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Association of cystatin C proteoforms with estimated glomerular filtration rate

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ABSTRACT

Background: Cystatin C (CysC), a marker for chronic kidney disease, exists as three sequence proteoforms, in addition to the wild-type sequence: one contains hydroxyproline at position 3 (3Pro-OH), the two others have truncated sequences (des-S and des-SSP). Here, we examine correlations between each of these CysC proteoforms and estimated glomerular filtration rate (eGFR), a diagnostic criterion for chronic kidney disease (CKD).

Methods: CysC proteoform concentrations were determined from the plasma of 297 diabetes patients at a baseline time point and nine-months later, using a mass spectrometric immunoassay, and were correlated with eGFR calculations.

Results: In all samples, 3Pro-OH was the most abundant CysC proteoform, followed by the wild-type proteoform. Least abundant were the truncated CysC proteoforms, des-S and des-SSP, although they demonstrated stronger negative correlation with eGFR than the 3Pro-OH and wild-type proteoforms. The des-SSP truncated proteoform exhibited negative predictive value for eGFR.

Conclusions: The truncated CysC proteoforms show potential for clinical and prognostic utility in CKD staging. This could be useful in populations where current methods do not provide satisfactory solutions. © 2016 The Association for Mass Spectrometry: Applications to the Clinical Lab (MSACL). Published by Elsevier B.V. This is an open access article under the CC BY license (http://creativecommons.org/licenses/by/ 4.0/).

1. Introduction

Cystatin C (CysC) is a cysteine proteinase inhibitor encoded by the CST3 gene, a member of the type 2 cystatin gene family [1]. Following cleavage of an N-terminal signal peptide, the mature form of CysC consists of a 120 amino acid polypeptide chain, with two disulfide bridges and a molecular mass of 13,343 Da. CysC exists *in vivo* as several posttranslationally-modified variants, or "proteoforms" [2]: the first proteoform has hydroxyproline at position 3 (3Pro-OH), while the other two variants have truncated sequences – one missing the N-terminal serine (des-S), and the other missing the first three N-terminal residues (des-SSP). Our group has detected these proteoforms in human plasma [3,4] and urine [5] using a mass spectrometric immunoassay (MSIA) method that combines protein affinity retrieval using antibodies and MALDI-TOF for proteoform identification [6,7]. This method differs from standard ELISAs in that it uses mass spectrometry (MS) to provide direct readout of the intact molecular mass – an intrinsic property of each proteoform. In contrast, the secondary (i.e., labeled) reporter antibody in ELISAs cannot discriminate between structural protein modifications; hence, the resulting quantitative signal is a summation of signals from all proteoforms for a given protein that are captured by the primary antibody.

Recently, we have developed and validated a quantitative version of the MSIA for CysC that uses beta-lactoglobulin (BL) as an internal reference standard [8]. BL was chosen, because it is not a human protein, and has a molecular mass of 18,281 Da, which is near that of human CysC. Quantitation was achieved by affinity co-purifying BL along with CysC using antibody-derivatized affinity pipettes, normalizing the CysC proteoform values based on BL recovery, and, determining protein concentration from the normalized signal intensity based on a standard curve. We have applied this quantitative MSIA to a cohort of 500 healthy individuals to determine the concentration ranges of these three proteoforms and the wild-type CysC in the healthy population [9].

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It was determined that 3Pro-OH was the most abundant CysC proteoform, followed by the wild-type protein; least abundant were the two truncated CysC proteoforms, des-S and des-SSP. The existence and percentages of these proteoforms in human sera were recently confirmed independently by another research group using immuno-MALDI-MS [10]. The CysC proteoforms have also been detected in fresh human plasma samples [9], indicating they are most likely created in vivo, through protein/enzymatic processing [11]. While it has been shown that leucocyte elastase cleaves the Val10-Gly11 bond of CysC [12], and cathepsin L cleaves the Gly11-Gly12 bond [13], the exact mechanism for the creation of the CysC truncated proteoforms analyzed here remains unknown. The N-terminus of CysC is unstructured and highly mobile [14], and it contains the inhibitory binding site for the cysteine proteinases. Removal of the first ten amino acids drastically reduces CvsC inhibition of several cvsteine proteinases as a result of decreased affinity [12,15]. Hence, the truncated des-S and des-SSP CysC proteoforms may also exhibit reduced biological activity.

CysC is an emerging biomarker for the diagnosis and monitoring of chronic kidney disease (CKD) [16]. CKD is defined by a persistent reduction in the glomerular filtration rate (GFR) to below 60 mL/min per 1.73 m^2 for three or more months [17]. The level of GFR and the magnitude of its change over time are important for CKD detection and severity assessment. The urinary or plasma clearance of an ideal filtration marker, such as inulin, is the gold standard for the measurement of GFR [18]. This method is cumbersome, however, leading investigators to use serum levels of endogenous filtration markers, such as creatinine, to obtain an estimated GFR (eGFR) instead.

Two commonly used equations to calculate eGFR are the Modification of Diet in Renal Disease (MDRD) study equation and the Cockcroft-Gault equation [19,20]. Both equations use serum creatinine in combination with age, sex, weight or race to estimate GFR. Another equation that also uses serum creatinine is the Chronic Disease Epidemiology Collaboration (CKD-EPI) [21]. However, all of these equations have shortcomings. The MDRD Study equation works best in the lower ranges of GFR [22]. GFR estimates using the MDRD Study equation that are greater than 60 mL/min per 1.73 m² underestimate measured GFR and may lead to misdiagnosis or misclassification of CKD in individuals with mild renal insufficiency [23,24]. Additionally, all equations underestimate GFR in the elderly, especially at higher GFRs [25,26].

Because CysC levels are less influenced by muscle mass and diet than creatinine, CysC is being considered as a novel endogenous filtration marker for the estimation of GFR [27]. Some studies have reported GFR-estimating equations based on serum levels of CysC, either alone or in combination with serum creatinine [28]; however, those equations are not accurate for all populations [29]. CysC has been indicated as an improved predictor of adverse effects in the elderly, including mortality, heart failure, bone loss, peripheral arterial disease and cognitive impairment [30,31].

Because of the increased use of CysC in eGFR equations, we investigated the correlation of individual CysC proteoforms with eGFR. We have recently correlated proteoforms of other proteins with diseases and outcomes in a number of clinical cohorts [32–36]. In the current work, we apply the quantitative CysC proteoforms assay to a cohort of diabetes patients to examine the relationship between CysC proteoforms and eGFR.

2. Materials and methods

2.1. Study population

Human plasma samples were obtained from the Risk Factors, Atherosclerosis and Clinical Events in Diabetes (RACED) cohort, a seven-center sub-study of the Veterans Affairs Diabetes Trial [37]. Patients were all carefully assessed by medical history, physical exam and blood work at the time of study enrollment and only those without active problems needing diabetes treatment were enrolled. Samples from 297 patients with type 2 diabetes were available at a baseline time point (Visit 1, the beginning of the study) and at 9 months (Visit 2); this was a glucose lowering trial, and the HbA1c values decreased and stabilized by 9 months into the study. The current study and use of human samples were approved by Arizona State University's Institutional Review Board #1106006545.

2.2. Mass spectrometric immunoassay (MSIA)

CysC proteoform concentrations were determined by MSIA, as previously described [8]. Briefly, antibodies against CysC (Cat. No. GCYS-80A, ICLLab, Portland, OR, USA) and the internal reference standard, BL (Cat. No. GTX77272, Genetex, Irvine, CA, USA) were co-immobilized onto microcolumns at the entrance of pipette tips, in an empirically-determined optimal mass ratio of 4.5:1 (CysC: BL). For the standard curves, commercially available CysC (Cat. No. CRC173B, Cell Sciences, Canton, MA, USA) was serially diluted in PBS buffer containing 3 g/L BSA to final concentrations of 1.25, 0.625, 0.313, 0.156, 0.078, and 0.039 mg/L. The human plasma samples were diluted 5-fold in PBS, 0.1% Tween. Analytical samples were prepared by combining 20 μ L of the CysC standards or diluted plasma with 40 μ L of 1 mg/mL BL and 100 μ L PBS 0.1% Tween, into the wells of a microtiter plate.

The antibody-derivatized pipette tips were mounted on the head of an automated multichannel pipettor (Multimek 96, Beckman Coulter, Brea, CA, USA), immersed into microtiter plate wells containing 160 µL of the analytical samples, and 250 aspiration and dispense cycles were performed (100 µL volumes each) allowing for affinity capture of the proteins (room temperature, \sim 10 min). Following rinses with PBS 0.1% Tween buffer (100 cycles, 100 μ L volumes) and water (2 × 10 cycles, 100 μ L volumes, \sim 5 min), the affinity pipettes were immersed into microplate wells containing MALDI matrix solution (25 g/L sinapic acid in aqueous solution containing 33% (v/v) acetonitrile, and 0.4% (v/v) trifluoroacetic acid), and 6 µL were aspirated into the affinity pipettes. After a 10 s delay, to allow for the dissociation of the protein from the capturing antibody, the matrix containing the proteins was dispensed directly onto a 96-well formatted MALDI target. Linear mass spectra were acquired on a Bruker Autoflex III MALDI-TOF mass spectrometer, with a delayed extraction mode using a 1.4 kV draw out pulse, 150 ns delay, and a full accelerating potential of 20 kV. In order to obtain accurate mass spectra representation, 10,000 laser shots were collected from various areas of each sample and summed into a single mass spectrum. The mass spectra were processed (baseline subtracted and smoothed) using Flex Analysis software (Bruker Daltonics).

Protein signals in the mass spectra were assigned based on accurately measured m/z values and our previous work, which included tryptic digestion of the eluted proteins to confirm the identities of the CysC proteoforms [8]. Peak heights for the CysC and BL signals were measured, and the ratios of the CysC/BL peak heights were calculated for the CysC standards samples. Standard curves were generated by plotting the CysC/BL ratios against the concentration of the human CysC standards. The CysC/BL peak heights ratios for each CysC proteoform in the plasma samples were then calculated, and the sum of the ratios was used to determine the total cystatin C concentration using the standard curves. The concentrations of the individual proteoforms were calculated based on their percentage representation of total CysC concentration.

2.3. Analysis of population data

eGFR at both Visits 1 and 2 was calculated from creatinine measurements as described [37], using the Cockcroft-Gault equation. For the Spearman correlations with eGFR, the CysC proteoform concentrations, determined from the MSIA, were used, in addition to their log ratios with respect to the wild-type CysC proteoform. An analysis of covariance (ANCOVA) for the eGFR values at Visit 2 was performed, with baseline values of the CysC proteoforms (concentrations and log ratios to wild-type CysC), using values for eGFR at Visit 1 as covariates.

3. Results and discussion

A representative mass spectrum resulting from the analysis of the human plasma samples is shown in Fig. 1. Present in the spectrum are only signals for the two targeted proteins, CysC and BL (i.e. internal reference standard), without any other signals resulting from non-specifically bound proteins, demonstrating the specificity of the assay and the selectivity of the antibodies. The inset spectrum in Fig. 1 shows the wild-type CysC protein signal, along with the three proteoforms: 3Pro-OH, des-S, and des-SSP. These proteoforms were observed in every sample from the cohort.

The concentrations of the individual CysC proteoforms were determined using the quantitative MALDI-TOF MSIA and the standard curves. While some doubts have been raised regarding the quantitative nature of MALDI MS, our group [38-42] and others [43-47] have successfully developed MALDI-based protein guantification assays with excellent CVs and reproducibility. Fundamental to accurate quantitative analysis using MALDI are proper matrix preparation and application, as well as reduction of shotto-shot variation, which was achieve through summation of a large number of laser shots across each sample spot into a single mass spectrum. With these parameters optimized, peak areas or peak heights can be used for protein quantification. In this work, peak heights were used instead of peak areas to allow for the quantification of the wild-type protein and 3Pro-OH signals that were not baseline resolved in the mass spectra (Fig. 1 inset). Others have found peak heights to be more accurate than peak areas for CysC MALDI-TOF MS quantification [10], as well as for other analytes [48,49]. As previously determined, the assay exhibits good intraand inter-assay precision (<10% CVs), linearity and spikingrecovery in the range of 90-110%, and a slight positive bias when compared to an IVD-approved ELISA test, which may be attributed

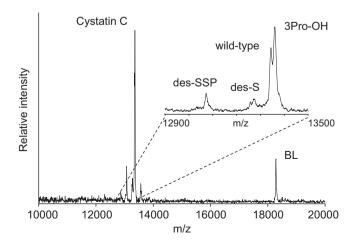


Fig. 1. Representative mass spectrum obtained with the CysC MSIA. CysC signals are observed form the wild-type, 3Pro-OH, des-S, and des-SSP proteoforms. BL – beta lactoglobulin, served as an internal reference standard.

to the possibility that the ELISA may not measure all of the proteoforms [8]. The concentrations of the CysC proteoforms in the entire cohort are shown in Fig. 2. Data from both visits were plotted together, as there was no statistically significant difference among the values for the two visits (results not shown). As observed previously in a healthy cohort of regular blood donors [9], 3Pro-OH is the most abundant proteoform; the wild-type protein was a smaller fraction of the total CysC. The 3Pro-OH to wildtype CysC ratio was ~1.25:1, which is consistent with previous studies that utilized amino acid sequencing/HPLC [50], and papain affinity chromatography/MALDI-TOF MS [51]. The total CysC concentration observed in this diabetes cohort (mean = 1.70 mg/L) was higher than that obtained for healthy individuals [9] (mean = 1.03 mg/L). Similar trends were observed in another study of CysC concentrations in type 2 diabetes patients [52].

Spearman correlation analyses were performed to examine the relationship between the concentrations of the CysC proteoforms and eGFR, which was estimated from serum creatinine measurements. The data are shown in Table 1. The two truncated CysC proteoforms (des-S and des-SSP) show stronger negative correlation with eGFR than the 3Pro-OH and wild-type proteoforms. When log ratios of the three proteoforms, with respect to wild-type CysC, are used (normalizing to the wild-type protein signal), the correlations become weaker; however, the negative association between des-SSP and eGFR remains strong. Notably, correlations were not detected in this study between the CysC proteoform concentrations and other clinical measures and characteristics of the cohort [53].

Similar results were observed in another recently published study [54]. Using the same quantitative MSIA on a cohort of diabetic CKD patients and controls, it was determined that the CysC proteoforms had higher concentrations in the diabetic CKD group than the healthy control group. Similar to what was observed in the current study, the negative association with eGFR was stronger for the truncated CysC proteoforms than wild-type. These results suggest that the truncated proteoforms may not be cleared through the renal glomeruli as quickly as wild-type CysC in CKD.

Because the cohort used in the current study consisted of samples obtained from two visits, nine months apart, we further examined the role of the CysC proteoforms as eGFR predictors. We analyzed the eGFR values at Visit 2 as a function of the baseline

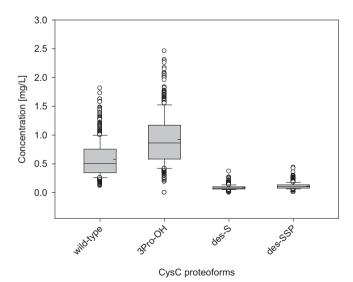


Fig. 2. Individual concentrations of cystatin C proteoforms. Box – 25–75th percentile; Solid line – median concentration; Short dash line – mean concentration; Error bars – 10th and 90th percentile; Symbols – outlying points.

Table 1

Spearman correlations for cystatin C proteoforms (concentrations, and log ratios to wild-type cystatin C) with eGFR, at baseline (Visit 1) and after nine months (Visit 2).

	eGFR	
	Visit 1	Visit 2
Concentrations		
des-SSP	-0.43 (p<0.001)	-0.38 (p < 0.001)
des-S	-0.37 (p < 0.001)	-0.37 (p < 0.001)
3Pro-OH	-0.21 (p < 0.001)	-0.15 (p = 0.02)
wild-type	-0.21 (p < 0.001)	-0.13 (p = 0.05)
Log ratio to wild-ty	rpe	
des-SSP	-0.33 (p < 0.001)	-0.38 (p < 0.001)
des-S	-0.19 (p < 0.001)	-0.22 (p < 0.001)
3Pro-OH	-0.03 (p = 0.6)	-0.06 (p = 0.3)

values of the CysC proteoforms, adjusted for the initial eGFR (Visit 1, baseline). The only proteoform that showed a negative association with eGFR was the des-SSP truncated form. The strength of the association was maintained throughout the concentration range (slope coefficient of -0.06) and the log ratio to wild-type values (slope coefficient of -0.08), all with p < 0.05. These data suggest a negative relationship between des-SSP levels and future GFR values, indicating a possible role of this proteoform as a predictor of GFR changes and CKD progression.

4. Conclusion

Proteoforms could have undiscovered pathophysiological implications and potential clinical utility for specific diseases. With the capability to detect and quantify a number of CysC proteoforms, we are able to explore their relationship with specific clinical metrics and outcomes. In this work, we examined the relationship between CysC proteoforms and eGFR in a cohort of diabetic patients, identifying a strong negative association for truncated proteoforms and predictive power for changes in GFR for the most-truncated proteoform. The truncated CysC proteoforms could have clinical and prognostic significance in CKD staging, especially in populations where the current equations do not provide satisfactory solutions. Studies with larger longitudinal cohorts are warranted to further explore these associations. If, and when, such associations are identified and validated, the CysC MSIA could readily be adapted for clinical use [55].

Author contributions

All the authors have accepted responsibility for the entire content of this submitted manuscript and approved submission.

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Employment or leadership

None declared.

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Competing interests

The funding organization(s) played no role in the study design; in the collection, analysis, and interpretation of data; in the writing of the report; or in the decision to submit the report for publication.

References

- M. Jarvinen, A. Rinne, V.K. Hopsu-Havu, Human cystatins in normal and diseased tissues-a review, Acta Histochem. 82 (1) (1987) 5–18.
- [2] L.M. Smith, N.L. Kelleher, Proteomics CfTD. Proteoform: a single term describing protein complexity, Nat. Methods 10 (3) (2013) 186–187.
- [3] D. Nedelkov, U.A. Kiernan, E.E. Niederkofler, K.A. Tubbs, R.W. Nelson, Investigating diversity in human plasma proteins, Proc. Natl. Acad. Sci. USA. 102 (31) (2005) 10852–10857.
- [4] D. Nedelkov, D.A. Phillips, K.A. Tubbs, R.W. Nelson, Investigation of human protein variants and their frequency in the general population, Mol. Cell. Proteomics 6 (7) (2007) 1183–1187.
- [5] U.A. Kiernan, K.A. Tubbs, D. Nedelkov, E.E. Niederkofler, E. McConnell, R.W. Nelson, Comparative urine protein phenotyping using mass spectrometric immunoassay, J. Proteome Res. 2 (2) (2003) 191–197.
- [6] D. Nedelkov, Mass spectrometry-based protein assays for in vitro diagnostic testing, Expert Rev. Mol. Diagn. 12 (3) (2012) 235–239.
- [7] R.W. Nelson, C.R. Borges, Mass spectrometric immunoassay revisited, J. Am. Soc. Mass Spectrom. 22 (6) (2011) 960–968.
- [8] O. Trenchevska, D. Nedelkov, Targeted quantitative mass spectrometric immunoassay for human protein variants, Proteome Sci. 9 (1) (2011) 19.
- [9] O. Trenchevska, D.A. Phillips, R.W. Nelson, D. Nedelkov, Delineation of concentration ranges and longitudinal changes of human plasma protein variants, PLoS One 9 (6) (2014) e100713.
- [10] K. Meyer, P.M. Ueland, Targeted quantification of C-reactive protein and cystatin c and its variants by immuno-MALDI-MS, Anal. Chem. 86 (12) (2014) 5807–5814.
- [11] N. Fortelny, P. Pavlidis, C.M. Overall, The path of no return-Truncated protein N-termini and current ignorance of their genesis, Proteomics 15 (14) (2015) 2547–2552.
- [12] M. Abrahamson, R.W. Mason, H. Hansson, D.J. Buttle, A. Grubb, K. Ohlsson, Human cystatin C. role of the N-terminal segment in the inhibition of human cysteine proteinases and in its inactivation by leucocyte elastase, Biochem. J. 273 (Pt 3) (1991) 621–626.
- [13] T. Popovic, N. Cimerman, I. Dolenc, A. Ritonja, J. Brzin, Cathepsin L is capable of truncating cystatin C of 11 N-terminal amino acids, FEBS Lett. 455 (1–2) (1999) 92–96.
- [14] I. Ekiel, M. Abrahamson, D.B. Fulton, P. Lindahl, A.C. Storer, W. Levadoux, et al., NMR structural studies of human cystatin C dimers and monomers, J. Mol. Biol. 271 (2) (1997) 266–277.
- [15] I. Björk, E. Pol, E. Raub-Segall, M. Abrahamson, A.D. Rowan, J.S. Mort, Differential changes in the association and dissociation rate constants for binding of cystatins to target proteinases occurring on N-terminal truncation of the inhibitors indicate that the interaction mechanism varies with different enzymes, Biochem. J. 299 (Pt 1) (1994) 219–225.
- [16] L.A. Inker, A. Okparavero, Cystatin C as a marker of glomerular filtration rate: prospects and limitations, Curr. Opin. Nephrol. Hypertens. 20 (6) (2011) 631– 639.
- [17] A.S. Levey, J. Coresh, E. Balk, A.T. Kausz, A. Levin, M.W. Steffes, et al., National Kidney Foundation practice guidelines for chronic kidney disease: evaluation, classification, and stratification, Ann. Intern. Med. 139 (2) (2003) 137–147.
- [18] L.A. Stevens, A.S. Levey, Measured GFR as a confirmatory test for estimated GFR, J. Am. Soc. Nephrol. 20 (11) (2009) 2305–2313.
- [19] A.S. Levey, J. Coresh, T. Greene, L.A. Stevens, Y.L. Zhang, S. Hendriksen, et al., Using standardized serum creatinine values in the modification of diet in renal disease study equation for estimating glomerular filtration rate, Ann. Intern. Med. 145 (4) (2006) 247–254.
- [20] D.W. Cockcroft, M.H. Gault, Prediction of creatinine clearance from serum creatinine, Nephron 16 (1) (1976) 31–41.
- [21] A.S. Levey, L.A. Stevens, C.H. Schmid, Y.L. Zhang, A.F. Castro, H.I. Feldman, et al., A new equation to estimate glomerular filtration rate, Ann. Intern. Med. 150 (9) (2009) 604–612.
- [22] A.D. Rule, T.S. Larson, E.J. Bergstralh, J.M. Slezak, S.J. Jacobsen, F.G. Cosio, Using serum creatinine to estimate glomerular filtration rate: accuracy in good health and in chronic kidney disease, Ann. Intern. Med. 141 (12) (2004) 929– 937.
- [23] L.A. Stevens, J. Coresh, T. Greene, A.S. Levey, Assessing kidney functionmeasured and estimated glomerular filtration rate, N. Engl. J. Med. 354 (23) (2006) 2473–2483.
- [24] M. Froissart, J. Rossert, C. Jacquot, M. Paillard, P. Houillier, Predictive performance of the modification of diet in renal disease and Cockcroft-Gault

equations for estimating renal function, J. Am. Soc. Nephrol. 16 (3) (2005) 763–773.

- [25] C. Pedone, A. Corsonello, R.A. Incalzi, G. Investigators, Estimating renal function in older people: a comparison of three formulas, Age Ageing 35 (2) (2006) 121–126.
- [26] T.C. Dowling, E.S. Wang, L. Ferrucci, J.D. Sorkin, Glomerular filtration rate equations overestimate creatinine clearance in older individuals enrolled in the Baltimore longitudinal study on aging: impact on renal drug dosing, Pharmacotherapy 33 (9) (2013) 912–921.
- [27] M.G. Shlipak, M.D. Mattes, C.A. Peralta, Update on cystatin C: incorporation into clinical practice, Am. J. Kidney Dis. 62 (3) (2013) 595–603.
- [28] L.A. Stevens, J. Coresh, C.H. Schmid, H.I. Feldman, M. Froissart, J. Kusek, et al., Estimating GFR using serum cystatin C alone and in combination with serum creatinine: a pooled analysis of 3,418 individuals with CKD, Am. J. Kidney Dis. 51 (3) (2008) 395–406.
- [29] L.A. Stevens, S. Padala, A.S. Levey, Advances in glomerular filtration rateestimating equations, Curr. Opin. Nephrol. Hypertens. 19 (3) (2010) 298–307.
- [30] M.J. Sarnak, R. Katz, C.O. Stehman-Breen, L.F. Fried, N.S. Jenny, B.M. Psaty, et al., Cystatin C concentration as a risk factor for heart failure in older adults, Ann. Intern. Med. 142 (7) (2005) 497–505.
- [31] M.G. Shlipak, M.J. Sarnak, R. Katz, L.F. Fried, S.L. Seliger, A.B. Newman, et al., Cystatin C and the risk of death and cardiovascular events among elderly persons, N. Engl. J. Med. 352 (20) (2005) 2049–2060.
- [32] I. Azizkhanian, O. Trenchevska, Y. Bashawri, J. Hu, J. Koska, P.D. Reaven, et al., Posttranslational modifications of apolipoprotein A-II proteoforms in type 2 diabetes, J. Clin. Lipidol. 10 (4) (2016) 808–815.
- [33] J. Koska, H. Yassine, O. Trenchevska, S. Sinari, D.C. Schwenke, F.T. Yen, et al., Disialylated apolipoprotein C-III proteoform is associated with improved lipids in prediabetes and type 2 diabetes, J. Lipid Res. 57 (5) (2016) 894–905.
- [34] H.N. Yassine, O. Trenchevska, H. He, C.R. Borges, D. Nedelkov, W. Mack, et al., Serum amyloid a truncations in type 2 diabetes mellitus, PLoS One 10 (1) (2015) e0115320.
- [35] H.N. Yassine, O. Trenchevska, A. Ramrakhiani, A. Parekh, J. Koska, R.W. Walker, et al., The Association of Human Apolipoprotein C-III Sialylation Proteoforms with Plasma Triglycerides, PLoS One 10 (12) (2015) e0144138.
- [36] H.N. Yassine, A.M. Jackson, P.D. Reaven, D. Nedelkov, R.W. Nelson, S.S. Lau, et al., The application of multiple reaction monitoring to assess Apo A-I methionine oxidations in diabetes and cardiovascular disease, Transl. Proteome 4–5 (2014) 18–24.
- [37] L. Agrawal, N. Azad, N.V. Emanuele, G.D. Bahn, D.G. Kaufman, T.E. Moritz, et al., Observation on renal outcomes in the veterans affairs diabetes trial, Diabetes Care 34 (9) (2011) 2090–2094.
- [38] O. Trenchevska, E. Kamcheva, D. Nedelkov, Mass spectrometric immunoassay for quantitative determination of protein biomarker isoforms, J. Proteome Res. 9 (11) (2010) 5969–5973.
- [39] U.A. Kiernan, D.A. Phillips, O. Trenchevska, D. Nedelkov, Quantitative mass spectrometry evaluation of human retinol binding protein 4 and related variants, PLoS One 6 (3) (2011) e17282.
- [40] N.D. Sherma, C.R. Borges, O. Trenchevska, J.W. Jarvis, D.S. Rehder, P.E. Oran, et al., Mass spectrometric immunoassay for the qualitative and quantitative

analysis of the cytokine Macrophage Migration Inhibitory Factor (MIF), Proteome Sci. 12 (1) (2014) 52.

- [41] P.E. Oran, O. Trenchevska, D. Nedelkov, C.R. Borges, M.R. Schaab, D.S. Rehder, et al., Parallel workflow for high-throughput (>1,000 samples/day) quantitative analysis of human insulin-like growth factor 1 using mass spectrometric immunoassay, PLoS One 9 (3) (2014) e92801.
- [42] O. Trenchevska, N.D. Sherma, P.E. Oran, P.D. Reaven, R.W. Nelson, D. Nedelkov, Quantitative mass spectrometric immunoassay for the chemokine RANTES and its variants, J. Proteomics. 116 (2015) 15–23.
- [43] M.W. Duncan, H. Roder, S.W. Hunsucker, Quantitative matrix-assisted laser desorption/ionization mass spectrometry, Brief Funct. Genomic Proteomic 7 (5) (2008) 355–370.
- [44] J.D. Reid, D.T. Holmes, D.R. Mason, B. Shah, C.H. Borchers, Towards the development of an immuno MALDI (iMALDI) mass spectrometry assay for the diagnosis of hypertension, J. Am. Soc. Mass Spectrom. 21 (10) (2010) 1680– 1686.
- [45] L. Hyzak, R. Moos, F. von Rath, V. Wulf, M. Wirtz, D. Melchior, et al., Quantitative matrix-assisted laser desorption ionization-time-of-flight mass spectrometry analysis of synthetic polymers and peptides, Anal. Chem. 83 (24) (2011) 9467–9471.
- [46] C. Lange, S. Schubert, J. Jung, M. Kostrzewa, K. Sparbier, Quantitative matrixassisted laser desorption ionization-time of flight mass spectrometry for rapid resistance detection, J. Clin. Microbiol. 52 (12) (2014) 4155–4162.
- [47] M.L. Vestal, Evolution of quantitative MALDI-TOF mass spectrometry for clinical applications, Clin. Chem. 62 (1) (2016) 20–23.
- [48] K.L. Howard, G.L. Boyer, Quantitative analysis of cyanobacterial toxins by matrix-assisted laser desorption ionization mass spectrometry, Anal. Chem. 79 (15) (2007) 5980–5986.
- [49] M. Bucknall, K.Y. Fung, M.W. Duncan, Practical quantitative biomedical applications of MALDI-TOF mass spectrometry, J. Am. Soc. Mass Spectrom. 13 (9) (2002) 1015–1027.
- [50] A. Grubb, H. Lofberg, Human gamma-trace, a basic microprotein: amino acid sequence and presence in the adenohypophysis, Proc. Natl. Acad. Sci. USA. 79 (9) (1982) 3024–3027.
- [51] B. Asgeirsson, S. Haebel, L. Thorsteinsson, E. Helgason, K.O. Gudmundsson, G. Gudmundsson, et al., Hereditary cystatin C amyloid angiopathy: monitoring the presence of the Leu-68->GIn cystatin C variant in cerebrospinal fluids and monocyte cultures by MS, Biochem. J. 329 (Pt 3) (1998) 497–503.
- [52] S.H. Lee, S.A. Park, S.H. Ko, H.W. Yim, Y.B. Ahn, K.H. Yoon, et al., Insulin resistance and inflammation may have an additional role in the link between cystatin C and cardiovascular disease in type 2 diabetes mellitus patients, Metabolism 59 (2) (2010) 241–246.
- [53] W. Duckworth, C. Abraira, T. Moritz, D. Reda, N. Emanuele, P.D. Reaven, et al., Glucose control and vascular complications in veterans with type 2 diabetes, N. Engl. J. Med. 360 (2) (2009) 129–139.
- [54] H.N. Yassine, O. Trenchevska, Z. Dong, Y. Bashawri, J. Koska, P.D. Reaven, et al., The association of plasma cystatin C proteoforms with diabetic chronic kidney disease, Proteome Sci. 14 (2016) 7.
- [55] M.W. Duncan, D. Nedelkov, R. Walsh, S.J. Hattan, Applications of MALDI mass spectrometry in clinical chemistry, Clin. Chem. 62 (1) (2016) 134–143.