

1 Central masked adjudication of stroke diagnosis at trial entry offered no  
2 advantage over diagnosis by local clinicians: secondary analysis and  
3 simulation

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21

22 **Abstract:**

23 **Background:** Central adjudication of stroke type is commonly implemented in large multicentre  
24 clinical trials. We investigated the effect of central adjudication of diagnosis of stroke type at trial entry  
25 in the Efficacy of Nitric Oxide in Stroke (ENOS) trial.

26 **Methods:** ENOS recruited patients with acute ischaemic or haemorrhagic stroke, and diagnostic  
27 adjudication was carried out using cranial scans. For this study, diagnoses made by local site  
28 clinicians were compared with those by central, masked adjudicators using kappa statistics. The trial  
29 primary analysis and subgroup analysis by stroke type were re-analysed using stroke diagnosis made  
30 by local clinicians, and simulations were used to assess the impact of increased non-differential  
31 misclassification and subgroup effects.

32 **Results:** Agreement on stroke type (Ischaemic, Intracerebral Haemorrhage, Unknown stroke type,  
33 No-stroke) was high ( $\kappa=0.92$ ). Adjudication of stroke type had no impact on the primary outcome or  
34 subgroup analysis by stroke type. With misclassification increased to 10 times the level observed in  
35 ENOS and a simulated subgroup effect present, adjudication would have affected trial conclusions.

36 **Conclusions:** Stroke type at trial entry was diagnosed accurately by local clinicians in ENOS.  
37 Adjudication of stroke type by central adjudicators had no measurable effect on trial conclusions.  
38 Diagnostic adjudication may be important if diagnosis is complex and a treatment-diagnosis  
39 interaction is expected.

40 **Keywords:** Adjudication, Diagnosis, Clinical Trial, Stroke

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## 46 **Introduction**

47 Clinical trials in acute stroke often recruit many thousands of participants making them complex,  
48 lengthy, and expensive. In many stroke trials, key endpoints, adverse events, or diagnoses qualifying  
49 for trial entry are adjudicated by independent experts. Independent, central adjudication may be  
50 conducted by one individual or a panel of experts, who may work independently or convene as a  
51 committee, with agreed procedures for assigning definitive values, usually blinded to treatment  
52 allocation whenever possible<sup>(1)</sup>. The adjudication procedure is believed to protect against bias  
53 resulting from differential misclassification<sup>(2, 3)</sup>, and to improve precision of treatment estimates by  
54 reducing 'noise' from random errors. This is especially important in trials where events are rare, in  
55 which a small degree of misclassification can have a large impact on study findings<sup>(2, 3)</sup> or where the  
56 event is subjective such as some clinical diagnoses. Adjudication also introduces a level of quality  
57 control to detect poorly trained or performing investigators.

58 Central adjudication is commonly included in cardiovascular studies<sup>(4, 5)</sup>, with conflicting evidence as  
59 to the value of adjudication of endpoints<sup>(6-13)</sup> compared with simply using endpoints assigned by local  
60 clinicians or investigators at participating research sites. There is little research evidence regarding  
61 the importance of diagnostic adjudication, where diagnosis is not used as an endpoint, but is used to  
62 diagnose patients at trial entry. Diagnoses made at trial entry can be used to define eligibility, as a  
63 stratification or minimisation factor, as a covariate in a regression model, or to specify categories in a  
64 subgroup analysis.

65 Stroke is a clinical diagnosis that can be further subclassified based on the results of further  
66 investigations, including brain and vessel imaging and cardiac examinations. Given the complex  
67 nature of stroke subtypes<sup>(14)</sup>, stroke diagnoses are commonly adjudicated by independent experts in  
68 clinical trials. Ninomiya et al.<sup>(11)</sup> found that adjudication of stroke type and cause of death as study  
69 endpoints had no substantive impact on treatment effect estimates in their trial. However, stroke  
70 diagnosis was an endpoint, rather than a criterion for inclusion. While adjudication of endpoints has  
71 the greatest potential to influence trial results and therefore has received greatest attention as to its  
72 value, misclassification of entry criteria might also introduce bias, affect the precision of effect  
73 estimates or reduce statistical power<sup>(15)</sup>. However, we are not aware of any such investigation of the  
74 value of central adjudication of the diagnosis qualifying for trial inclusion.

75 The aim of this study was to investigate the value of central adjudication of stroke type at trial entry in  
76 a secondary analysis of a large acute stroke trial. The three objectives were to: (1) compare stroke  
77 diagnoses made by local clinicians and central masked adjudicators; (2) assess the impact of  
78 adjudication on the primary analysis and the subgroup analysis by stroke type; (3) using simulation,  
79 explore the effects of increasing levels of misclassification of diagnosis and introducing a subgroup  
80 effect by stroke type on analyses.

81

## 82 **Materials and Methods**

### 83 **Efficacy of Nitric Oxide in Stroke (ENOS) Trial**

84 The Efficacy of Nitric Oxide in Stroke (ENOS) trial examined the safety and efficacy of glyceryl  
85 trinitrate (GTN) versus no GTN in patients with acute ischaemic or haemorrhagic stroke. Independent  
86 expert assessors, referred to in this paper as adjudicators, who were masked to treatment allocation,  
87 centrally assessed CT and MRI scans to inform diagnosis of stroke type. The primary outcome was  
88 functional outcome after stroke, measured using the modified Rankin Scale (mRS) at day 90 by  
89 outcome assessors who were masked to treatment allocation. The trial recruited 4011 patients from  
90 173 sites, across 23 countries on five continents. The primary outcome was analysed using ordinal  
91 logistic regression, and the adjusted common odds ratio (OR) for worse outcome with GTN versus no  
92 GTN was 1.01 (95% CI 0.91 to 1.13;  $p=0.83$ ). The protocol, statistical analysis plan, and main results  
93 for ENOS have been described in detail elsewhere<sup>(16-18)</sup>.

94

### 95 **Diagnosis of Stroke Type**

96 After enrolment into the ENOS trial, all participants had a CT (or MRI) scan at baseline or within  
97 seven days (referred to as baseline scan), and if possible again after seven days (referred to as  
98 follow-up scan) to assess evolution of the stroke lesion. Each scan was analysed by local clinicians,  
99 who then used information from the baseline scan, follow-up scan if available, input from the local  
100 radiology team, and clinical history and assessment of the participant between admission and  
101 discharge, in order to assign a clinical diagnosis for each participant (referred to as Local clinician

102 diagnosis). The following diagnoses were made: Ischaemic stroke, intracerebral haemorrhage,  
103 unknown stroke and no stroke. All scans were then sent electronically to the central trial team.

104 A team of independent, central adjudicators, masked to treatment allocation and Local clinician  
105 diagnosis, assessed all brain scans. They recorded their assessment using a specially designed  
106 questionnaire that captured information on the presence of stroke, haemorrhage, occluded arteries,  
107 Alberta stroke program early CT score<sup>(19)</sup>, mass effect, white matter disease, atrophy, and other  
108 visible lesions. This information was used to determine an adjudicator diagnosis of stroke type for  
109 both baseline and follow-up scans. A final diagnosis of stroke type for each participant (referred to as  
110 Central adjudication diagnosis) was assigned using an algorithm that assessed whether diagnoses  
111 from local clinicians and adjudicators sufficiently agreed, otherwise stroke diagnosis was allocated on  
112 a case-by-case basis.

113 Central adjudication diagnosis was assigned using all available information from both local clinicians  
114 and adjudicators, and was thus considered in this study as the 'gold standard'. Local clinician  
115 diagnosis represents the diagnosis of stroke type in ENOS if no central adjudication had taken place.

116 In the ENOS analyses, stroke type at trial entry was included in between-group comparisons as a  
117 baseline covariate, and as a subgroup variable to investigate any differential effects of the  
118 interventions according to stroke type. The main ENOS analyses used Central adjudication diagnosis  
119 of stroke. The analyses presented here compared the main ENOS analyses with analyses conducted  
120 using Local clinician diagnosis of stroke, thus allowing an investigation into the value of adjudication  
121 of a baseline variable in ENOS.

122

### 123 **Simulated misclassification of stroke type and simulated subgroup effect**

124 Statistical simulations were created to: (1) increase the extent of misclassification of Local clinician  
125 diagnosis of stroke compared with the gold standard Central adjudication diagnosis; (2) introduce an  
126 interaction (subgroup effect) between ENOS treatment arm and stroke type. These simulations  
127 enabled us to investigate the effects of misclassification on the ENOS primary analysis and on  
128 subgroup analysis, for both the subgroup effect observed in ENOS and for a subgroup effect  
129 introduced by simulation. The magnitude of the treatment-stroke type interaction was increased in  
130 simulation as there was no statistical evidence of a subgroup effect in the observed ENOS dataset.

131 In simulated datasets, the misclassification of Local clinician diagnosis observed in ENOS was  
132 increased by factors of 3, 5, 10, 15 and 20 (referred to as SX3, SX5, SX10, SX15 and SX20  
133 respectively). We also introduced a subgroup effect by reducing mRS score by 1 point for 10% of  
134 participants with an Ischaemic stroke, and increasing mRS score by 1 point for 30% of participants  
135 with an Intracerebral Haemorrhage, with mRS scores for all participants constrained to be in the  
136 normal range 0 to 6. All participants with an altered mRS score were in the GTN arm of the trial. For  
137 more detailed simulation methods, please consult Supplementary File S1.

138

## 139 **Statistical Methods**

140 Categorical variables were described using N (%). Observed agreement between Local clinician and  
141 Central adjudication diagnoses was quantified using unweighted kappa statistics.

142 Using observed ENOS data, the effect of GTN treatment on mRS score was estimated as in the  
143 ENOS trial main report, using ordinal logistic regression models, adjusted for stratification and  
144 minimisation variables. Models including Local clinician and Central adjudication diagnosis of stroke  
145 type as a covariate were fitted separately and the estimated effects of GTN treatment from the two  
146 models were compared using a test of homogeneity. Similarly, subgroup effects were estimated by  
147 fitting an interaction term between GTN treatment and stroke type according to either Local clinician  
148 or Central adjudication diagnosis.

149 The primary trial analysis was then repeated using each simulated level of Local clinician diagnosis  
150 misclassification (SX3 to SX20). The subgroup analysis was also repeated for each simulated level of  
151 Local clinician diagnosis misclassification for both the subgroup effect observed in the ENOS dataset,  
152 and for the increased subgroup effect created using simulation. Regression model coefficients and  
153 standard errors are presented on the log scale for ease of comparison.

154

## 155 **Results**

156 Of 4011 participants randomised, 3857 (96%) and 1025 (26%) had baseline and follow up scans  
157 respectively that were assessed by adjudicators. A total of 35 participants had a missing Local

158 clinician diagnosis, and all participants had a Central adjudication diagnosis assigned after the  
159 combined information from the hospital and central adjudicators was reviewed (Figure 1).

160 The proportion of participants with each stroke type was similar for those that did or did not have a  
161 follow-up scan, indicating no evidence of bias in the selection of participants for a follow up scan and  
162 therefore having more information with which to assign a diagnosis (*see Supplementary File, S2*).

163 Agreement was high in ENOS, with local clinicians and central adjudicators agreeing on 79% of  
164 diagnoses at baseline. There was excellent agreement between Local clinician and Central  
165 adjudication diagnoses (crude agreement 98%, unweighted kappa,  $\kappa=0.92$ ) for the 3976 (99%)  
166 participants who could be included in this analysis (Table 1).

167 Misclassification of Local clinician diagnosis resulted in kappa statistics for agreement between  
168 Central adjudication and Local clinician diagnoses of 0.78, 0.67, 0.46, 0.32 and 0.21 for SX3-SX20  
169 respectively. As expected due to strong agreement between Central adjudication and Local clinician  
170 diagnoses of stroke type, it made little difference which one was used as a covariate in the primary  
171 analysis of observed ENOS data (p-value for homogeneity  $p=0.95$ , *see Supplementary File, S3*).  
172 Similarly, coefficients and standard errors for the interaction between GTN and stroke type were very  
173 similar regardless of whether Local clinician or Central adjudication diagnosis of stroke type was used  
174 (*data not shown*).

175 Increased levels of non-differential misclassification of stroke diagnosis introduced by simulation  
176 made no material difference to the estimated treatment effect of GTN or the precision of the estimate  
177 (Table 2). Table 3 shows the effect of GTN separately for each stroke type using the magnitude of  
178 subgroup effect observed in the ENOS data, and where non-differential misclassification of stroke  
179 type is increased by simulation. The number of participants diagnosed with ischaemic stroke  
180 decreased, whilst each of the other types of stroke increased, respectively, with increasing  
181 misclassification. The effects of misclassification on stroke-specific estimates of GTN treatment were  
182 not wholly consistent, although increasing misclassification tended to give treatment effects closer to  
183 zero and standard errors that increased or decreased inversely with stroke-specific sample size  
184 accordingly.

185 Simulation of a subgroup effect, whereby GTN was beneficial among participants with an ischaemic  
186 stroke, and harmful among participants with a haemorrhagic stroke, attenuated the treatment effects  
187 even further (Table 4). After stroke type was increasingly misclassified using simulation, statistical  
188 evidence of a subgroup effect was reduced and the effects of subgroup sample size on precision  
189 were as expected (Tables 4 and 5).

190

## 191 **Discussion**

192 Misclassification of stroke type by local trial site clinicians was low, with excellent agreement found  
193 between the Central adjudication and Local clinician diagnosis. Due to the level of agreement, there  
194 was little impact of adjudication of stroke type at trial entry on the primary analysis or subgroup  
195 analysis of ENOS. Increased levels of non-differential misclassification produced little change in the  
196 primary outcome. After simulating a strong subgroup effect by stroke type, increased misclassification  
197 resulted in reduction of the subgroup effect, suggesting that in this situation adjudication may be  
198 important to ensure robust results.

199 In ENOS, due to blinding, differential misclassification of stroke type was unlikely, which was why we  
200 introduced non-differential misclassification using simulation. Even with non-differential  
201 misclassification increased by 20 times the observed level, there was little effect on both the primary  
202 and subgroup analyses. Only when a substantial subgroup effect ( $p < 0.01$ ) and marked  
203 misclassification of stroke diagnosis by local investigators were simulated would adjudication have  
204 resulted in differing conclusions. These extreme, and thus arguably unlikely, conditions before central  
205 adjudication is seen to add value are likely due to the fact that in our analyses, diagnosis of stroke  
206 type is a baseline variable rather than a study endpoint. However a recent Cochrane review<sup>(20)</sup> that  
207 assessed endpoint adjudication of subjective binary events across a range of clinical areas, including  
208 47 RCTs, also found that adjudication did not affect the treatment effect estimates (Ratio of Odds  
209 Ratios: 1.00, 95% C.I.: [0.97 to 1.04]). The review suggested that adjudication 'may be most important  
210 when onsite assessors are not blinded and the risk of misclassification is high'.



211 It is worth noting that in ENOS, diagnostic adjudication was used for purposes in addition to informing  
212 the diagnosis. The adjudication process provided a large amount of extra information which hospital  
213 scan results would not have recorded. This information can be used to carry out imaging-based  
214 subgroup analyses or help to improve any subsequent sub-studies. Furthermore, the central  
215 adjudication process meant that each scan had been rated using a central, standard approach,  
216 enabling data to be pooled with other trials that have used a similar method. Therefore, the ENOS  
217 data can be utilised further, alongside existing data, to provide a larger sample size to test the  
218 independent prognostic value and potential treatment implications of the scan signs raised in various  
219 studies, as well as assisting in confirming or refuting ideas about not treating certain types of infarct or  
220 effects on infarct swelling.

221 The diagnostic adjudication process in ENOS resulted in increased complexity, and monetary and  
222 time costs. These included payments to adjudicators, resources associated with handling adjudicator  
223 data (data entry, database programming, and statistical analysis), the time taken by the trial team to  
224 determine the trial diagnosis, and data queries. Although this is the first study we are aware of to  
225 investigate diagnostic adjudication in stroke trials, where diagnosis is not used as an endpoint,  
226 previous studies which have looked at adjudication of endpoints have found similar conclusions.  
227 Slight benefits of improving accuracy and reducing misclassification were outweighed by the cost and  
228 complications introduced by an adjudication committee<sup>(2, 11)</sup>. However, there may be some  
229 unmeasurable benefits of an adjudication process, and adjudication could have indirectly  
230 strengthened local assessment due to a policing effect. This effect could have resulted in improved  
231 site performance as investigators would have been aware that diagnoses would have been checked  
232 centrally, and thus perform more carefully.

233 One strength of this study is that we used a large, well conducted, randomised trial to provide data  
234 from over 4000 participants for analysis. Furthermore, the data completeness was extremely high,  
235 minimising the risk of bias due to partially completed data. The simulation undertaken in this study  
236 allowed an investigation into the robustness of observed results to more extreme data scenarios. This  
237 was important to understand how adjudication of diagnosis at trial entry could affect a similar trial  
238 where agreement was not as good as observed in ENOS. This approach, using a combination of

239 observed and simulated data, can be readily applied to secondary analyses of other trials, notably on  
240 outcome variables as well as baseline variables, in order to inform future studies.

241 A limitation of this study is that the potential for adjudication to have an important effect is likely to be  
242 less for a baseline variable, as seen in ENOS, rather than a primary outcome as in Ninomiya et al.<sup>(11)</sup>.  
243 Therefore, we also looked into the impact of adjudication on subgroup analyses involving stroke type,  
244 to allow a thorough investigation into the value of central adjudication of a baseline variable had on  
245 ENOS. Furthermore, the treatment estimates for GTN for both ischaemic and haemorrhagic stroke  
246 were similar, so increased misclassification in this situation had limited impact, although this may not  
247 be the case in other studies where there is a treatment-diagnosis interaction. Simulation allowed us to  
248 explore this setting, but a further investigation using data from another large trial would be beneficial  
249 to reinforce our findings.

250

## 251 **Conclusions**

252 This study found that clinicians at ENOS trial sites largely were correct in their diagnosis of stroke and  
253 adjudication did not impact on the trial results. Adjudication of stroke type at trial entry would have  
254 altered conclusions had there been strong evidence of a subgroup effect by stroke type, and where  
255 misclassification was at least ten times that observed in ENOS. In pilot or feasibility studies,  
256 misclassification could be estimated in order to inform whether adjudication would be useful in that  
257 particular trial. Researchers should consider the value adjudication could bring to their study before its  
258 implementation in a clinical trial to avoid wasted time and unneeded expenditure.

259

## 260 **List of abbreviations:**

261 ENOS – Efficacy of Nitric Oxide in Stroke

262 GTN – Glyceryl trinitrate

## 263 **Ethics approval and consent to participate:**

264 Ethical approval was not required for this research due to it being a secondary analysis of the ENOS  
265 trial where ethical approval was attained. Written informed consent for the ENOS trial was obtained  
266 from each patient, or, in the case when the patient did not have capacity, from a relative or  
267 independent physician.

268 **Availability of data and materials:**

269 The datasets used and/or analysed during the current study are available from the corresponding  
270 author on reasonable request.

271 **Competing interests:**

272 The authors declare that they have no competing interests

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280 and interpretation of data.

281 **Authors' contributions:**

282 PJG, PMB and AAM conceived the study. PJG prepared the analysis plan and conducted the  
283 analysis. AAM and PMB reviewed the analysis plan. All authors interpreted the data. PJG wrote the  
284 first draft of the manuscript. All authors reviewed and edited the manuscript for important intellectual  
285 content. All authors approved the final manuscript.

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290

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356 **Supplementary Files**

357 File name: File S1

358 Title of data: S1 – Simulation methods

359 Description of data: Detailed description of the simulation process used in this study, including how  
360 the simulated datasets were generated, how misclassification was increased and how the number of  
361 simulations was calculated.

362 File name: File S2

363 Title of data: S2 - Supplementary Table 1

364 Description of data: Table which shows the diagnosis of stroke type made by the adjudicators for all  
365 scans that were assessed. This table is further split for participants that did and did not have a follow  
366 up scan available, which show that adjudicator's diagnoses were similar for those participants that did  
367 and did not receive a follow up scan.

368 File name: File S3

369 Title of data: S3 – Supplementary Table 2

370 Description of data: Table which shows the primary outcome measure for ENOS and how this result  
371 would be affected with and without adjudication. A p-value for homogeneity is given which tests the  
372 null hypothesis that the estimates from both analyses are the same.

373

374 **Tables and Figures Legends:**

375 Figure 1: Flow diagram showing diagnosis of stroke type in ENOS

376 Table 1: Agreement between Local clinician and Central adjudication diagnosis

377 Table 2: Effect of increased misclassification of stroke type at trial entry on ENOS primary analysis

378 Table 3: Effect of misclassification of stroke type at trial entry on subgroup analysis: based on

379 subgroup effect observed in ENOS data

380 Table 4: Effect of misclassification of stroke type on subgroup analysis: based on simulated subgroup

381 effect

382 Table 5: P-values for interaction tests between GTN and stroke type based on observed and

383 simulated ENOS data

384

385



386 **Tables:**

387 Table 1: Agreement between Local clinician and Central adjudication diagnosis

<b>Central adjudication diagnosis</b>					
<b>Local clinician diagnosis</b>	Ischaemic	Intracerebral Haemorrhage	Unknown stroke type	No-stroke	<b>Total</b>
Ischaemic	3233	6	0	0	<b>3239</b>
Intracerebral Haemorrhage	18	615	0	1	<b>634</b>
Unknown stroke type	63	0	0	0	<b>63</b>
No-stroke	2	2	0	36	<b>40</b>
<b>Total</b>	<b>3316</b>	<b>623</b>	<b>0</b>	<b>37</b>	<b>3976</b>

388 Crude agreement = 3884/3976 = 98%

389 Unweighted kappa = 0.92

390

391 Table 2: Effect of increased misclassification of stroke type at trial entry on ENOS  
 392 primary analysis

Source of diagnosis of stroke type at trial entry	Results from regression model comparing effect of GTN versus no GTN	
	Log OR	SE log OR
Central adjudication	-0.02473	0.05565
SX3	-0.02446	0.05563
SX5	-0.02426	0.05563
SX10	-0.02426	0.05561
SX15	-0.02411	0.05561
SX20	-0.02415	0.05561

406 SX3-SX20 refer to the misclassified Local clinical diagnoses. Kappa statistics showing the agreement  
 407 between each diagnosis and Central adjudication diagnosis are 0.78, 0.67, 0.46, 0.32 and 0.21 for  
 408 SX3-SX20 respectively.

409

410 Table 3: Effect of misclassification of stroke type at trial entry on subgroup analysis:  
 411 based on subgroup effect observed in ENOS data

Stroke Type	Source of diagnosis of stroke type at trial entry	N	Subgroup-specific estimated effect of GTN versus no GTN	
			Log OR	SE log OR
<i>Ischaemic</i>	Central adjudication	3338	-0.03048	0.06085
	SX3	3096	-0.03114	0.06003
	SX5	2935	-0.02953	0.06491
	SX10	2531	-0.02503	0.06987
	SX15	2129	-0.03130	0.07618
	SX20	1725	-0.03043	0.08476
<i>Haemorrhagic</i>	Central adjudication	623	0.02699	0.14110
	SX3	657	0.02761	0.13717
	SX5	682	0.01898	0.13474
	SX10	739	0.01443	0.12943
	SX15	798	-0.00496	0.12456
	SX20	855	0.00832	0.12027
<i>Unknown</i>	Central adjudication	1	-	
	SX3	196	0.00091	0.25320
	SX5	325	-0.01070	0.19592
	SX10	652	-0.03348	0.13830
	SX15	975	-0.00867	0.11286
	SX20	1302	-0.02353	0.09743
<i>No-stroke</i>	Central adjudication	38	0.18475	0.65491
	SX3	51	0.16043	0.54100
	SX5	58	0.05159	0.48825
	SX10	78	-0.00907	0.40673

SX15	98	-0.00359	0.36080
SX20	118	-0.00448	0.32877

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412 Simulations produced datasets containing 4000 observations.

413 SX3-SX20 refer to the misclassified Local clinical diagnoses. Kappa statistics showing the agreement  
414 between each diagnosis and Central adjudication diagnosis are 0.78, 0.67, 0.46, 0.32 and 0.21 for  
415 SX3-SX20 respectively.

416

417 Table 4: Effect of misclassification of stroke type on subgroup analysis: based on  
 418 simulated subgroup effect

Stroke Type	Source of diagnosis of stroke type at trial entry	N	Subgroup-specific estimated effect of GTN versus no GTN	
			Log OR	SE log OR
<i>Ischaemic</i>	Central adjudication	3338	-0.14122	0.06085
	SX3	3096	-0.13885	0.06320
	SX5	2935	-0.13576	0.06493
	SX10	2531	-0.12591	0.06988
	SX15	2129	-0.11477	0.07624
	SX20	1725	-0.11388	0.08478
<i>Haemorrhagic</i>	Central adjudication	623	0.29183	0.14156
	SX3	657	0.25154	0.13760
	SX5	682	0.22485	0.13522
	SX10	739	0.17519	0.12990
	SX15	798	0.11796	0.12482
	SX20	855	0.08150	0.12015
<i>Unknown</i>	Central adjudication	1	-	
	SX3	196	-0.09800	0.25358
	SX5	325	-0.12723	0.19472
	SX10	652	-0.15382	0.13795
	SX15	975	-0.13565	0.11250
	SX20	1302	-0.12854	0.09744
<i>No-stroke</i>	Central adjudication	38	0.25680	0.68781
	SX3	51	0.14555	0.53944
	SX5	58	0.05832	0.49047
	SX10	78	0.07533	0.40879

SX15	98	-0.05162	0.36561
SX20	118	0.02436	0.33095

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419 Simulations produced datasets containing 4000 observations.

420 SX3-SX20 refer to the misclassified Local clinical diagnoses. Kappa statistics showing the agreement  
421 between each diagnosis and Central adjudication diagnosis are 0.78, 0.67, 0.46, 0.32 and 0.21 for  
422 SX3-SX20 respectively.

423

424

425 Table 5: P-values for interaction tests between GTN and stroke type based on  
 426 observed and simulated ENOS data

427

428	<b>Data source</b>	<b>Source of diagnosis</b>	<b>Median p-value from 100 simulated analyses (IQR)</b>
429			
430	<i>Subgroup effect based on observed ENOS data</i>	Central adjudication	0.38592 (0.17160, 0.61673)
431			
432			
433		SX3	0.39858 (0.15347, 0.65161)
434		SX5	0.46350 (0.16563, 0.72459)
435			
436		SX10	0.43609 (0.22501, 0.78923)
437		SX15	0.54638 (0.32173, 0.79183)
438		SX20	0.46829 (0.23353, 0.70323)
439			
440	<i>Subgroup effect based on simulated ENOS data</i>	Central adjudication	0.00882 (0.00096, 0.06394)
441			
442			
443		SX3	0.02801 (0.00699, 0.14882)
444		SX5	0.04675 (0.00677, 0.24457)
445			
446		SX10	0.10912 (0.01997, 0.31892)
447		SX15	0.16117 (0.05030, 0.47707)
448		SX20	0.24764 (0.06521, 0.54910)

449 SX3-SX20 refer to the misclassified Local clinical diagnoses. Kappa statistics showing the agreement  
 450 between each diagnosis and Central adjudication diagnosis are 0.78, 0.67, 0.46, 0.32 and 0.21 for  
 451 SX3-SX20 respectively.

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