External Validation of the DCD-N Score and a Linear Prediction Model to Identify Potential Candidates for Organ Donation After Circulatory Death: A Nationwide Multicenter Cohort Study

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Financial Disclosure: The authors received no funding to perform the research presented in this manuscript.

Disclaimer: The authors declare no conflicts of interest.
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Abbreviations:

DCD  Donation after circulatory death
WLST Withdrawal of life-sustaining treatment
ICU  Intensive care unit
OI   Oxygenation index
LPDCD Linear prediction model for donation after circulatory death
DBD  Donation after brain death
APACHE Acute physiology, age, chronic health evaluation
pPeak Peak airway pressure
PEEP Positive end expiratory pressure
ROC Receiver-operator curve
AUC  Area under the curve
HL   Hosmer and Lemeshow
UMCG  University Medical Center Groningen
PaO₂  Partial pressure of oxygen in arterial blood
CI    Confidence Interval
Abstract

Background - Donation after circulatory death (DCD) is a procedure in which after planned withdrawal of life-sustaining treatment (WLST) the dying process is monitored. A DCD procedure can only be continued if the potential organ donor dies shortly after WLST. This study performed an external validation of 2 existing prediction models to identify potentially DCD candidates, using one of the largest cohorts.

Methods - This multicenter retrospective study analyzed all patients eligible for DCD donation from 2010-2015. The first model (DCD N-score) assigned points for absence of neurological reflexes and oxygenation index (OI). The second model, a linear prediction model (LPDCD), yielded the probability of death within 60 min. This study determined discrimination (c-statistic) and calibration (HL test) for both models.

Results - This study included 394 patients, 283 (72%) died within 60 min after WLST. The DCD-N score had a c-statistic of 0.77 (95% CI 0.71 – 0.83) and the LPDCD model 0.75 (95% CI 0.68 – 0.81). Calibration of the LPDCD 60 min model proved to be poor (HL-test p<0.001).

Conclusions - The DCD-N score and the LPDCD model showed good discrimination, but poor calibration for prediction the probability of death within 60 min. Construction of a new prediction model on a large data set is needed to obtain better calibration.
Introduction

Organ transplantation improves quality of life and increases life expectancy of patients with end-stage organ failure. The demand for organ transplantations is likely to increase because of an ageing population and an increased patient survival, leading to more retransplantations. Donation after circulatory death (DCD) is a procedure in which after planned withdrawal of life-sustaining treatment (WLST) the course of the dying process is precisely monitored at an intensive care unit (ICU). DCD exposes organs to accumulate warm ischemic injury during the period between withdrawal of treatment and actual cardiac arrest, the agonal phase. The agonal phase refers to a time period that begins after decrease of oxygen saturation below SpO2 < 80% or systolic blood pressure below SBP < 80mmHg. After the circulatory arrest, there is a no-touch period of 5 minutes. After this 5 minutes the donor will be transported to the operation room. Identification of patients who will die shortly (within the prespecified time interval) after withdrawal of treatment is therefore of great importance, as longer time periods will result in organs too damaged for transplantation. In many countries, donors who do not fulfil the legal criteria for donation after brain death (DBD) are considered for DCD donation. DCD has proven to be a valuable source of donor organs in times of shortage. The Netherlands have a relatively high percentage (>50%) of DCD donors compared to surrounding Eurotransplant countries. In the Netherlands, the maximum duration of the agonal phase is 60 min for lung, liver and pancreas retrieval and 120 min for kidney procurement. In the United States 30 minutes is a common threshold for liver donation. In approximately 20% of potential DCD donors, circulation persists for more than 120 min after WLST and the donation procedure is cancelled, because of the expected organ damage. Avoiding the initiation of unsuccessful DCD procedures may reduce the discomfort for the patients’ relatives and for the patients who are already called into hospital in anticipation of a donor organ. Early
identification of the probability that a patient will die within 120 minutes will also ensure logistic benefits for transplant care teams. Additionally, reliably predicting the duration of this period may avoid unnecessary costs as well. The challenges of DCD include: screening and selecting patients, supporting and maintaining the trust of bereaved families and managing and minimizing the consequences of warm ischemia in such a way that is both acceptable professionally and ethically, and according to national laws. In practice, ICU clinicians estimate the time to death by making use of physiological and neurological tests. Currently, they do not use a predictive scoring system to assess the likelihood of the patient dying within an acceptable time period. The APACHE (Acute Physiology, Age, Chronic Health Evaluation) tool is frequently used to predict the mortality risk for DCD donors although this does not predict the length of the agonal phase. In 2012 a scoring system for circulatory death in patients in neurocritical state (DCD-N score) has been developed by Rabinstein and colleagues. This model was aimed at predicting the probability of death within 60 min after WLST, which is in general the maximum allowed length of the agonal phase in the United States. Some centers have been using the maximum duration of 120 min for kidney procurement. The DCD-N score takes 4 neurological variables into account, known to be associated with the time of death after WLST, absence of brainstem reflexes (cough reflex, cornea reflex and extensor or motor reflex to pain) which, combined with the oxygenation index (OI) > 3.0, are used as predictors for the calculation of the probability of the patient’s death within 60 minutes. The same group also constructed a second model, the ‘linear prediction for donation after circulatory death model’ (LPDCD-model), in which they adjusted the weight of the variables on the basis of strength of associations identified in earlier studies. Although this study has many strengths, no proper validation was performed, an essential step to assess reliability and applicability. The previous study also included nonpotential donors and this study wanted to validate these
models using only potential DCD donors. The aim of this study was to perform a proper external validation of the 2 models using one of the largest DCD cohorts to date.

**Materials and Methods**

*Study design*

We set up a multicenter retrospective observational cohort study with ICU patients aged 18 years and older who were medically and legally eligible for DCD. All procedures took place between 2010 and 2015 at 6 participating hospitals in the Netherlands. All potential donors who were at the Intensive Care Unit, waiting for withdrawal of life-sustaining treatment, were included in the database. The retrospective inclusion of patients in our study ended in 2015 due to the start of a new large multicenter prospective study which aims at developing a novel prediction model for the length of the agonal phase after WLST. Exclusion criteria were infections, lack of mechanical ventilation, and euthanasia. All variables were collected from medical files or electronic patient data management systems (EPIC, Epic systems corporation, Wisconsin, USA, ChipSoft, HiX software, Amsterdam, the Netherlands and Metavision provided by iMDsoft, Needham, Massachusetts, USA). Demographic characteristics included: age, gender, body mass index, admission date and time of the ICU and WLST date and time. Neurological diagnoses were classified according to the International Statistical Classification of Diseases and Related Health Problems by the World Health Organization. Hemodynamic parameters, necessary for the calculating of the OI, were assessed just before the patient’s WLST. Brainstem, motor and extensor reflexes were tested by a neurologist according to the Dutch protocol for determining brain death.

Our study was registered in the University Medical Center Groningen research register. Due to the descriptive character of this study, our institution’s Medical Ethics Committee granted dispensation for the Dutch law regarding patient-based medical research (WMO) obligation. Patient data were processed and electronically stored according to the Declaration of Helsinki.
for medical research involving human subjects. The clinical and research activities were consistent with the Principles of the Declaration of Istanbul as outlined in the ‘Declaration of Istanbul on Organ Trafficking and Transplant Tourism’.

**Measurements**

The OI was computed as: \(100 \times \left( F_{1}O_{2} \times \text{mean airway pressure in cm H}_2\text{O} \right) / P_{a}O_{2} \text{ in torr}\).² The mean airway pressure is half the combination of peak airway pressure (pPeak) in cm H₂O and the positive end expiratory pressure (PEEP) in Torr. An OI > 3.0 was defined as elevated.² These variables were assessed at the last examination before WLST. For the external validation of the LPDCD-model this study used the following formula:

\[
\text{Exp(logit)}/1+\text{Exp(logit)} = -2.49 + (0.90 \times \text{Absent cornea reflex}) + (1.65 \times \text{Absent cough reflex}) + (0.98 \times \text{Absent extensor or motor reflex}) + (0.12 \times \text{Oxygenation index})
\]

² Receiver-operator curve (ROC) analysis was used for the external validation of the DCD-N score.⁸ For validation of the LPDCD model all reflexes were used dichotomized and the OI as a categorical variable. The validation of the DCD-N score required absence of the neurological reflexes and the OI as dichotomized variables.

**Statistical analysis**

Univariate and multivariate logistic regression analysis were used to calculate the odds ratio for death within 60 minutes as a binary outcome variable. ROC DeLong analysis was used to assess how well the DCD-N score predicts the chance for the potential donor to die within 60 or 120 minutes after WLST. The corresponding area under the curve (AUC) is then calculated, as a measure for each model’s discrimination. An AUC of 0.7-0.8 was regarded as acceptable. An area of 0.8-0.9 was considered excellent, and more than 0.9 as outstanding.²⁸

Discrimination is the degree to which risk estimates from a model characterize different patient prognoses.⁵ For calibration, this study used Hosmer and Lemeshow goodness of fit (HL) test to predict the chance of dying within 30, 60 or 120 min after WLST. The database
had 2.6% missing data. All analyses were repeated after applying multiple imputation in order to deal with missing data.

**Results**

*Baseline characteristics*

This study enrolled a total of 406 patients at the University Medical Center Groningen (UMCG), Elisabeth Twee Steden Hospital (Tilburg), Isala Hospital (Zwolle), Catharina Hospital (Eindhoven), Radboud University Medical Center (Nijmegen) and at Medisch Spectrum Twente (Enschede), all located in the Netherlands. Twelve patients who did not fulfill the inclusion criteria were excluded. Hence, analyses in this study are based on a total of 394 patients. All donations took place between 2010 and 2015. Two hundred eighty-three (72%) patients died within 60 min after WLST. Time to death after WLST ranged from 2 to 1253 minutes with a median of 32 minutes. There was no significant difference in age, sex and physiological diagnosis between patients that died within or after 60 minutes. Within this cohort, 2.6% of data could not be reliably retrieved from the records. This concerned the variables of neurological reflexes and variables needed for the calculation of the oxygenation index. The scorings systems attributed points to: absence of cough reflex (2 points), absence of cornea reflex (1 point), absent extensor or motor reflex to pain (1 point) and an oxygenation index > 3.0 (1 point).

Table 1a summarizes the demographic variables of the study population divided into 2 cohorts: death within 60 minutes and death after 60 minutes. Table 1b summarizes the demographics divided into death within 120 minutes and death after 120 minutes. The LPDCD model with a cutoff score of a probability of 0.80 showed a sensitivity of 83.1%, a specificity of 50.8%, a positive predictive value of 37.5%, and a negative predictive value of 89.4% to predict death within 60 min after WLST.
The DCD-N score model showed with the same cutoff score a sensitivity of 87.6%, a specificity of 45.6%, a positive predictive value of 36.4%, and a negative predictive value of 91.7% to predict death within 60 min after WLST.

The binary regression analysis is shown in Table 2 with all the variables combined with death within 60 minutes. It showed that absence of cough reflex, cornea reflex, extensor – or motor reflex and an oxygenation index > 3.0 are associated with a higher probability of death within 60 minutes. ROC analysis for the DCD-N scorings system showed an AUC of 0.71 (95% CI 0.66 – 0.77) for prediction death within 30 minutes (Figure 1a). ROC analysis for the DCD-N scorings system showed an AUC of 0.77 (95% CI 0.71 – 0.83) for prediction death within 60 minutes (Figure 1b). The ROC analysis for the DCD-N scorings system of death within 120 minutes showed an AUC of 0.80 (95% CI 0.74 – 0.86) (Figure 1c). ROC analysis for the LPDCD 30 min model showed an AUC score of 0.71 (95% CI 0.65 – 0.76) (Figure 2a).

Calibration of the LPDCD 30 min model showed that the model underpredicted and overpredicted the probability of death (HL-test p<0.001) (Figure 2b). ROC analysis for the LPDCD 60 min model showed an AUC score of 0.75 (95% CI 0.68 – 0.81) (Figure 2c).

Calibration of the LPDCD 60 min model showed that the model underpredicted and overpredicted the probability of death (HL-test p<0.001) (Figure 2d). ROC analysis for the LPDCD 120 min model showed an AUC score of 0.83 (95% CI 0.77 – 0.88) (Figure 2e).

Calibration of the LPDCD model showed that the model under predicted the probability of death (HL-test p<0.001) (Figure 2f). As the DCD-N score does not result in an actual probability, but merely stratifies risk in terms of an integer between 0 and 5, no calibration measures could be calculated for this model.
Discussion

The DCD-N and LPDCD models are originally made for prediction time to death within 60 minutes after WLST. However, this study shows that the DCD-N and LPDCD models can make an acceptable prediction for death within 30, 60, but also 120 minutes after WLST. These models show good and excellent discrimination, which means that the models are able to predict which patients will die within 60 minutes. Inadequate discrimination is a more important failing than poor calibration because calibration can be improved by updating the model.\textsuperscript{9} Calibration of the LPDCD model shows that the validation model underpredicted and overpredicted the probability of death. This can be due to the fact that more variables have to be combined as strong predictors for time to death within 60 minutes. Better calibration is necessary for making the model more suitable for daily practice. Construction of a new prediction model on this large data set is needed to obtain better calibration.

In order to obtain a better reflection of the clinical conditions in practice this study included one of the largest cohorts of 394 patients. Previous external validation studies included smaller cohorts with a maximum of 211 patients.\textsuperscript{2,5,10-12} These previous studies included also nonpotential donors or included a very selected population, whereas this study focused on validating these models using nothing but potential DCD-donors.

The DCD program inevitably includes a number of potential donors who do not die within the established period of 60 minutes.\textsuperscript{10} In these situations, identification of appropriate DCD candidates is essential. This large retrospective multicenter study confirms that loss of brainstem reflexes and an oxygenation index > 3.0 are associated with death within 60 minutes after WLST. These results can support future donor management and provide information to relatives. In addition to the DCD-N and the LPDCD model, various other variables may have to be taken into account when prediction of death within 60 minutes is attempted. Several studies have demonstrated both a positive and negative effect of the use of
sedatives and analgesics.\textsuperscript{10,13,14} One study based on a Dutch population concluded that higher dosages of sedatives and opioids were associated with death in more than 60 minutes and concluded that it is useful as a predictor for death after 60 minutes.\textsuperscript{10}

The APACHE prognostic system is the current tool to predict mortality risk for critically ill adults. This system is primarily used to determine the required level of care in ICU patients but is neither developed nor validated to predict the agonal phase in DCD patients. Nonetheless, the APACHE score is often applied for this purpose. Other studies set up to identify appropriate DCD candidates include the UNOS and the University of Wisconsin criteria.\textsuperscript{4} Interestingly, neither the University of Wisconsin criteria nor the UNOS criteria model incorporate the degree of neurological injury. These scoring systems cover neurological information at the level of consciousness which is not a predictor of early death after WLST in neurocritical patients.\textsuperscript{15} Moreover, these scoring systems require disconnection from mechanical ventilation longer than 10 minutes.\textsuperscript{4,16} One of these studies used the UNOS criteria for predicting time to death within 60 min after WLST.\textsuperscript{16} This study showed that the absence of any criterion was associated with a low probability of death within 60 min.\textsuperscript{16} However, such prolonged ventilator disconnection may cause distress for patients and relatives. In contrast, both the DCD-N score and the LPDCD model facilitate assessment of patients while they remain supported by mechanical ventilation. Because of this difference, these models appear to provide a more efficient, easily applicable and possibly more acceptable way of predicting time to death.

The absence of neurological variables such as corneal and cough reflexes and motor- or extensor response to pain as predictors for time to death have been reported in previous studies.\textsuperscript{12,17} A study among 149 patients showed that absence of pupil reflexes was a significant predictor for the course after WLST.\textsuperscript{12} Unfortunately, external validation with a
smaller sample size (n=82) showed no statistical differences, which was most likely an effect of low power or differences in sample size.\textsuperscript{17}

Prediction of time to death solely based on clinical judgement is a proven inaccurate method with a fairly high sensitivity but low specificity.\textsuperscript{2,10} Based on these outcomes, it was even concluded that for each medically and legally eligible potential DCD donor a DCD procedure should be started to avoid loss of potential organs. Although this approach is very understandable given the increasing shortage of suitable organs, each noneffectuated donation procedure causes discomfort for the relatives of the donor and the patients who are waiting for the anticipated transplantation. Also, the effort in donor recruitment and management, as well as the resulting costs, are considerably higher, with a lower and more uncertain organ yield when no selection is applied to avoid noneffectuated DCD donors. Not-effectuated donations have a deep impact on already grieving families and put psychological and physical strain on procurement teams and ICU-staff.\textsuperscript{18,19} This study will support the management of expectations of both the donor family and the treating physician and may support clinical decisions on the feasibility of planning a certain DCD procedure.

Given the current shortage of deceased donor organs, the latter should be done with great caution, as to avoid an increase in donor nonutilization due to predicted, but not fully reliable high odds of a prolonged agonal phase.

In conclusion, validation of both existing models showed acceptable discrimination, but poor calibration with underestimation and overestimation of the probability of death within 30, 60 and 120 minutes. Our external validation of the DCD-N and LPDCD model is the first step in the process of developing a new predictive model using a large prospective cohort that can more accurately identify potential DCD candidates without losing available viable donors.

**Acknowledgments**

No acknowledgements
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Figure legends

Figure 1a. Receiver operating characteristic curve based on the multivariable DCD-N model for prediction of death within 30 minutes.

Figure 1b. Receiver operating characteristic curve based on the multivariable DCD-N model for prediction of death within 60 minutes.

Figure 1c. Receiver operating characteristic curve based on the multivariable DCD-N model for prediction of death within 120 minutes.

Figure 2a. Receiver operating characteristic curve based on the LPDCD multivariable model for prediction of death within 30 minutes.

Figure 2b. Calibration plot with the combination of the observed proportion and the predicted mortality for the LPDCD 30 minutes model.

Figure 2c. Receiver operating characteristic curve based on the LPDCD multivariable model for prediction of death within 60 minutes.

Figure 2d. Calibration plot with the combination of the observed proportion and the predicted mortality for the LPDCD 60 minutes model.

Figure 2e. Receiver operating characteristic curve based on the LPDCD multivariable model for prediction of death within 120 minutes.

Figure 2f. Calibration plot with the combination of the observed proportion and the predicted mortality for the LPDCD 120 minutes model.
### Patients (n=394)

| Age, years | 54.3 (18 – 75) |
|------------|----------------|
| Sex, female | 140 (36%) |

| Death within 60 min (n=283) | Death after 60 min (n=111) | T-test (P-value) | Chi-square (P-value) |
|-----------------------------|-----------------------------|------------------|----------------------|
| Age, years                  | 53.99 (18 – 75)             | 54.86 (22–75)    | 0.748                | 0.197                |
| Sex, female                 | 97 (34%)                    | 43 (39%)         | 0.087                | 0.348                |
| Sex, male                   | 186 (66%)                   | 68 (61%)         |                      |                      |

**Primary diagnosis**

| Traumatic brain injury (TBI) | 74 (26%) | 27 (24%) |
|-----------------------------|----------|----------|
| Intracranial hemorrhage (ICH) | 23 (8%) | 14 (13%) |
| Subarachnoid hemorrhage (SAH) | 46 (16%) | 23 (21%) |
| Ischemic cerebrovascular accident (CVA) | 35 (12%) | 9 (8%) |
| Anoxic damage after CRP/Cardiac arrest | 70 (25%) | 34 (31%) |
| Spinal cord injury | 24 (9%) | 1 (1%) |
| Respiratory failure | 3 (1%) | 0 (0%) |
| Intoxication, suicide | 5 (2%) | 2 (2%) |
| Unknown | 3 (1%) | 1 (1%) |

**Tables** Table 1a. Baseline demographics stratified into death within 60 minutes and death after 60 minutes.
Table 1b. Baseline demographics stratified into death within 120 minutes and death after 120 minutes.

| Patients (n=394) | Death within 120 min (n=319) | Death after 120 min (n=75) | T-test (P-value) | Chi-square (P-value) |
|------------------|-------------------------------|----------------------------|------------------|-----------------------|
| Age, years       | 53.92 (18 – 75)              | 55.39 (22– 75)            | 0.503            | 0.350                |
| Sex, female      | 110 (34%)                    | 30 (40%)                  | 0.206            | 0.483                |
| Sex, male        | 210 (66%)                    | 44 (60%)                  |                  |                       |
| Primary diagnosis|                               |                            | 0.151            | 0.296                |
| Traumatic brain injury (TBI) | 86 (27%) | 15 (20%)               |                  |                       |
| Intracranial hemorrhage (ICH) | 29 (9%) | 8 (11%)              |                  |                       |
| Subarachnoid hemorrhage (SAH) | 56 (18%) | 15 (20%)        |                  |                       |
| Ischemic cerebrovascular accident (CVA) | 37 (12%) | 7 (9%)            |                  |                       |
| Anoxic damage after CRP/Cardiac arrest | 76 (24%) | 26 (35%)        |                  |                       |
| Spinal cord injury | 24 (8%)             | 1 (1%)                    |                  |                       |
| Respiratory failure | 3 (1%)        | 0 (0%)                   |                  |                       |
| Intoxication, suicide | 5 (2%)       | 2 (2%)                   |                  |                       |
| Unknown          | 3 (1%)                       | 1 (1%)                    |                  |                       |
| Variable                        | Death within 60 min (n=283) | Death after 60 min (n=111) | Odds ratio (95% CI) | p value |
|--------------------------------|------------------------------|----------------------------|---------------------|---------|
| Absent corneal reflex          | 212 (77%)                    | 35 (35%)                   | 3.237 (1.820 – 5.756) | <0.0005 |
| Absent cough reflex            | 218 (80%)                    | 38 (37%)                   | 4.306 (2.419 – 7.664) | <0.0005 |
| Extensor or absent motor response | 267 (94%)                  | 92 (84%)                   | 2.468 (0.995 – 6.123)  | 0.051   |
| Oxygenation index > 3.0        | 187 (70%)                    | 76 (74%)                   | 0.688 (0.371 – 1.276)  | 0.236   |

Table 2. Binary logistic regression analysis with distribution of variables of interest according to time to death after WLST.
Figure 1a

ROC for DCD-N (30 min model)
Figure 1b

ROC for DCD-N (60 min model)
Figure 1c
Figure 2a

ROC for LPDCD (30 min model)
Figure 2b

Calibration plot of the LPDCD 30 minutes model

Day: 0.414
C (PROC): 0.767
K2: -0.015
D: -0.029
U: 0.138
G: -0.167
B1er: 0.228
Intercept: -0.260
Slope: 0.517
Emax: 0.325
E90: 0.033
Eavg: 0.120
Siz: 5.975
Sp: 0.000
Figure 2c
Figure 2d

Calibration plot of the LPDCD 60 minutes model

- Day: 0.499
- C (ROCR): 0.748
- R2: 0.111
- D: 0.075
- U: 0.093
- Q: 0.021
- Eirr: 0.172
- Intercept: 0.602
- Slope: 0.666
- Emax: 0.172
- ES0: 0.165
- Eavg: 0.090
- Bz: 0.687
- Sp: 0.492
Figure 2e
Figure 2f

Calibration plot of the LPDCD 120 minutes model

- Dxy: 0.851
- C (ROC): 0.825
- R2: 0.097
- D: 0.052
- U: 0.160
- Q: 0.106
- E(r): 0.143
- Intercept: 1.119
- Slope: 1.064
- Emax: 0.272
- E90: 0.256
- Eavg: 0.150
- Ez: -1.783
- Ep: 0.075