Personalized Prehospital Triage in Acute Ischemic Stroke
A Decision-Analytic Model

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Background and Purpose—Direct transportation to a center with facilities for endovascular treatment might be beneficial for patients with acute ischemic stroke, but it can also cause harm by delay of intravenous treatment. Our aim was to determine the optimal prehospital transportation strategy for individual patients and to assess which factors influence this decision.

Methods—We constructed a decision tree model to compare outcome of ischemic stroke patients after transportation to a primary stroke center versus a more distant intervention center. The optimal strategy was estimated based on individual patient characteristics, geographic location, and workflow times. In the base case scenario, the primary stroke center was located at 20 minutes and the intervention center at 45 minutes. Additional sensitivity analyses included an urban scenario (10 versus 20 minutes) and a rural scenario (30 versus 90 minutes).

Results—Direct transportation to the intervention center led to better outcomes in the base case scenario when the likelihood of a large vessel occlusion as a cause of the ischemic stroke was >33%. With a high likelihood of large vessel occlusion (66%, comparable with a Rapid Arterial Occlusion Evaluation score of 5 or above), the benefit of direct transportation to the intervention center was 0.10 quality-adjusted life years (=36 days in full health). In the urban scenario, direct transportation to an intervention center was beneficial when the risk of large vessel occlusion was 24% or higher. In the rural scenario, this threshold was 49%. Other factors influencing the decision included door-to-needle times, door-to-groin times, and the door-in-door-out time.

Conclusions—The preferred prehospital transportation strategy for suspected stroke patients depends mainly on the likelihood of large vessel occlusion, driving times, and in-hospital workflow times. We constructed a robust model that combines these characteristics and can be used to personalize prehospital triage, especially in more remote areas. (Stroke. 2019;50:313-320. DOI: 10.1161/STROKEAHA.118.022562.)

Key Words: decision tree • emergency medical services • quality-adjusted life years • thrombectomy • transportation

Interhospital transfer for endovascular treatment (EVT) of patients with acute ischemic stroke because of an intracranial large vessel occlusion (LVO) is one of the major causes of treatment delay.1-5 Delay of EVT is associated with poor functional outcome.6,7 Early identification of patients eligible for EVT followed by direct transportation to an endovascular-capable center might reduce transfer-related delay and thereby improve outcome. For this purpose, several prehospital stroke scales have been developed to identify patients who are at high risk of having an intracranial LVO based on their clinical symptoms. Currently, there is no evidence on superiority of one of these scales and more prospective validation in the prehospital setting is needed to reliably assess their accuracy.8,9 Other factors that might be of importance for the prehospital triage of suspected stroke patients include the prognosis of an individual patient and the expected benefit of EVT. Especially because of the large majority of ischemic stroke patients is not eligible for EVT and only benefits from rapid treatment with intravenous thrombolitics (IVT), the harm of delaying IVT should be taken into account as well. The
time-dependent effect of both treatments requires a tradeoff between reducing delay of IVT by transportation to the nearest primary stroke center and avoiding transfer-related delay of EVT by direct transportation to an endovascular-capable intervention center.

Our aim was to determine the optimal prehospital transportation strategy for individual patients with suspected ischemic stroke and to assess which factors influence this decision.

**Methods**

**Decision Model**

We constructed a decision tree model to compare outcome of patients with ischemic stroke after 2 different prehospital strategies: transportation to the nearest primary stroke center versus direct transportation to a more distant intervention center (Figure 1). The outcome of each strategy was estimated based on individual patient characteristics, geographic location, and treatment times. We combined a short-run model including 3 months outcome data from randomized controlled trials and a long-run Markov model that simulated 40 annual cycles. The benefit of direct transportation to the intervention center was defined as the average amount of quality-adjusted life years (QALYs) gained by this strategy. We considered a difference of >0.02 QALYs (=1 week in full health) to be clinically relevant.

Our study did not need approval by an ethics committee because we did not use individual patient data. Analytic methods and study materials that support the findings of this study are available from the corresponding author on reasonable request.

**Model Parameters**

Individual input parameters consisted of age, sex, time since symptom onset, likelihood of LVO, driving time from scene to the primary stroke center, driving time from scene to the intervention center, and driving time between the primary stroke center and the intervention center. The base case used in our analyses was a 68-year-old man with suspected stroke symptoms since 1 hour and an average risk of 30% to have an LVO causing the ischemic stroke. The nearest primary stroke center in the base case scenario was located at a 20-minute drive by ambulance, while the intervention center was located at 45 minutes. Driving time between the centers was 35 minutes. Total time to treatment was calculated based on the driving times and in-hospital workflow characteristics (Tables 1 through 3). These parameters were varied in several sensitivity analyses.

Other model parameters were estimated based on previous literature and expert opinion of 2 neurovascular specialists (Drs. Dippel and Roozenbeek; Table I in the online-only Data Supplement). The coefficient of treatment effect and time-dependent decline of treatment effect were estimated with an ordinal logistic regression model using previously reported outcome distributions of treated patients and control patients in different time intervals. The effect of the uncertainty around these estimates was assessed with a probabilistic sensitivity analysis.

**Likelihood of LVO**

We modeled the entire range from 0% to 100% likelihood of having an LVO as a cause of the ischemic stroke, and we calculated the threshold at which direct transportation to the intervention center would be beneficial. The average prevalence of LVO ranges mostly between 20% and 40% in different populations of ischemic stroke patients, but prehospital stroke scales can be used in individual patients to distinguish between lower and higher risk. To illustrate this, we calculated the likelihood of LVO in case of a positive test (positive predictive value) and in case of a negative test (1 minus the negative predictive value) for the Los Angeles Motor Scale, the Rapid Arterial Occlusion Evaluation, and the 3-Item Stroke Scale using the pooled sensitivity and specificity rates reported in a previous meta-analysis (Table 4). In the sensitivity analyses, we used the Rapid Arterial Occlusion Evaluation scale for examples of low risk (14%) and high risk (66%).

![Figure 1. Schematic overview of the model structure. The model starts with the initial decision of transportation to the primary stroke center or to the nearest endovascular-capable intervention center. The short-run model calculates the probability of every possible pathway and the associated distribution of the modified Rankin Scale (mRS) score after 3 mo. It takes into account driving times, in-hospital workflow characteristics, and time-dependent treatment effects. In each annual cycle of the following Markov model, patients can remain in the same health state or die. These probabilities are based on the age and sex-dependent annual mortality rates, adjusted for previously reported death hazard rate ratios of stroke patients. The decision node is represented with a square. The circles represent chance nodes, the circles marked with an M represent Markov models and the triangles represent terminal nodes. EVT indicates endovascular treatment; IVT, treatment with intravenous thrombolitics; and LVO, large vessel occlusion.](http://ahajournals.org)
Outcome Measures

The primary outcome of the short-run model was the distribution of modified Rankin Scale (mRS) scores after 3 months. The mRS is a 7-point scale to assess functional outcome and disability and ranges from 0 (no symptoms) to 6 (death). The outcome distributions of untreated ischemic stroke patients with and without an LVO were based on the control groups of previously reported randomized clinical trials (Figure I in the online-only Data Supplement).10,12 These distributions were shifted for different outcomes based on the estimated treatment effect. To do so, we first transformed the probabilities of each of the baseline mRS scores to the log-odds scale, using the formula log-odds(p) = log(p/(1−p)). We then adjusted these log-odds for the effect of IVT and EVT by adding the β-coefficient (=log odds ratio) for the treatment effect at time 0 minus the decline in treatment effect over time (Tables 5 and 6). The adjusted log-odds were then converted back to probabilities using the inverse logit function: p(log-odds) = 1/(1+exp(−log-odds)). This resulted in the adjusted mRS distributions for treated patients.

All mRS scores were considered to be separate health states with an associated utility score to calculate the average amount of QALYs per strategy. This is a commonly used measure to assess both the time (Tables 5 and 6). The adjusted log-odds were then converted back to probabilities using the inverse logit function: p(log-odds) = 1/(1+exp(−log-odds)). This resulted in the adjusted mRS distributions for treated patients.

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Model Assumptions

The model was used for scenarios in which the nearest hospital is not an intervention center because there has to be decisional uncertainty about the optimal transportation strategy. Outcome was modeled for patients with ischemic stroke only, disregarding other diagnoses. A substantial proportion of all patients suspected of ischemic stroke has an intracerebral hemorrhage, a transient ischemic attack or a stroke-mimic, but we assumed the outcome of these patients to be unrelated to the transportation strategy. Because our analyses concerned the tradeoff between IVT and EVT, we did not consider patients presenting >4.5 hours after onset of symptoms.

We assumed that IVT could be given within 4.5 hours after onset of symptoms and EVT within 6 hours after onset. We modeled the decrease in treatment effect of IVT and EVT consistently over time and similar for all patients. Because large pooled analyses showed no significant interaction between age or sex and treatment effect, we did not include age- or sex-specific treatment effects in our model.12,16 The relative treatment effect of IVT was not influenced by stroke severity, which implies a smaller absolute treatment effect for patients with a more severe stroke. The importance of these model assumptions was assessed in several sensitivity analyses.

Sensitivity Analyses

In addition to the base case scenario, we also assessed examples of an urban scenario (10 minutes from scene to primary stroke center; 20 minutes from scene to intervention center; and 15 minutes intercenter driving time) and a more rural scenario (30, 90, and 75 minutes, respectively). We used a Tornado-analysis to explore the relative importance of the model parameters, by varying each parameter at a time while the others were held constant. Additional sensitivity analyses were performed with: increased workflow times in the primary stroke center (Table 2), a female patient, a patient with contra-indications for IVT, an absent effect of IVT for patients with an LVO, and utility weights as defined in a study of Chaisanumkul et al.17

We performed a probabilistic sensitivity analysis using second-order Monte-Carlo simulations to assess decisional uncertainty around the model parameters. This involves running the model 10000 times and calculating the optimal transportation strategy

Table 1. Driving Times Used in the Various Analyses, in Minutes

<table>
<thead>
<tr>
<th>Driving Times</th>
<th>Base Case Scenario</th>
<th>Urban Scenario</th>
<th>Rural Scenario</th>
</tr>
</thead>
<tbody>
<tr>
<td>From scene to primary stroke center</td>
<td>20</td>
<td>10</td>
<td>30</td>
</tr>
<tr>
<td>From scene to intervention center</td>
<td>45</td>
<td>20</td>
<td>90</td>
</tr>
<tr>
<td>From primary stroke center to intervention center</td>
<td>35</td>
<td>15</td>
<td>75</td>
</tr>
</tbody>
</table>

Primary stroke center: the nearest nonendovascular-capable stroke center and intervention center: the nearest endovascular-capable stroke center. EVT indicates endovascular treatment; and IVT, treatment with intravenous thrombolytics.

Table 2. In-Hospital Workflow Characteristics Used in the Various Analyses, in Minutes

<table>
<thead>
<tr>
<th>In-Hospital Workflow Characteristics</th>
<th>Base Case Analysis</th>
<th>Sensitivity Analysis</th>
</tr>
</thead>
<tbody>
<tr>
<td>Door-to-needle time (primary stroke center)</td>
<td>30</td>
<td>60</td>
</tr>
<tr>
<td>Door-to-needle time (intervention center)</td>
<td>30</td>
<td>30</td>
</tr>
<tr>
<td>Door-in-door-out time (primary stroke center)</td>
<td>60</td>
<td>90</td>
</tr>
<tr>
<td>Door-to-groin time (directly admitted)</td>
<td>80</td>
<td>80</td>
</tr>
<tr>
<td>Door-to-groin time (transferred)</td>
<td>50</td>
<td>50</td>
</tr>
</tbody>
</table>

Primary stroke center: the nearest nonendovascular-capable stroke center and intervention center: the nearest endovascular-capable stroke center.

Table 3. Calculations of Treatment Times in the Model, in Minutes

<table>
<thead>
<tr>
<th>Treatment Times</th>
<th>Time to IVT (primary stroke center)</th>
<th>Time since onset of symptoms+driving time from scene to primary stroke center+door-to-needle time in primary stroke center</th>
</tr>
</thead>
<tbody>
<tr>
<td>Time to IVT (intervention center)</td>
<td>Time since onset of symptoms+driving time from scene to intervention center+door-to-needle time in intervention center</td>
<td></td>
</tr>
<tr>
<td>Time to EVT (directly admitted)</td>
<td>Time since onset of symptoms+driving time from scene to intervention center+door-to-groin time for directly admitted patients</td>
<td></td>
</tr>
<tr>
<td>Time to EVT (transferred)</td>
<td>Time since onset of symptoms+driving time from scene to primary stroke center+door-in-door-out time+driving time from primary stroke center to intervention center+door-to-groin time for transferred patients</td>
<td></td>
</tr>
</tbody>
</table>

Primary stroke center: the nearest nonendovascular-capable stroke center and intervention center: the nearest endovascular-capable stroke center. EVT indicates endovascular treatment; and IVT, treatment with intravenous thrombolytics.
every single time. In each simulation, the estimates of the model parameters were randomly sampled from the prespecified distribution of each parameter (β, (log)normal or triangular; Table I in the online-only Data Supplement). We reported the percentage of simulations in which transportation to the primary stroke center or direct transportation to the intervention center was preferred, and the median benefit of direct transportation to an intervention center with a 95% credible interval.

We constructed an online tool in which input factors and regional workflow times can be adjusted for an individual patient in a specific scenario. We used R statistical software (version 3.4.4) with the dampack package (version 0.0.0.9) and the R Shiny package (version 1.0.5).

### Results

**Base Case Analysis**

Direct transportation to the intervention center was preferred for the 68-year-old man in the base case scenario, with a primary stroke center located at 20 minutes and an intervention center at 45 minutes, when the likelihood of LVO was 34% or above (Figure 2). When the risk of having an LVO was 66%, comparable with a Rapid Arterial Occlusion Evaluation score of 5 or more, the benefit of direct transportation to the intervention center was 0.10 QALYs (=36 days). When the risk of having an LVO was 14%, comparable with a Rapid Arterial Occlusion Evaluation score below 5, transportation to the primary stroke center was preferred with a difference of 0.03 QALYs (=11 days).

**Sensitivity Analyses**

In the urban scenario (10 minutes to the primary stroke center and 20 minutes to the intervention center), the threshold for direct transportation to the intervention center was lower than in the base case scenario (24%; Figure 2). In the rural scenario (30 minutes to the primary stroke center and 90 minutes to the intervention center), direct transportation to the intervention center was only preferred for patients with an LVO likelihood of 49% or above.

Other factors that strongly affected the decision threshold in the Tornado-analysis were the in-hospital workflow characteristics (Figure III in the online-only Data Supplement). While remaining the other parameters of the base case constant, transportation to the primary stroke center was beneficial when the door-to-groin time for directly transported patients was above 102 minutes, the door-to-groin time for transferred patients was below 28 minutes, or the door-in-door-out time in the primary stroke center was <38 minutes. Age did not influence the preferred strategy, although it was important for the general prognosis of individual patients.

In the analysis with increased workflow times in the primary stroke center, transportation to the intervention center was more favorable (Figure IV in the online-only Data Supplement). As might be expected, we found that transportation to the primary stroke center was never preferred for a patient with contra-indications for IVT. In all other exploratory sensitivity analyses, the decision threshold remained relatively unchanged.

### Probabilistic Sensitivity Analysis

The uncertainty around the estimates of the model parameters caused little decisional uncertainty in the probabilistic sensitivity analysis (Table II in the online-only Data Supplement). When the risk of having an LVO in the base case scenario was high, transportation to the intervention center was associated with a clinically relevant benefit in 94% of the simulations with a median difference of 0.09 QALYs (=33 days in full health, 95% credible interval: 0.00–0.20). When the risk of having an LVO was low, direct transportation to the intervention center was preferred in only 6% of the simulations, with a median difference of −0.03 QALYs (95% credible interval: −0.09 to 0.03).

### Table 4. Likelihood of LVO Based on Several Prehospital Stroke Scales

<table>
<thead>
<tr>
<th>Stroke Scale</th>
<th>Sensitivity*</th>
<th>Specificity*</th>
</tr>
</thead>
<tbody>
<tr>
<td>LAMS ≥4</td>
<td>38%</td>
<td>87%</td>
</tr>
<tr>
<td>RACE ≥5</td>
<td>67%</td>
<td>85%</td>
</tr>
<tr>
<td>3I-ISS ≥4</td>
<td>19%</td>
<td>97%</td>
</tr>
</tbody>
</table>

The different prior probabilities illustrate the prevalence of LVO in different populations of ischemic stroke patients. LAMS, Los Angeles Motor Scale; LVO, large vessel occlusion; NPV, negative predictive value; PPV, positive predictive value; and RACE, Rapid Arterial Occlusion Evaluation.

*Pooled sensitivity and specificity rates as reported previously.

### Table 5. Calculations of Treatment Effect in the Model

<table>
<thead>
<tr>
<th>Treatment Characteristics</th>
<th>Treatment effect EVT (β coefficient)*</th>
</tr>
</thead>
<tbody>
<tr>
<td>Probability of receiving IVT</td>
<td>If treatment is possible within 4.5 h after onset of symptoms: 55%, otherwise: 0%</td>
</tr>
<tr>
<td>Probability of early reperfusion</td>
<td>If presenting with an LVO and treated with IVT: 11%, otherwise: 0%</td>
</tr>
<tr>
<td>Probability of receiving EVT</td>
<td>If treatment is possible within 6 h after onset of symptoms: 85%, otherwise 0%</td>
</tr>
<tr>
<td>Treatment effect IVT (β coefficient)*</td>
<td>0.56 (=odds ratio 1.75) at time 0, minus 0.0019× time to EVT in minutes</td>
</tr>
<tr>
<td>Treatment effect EVT (β coefficient)*</td>
<td>1.35 (=odds ratio 3.85) at time 0, minus 0.0026× time to EVT in minutes</td>
</tr>
</tbody>
</table>

EVT indicates endovascular treatment; IVT, treatment with intravenous thrombolitics; and LVO, large vessel occlusion.

*The time-dependent decrease in treatment effect is illustrated in Figure II in the online-only Data Supplement.
However, prospective prehospital validation is required to ported cutoffs of one or multiple prehospital stroke scales. For the likelihood of LVO, these models used previously re-

As a measure for EVT, as was described by other models. As a measure

incorporated in our model. It is likely that unnecessary transportation to an intervention center will cause crowding at the emergency department, which might have financial consequences and a negative effect on the in-hospital workflow. It is, therefore, important to further explore the cost-effectiveness of different prehospital triage strategies.

assess and compare the accuracy of these scales when used by emergency medical services in a broad population of suspected stroke patients. The full dependence of these models on insufficiently validated prehospital stroke scales limits the validity of their results. Therefore, we did not depend our model on a specific prehospital stroke scale but assessed the entire range from 0% to 100% likelihood of LVO. Several prehospital stroke scales may be used to estimate this likelihood, preferably after more extensive validation.

Our study also has some limitations. Decision-analytic modeling requires the use of estimated model parameters and multiple assumptions. In extensive sensitivity analyses, we found no substantial decisional uncertainty. Nevertheless, the assumptions we made about the treatment effect of IVT and EVT might have influenced our results. We used reported outcomes and effect sizes of IVT based on studies that did not assess LVO status because randomized trials on the effect of IVT were performed before the introduction of CT angiography as standard of care for acute stroke patients. To calculate a common odds ratio for the decay in effect of IVT, we used mRS distributions at different time points provided by a pooled analysis from 2004. A more recent study with a larger sample size only reported odds ratios specific for the cutoff of mRS score of 0 to 1, which are not valid to use in shift-

The effect of simultaneous changes in individual patient characteristics and regional workflow characteristics on the optimal prehospital transportation strategy for individual patients can be further explored by using the online tool at https://mrpredicts.shinyapps.io/triage/ (Figure 3).

Online Tool
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Discussion
Our decision model shows that direct transportation to an intervention center can be beneficial for patients with a high risk of having an LVO, but will likely lead to worse outcomes when the risk is low, especially in scenarios with longer driving times. Combining individual likelihood of LVO and estimated driving times on a case-by-case basis could improve prehos-

Figure 2. The optimal transportation strategy based on the likelihood of large vessel occlusion. Primary stroke center: the nearest nonendovascular-capable stroke center; intervention center: the nearest endovascular-capable stroke center. A, Represents the base case scenario (primary stroke center at 20 min and intervention center at 45 min); (B) the urban scenario (10 and 20 min, respectively); and (C) the rural scenario (30 and 90 min).
We constructed a decision model that predicts the optimal strategy based on individual patient characteristics and that can easily be adjusted for differences in geographic location or in-hospital workflow characteristics. Our online tool forms the basis of personalized prehospital triage of suspected ischemic stroke patients. Clinicians and researchers can plug in the specific characteristics of their own region into this tool to guide local prehospital triage policies. Although a fixed threshold the likelihood of LVO might be suitable for triage in urbanized regions with small distances between centers, our model showed that a higher threshold should likely be considered in more remote areas. After more extensive validation of the prehospital stroke scales, the tool can be used to combine the individual’s likelihood of LVO with estimated driving and local workflow times to improve prehospital triage decisions on a case-by-case basis. Future integration with a GPS-controlled navigation application will further facilitate personalized triage based on real-time information about driving and local workflow times.

Conclusions
The preferred prehospital transportation strategy for suspected stroke patients depends mainly on the likelihood of LVO, driving times, and in-hospital workflow times. We constructed a robust model that combines these characteristics and can be used to personalize prehospital triage, especially in more remote areas.

Sources of Funding
This study was funded by the Erasmus MC program for Cost-Effectiveness Research.
Table 6. Calculations of Outcome in the Model

<table>
<thead>
<tr>
<th>Event</th>
<th>Primary Stroke Center</th>
<th>Intervention Center</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Outcome with large vessel occlusion (log-odds)</strong>*</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Early reperfusion after IVT</td>
<td>1.35+ (−0.0026× time to IVT [primary stroke center]) +log-odds from baseline mRS distribution with LVO</td>
<td>1.35+ (−0.0026× time to IVT [intervention center]) +log-odds from baseline mRS distribution with LVO</td>
</tr>
<tr>
<td>EVT (with or without prior IVT)</td>
<td>1.35+ (−0.0026× time to EVT [transferred] +log-odds from baseline mRS distribution with LVO)</td>
<td>1.35+ (−0.0026× time to EVT [directly admitted] +log-odds from baseline mRS distribution with LVO)</td>
</tr>
<tr>
<td>No EVT</td>
<td>Log-odds from baseline mRS distribution with LVO</td>
<td>Log-odds from baseline mRS distribution with LVO</td>
</tr>
<tr>
<td><strong>Outcome without large vessel occlusion (log-odds)</strong>*</td>
<td></td>
<td></td>
</tr>
<tr>
<td>IVT</td>
<td>0.56+ (−0.0019× time to IVT [primary stroke center]) +log-odds from baseline mRS distribution without LVO</td>
<td>0.56+ (−0.0019× time to IVT [intervention center]) +log-odds from baseline mRS distribution without LVO</td>
</tr>
<tr>
<td>No IVT</td>
<td>Log-odds from baseline mRS distribution without LVO</td>
<td>Log-odds from baseline mRS distribution without LVO</td>
</tr>
</tbody>
</table>

EVT indicates endovascular treatment; IVT, treatment with intravenous thrombolitics; LVO, large vessel occlusion; and mRS, modified Rankin Scale.

*The baseline mRS distribution with and without LVO is illustrated in Figure I in the online-only Data Supplement.

**Disclosures**

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