



ORIGINAL ARTICLE

Proteomic analysis of machine perfusion solution from brain dead donor kidneys reveals that elevated complement, cytoskeleton and lipid metabolism proteins are associated with 1-year outcome

ESOT 2021
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SUMMARY

Assessment of donor kidney quality is based on clinical scores or requires biopsies for histological assessment. Noninvasive strategies to identify and predict graft outcome at an early stage are, therefore, needed. We evaluated the perfusate of donation after brain death (DBD) kidneys during nonoxygenated hypothermic machine perfusion (HMP). In particular, we compared perfusate protein profiles of good outcome (GO) and suboptimal outcome (SO) 1-year post-transplantation. Samples taken 15 min after the start HMP (T1) and before the termination of HMP (T2) were analysed using quantitative liquid chromatography–tandem mass spectrometry (LC-MS/MS). Hierarchical clustering of the 100 most abundant proteins showed discrimination between grafts with a GO and SO at T1. Elevated levels of proteins involved in classical complement cascades at both T1 and T2 and a reduced abundance of lipid metabolism at T1 and of cytoskeletal proteins at T2 in GO versus SO was observed. ATP-citrate synthase and fatty acid-binding protein 5 (T1) and immunoglobulin heavy variable 2-26 and desmoplakin (T2) showed 91% and 86% predictive values, respectively, for transplant outcome. Taken together, DBD kidney HMP perfusate profiles can distinguish between outcome 1-year post-transplantation. Furthermore, it provides insights into mechanisms that could play a role in post-transplant outcomes.

Transplant International 2021; 34: 1618–1629

Key words

biomarker discovery, complement activation, donation after brain death, hypothermic machine perfusion, kidney preservation, proteomics

Received: 25 June 2021; Revision requested: 14 July 2021; Accepted: 15 July 2021

Introduction

Kidney transplantation is the only curative treatment for patients suffering from end-stage renal disease (ESRD) as it can permanently replace a patient's kidney function. In 2019, almost 100 000 kidney transplantations were performed globally, and it is the most transplanted organ worldwide. [1]

Currently, donation after brain death (DBD) donors are the main source for donor kidneys [2]. Brain death causes various pathophysiological effects in the donor such as haemodynamic instability due to a severe reduction in blood flow, immune system activation including a cytokine storm and complement activation followed by leucocyte infiltration [3–4]. These effects have shown to have a negative impact on graft outcome [5].

Standard criteria donors (SCD) are preferred as they show favourable outcomes compared with other donor types [6–7]. The continuing shortage of donor organs forces the use of organs from donors that do not meet the standard criteria such as extended criteria donors (ECD). ECD are DBD with an age above 60, or an age above 50 with two of the following characteristics: history of hypertension, terminal serum creatinine levels of ≥ 1.5 mg/dl or cerebrovascular cause of death [8–9]. Unfortunately, ECD kidney grafts have worse outcomes compared with SCD [9–12].

With the increased use of organs from ECD, the decision-making process for suitable donor organs is becoming even more critical. Current strategies to assess donor organ quality are based on clinical characteristics like the kidney donor risk score or kidney donor profile index (KDPI), [13] 14 or require a biopsy for histological assessment. However, taking biopsies may cause unnecessary injury to the kidney, and the obtained information is intrinsically representative of a very focal site. Hence, quality assessment using biopsies is under debate [15]. Due to such limitations, the quality of the graft cannot always be correctly assessed, causing many organs to be discarded, possibly unnecessarily. Therefore, noninvasive strategies to identify and predict graft outcome are highly needed.

As of 2016, all donor kidneys derived from deceased donors in the Netherlands are preserved by means of nonoxygenated hypothermic machine perfusion (HMP) instead of static cold storage (SCS) [16]. HMP is a preservation method that allows continuous perfusion of the graft with perfusion solution in the time frame between donation and transplantation. The use of HMP itself is associated with improved graft outcome [17–22]. Furthermore, it creates a new horizon for

measuring secreted predictive kidney graft biomarkers in a noninvasive manner, providing the opportunity to evaluate graft quality before transplantation. Several studies have demonstrated that organ preservation solution is a potential source for biomarkers associated with post-transplantation organ function, although these studies did not look at HMP perfusate or only quantified selected proteins [23–25].

Our aim was to perform a proteomic perfusate analysis of DBD kidneys preserved using HMP by identifying the differences between proteomic profiles of kidneys with a good and suboptimal outcome 1-year post-transplantation. Furthermore, our aim was to identify potential biomarkers for organ quality and unravel underlying pathways that influence transplant outcome using an unbiased proteomics approach.

Materials and methods

Characteristics of donor cohort

Twenty-two DBD kidneys with a good or suboptimal outcome 1-year post-transplantation were selected ($n = 11$ per group) that were transplanted between 11 January 2016 and 31 December 2017. Good outcome (GO) was defined as kidney function with an estimated glomerular filtration rate (eGFR) ≥ 45 ml/min/1.73 m² and a serum creatinine level of ≤ 120 μ mol/l, whereas suboptimal outcome (SO) was defined as kidneys with an eGFR ≤ 45 ml/min/1.73 m² and a serum creatinine level of ≥ 120 μ mol/l 1-year post-transplantation (Fig. 1). Selected kidneys were matched on various donor and recipient characteristics (Table 1). Characteristics were analysed using SPSS statistics 28. Noncategorical variables were tested for normal distribution using the Shapiro–Wilk normality test. Two-sample *t* test or Mann–Whitney U test was used to test for two independent groups of continuous variables. Chi-square test or Fisher's exact test were used for discrete variables. Exclusion criteria were kidneys allocated to recipients receiving a multi-organ transplant and donors with acute kidney injury.

Sample collection and storage

Nonoxygenated HMP was performed using either the LifePort Kidney Transporter machine (organ recovery systems) with a mean arterial pressure of 30 mmHg or the kidney assist transporter (organ assist) with a mean arterial pressure of 25 mmHg and a temperature between 1 and 8°C. Five hundred millilitres of

University of Wisconsin machine perfusion solution (UW-MPS; Bridge to Life) was used as the preservation solution. A total of 44 perfusate samples were collected; after 15 min of HMP (T1) and just before ending HMP (T2). The perfusate was centrifuged at 1000 g. for 12 min at 4°C. The supernatant was aliquoted and stored at −80°C until proteomic analysis (Fig. 1).

Ethical approval

The audit committee of the Dutch Transplant Foundation has approved the collection of HMP perfusate material and analysis of data as described within this manuscript. The research register number is 201600608 for Biobank perfusate donated kidneys.

Preparation of kidney perfusate samples

Perfusate samples were depleted of albumin by means of chemical depletion. In short, a 500 µl aliquot of each sample was precipitated with 42% ethanol 4°C for 1 h. Samples were centrifuged at 16 000 × g. for 45 min at 4°C, supernatant was removed and pellets were dried. Pellets were then resuspended in 50 µl 6 M urea buffer and 450 µL phosphate-buffered solution, and the

chemical depletion process was repeated as described above. Final pellets were then resuspended in 6 M urea and 100mM Tris/HCl (pH 8, 2). Protein concentration was determined using the bicinchoninic acid (BCA) Protein Assay Kit (Thermo Scientific) according to manufacturer's protocol.

Next, 10 µg of protein per sample was applied on an SDS-PAGE gel and stained with Coomassie Blue, after which the albumin-containing bands were removed. The remaining bands were prepared for in-gel digestion as previously described by Shevchenko *et al.* [26]. Protein digestion was performed using 20 ng/µl trypsin in ice-cold 50 mM ammonium bicarbonate at 37°C overnight. Tryptic peptides were extracted using 50% acetonitrile, 5% formic acid in MilliQ-H₂O, followed by 85% acetonitrile, 5% formic acid in MilliQ-H₂O. Peptides were dried using a MiVac DUO concentrator (SP Scientific) and resuspended in 20 µl 2% acetonitrile, 0.1% TFA in MilliQ-H₂O. LC-MS/MS analysis was performed using a Dionex Ultimate 3000 nano-ultra high-pressure reverse phase chromatography coupled online to a Q Exactive high field (HF) mass spectrometer (Thermo Scientific) as previously described by Fye *et al.* [27]. In short, samples were separated on an EASY-Spray PepMap RSLC C18 column (500 mm × 75 mm, 2 mm particle size, Thermo Scientific) over a 60-min gradient of 2–35% acetonitrile

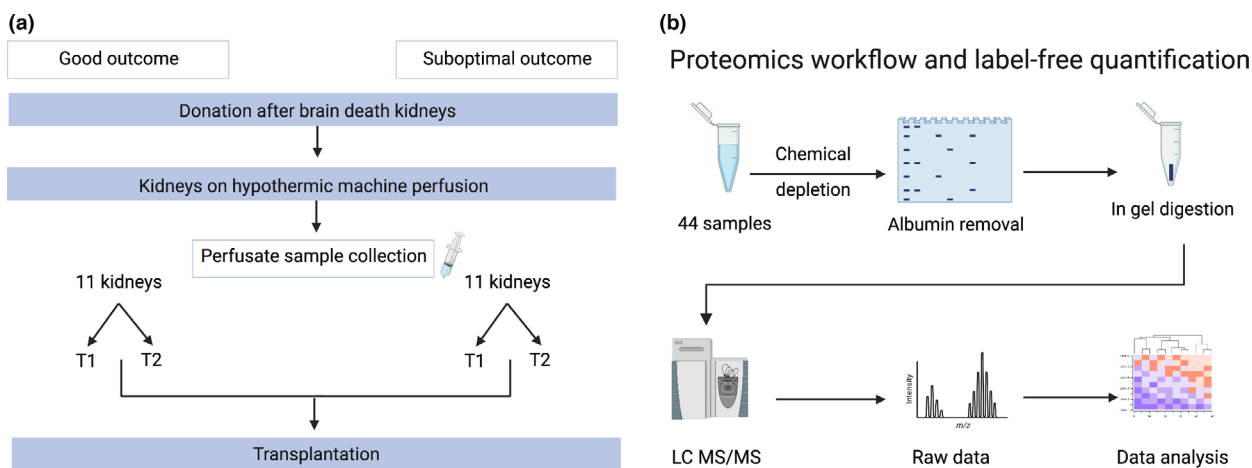


Figure 1 Experimental design and workflow. (a) Overview of kidney perfusate sample collection. Perfusate samples were selected on the basis of 1-year transplant outcomes. Good outcome (GO) was defined as kidney function with an estimated glomerular filtration rate (eGFR) ≥ 45 ml/min/1.73 m² and a serum creatinine level of ≤ 120 µmol/l, whereas suboptimal outcome (SO) was defined as kidneys with an eGFR ≤ 45 ml/min/1.73 m² and a serum creatinine level of ≥ 120 µmol/l. Samples from each kidney were taken at 2 time points: 15 min after the start of HMP (T1) and at the end of HMP (T2). (b) Workflow for proteomics analysis. A total of 44 samples were analysed (11 kidneys per group, 22 samples per outcome). Perfusate samples were chemically depleted and loaded onto a gel for SDS-PAGE. Albumin bands were removed, and samples were prepared for in-gel digestion, desalted and analysed using LC-MS/MS. Raw peaks and spectra were searched against Homo sapiens database and data were analysed. eGFR, estimated glomerular filtration rate; LC-MS/MS, liquid chromatography–tandem mass spectrometry. (Image created using biorender.com).

Table 1. Characteristics of donor and recipients.

Donor	Good Outcome	Suboptimal Outcome	P-value
Age y	53.6 (6.2)	57.2 (6.2)	0.195
Male [<i>n</i> (%)]	7 (63.6%)	3 (27.3%)	0.198
BMI [kg/m ⁻²]	25.9 (4.9)	26.8 (4.8)	0.689
History of hypertension [<i>n</i> (%)]	5 (45.5%)	4 (36.4%)	1.000
Terminal serum creatinine	60.2 (16.6)	68.6 (24.1)	0.349
Recipient	Good Outcome	Suboptimal Outcome	P-value
Age y	51.6 (9.1)	57.6(10.8)	0.174
Male [<i>n</i> (%)]	7 (63.6%)	8 (72.7%)	1.000
BMI [kg/m ⁻²]	24.0 (3.0)	25.5 (3.1)	0.373
≥3 HLA mismatches [<i>n</i> (%)]	9 (81.8%)	6 (54.5%)	0.361
Ischaemia times			
Total CIT (minutes)	969.9 (386.6)	689.18 (323.7)	0.086
WIT 2 (minutes)	32.4	37.8 (10.3)	0.264
Renal Outcome	Good Outcome	Suboptimal Outcome	P-value
DGF	0 (0%)	5 (45.5%)	0.035
3-month serum creatinine	95 (88–116)	189 (176–189)	0.000
3-month eGFR	68.6 (17.0)	28 (12.0)	0.000
12-month serum creatinine	103 (95–117)	196 (182–230)	0.000
12-month eGFR	67.4 (15.5)	27.6 (9.7)	0.000

Data are given as mean (SD), median (IQR) or *n* (%). *n*, number in group; BMI, body mass index; HLA, human leukocytes antigen; CIT, cold ischaemia time; WIT2, warm ischaemia time 2; DGF, delayed graft function.

in 5% DMSO, 0.1% formic acid at 250 nl/min. MS1 scans were acquired at a resolution of 60 000 at 200 m/z and the top 12 most abundant precursor ions were selected for HCD fragmentation.

MS data analysis and label-free quantitation

Raw MS data were searched against the Uniprot human sequence database (*Homo sapiens*) retrieved on 02.05.2017, and label-free quantitation was performed using MaxQuant software v 1.6.0.16 [28]. The following parameters were used: Trypsin/P for enzyme specificity, allowing up to 2 missed cleavages and a minimum peptide length of 7, carbamidomethyl (C) was set as fixed a modification, and oxidation (M) and deamidation (NQ) were chosen as variable modifications included in protein quantification, precursor mass tolerance at 7 ppm and fragment mass tolerance at 0.01 Da and quantification with a false discovery rate (FDR) of 1%. Match between runs was used. Label-free quantitation (LFQ) values were extracted from MS spectra in MaxQuant using the MaxLFQ algorithm and analysed using Perseus v1.6.12.0 [29]. In short, rows were log₂ transformed, filtered based on 70% valid values and missing values were replaced using imputation of the normal distribution. Hierarchical clustering and heat maps were

generated using Z-score and Pearson's correlation. For analysing the differences between the two groups, two-sample *t* tests were performed. Volcano plots were generated using obtained *p*-values and log₂ intensity differences using GraphPad Prism v8 (GraphPad Software, USA). Analyses were considered statistically significant when *P*-values < 0.05. STRING v11 [30] and Cytoscape v3.8.2 [31] were used to visualize and interpret the molecular function of proteins and pathways represented by the significant different proteins. Prediction models were made in SPSS statistics 28 using multivariable logistic regression, with forward and backward stepwise inclusion of model factors. Different models were compared with a chi-square test, the model discriminative potential was assessed by receiver operating characteristic (ROC) curves as previously described by van Balkom *et al.* [23], and *P*-values < 0.05 were considered statistically significant.

Results

Clinical characteristics

There were no significant differences observed in donor and recipient characteristics (Table 1). DGF, 3-month serum creatinine, 3-month eGFR, 12-month serum

creatinine and 12-month eGFR were significantly different between experimental groups (Table 1).

Kidney perfusate profiles

All 44 kidney perfusate samples were analysed using label-free quantitative mass spectrometry analysis (LFQ MS). In total, 1255 proteins were identified and quantified across all different groups. These proteins were filtered based on valid values set at a minimum of 70%, leading to a total of 498 quantified proteins (Table S1).

Perfusate protein profiles after 15 min of HMP discriminate between good and suboptimal outcome and reveal enriched complement in good outcome group

Perfusate samples collected at T1 were first analysed (Fig. 2). Unsupervised hierarchical clustering for Pearson correlation distance using 100 most abundant proteins detected at T1 was performed to examine data uniformity. Distinctive clustering was observed between the GO and SO group with the exception of 2 samples per group (Fig. 2A). STRING pathway analysis and functional enrichment of these 100 most abundant proteins showed that 66 of these proteins are located in extracellular regions (FDR: $1.39\text{E-}40$) and that 23 of these proteins belong to complement and coagulation cascades (FDR: $2.31\text{E-}32$) (Fig. 2B).

Figure 2C shows significant differentially expressed proteins between GO and SO at T1 (Table S1). Eighteen proteins were significantly upregulated in GO vs. SO. STRING pathway analysis and functional enrichment revealed that all significant proteins are located in extracellular regions (FDR: $9.01\text{E-}7$) (Fig. 2D). Furthermore, 5 of these proteins belong to complement and coagulation cascades including complement C1q subcomponent subunit B (C1QB), complement C1s subcomponent (C1S), complement C1r subcomponent (C1R) and C4b-binding protein alpha chain (C4BPA) (FDR: $1.36\text{E-}9$) (Fig. 2D). Ten proteins were significantly downregulated in GO vs. SO. STRING pathway analysis and functional enrichment revealed that 2 of these proteins, fatty acid-binding protein (FABP) 4 and 5, are involved in PPAR signalling (FDR: 0.004) (Fig. 2E).

Perfusate profiles at the end of HMP reveal enriched complement in good outcome group and enriched cytoskeletal proteins in suboptimal group

Perfusate samples at T2 were analysed (Fig. 3; Table S1). Unsupervised hierarchical clustering of 100

most abundant proteins at T2 showed no distinction between groups (Fig. 3A). STRING pathway analysis and functional enrichment of these 100 most abundant proteins showed that 67 of these proteins were located in extracellular regions (FDR: $2.28\text{E-}41$) and that 24 of these proteins belong to the complement and coagulation cascades (FDR: $4.2\text{E-}34$). (Fig. 3B).

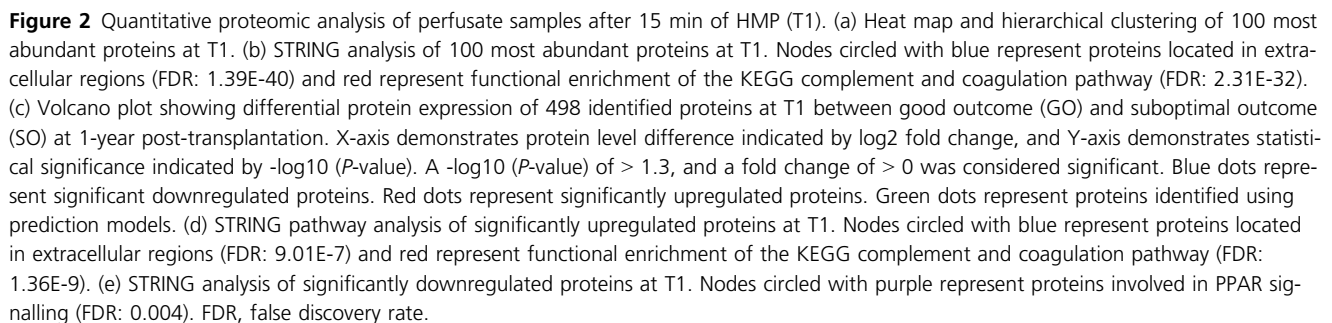
Figure 3C shows significant differentially expressed proteins between GO and SO at T2 (Table S1). Twenty-two proteins were significantly upregulated in GO vs. SO. STRING pathway analysis and functional enrichment revealed that 5 of these proteins belong to complement and coagulation cascades (FDR: $1.21\text{E-}8$) (Fig. 3D), of which 4 are involved in complement activation of the classical pathway including complement C1r subcomponent-like protein (C1RL), C1S and C1R (FDR: $1.6\text{E-}6$) (Fig. 3D). Twenty-six proteins were significantly downregulated in GO vs. SO. STRING pathway analysis and functional enrichment revealed that 14 of these proteins are cytoskeletal proteins including desmoplakin (DSP), talin-1 (TLN1) and alpha actinin-1 (ACTN1) (FDR: $1.33\text{E-}5$) (Fig. 3E). Thirteen of the downregulated proteins are affiliated with the immune system (FDR: $4.28\text{E-}5$) (Fig. 3D).

Outcome prediction based on differentially abundant proteins

To determine whether differentially abundant proteins are independent predictors of transplantation outcome, we performed a stepwise logistic regression analysis. At T1, ATP-citrate synthase (ALCY) together with FABP5 provided a predictive value of 91%, corresponding to a ROC with an area under the curve (AUC) of 0.97 (Fig. 4A). Furthermore, immunoglobulin heavy variable 2-26 (IGHV2-26) together with DSP provided a predictive value of 86%, a corresponding AUC of 0.95 in the ROC analysis at T2 (Fig. 4B). When adding donor sex as an extra factor to logistic regression analysis, the predictive value at T2 becomes 100% with a corresponding AUC of 1 (data not shown).

Discussion

The use of ECD organs for transplantation is increasing due to the continuing donor organ shortage, urging novel noninvasive strategies to identify and predict graft outcome. The aim of this study was to evaluate the flushed-out tissue-resident proteins that end up in the perfusate of DBD kidneys preserved using HMP via an unbiased proteomics



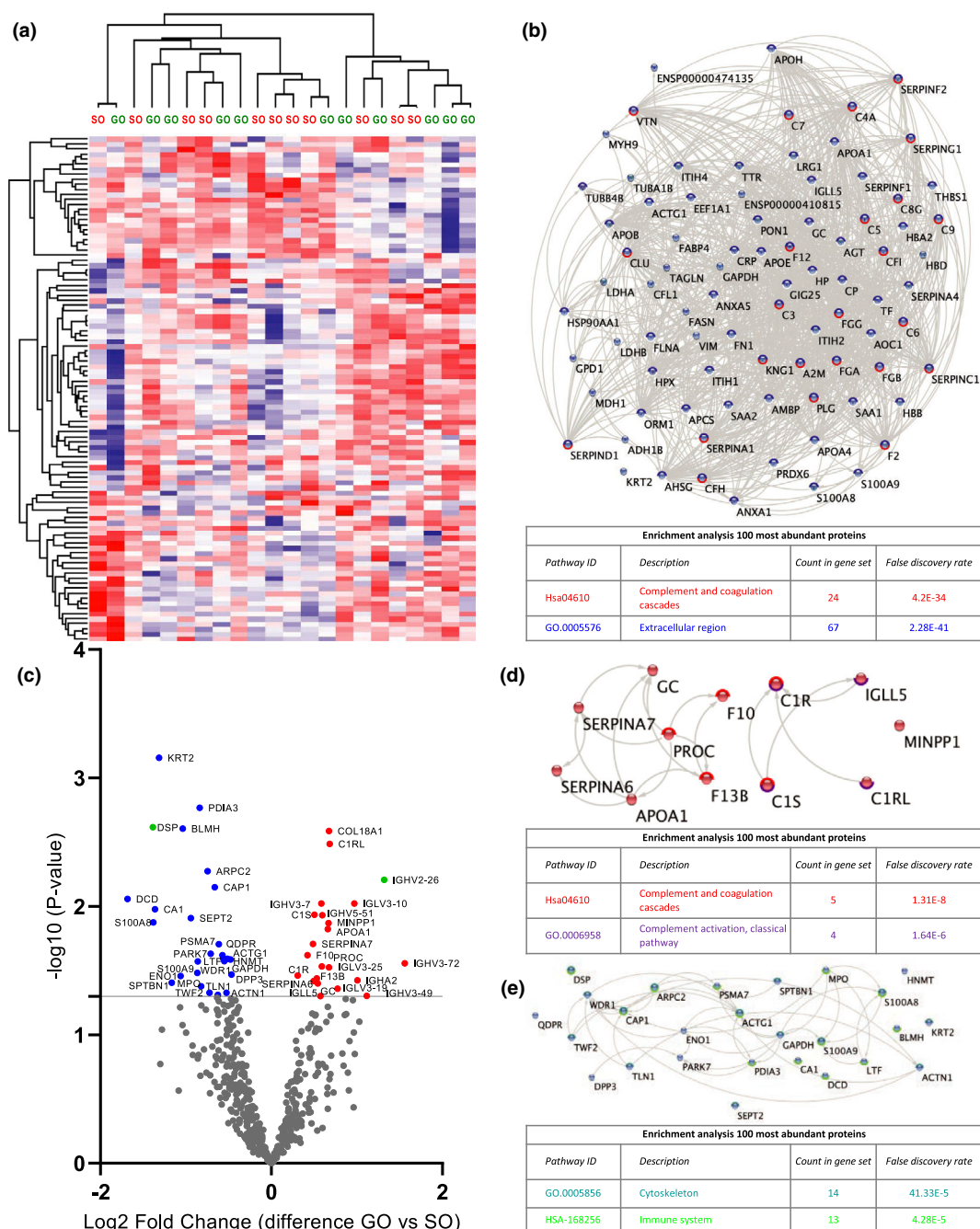


Figure 3 Quantitative proteomic analysis of perfusate samples taken before the termination of HMP (T2). (a) Heat map and hierarchical clustering of 100 most abundant proteins at T2. (b) STRING analysis of 100 most abundant proteins at T2. Nodes circled with blue represent proteins located in extracellular regions (FDR: 2.28E-41) and red represent functional enrichment of the KEGG complement and coagulation pathway (FDR: 4.2E-34). (c) Volcano plot showing differential protein expression of 498 identified proteins at T2 between good outcome (GO) and suboptimal outcome (SO) at 1-year post-transplantation. X-axis demonstrates protein level difference indicated by log2 fold change; Y-axis demonstrates statistical significance indicated by $-\log_{10}(P\text{-value})$. A $-\log_{10}(P\text{-value})$ of > 1.3 , and a fold change of > 0 was considered significant. Blue dots represent significant downregulated proteins. Red dots represent significantly upregulated proteins. Green dots represent proteins identified using prediction models. (d) STRING pathway analysis of significantly upregulated proteins at T2. Nodes circled with red represent functional enrichment of the KEGG complement and coagulation pathway (FDR: 1.21E-8) and purple represents proteins involved in complement activation of the classical pathway (FDR: 1.64E-6). (e) STRING analysis of significantly downregulated proteins at T2. Nodes circled with turquoise represent functional enrichment GO component cytoskeleton (FDR: 1.33E-5) and green represents proteins affiliated with the immune system (FDR: 4.28E-5). FDR, false discovery rate.

approach. We show that protein secretion profiles after 15 min of HMP results in distinctive clustering between DBD kidneys with a good and suboptimal outcome 1-year post-transplantation. To our knowledge, this is the first study that distinguishes between a good and suboptimal transplant outcome based on protein profiles identified in HMP perfusates of DBD kidneys.

Increase in complement proteins in HMP perfusate is associated with good kidney function at 1-year post-transplantation

It has been shown that brain death is associated with complement activation [32], and that the complement system is involved in transplant-related pathologies [33–34]. Although the main function of the complement system is to aid the immune system to attack and clear pathogens, it also plays a role in apoptotic cell removal. Excessive complement activation can lead to cell lysis and necrosis and may, therefore, be linked to the allograft induced immune attack and apoptotic clearance. [35–36] However, the exact underlying mechanisms of complement activation pathways during transplantation have been the topic of many studies and is still not fully resolved [32,37–43].

Therefore, it was somewhat surprising to find significantly elevated levels of C1r, C1s, C1q and C4BPA in perfusate samples of donor kidneys with a good outcome 1-year post-transplantation. C1r, C1s and C1q are the three subcomponents of the C1 complex that represent upstream proteins of the classical complement system. The classical pathway is initiated when the C1 complex binds to an activator. One of the main roles of C1q is the removal of apoptotic cells and debris [44]. The deficiency of these three proteins, especially C1q, has been highly associated with renal failure caused by autoimmune diseases such as systemic lupus erythematosus [44]. Furthermore, deposition of C1q in the kidney is associated with C1q nephropathy [45]. Interestingly, studies have shown that treatment with C1 inhibitor has potential protective effects during renal transplantation [37–38]. However, C1 inhibitor also has anti-inflammatory functions and anti-apoptotic effects and inhibits coagulation, contact and kinin systems [46]. Therefore, these protective effects could also be the consequence of reduced inflammation, attenuated oedema and decreased formation of microthrombi during transplantation. C4BPA is a complement regulator that inhibits the complement system via the classical and the lectin pathway. Furthermore, it protects

apoptotic cells from complement CD40-dependent stimulation of B cells [47]. During ischaemia, the endothelium becomes activated by cytokines, resulting in the breakdown of the glycocalyx and subsequent loss of regulators of the complement system such as C4b-binding protein [33]. The loss of these regulators means that the endothelial cell surface is no longer protected from attack by the complement, coagulation and contact systems. Our results suggest that a wash out of tissue-resident complement proteins detected in perfusates is associated with a good outcome, perhaps due to detachment of remnant unproductive complement attack complexes.

The role of complement activation during brain death appears to be controversial. For instance, Brown *et al.* (2006) showed that differential expression of C3 alleles by donor renal cells appears to have a negative effect on late graft outcome [41]. However, these findings could not be reproduced in larger cohort studies, as Varaguman *et al.* (2009) did not show a difference in transplant outcomes between donors with different C3 allotypes [42]. Furthermore, Damman *et al.* (2012) showed that the whole lectin complement pathway gene profile of the donor does not influence outcome [43]. Damman *et al.* (2011) showed that elevated complement C5b-9 levels in deceased donor plasma samples is associated with acute rejection in the recipient after renal transplantation. However, they did not find any significant association between donor complement activation levels and primary nonfunction (PNF), delayed graft function (DGF) or graft survival in the first year after transplantation [32]. Importantly, all of these reports were mainly focused on distinct components of the complement system in tissue and plasma, whereas our study was conducted in an unbiased manner using perfusate samples. Complement activation will inevitably occur in every donor kidney, but the amount of activation and/or complement composition may be a key determinant of the level of injury and repair. Understanding these aspects in the context of donor organ function warrants further investigation.

Disruption of cytoskeletal proteins is associated with a suboptimal transplant outcome

Proteomic perfusate analysis at the end of HMP revealed significant upregulation of cytoskeletal proteins in the suboptimal outcome group. This indicates that enriched cytoskeletal proteins in the perfusate are associated with a suboptimal outcome 1-year post-transplantation. These results are in line with a

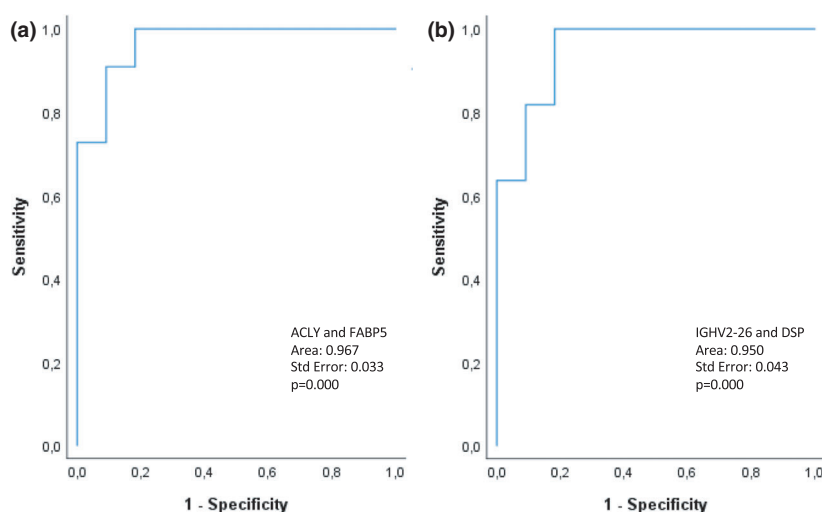


Figure 4 Receiver-operated (ROC) curves for the prediction models based on T1 and T2. (a) Perfusate analysis at T1 shows a 91% predictive value for ATP-citrate synthase (ALCY) and fatty acid-binding protein 5 (FABP5). (b) Perfusate analysis at T2 shows an 86% predictive value for immunoglobulin heavy variable 2-26 (IGHV2-26) and desmoplakin (DSP).

proteomics analysis performed by Coskun *et al.* (2016), who found secretion of cytoskeletal proteins of DBD donor kidneys during SCS. However, they did not include transplant outcome into their analysis [24].

Ischaemia has been associated with disruption of the cytoskeleton, leading to loss of tissue integrity [48–49]. However, the effects of brain death specifically on renal tissue integrity are still unknown. We observed enrichment of actinin-1 and talin-1 in the perfusate of the suboptimal outcome group. Actinin-1 is one of four actin-bundling proteins that seems to play a crucial role in podocyte architecture. Studies have shown that both the podocyte actin cytoskeleton, as well as podocyte-associated talin-1, play a crucial role in preserving an intact glomerular filtration barrier [50–51]. Enhanced excretion of cytoskeletal proteins could be the cause of lower eGFR rates in our suboptimal group due to loss of tissue integrity and disruption of the glomerular filtration barrier.

Desmoplakin together with IGHV2-26 showed a predictive value in which an increased expression of desmoplakin in the perfusate was correlated with a suboptimal transplant outcome. Desmoplakin is the most abundant part of the desmosome or intercellular junctions of epithelia and cardiac muscle, providing mechanical strength to tissues. Desmoplakin, therefore, plays a key role in tissue morphogenesis and integrity [52], but studies on the role of desmoplakin in renal function and tissue integrity are scarce. It has been shown that structural alterations in desmosomal junctions could potentially influence renal function [53]. However, more research into the role of desmoplakin should be conducted.

Lipid metabolism and renal function

ALCY, FABP4 and FABP5 were significantly upregulated in the perfusate collected after 15 min of HMP of kidneys with a suboptimal outcome. Furthermore, FABP5 together with ALCY showed a predictive outcome of 91%. ALCY is a key enzyme of *de novo* fatty acid synthesis [54]. FABP4 and FABP5 are key proteins in lipid metabolism and play an important role in inflammation via the PPAR- γ signalling pathway [55]. The effect of FABP5 and ALCY on renal function is unclear, however, overexpression of both proteins has been associated with renal carcinoma [56–57]. Excreted FABP4 in the serum has been associated with renal dysfunction [58]. Furthermore, Huang *et al.* (2018) revealed elevated levels of FABP4 in rat kidneys exposed to ischaemia/reperfusion injury (IRI) [59]. Inhibition of FABP4 in an acute renal injury mouse model has shown to attenuate renal structural damage and improve renal function [60]. Several studies have shown that that renal transplant-related IRI leads to disturbed metabolism due to energy deficit [59,61]. Our results suggest that this disturbed lipid metabolism plays a role in post-transplant outcome, and oxygenated HMP may be the key to limit this metabolic disturbance and improve outcome.

Impact of donor characteristics, sample time point and total CIT

Perfusate samples obtained at T1 resulted in distinctive clustering between a good and suboptimal outcome. However, we did not observe distinctive clustering at T2. This

could be caused by the variation in time point of sample collection at T2 between groups as shown by the total CIT (Table 1). Interestingly, kidneys with a good outcome had an average longer total CIT although these differences were not significant. Furthermore, samples collected at T1 better represent kidney quality as these samples can be considered a second flush-out. Samples collected at T2 possibly also represent the effect of preservation.

Donor sex seems to influence the predictive values of proteins identified at T2, although there was no significant difference in male/female ratio between experimental groups (Table 1).

Limitations

This pilot study consisted of a small cohort, and the results are based on proteomics analysis only. Validation in a larger patient cohort will be needed to determine the impact of our discovered biomarkers and their predictive value for determining the level of injury of donor kidneys prior to transplantation.

Conclusion

In conclusion, this is the first study that shows that the proteomic perfusate analysis of DBD kidneys during HMP preservation can discriminate between a good and suboptimal post-transplant outcome. Our results provide novel insights into pathophysiological mechanisms underlying brain death and organ preservation. The presence of markers in the perfusate reflecting complement activation and loss of epithelial and tissue integrity may be indicators for transplant outcome. However, additional research needs to be conducted to confirm these results. Ultimately, these insights provide a noninvasive way to assess donor organ quality and provide potential inroads for repair. Novel preservation techniques such as HMP are making their way

into the clinics, and with this study, we show that HMP can provide a potential platform for noninvasive graft assessment and outcome prediction. Furthermore, HMP could be used as a platform for pharmaceutical interventions and repair of donor organs with a suboptimal quality.

Acknowledgements

We thank Letizia Lo Faro and members of the Ploeg and Kessler group for insightful discussions. We thank the transplant technicians for collecting the perfusate samples. We thank members of the Discovery Proteomics Facility (DPF), led by Roman Fischer, for expert help with mass spectrometry analysis. Honglei Huang and the experimental costs were supported by a grant from the Biomedical Research Centre (NIHR) awarded to Rutger Ploeg. Aukje Brat was supported by the Tekke Huizinga Foundation in the Netherlands.

CONFLICT OF INTEREST

The authors have declared no conflicts of interest.

Data Availability Statement

The mass spectrometry proteomics data have been deposited to the ProteomeXchange Consortium (<http://proteomecentral.proteomexchange.org>) via the PRIDE partner repository [62] with the dataset identifier px-submission PXD027127.

SUPPORTING INFORMATION

Additional supporting information may be found online in the Supporting Information section at the end of the article.

Table S1

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