemorrhage onset, and patients for whom both information about anticoagulant therapy at intracerebral haemorrhage onset and spot sign on acute CT angiography were known.

Findings Of 4191 studies identified, 77 were eligible for inclusion. Overall, 36 (47%) cohorts provided data on 5435 eligible patients. 5076 of these patients were not taking antiplatelet therapy at symptom onset (median age 67 years, IQR 56–76), of whom 1009 (20%) had intracerebral haemorrhage growth. Multivariable models of patients with data on antiplatelet therapy use, data on anticoagulant therapy use, and assessment of CT angiography spot sign at symptom onset showed that time from symptom onset to baseline imaging (odds ratio 0·50, 95% CI 0·36–0·70; p<0·0001), intracerebral haemorrhage volume on baseline imaging (7·18, 4·46–11·60; p<0·0001), antplatelet use (1·68, 1·06–2·66; p=0·026), and anticoagulant use (3·48, 1·96–6·16; p<0·0001) were independent predictors of intracerebral haemorrhage growth (C-index 0·78, 95% CI 0·75–0·82). Addition of CT angiography spot sign (odds ratio 4·46, 95% CI 2·95–6·75; p<0·0001) to the model increased the C-index by 0·05 (95% CI 0·03–0·07).

Interpretation In this large patient-level meta-analysis, models using four or five predictors had acceptable to good discrimination. These models could inform the location and frequency of observations on patients in clinical practice, explain treatment effects in prior randomised trials, and guide the design of future trials.

Funding UK Medical Research Council and British Heart Foundation.

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Research in context

Evidence before this study

We did a systematic review of studies of the risk of intracerebral haemorrhage growth, and associations with it, published in OVID MEDLINE (from Jan 1, 1970, to Dec 31, 2015) using a comprehensive search strategy, limited to humans, combining terms for intracerebral haemorrhage (“exp basal ganglia hemorrhage”), “intracranial hemorrhages/”, “cerebral hemorrhage/”, “intracranial hemorrhage, hypertensive/”, and other text words) with text words suggesting growth (“expansion”, “growth”, or “enlargement”), with no language restrictions. When we updated the search to March 1, 2018, we identified reports of five new cohorts, representing a maximum of a 10% increase in the number of eligible patients compared with those from the 36 cohorts that provided individual patient data in this meta-analysis. We did not include these five new cohorts in our analyses. Intracerebral haemorrhage growth risk is known to be highest soon after intracerebral haemorrhage symptom onset, but its absolute risks over time and by baseline volume are unclear. Studies have identified several risk factors associated with intracerebral haemorrhage growth, but many associations are not consistent across studies, and the predictive values of these risk factors remain to be determined.

Added value of this study

Our systematic review led to the pooling of 5435 eligible patients from 36 cohorts, which is, to the best of our knowledge, the largest patient-level meta-analysis to explore the absolute risk and predictors of intracerebral haemorrhage growth. We found that the risks of growth over time and by baseline intracerebral haemorrhage volume were not linear. The sample size enabled us to model these associations with good precision and construct and validate multivariable models adjusted for 13 categorical or continuous covariates. Four predictors (time from symptom onset to baseline imaging, intracerebral haemorrhage volume on baseline imaging, antiplatelet use, and anticoagulant use) were independent predictors of intracerebral haemorrhage growth (C-index 0·78, 95% CI 0·75–0·82). Addition of information about the presence of spot sign on CT angiography to the model increased the C-index by just 0·05 (95% CI 0·03–0·07).

Implications of all the available evidence

Models using four or five predictors that are simple to collect had acceptable to good discrimination for predicting intracerebral haemorrhage growth, which was slightly improved by the addition of information on spot sign from CT angiography. These models could guide the monitoring of patients at risk of clinical deterioration as well as the interpretation and investigation of treatment effects in randomised trials.

Introduction

Haemorrhagic stroke is responsible for around 11% of strokes in high-income countries but 22% of strokes in low-income and middle-income countries,1 where 75% of deaths due to haemorrhagic stroke occur.2 Spontaneous (non-traumatic) intracerebral haemorrhage is the most frequent type of haemorrhagic stroke and has the worst outcome: almost half of patients die within the first month and 80% of survivors are dependent on a caregiver.3

Intracerebral haemorrhage volume increases after vessel rupture and growth can continue after intracerebral haemorrhage is first diagnosed on brain imaging. Intracerebral haemorrhage growth is associated with poor clinical outcome.4 Therefore, immediately after confirmation of intracerebral haemorrhage diagnosis on brain imaging, accurate prediction of the risk of later intracerebral haemorrhage growth could help to target patients’ monitoring, treatment and transfer to specialist care, and the design and interpretation of randomised trials of treatments to limit intracerebral haemorrhage growth.5

The timing of the first brain imaging done after intracerebral haemorrhage onset and the intracerebral haemorrhage volume found on imaging are two consistently identified risk factors for intracerebral haemorrhage growth, although the association of other potential risk factors has been inconsistent in many small observational studies. Interest has grown in whether a so-called spot sign due to contrast extravasation on additional angiography at the time of diagnostic imaging is a predictor of intracerebral haemorrhage growth.6 There are several multivariable prediction models for intracerebral haemorrhage growth,7–11 but the identified predictors have varied across models, and several have relied on CT angiography, which is not readily available in low-income and middle-income countries. Identifying more accurate predictors of intracerebral haemorrhage growth is recognised to be a research priority.12

Therefore, we aimed to identify the risk and predictors of acute intracerebral haemorrhage growth, develop and validate prediction models that could be used worldwide, and evaluate the added value of CT angiography.

Methods

Search strategy and selection criteria

We conducted a systematic review to identify studies of intracerebral haemorrhage growth that would share individual patient data for a patient-level meta-analysis of the absolute risks and predictors of intracerebral haemorrhage growth.13 A prespecified protocol (finalised on June 20, 2013, and not registered; appendix) guided our data collection and analyses.
One author (JF) identified potentially eligible cohorts by searching OVID MEDLINE from Jan 1, 1970, to Dec 31, 2015, using a comprehensive search strategy (appendix); hand-searching relevant studies’ bibliographies; contacting authors of collaborating studies; and accessing patient-level data from eligible cohorts in the Virtual International Stroke Trials Archive. We included the largest single report of any observational or randomised cohort—regardless of language of publication—if it included at least ten eligible patients with acute intracerebral haemorrhage who had brain imaging (by CT with or without angiography or by MRI) to diagnose intracerebral haemorrhage and used a predefined protocol for repeat imaging (done regardless of clinical need), which would minimise the risks of selection and information biases about intracerebral haemorrhage growth.

We included patients from these cohorts if they were aged 18 years or older; had non-traumatic intracerebral haemorrhage that was probably due to cerebral small vessel disease and not secondary to an underlying structural cause identified by brain imaging; had data available from brain imaging initially done 0·5–24 h and repeated fewer than 6 days after symptom onset; had baseline intracerebral haemorrhage volume of less than 150 mL; and did not undergo acute treatment that might reduce intracerebral haemorrhage volume (ie, surgical evacuation,15 haemostatic therapy,3 or blood pressure lowering30). We excluded patients if the time from symptom onset to baseline imaging was not known in hours or if they had not been included in the published report of their cohort.

We emailed our protocol and an invitation to collaborate to the corresponding authors of cohorts that were eligible for inclusion, followed by one reminder. We included cohorts if corresponding authors of studies reporting them confirmed their eligibility and provided patient-level data on eligibility criteria and other variables at baseline, information on type and timing of baseline and repeat brain imaging, intracerebral haemorrhage characteristics (location, volume on baseline and repeat imaging, presence of intraventricular haemorrhage), and the presence of the spot sign on CT angiography if done (appendix).35

Research ethics committees or other entities overseeing the use of patients’ data had approved the collaborating cohorts. Cohorts shared only anonymised data, so neither individual consent nor specific approval for this individual patient data meta-analysis were required.

Data analysis
We used reports of the included cohorts to categorise their method of intracerebral haemorrhage volume measurement as a cohort-level characteristic into either the manual ABC/2 method36 or an automated or semi-automated planimetric method.37 We assessed risk of bias across cohorts by identifying the studies that did not meet our eligibility criteria, did not share data, or did not provide data on a sufficient number of the variables of interest (appendix). We checked data completeness and consistency within each cohort and resolved any queries directly with the relevant collaborators. We standardised the format, coding, and units of measurement of variables to maximise the number available for analysis in all cohorts. We did not use or request aggregate data from cohorts that did not share patient-level data.

We prespecified that the primary outcome measure of intracerebral haemorrhage growth would be an increase in intracerebral haemorrhage volume between baseline and repeat imaging of more than 6 mL; we chose an absolute measure of intracerebral haemorrhage growth in volume because such measures seem to have higher positive predictive values for more severe clinical outcomes than does the combination of absolute or relative increases in intracerebral haemorrhage volume (eg, >33%).19

We prespecified the variables that might be predictors of intracerebral haemorrhage growth in our protocol (appendix) on the basis of their clinical relevance, likelihood of being associated with outcome, and reliability and accuracy of measurement (appendix). To these variables, we added history of liver disease and history of stroke; we also added CT angiography spot sign in view of the increasing interest in its role as a predictor since the protocol had originally been written (appendix).4 Of these prespecified variables, we selected potential predictors on the basis of their completeness and availability at the time of diagnosis in the available cohorts and the extent to which their selection maximised the total sample size available for multivariable analyses. Many cohorts excluded patients taking anticoagulant therapy at onset and only a few cohorts conducted CT angiography, so we took a hierarchical approach to investigating univariable and multivariable associations and predictors of intracerebral haemorrhage growth.

First, we analysed patients not taking anticoagulant therapy at intracerebral haemorrhage symptom onset because they constituted the vast majority of the included cohorts. In this dataset, we examined the associations between intracerebral haemorrhage growth and a subset of the variables, which were chosen on the basis of their completeness and availability at the time of intracerebral haemorrhage diagnosis in the participating cohorts. We visually inspected plots of cohort-specific estimates of association for each variable to exclude major heterogeneity. We then used a one-stage approach to meta-analysis to obtain unadjusted and adjusted estimates pooled across the cohorts using logistic regression models with random intercepts and random coefficients. For all continuous predictors, we used either a linear term or, where there was strong evidence (p<0·01) of non-linearity on the log-odds scale, a fractional polynomial. We described the univariable associations between intracerebral haemorrhage growth and the two of the continuous variables (time to baseline...
imaging and intracerebral haemorrhage volume at baseline) by plotting the predicted probability of intracerebral haemorrhage growth derived from the model against the predictor. For the remaining continuous variables, we quantified the unadjusted and adjusted associations using the odds ratio for the upper quartile compared with the lower quartile based on the fitted linear or fractional polynomial terms in the logistic regression model. We had a sufficient sample size to split those patients who were not taking anticoagulant therapy by contributing cohort into two datasets: one to develop a prediction model and another to validate its performance. We did this temporal validation with patients from earlier cohorts (1994–2007) allocated to the development dataset and patients from more recent cohorts (2008–15) allocated to the validation dataset. We chose a subset of potential predictors for entry into a multivariable model on the basis of their combined availability in the development dataset and the number of patients with intracerebral haemorrhage growth (to avoid overfitting), without considering the results of the unadjusted and adjusted associations between each predictor and intracerebral haemorrhage growth. We did not examine interactions between other covariates and these associations. We derived a prediction index for intracerebral haemorrhage growth with the predictors that remained in a multivariable logistic regression model after backwards elimination. We assessed the performance of the prediction model using calibration plots of predicted versus observed probabilities, receiver operating characteristic curves, and the C-index to assess discrimination in both the development and validation datasets and in patients from cohorts that included patients taking anticoagulant therapy at intracerebral haemorrhage onset.

Second, we assessed the performance of the prediction model in patients taking anticoagulant therapy at intracerebral haemorrhage onset.

Third, we split by contributing cohort those patients from cohorts that included at least some patients taking anticoagulant therapy at intracerebral haemorrhage onset into one dataset to develop a prediction model and another to validate its performance (using temporal validation, as described above). We considered the same subset of potential predictors as for the first prediction model, with the addition of anticoagulant therapy use in the development dataset and patients from more recent cohorts of ten patients or fewer. We also included and distinguished patients taking anticoagulant therapy at onset, we assessed whether spot sign presence was independently associated with intracerebral haemorrhage growth and assessed the performance using the same approaches as for the first prediction model.

Fourth, in cohorts that included at least some patients with data available on the spot sign identified by CT angiography and that also included and distinguished patients taking anticoagulant therapy at onset, we assessed whether spot sign presence was independently associated with intracerebral haemorrhage growth and the predictive performance when it was added to the predictors in the second prediction model.

We did a prespecified sensitivity analysis to compare our findings using a definition of intracerebral haemorrhage growth as an absolute increase of more than 6 mL versus an absolute increase of more than 6 mL.
or a relative increase of more than 33% in intracerebral haemorrhage volume. We did post-hoc sensitivity analyses to compare associations between time from intracerebral haemorrhage symptom onset to baseline brain imaging and intracerebral haemorrhage volume on baseline imaging with intracerebral haemorrhage growth in cohorts using ABC/2 versus planimetric methods of measuring intracerebral haemorrhage volume and in cohorts from earlier versus later time periods.

Analyses were done using SAS software version 9.4 (SAS Institute) and Stata version 12.1 (StataCorp).

Table 1: Characteristics of patients included in the four datasets for meta-analysis

<table>
<thead>
<tr>
<th></th>
<th>Not taking anticoagulant therapy (n=5076)</th>
<th>Taking anticoagulant therapy (n=351)</th>
<th>From cohorts with some patients taking anticoagulant therapy (n=3550)</th>
<th>CT angiography (n=868)</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Sex</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Female</td>
<td>1971/4884 (40%)</td>
<td>135 (38%)</td>
<td>1449 (41%)</td>
<td>379 (44%)</td>
</tr>
<tr>
<td>Male</td>
<td>2913/4884 (60%)</td>
<td>216 (62%)</td>
<td>2101 (59%)</td>
<td>489 (56%)</td>
</tr>
<tr>
<td><strong>Age, years</strong></td>
<td>67 (56–76)</td>
<td>76 (69–82)</td>
<td>69 (58–78)</td>
<td>70 (57–79)</td>
</tr>
<tr>
<td><strong>Previous stroke</strong></td>
<td>607/4560 (13%)</td>
<td>77/317 (24%)</td>
<td>483/3308 (15%)</td>
<td>97/829 (12%)</td>
</tr>
<tr>
<td><strong>Previous intracerebral haemorrhage</strong></td>
<td>179/2753 (7%)</td>
<td>9/246 (4%)</td>
<td>113/2051 (6%)</td>
<td>29/805 (4%)</td>
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<tr>
<td><strong>Previous ischaemic stroke</strong></td>
<td>246/2755 (9%)</td>
<td>53/246 (22%)</td>
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<td>70/805 (9%)</td>
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<tr>
<td><strong>History of hypertension</strong></td>
<td>3787/5050 (75%)</td>
<td>291 (83%)</td>
<td>2739/3547 (77%)</td>
<td>137/807 (17%)</td>
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<tr>
<td><strong>History of diabetes mellitus</strong></td>
<td>727/4197 (17%)</td>
<td>82/343 (24%)</td>
<td>626/3475 (18%)</td>
<td>73/554 (13%)</td>
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<tr>
<td><strong>History of liver disease</strong></td>
<td>256/3360 (8%)</td>
<td>14/220 (6%)</td>
<td>96/1946 (5%)</td>
<td>15/360 (4%)</td>
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<tr>
<td><strong>History of excessive alcohol consumption†</strong></td>
<td>568/3091 (18%)</td>
<td>20/177 (11%)</td>
<td>221/1455 (15%)</td>
<td>72/866 (8%)</td>
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<tr>
<td><strong>Anticoagulant therapy at symptom onset</strong></td>
<td>0</td>
<td>351 (100%)</td>
<td>349/3547 (10%)</td>
<td>87/841 (10%)</td>
</tr>
<tr>
<td><strong>Systolic blood pressure at presentation, mm Hg</strong></td>
<td>177 (158–198); n=4882</td>
<td>170 (147–190); n=320</td>
<td>177 (157–197); n=3303</td>
<td>175 (150–200); n=860</td>
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Data are n (%), n/N (%), or median (IQR). NIHSS=National Institutes of Health Stroke Scale. †Definition of excessive consumption varied across cohorts.

**Results**

We screened 4191 studies identified by our searches, assessed 167 for eligibility, invited 77 eligible cohorts to share data, and obtained patient-level data from 36 (47%) cohorts involving 6428 patients with repeat brain imaging after intracerebral haemorrhage between 1985 and 2015 (no data up to 1984 were obtained; figure 1; appendix). Countries classified as high income by the World Bank contributed to 26 (72%) of 36 collaborating cohorts versus 30 (73%) of 41 eligible cohorts that did not collaborate. Planimetric methods of measuring intracerebral haemorrhage volume were used by 19 (53%) for publication. The data were available to all authors on request. The corresponding author had final responsibility for the decision to submit for publication.
After confirming the integrity of the data from eligible cohorts and excluding patients who were ineligible, we created a dataset of 3435 patients (appendix), from which we identified four groups of patients for further analysis: 3076 patients not taking anticoagulant therapy at intracerebral haemorrhage onset, 351 patients taking anticoagulant therapy at intracerebral haemorrhage onset, 3550 patients from cohorts that included at least some patients taking anticoagulant therapy at intracerebral haemorrhage onset and spot sign on acute CT angiography. More than 80% of patients in all groups had repeat imaging done within 48 h of intracerebral haemorrhage onset and less than 2% of patients had repeat imaging done more than 4 days after intracerebral haemorrhage onset (appendix).

When assessing the two variables with non-linear associations, we found that in patients not taking anticoagulant therapy at intracerebral haemorrhage onset, the predicted probability of intracerebral haemorrhage growth declined with increasing time from intracerebral haemorrhage symptom onset to baseline imaging: the rate of decline was steepest 0·5–3 h after intracerebral haemorrhage symptom onset (figure 2A). The predicted probability of intracerebral haemorrhage growth increased with increasing intracerebral haemorrhage volume on baseline brain imaging and peaked at about 75 mL, above which it declined (figure 2B). We aimed to quantify the associations between 17 additional variables and the occurrence of intracerebral haemorrhage growth (appendix). There were too few patients with data for six variables (previous intracerebral haemorrhage, previous ischaemic stroke, history of liver disease, history of excessive alcohol consumption, platelet count at presentation, and National Institutes of Health Stroke Scale [NIHSS] score at presentation). Therefore, we selected 13 of the 19 variables as potential predictors for a multivariable model in patients not taking anticoagulant therapy, on the basis of maximising the number of predictors being considered while also maximising the number of patients with complete data for all the predictors chosen for the subset: time from symptom onset to baseline imaging, intracerebral haemorrhage volume on baseline imaging, sex, age, previous stroke, history of hypertension, history of diabetes, antiplatelet therapy at symptom onset, systolic blood pressure at presentation, blood glucose at presentation, Glasgow Coma Scale score at presentation, intracerebral haemorrhage location on baseline scan, and intraventricular haemorrhage on baseline scan. We restricted all further analyses to datasets of patients with complete data on these 13 potential predictors.

3479 patients who were not taking anticoagulant therapy at intracerebral haemorrhage onset had data available for the 13 predictors. We developed a prediction model for intracerebral haemorrhage growth using a dataset of 2534 (73%) of these patients from 18 earlier cohorts (i.e., 1994–2007; appendix). From the 13 potential predictors considered, three significant predictors constituted the final model (table 2):

Predicted probability of intracerebral haemorrhage growth

$$\frac{1}{1 + e^{-[\text{predicted probability}]}}$$
where the predictive index (PI) is given by

\[-4.254 - 0.196 \text{time} - 0.0754 \text{volume} + 1.186 \text{volume} + 0.320 \text{antiplatelet}\]

with time measured in hours, volume measured in mL, and antiplatelet an indicator variable for antiplatelet therapy at intracerebral haemorrhage onset taking values 1 for yes and 0 for no.

This first prediction model had good calibration (appendix) and its discrimination was good in both the development dataset (C-index 0.75, 95% CI 0.72–0.77) and the temporal validation dataset of 945 (27%) patients from six later cohorts (0.76, 0.73–0.79). This prediction model, derived in patients who were not taking anticoagulant therapy at symptom onset, underestimated the probability of intracerebral haemorrhage growth in the 351 patients in 21 cohorts who were taking anticoagulant therapy at symptom onset (appendix), but its discrimination remained good (0.73, 0.68–0.79).

We also developed a prediction model for intracerebral haemorrhage growth using a dataset of 2381 patients from 10 cohorts that included at least some patients taking anticoagulant therapy at intracerebral haemorrhage onset (appendix). From the 13 potential predictors plus anticoagulant therapy at intracerebral haemorrhage onset, four predictors constituted the final model (table 2), where PI is given by

\[-4.426 - 0.230 \text{time} - 0.0776 \text{volume} + 1.196 \text{volume} + 0.310 \text{antiplatelet} + 1.065 \text{anticoagulant}\]

where anticoagulant is an indicator variable for anticoagulant therapy at intracerebral haemorrhage onset taking values 1 for yes and 0 for no.

This second prediction model was well calibrated (appendix) and its discrimination was good in both the development dataset (C-index 0.75, 95% CI 0.73–0.78) and the validation dataset of 895 patients from five cohorts (0.74, 0.71–0.78).

Finally, to assess the additional predictive value of spot sign on CT angiography, we assessed the performance of the second and third prediction models at different thresholds of predicted probability of intracerebral haemorrhage growth and found very few significant differences in sensitivity, specificity, positive predictive value, and negative predictive value (appendix).

In a prespecified sensitivity analysis, when we defined intracerebral haemorrhage growth as an absolute increase of more than 6 mL or a relative increase of more than 33% in intracerebral haemorrhage volume between baseline and follow-up imaging, the direction, strength, and significance of the adjusted associations between almost all predictors and intracerebral haemorrhage growth remained the same (appendix), and the C-index of our second prediction model improved from 0.71 (95% CI 0.67–0.75) to 0.76 (0.72–0.80) with the addition of information from CT angiography (appendix). In a post-hoc sensitivity analysis, we found no evidence that the risk of intracerebral haemorrhage growth according to time from symptom onset to baseline imaging or according to intracerebral haemorrhage volume on baseline imaging differed by cohort epoch or volumetric method used (appendix).

### Discussion

This collaborative meta-analysis evaluated 19 covariates in one or more analyses of predictors of intracerebral haemorrhage growth from 5435 eligible patients in 36 cohorts. We identified novel non-linear associations between the probability of intracerebral haemorrhage growth and both the time from symptom onset to baseline imaging and baseline intracerebral haemorrhage volume. We showed that only four predictors that are simple to collect (time from symptom onset to baseline imaging, mL†

Table 2: Multivariable models of predictors of intracerebral haemorrhage growth >6 mL.

<table>
<thead>
<tr>
<th>Comparison</th>
<th>Odds ratio (95% CI)</th>
<th>p value</th>
</tr>
</thead>
<tbody>
<tr>
<td>Time from symptom onset to baseline imaging, h†</td>
<td>4.3 vs 2</td>
<td>0.65 (0.51–0.82)</td>
</tr>
<tr>
<td>Intracerebral haemorrhage volume on baseline imaging, mL†</td>
<td>28 vs 7</td>
<td>4.73 (1.81–5.87)</td>
</tr>
<tr>
<td>Antiplatelet therapy at symptom onset</td>
<td>Yes vs no</td>
<td>1.38 (1.06–1.79)</td>
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*Data were calculated on 2534 patients from 18 cohorts (appendix). †The odds ratios for time from symptom onset to baseline imaging and intracerebral haemorrhage volume on baseline imaging are for upper quartile compared with lower quartile. ‡Data were calculated on 2381 patients from ten cohorts (appendix).
Since 2011, there has been growing interest in use of the CT angiography spot sign for predicting intracerebral haemorrhage growth, but the clinical utility of the spot sign on CT angiography has been controversial. We studied this association using a large number of widely available predictors from geographically diverse cohorts to develop and externally validate prediction models involving simple predictors that could be used in any health-care setting, as well as the added value of CT angiography in high-income countries. We minimised the risk of selection and information biases by restricting eligibility to cohorts that had defined when they would repeat brain imaging soon after intracerebral haemorrhage onset in all survivors and not according to clinical need alone.

Although our study was large, only half of the investigators of the available cohorts shared patient-level data. Most cohorts were assembled in high-income countries. A shortage of data on the following variables precluded their inclusion in our prediction models: previous intracerebral haemorrhage, previous ischaemic stroke, history of liver disease, history of excessive alcohol consumption, platelet count at presentation, and NIHSS score at presentation. Since the end of the literature search that defined inclusion in our analyses, our update of the search to March 1, 2018, identified reports of five new cohorts involving 669 patients, representing a maximum that defined inclusion in our analyses, our update of the search to March 1, 2018, identified reports of five new cohorts involving 669 patients, representing a maximum of a 10% increase over the 6428 patients from 36 cohorts that provided individual patient data. Nonetheless, the sample size we achieved allowed us to develop and validate prediction models involving a large number of widely available predictors, without omitting any predictors that had been identified by previous prediction models. Included cohorts with data collected in the 1990s might not have used multiple-row detector array technology and digitisation, which might have affected their accuracy of imaging, intracerebral haemorrhage volume on baseline imaging, antplatelet use, and anticoagulant use) were independently associated with intracerebral haemorrhage growth in multivariable models, and a prediction model that we developed using these predictors not only had good calibration and discrimination but also done well in an external validation dataset. The addition of information about the presence of spot sign on CT angiography to this prediction model gave a small increase in discrimination.

Although many studies have investigated unadjusted and adjusted associations between a wide variety of clinical, blood, genetic, imaging, and pharmacological factors and the occurrence of intracerebral haemorrhage growth, only a few prediction models have been developed and the predictors used have varied considerably.1,2,5-11,51 Since 2011, there has been growing interest in use of the spot sign on CT angiography for predicting intracerebral haemorrhage growth, but the clinical utility of the small increase in discrimination that resource-intensive advanced vascular imaging adds to simple clinical and imaging predictors that are available worldwide is unclear.

The strengths of this study include its large sample size and availability of many predictors from geographically diverse cohorts to develop and externally validate prediction models involving simple predictors that could be used in any health-care setting, as well as the added value of CT angiography in high-income countries. We minimised the risk of selection and information biases by restricting eligibility to cohorts that had defined when they would repeat brain imaging soon after intracerebral haemorrhage onset in all survivors and not according to clinical need alone.

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intracerebral haemorrhage volume measurement, although there was no evidence that our findings differed by cohort epoch in sensitivity analyses. 19 (53%) of 36 cohorts used planimetric methods to estimate intracerebral haemorrhage volume but 17 (47%) of 36 cohorts used the ABC/2 method (which can marginally overestimate intracerebral haemorrhage volume46), although we found no evidence that our findings differed by volumetric method in sensitivity analyses. Since these cohorts were studied, a variety of new imaging signs (eg, density, irregularity, fluid level, hypodensity, island, satellite, swirl56 blend, and black hole57) have been described, but we were unable to evaluate them because they were not collected by the collaborating cohorts and we could not re-evaluate patients’ imaging. However, our simple prediction models provide the basis upon which the added value of these new signs can be assessed, as we have done for the CT angiography spot sign.

We found that the rate of decline in the probability of intracerebral haemorrhage growth was steepest during the 0–5–3 h after intracerebral haemorrhage symptom onset and that the predicted probability of intracerebral haemorrhage growth peaked at an intracerebral haemorrhage volume of about 75 mL. These findings could in part explain the neutral results of recent randomised trials of acute interventions designed to limit intracerebral haemorrhage growth, which enrolled many patients towards or beyond the time of greatest risk of intracerebral haemorrhage growth and most patients had small intracerebral haemorrhages at low probability of growth. For example, the average time to randomisation after intracerebral haemorrhage symptom onset and average intracerebral haemorrhage volume were 3·7 h and 13 mL in TICH2,28 3·7 h and 11 mL in INTERACT2,21 3·1 h and 10 mL in ATACH2,19 and 2·7 h and 22–24 mL in FAST.45 In particular, our findings about the association between time after intracerebral haemorrhage symptom onset and the probability of intracerebral haemorrhage growth emphasise the importance of extremely rapid assessment, investigation, and randomisation in future trials of therapies to improve outcome by limiting intracerebral haemorrhage growth.

The prediction models that we have developed could be useful in clinical practice for predicting the risk of intracerebral haemorrhage growth, which is recommended in the emergency assessment of acute intracerebral haemorrhage. The clinically useful threshold for the predicted probability of intracerebral haemorrhage growth will vary according to its desired accuracy (appendix), the clinical setting, and future therapeutic advances, such that our models might help in determining patients’ place of care and frequency of observation.15 Contributers

RA-SS conceived and designed the project. JF designed the literature search strategies and searched the literature. RA-SS, JF, and RJL co-wrote the protocol and arbitrated cohort eligibility. JF described included studies and communicated with coauthors. RJL processed data and did the data analyses, with oversight from RA-SS. RA-SS, JF, and RJL wrote the first draft of the manuscript. PDL, TWKB, AMA, JNG, SAM, TS, XW, HA, HH, MO, DAG, LM, DD, DR-L, CAM, D-KJ, AD, JCa, YY, JCL, BV, SK, YO, SF, KT, QL, JK, PD, JAS, MH-G, LP-S, CC, MPK, RM, CV, MND, YL, HW, WCW, Cld-Fe, RIA, PR, YM, ARZ, KSB, SMS, JCG, J-MF, JM, JB, HY, DS, ESC, MS, RL, BHM, AMD, MDN, YF, CSa, and JR acquired data, revised the work critically for important intellectual content where required, approved the final version to be published, and agreed to be accountable for all aspects of the work.

Declaration of interests

RJL reports grants from the UK Medical Research Council, during the conduct of the study. JNG reports personal fees from CSL Behring and Octapharma; and grants from Pfizer, Boehringer Ingelheim, and Portola, outside of the submitted work. HA reports personal fees from Asuka, Bayer, Daiichi-Sankyo, and Takeda, outside of the submitted work. JCI reports grants from the DANA Foundation and personal fees from SAGE Therapeutics, outside of the submitted work. BP reports personal fees from Pfizer/Bristol-Myers Squibb and Bayer, outside of the submitted work. JR reports grants from the National Institutes of Health/National Institute of Neurological Disorders and Stroke, during the conduct of the study. CV reports grants from the Neurocritical Care Society, during the conduct of the study. MS reports grants from National Institutes of Health/National Institute of Neurological Disorders and Stroke and the American Heart Association, outside of the submitted work. CSA reports grants from the National Health and Medical Research Council of Australia, during the conduct of the study; and personal fees from Takeda and Agena, outside of the submitted work. JR reports grants from the National Institutes of Health and personal fees from Boehringer Ingelheim and Pfizer, outside of the submitted work. All remaining authors declare no competing interests.

Acknowledgments

This study was funded by the UK Medical Research Council (senior clinical fellowship CS/A03695 and the Edinburgh Hub for Trials Methodology Research (C0800803) and the British Heart Foundation (travel fellowship FS/11/72/30531).

References


