Abstract. Background: Secreted proteins play an important role in promoting cancer (PCa) cell migration and invasion. Proteogenomics helps elucidate the mechanism of diseases, discover therapeutic targets, and generate biomarkers for diagnosis through protein variations. Materials and Methods: We carried out a mass spectrometry-based proteomic analysis of the conditioned media (CM) from two human prostate cancer cell lines, belonging to different metastatic sites, to identify potential metastatic and/or aggressive factors. Results: We identified a total of 598 proteins, among which 561 were quantified based on proteomic analysis. Among the quantified proteins, 128 were up-regulated and 83 were down-regulated in DU145/PC3 cells. Six mutant peptides were identified in the CM of prostate cancer cell lines using proteogenomics approach. Conclusion: This is the first proteogenomics study in PCa aiming at exploring a new type of metastatic factor, which are mutant peptides, predicting a novel biomarker of metastatic PCa for diagnosis, prognosis and drug targeting.

Prostate cancer (PCa) is the second most frequent cancer in men worldwide, with lung cancer topping the list. The American Cancer Society has reported that PCa accounted for approximately 180,890 new cases in 2016, with an estimated 26,120 deaths only in the United States (1). Twenty-five percent of patients progress from initial hormone sensitivity to castrate-resistant prostate cancer (CRPC), that no longer accepts hormone therapy (2). Consequently, metastatic castration-resistant prostate cancer (mCRPC) is considered incurable; many of these patients develop recurrence with a poor outlook, eventually leading to death. Metastasis of cancer is a multi-step cascade that includes local migration, invasion, and colonization. PCa is able to metastasize to liver, lung, and brain, but almost proprietarily associates with a high incidence of bone metastases (3). Despite many studies on mCRPC, the metastatic mechanism and/or involved factors are still not clear. Hence, metastatic studies and identification of diagnostic factors in PCa are of increasingly high priority.

Secretory proteins play an important role in cancer metastasis by stimulating cancer cell migration and invasion capability, consequently increasing the cancer metastasis in the extracellular microenvironment (4). Several researchers have reported the relation of secretory proteins with cancer metastasis, suggesting their application as putative tumor biomarkers and/or in understanding pathophysiological pathways (5-9). A big challenge in the diagnosis of PCa is the lack of alternative screening to replace the existing PCa biomarker prostate-specific antigen (PSA). Although widely used, PSA cannot distinguish between indolent and aggressive PCa (10). For this reason, exploring biomarkers, especially metastasis-associated factors, is imperative for prostate cancer research.

Cancer- or onco-proteogenomics combines mass spectrometry-based proteomics with genomic information to study protein variations and cancer-specific mutant proteins. Proteogenomics can help researchers understand the underlying mechanisms of diseases, discover therapeutic targets, or generate biomarkers for diagnosis or tailored therapies (11, 12). Generally, identification of mutant proteins by mass spectrometry-based proteomics has been utilized by de novo sequencing or database search method.
with genomics and transcriptomics. *De novo* sequencing algorithms can directly indicate amino acid sequence without database matching, but its accuracy is not reliable enough, mostly due to ambiguous interpretations of MS/MS spectra (13). However, database search algorithms, associated with cancer-related mutation database, such as Unknown peptide-level Mutation Analysis (XMAn), Human Cancer Proteome Variation Database (CanProVar), and Cancer Mutation Proteome Database (CMPD), can be used to identify the relevant mutant proteins (14-16).

In this study, we adopted a proteomics approach for the conditioned media from two human prostate cancer cell lines of differing metastatic origins [DU145 (brain metastasis) and PC3 (bone metastasis)] to identify the putative metastatic factors, including secreted and/or mutant proteins, using quantitative mass spectrometry. We identified a total of 598 secretory proteins, out of which 561 were quantified based on proteomic analysis. Among the quantified proteins, 128 were up-regulated and 83 were down-regulated in DU145/PC3. Of these, laminin gamma 1 (LAMC1) was significantly elevated in DU145 cell media. Also, MS/MS spectra were re-analyzed by MaxQuant with CMPD to find out prostate cancer-specific mutant proteins in different cell lines. As a consequence, we discovered six mutant peptides, two of which are assumed to be associated with metastasis and/or aggressiveness in DU145 cells. We propose that LAMC1 and the six mutant peptides are not only able to distinguish metastatic sites, but are also likely to be used for drug targeting in prostate cancer.

**Materials and Methods**

**Cell culture.** The human prostate cancer cell lines, DU145 and PC3, were purchased from the Korean Cell Line Bank (KCLB, Seoul, Korea). Cells were cultured in RPMI-1640 (GE healthcare, Little Chalfont, UK) supplemented with 10% (v/v) fetal bovine serum (Gibco, Rockville, MD, USA) and 1% penicillin-streptomycin (Gibco) at 37˚C with 5% CO2 in a humidified incubator.

**Enrichment of secretory proteins from conditioned media.** DU145 and PC3 cells were grown to ~70% confluence (~3.5×106 cells) in 150 mm culture dishes (SPL Life Science, Pocheon, Republic of Korea). The cell monolayer was rinsed carefully with Dulbecco’s Phosphate-Buffered Saline (Gibco) and serum-free medium (SFM) to avoid serum contamination. The cells were incubated in SFM at 37˚C for 24 h. After incubation, the conditioned media from three plates were carefully collected and protease inhibitors (Thermo Fisher Scientific) were added. Floating cells and cellular debris were removed by centrifugation at 3,000 × g for 10 min at 4˚C. The supernatants were concentrated by ultrafiltration (Thermo Fisher Scientific). All protein samples were stored at –80˚C until further use.

**Western blotting.** Protein samples, containing 5 μg of cell lysate and secretory proteins from conditioned media, were separated by 10% SDS-PAGE and transferred to PVDF membranes (Roche, Basel, Switzerland). All membranes were blocked by 5% BSA, TBST buffer (20 mM Tris, 147 mM sodium chloride, 0.1% Tween-20) for 3 h at RT and incubated overnight at 4˚C with primary antibodies. All antibodies were diluted to 1:1,000 with 5% BSA in TBST. The membranes were washed thrice in TBST for 15 min each, incubated with corresponding IgG-HRP secondary antibodies at a dilution of 1: 2,000 for 1 h at RT, washed, and finally visualized with the ECL plus as chemiluminescent substrate (GE Healthcare, Uppsala, Sweden). The primary antibodies used include anti-β-actin from Cell Signaling Technology (Danvers, MA), anti-α-tubulin from Abcam (Cambridge, UK), anti-Serpin B3 from Thermo Fisher Scientific and anti-LamC1 from OriGene (Rockville, MD, USA). The secondary antibodies used were anti-mouse IgG-HRP (Cell Signaling Technology, Danvers, MA, USA) and anti-rabbit IgG-HRP (Cell Signaling Technology).

**Sample preparation for quantitative proteomics.** The protein samples (25 μg) were reduced with 200 mM Tris(2-carboxyethyl)phosphine (TCEP) at 55˚C for 1 h and alkylated with 60 mM iodoacetamide (IAA) for 30 min, protected from light at room temperature. Afterwards, the proteins were precipitated with six volumes of pre-chilled acetone for 4 h at −20˚C. After centrifugation, the pellet obtained was dissolved in 100 mM TEAB. The protein was digested by sequencing-grade modified trypsin (Promega, Madison, WI, USA) at 37˚C for 16 h. The ratio of enzyme to protein was 1:50.

The digested peptides were labeled with a six-plex TMT isobaric Mass Tagging Kit (Thermo Fisher Scientific) according to the manufacturer’s protocol. Briefly, 41 μl of anhydrous acetonitrile was added to each tube. Reagents were dissolved by vortexing for 5 min. Protein samples were labeled by adding 20 μl of TMT Label Reagent (PC3: 127, DU145: 129) and incubated for 1 h at room temperature. To quench the reaction, 5% hydroxylamine was added, followed by 15-min incubation at RT. The TMT labeled peptide mixtures were dried by vacuum centrifugation. The peptide was fractionated by High pH Reversed-Phase Peptide Fractionation Kit (Thermo Fisher Scientific) according to the manufacturer’s protocol. A total of 18 fractions were cleaned with C18 ZipTips (Merck Millipore, Burlington, MA) in accordance with the manufacturer’s instructions, followed by LC/MS analysis.

**Nano LC-MS/MS analysis.** Enriched peptides were dissolved in solvent A (0.1% formic acid and 2% acetonitrile in water). Peptide separation was performed using a homemade C18 reversed-phase analytical column (Reprosil C18 3 μm beads, Dr. Maisch GmbH, Germany) with a linear gradient of 2-23% solvent B (0.1% formic acid and 10% water in acetonitrile) for 44 min and 23-90% solvent B for 5 min at a constant flow rate of 200 nL/min using an EASY-nLC 1000 UPLC system (Thermo Fisher Scientific). The peptides were ionized using a nanospray ion source followed by tandem MS/MS in a Q-Exactive Plus Hybrid Quadrupole-Orbitrap Mass Spectrometer (Thermo Fisher Scientific) coupled online to the UPLC. Intact peptides were detected in the Orbitrap at a resolution of 70,000. Peptides were selected for MS/MS using 27% normalized higher-energy collisional dissociation (HCD); ion fragments were detected in the Orbitrap at a resolution of 17,500. A data-dependent procedure that alternated between one MS scan followed by 15 MS/MS scans was applied for the top 15 precursor ions above a threshold ion count of 3E6 in the MS survey scan with a
25 s dynamic exclusion. The electrospray voltage applied was 2.2 kV. Automatic gain control was used to prevent overfilling of the ion trap; 1E5 ions accumulated for the generation of MS/MS spectra. For MS scans, the m/z scan range was 300 to 1,400 Da.

**MS/MS data analysis.** The resulting MS/MS data were processed using MaxQuant with the integrated Andromeda search engine (v.1.5.1.0) (17). MS/MS spectra were searched against the Uniprot Human database from Uniprot (http://Uniprot.org) (18). To identify site-specific mutant peptides, only prostate cancer-related database at NCI-60 and TCGA were downloaded at CMPD (http://cgbc.cgu.edu.tw/cmpd/) (16) and applied in MaxQuant. For MaxQuant searching, Trypsin/P was specified as the cleavage enzyme and up to two missed cleavages, five modifications per peptide, and seven charges were allowed. Mass error was set at 10 ppm for precursor ions and 0.02 Da for fragment ions. Carbamidomethylation on Cys was specified as a fixed modification, whereas oxidation on Met and acetylation on protein N-termini were specified as variable modifications. False-discovery rate thresholds for protein, peptide, and modification sites were specified at 1%. Minimum peptide length was set at 7. For quantification of peptides by TMT, the reporter ions were calculated by six-plex TMT at peptide N-termini and lysine. All the other parameters in MaxQuant were set to default values.

**Bioinformatics analyses.** The identified proteins were annotated by the Kyoto Encyclopedia of Genes and Genomes (KEGG), InterPro, and Gene Ontology (GO), which consist of GO molecular function (GOMF), GO biological process (GOBP), and GO cellular component (GOCC) using Perseus 1.6 (19). For functional enrichment of differentially quantified proteins, DAVID 6.7 web-based software, which provides comprehensive set of functional annotation tools for investors to understand biological meaning, was employed. Threshold of EASE score, a modified Fisher’s exact test p-value, was smaller than 0.05 to be deliberated as high enrichment of annotation categories (20). We used the STRING software to trace the critical assessment and integration of protein-protein interactions from the identified secretory proteins (21). The STRING diagrams, with a corrected interaction score <0.9, is considered significant and masked disconnected proteins in the network. Visualization, such as scatter plot, for proteomics results were drawn using Perseus 1.6, following the protocol.

**Results**

**Quantitative secretome profiling of metastatic prostate cancer cell lines.** In this study, we performed a quantitative investigation of secretory and mutant proteins associated with prostate cancer in CM using a mass spectrometry-based proteomics approach with TMT labeling (Figure 1A). Secretory proteins from CM were enriched using 10K MWCO protein concentrators, which were able to concentrate and purify proteins easily. For quality check of the proteins from CM, we performed western blotting, with β-actin and α-tubulin antibodies serving as controls for the analysis. As expected, α-tubulin was not detected in the CM, whereas β-actin was detected in both cell lysate and CM (Figure 1B). Therefore, we could confirm that the protein samples from CM were not contaminated by cell debris.

For in-solution tryptic digestion, the proteins were reduced and alkylated by TCEP and IAA, respectively, followed by precipitation in ice-cold acetone for 4 h at –20°C to remove the excess TCEP and IAA. Trypsin was added to purified proteins and incubated overnight at 37°C. The peptides were labeled by TMT reagents to discriminate secretory proteins between DU145 and PC3 cells. To improve the identification of secretory proteins, we carried out high-pH reverse-phase peptide fractionation, which could separate eight fractions from a mixture of TMT-labeled peptides. High-resolution and accurate-mass spectrometry (HRAM), namely Q-Exactive Hybrid Quadrupole-Orbitrap MS, was utilized with EASY-nLC 1000 UPLC system in order to analyze the peptides. To quantify the secreted proteins from PC3 and DU145 cells, we performed MaxQuant database search of the obtained MS/MS spectra against the Uniprot Human database (including 70,630 proteins) (Figure 1C). To improve the accuracy of identification, search results were filtered, allowing a maximum false detection rate of 1% and minimum score of 40 for proteins and peptides. Also, our data were replicated to minimize random noise associated with TMT labeling and LC-MS/MS equipment. The Pearson correlation coefficient (R) of reporter ion intensity in 127 was 0.804 and that in 129 was 0.798 (Figure 2A).

Using the Uniprot Human database, we identified a total of 598 secretory proteins with at least two peptides in the combined CM of DU145 and PC3 cells, among which 561 proteins were quantified based on comparative proteomic analysis. 350 proteins (62.4%) remained unaltered, commonly detected in both cell lines. Among the quantified proteins, 128 (22.8%) were highly secreted (DU145 vs. PC3, log2 ratio ≥1) and 83 (14.8%) were less-secreted (DU145 vs. PC3, log2 ratio ≤ –1) in DU145 cells, respectively (Figure 2B).

**Verification and bioinformatic analysis of secretory proteins in DU145 and PC3.** We confirmed the secretory proteins (detected from MS-based proteomics results) by western blot assays. From the proteomics results, LamC1 (#P11047, ratio: 4.013) was observed to be secreted more into the CM of DU145, compared to that in PC3, whereas Serpin B3 (#P29508, ratio: 0.752) was detected to be slightly decreased in DU145 compared to that in PC3 (Figure 2C). Western blots of LamC1 and Serpin B3 conform to the conducted quantitative proteomics analysis.

To acquire in-depth understanding of the biological functions of differentially expressed proteins, we used the DAVID web-based software and conducted an enrichment analysis of different categories, including Gene Ontology (GO), KEGG pathway, and InterPro. The 211 differentially secreted proteins, including 128 predominant in DU145 and 83 predominant in PC3, were analyzed for functional annotation enrichment, with respect to GOBP, GOMF, GOCC, KEGG, and InterPro, enlisted according to the
Figure 1. Comparative profiling of secreted proteins in DU145 and PC3 cells using TMT-based proteomics (A). Western blot analysis of actin and tubulin in CM and CE of PC3 and DU145 cells (B). Schematic illustration of proteomics and proteogenomics for cancer-related mutant proteins via the MaxQuant search engine (C).
Figure 2. Pearson correlation coefficient of TMT reporter ion intensity 127 (left) and 129 (right) (A). Scatter plot and distribution of secretory proteins between PC3 and DU145 (B). Western blot analysis of SerpinB3 and LamC1 in CM and CE from PC3 and DU145 cells (C).
differentially secreted proteins. Among these, three highest data from Fisher exact test, for each category, were extracted and summarized in Figure 3. The functional annotation enrichment of GOCC was mainly related to extracellular terms in both cell lines, as expected. Based on the enrichment results of GOMF, laminin-, integrin-, and cadherin-binding proteins were enriched in DU145 cells, whereas growth factor-related proteins were mostly secreted from PC3. The enrichment analysis of GOBP indicated that the highly secreted proteins in PC3 were significantly enriched in response to angiogenesis and platelet degranulation. The KEGG pathway enrichment revealed that lysosome-related secretory proteins were significantly enhanced in DU145. The domain annotation enrichment from InterPro results indicated that EGF-like domain was significantly concentrated in DU145. Previous studies had revealed that one of the well-known EGF-like domains is laminin, which affects cellular growth and mitogenic events (22, 23). In Figure 4, the STRING database was used to figure out the integrated protein-protein interactions (PPI) of differentially secreted proteins; results showed that proteins such as LAMC1, AGRN, HSPG2, and APP were highly inter-connected, dominant secretory proteins in DU145 whereas FN1 and INHBA proteins were the predominant secretory proteins of PC3. Based on these bioinformatics analysis, we focused on laminin C and laminin-binding in GOMF, to understand the specific characteristics of DU145 cells.

Prostate-cancer related mutant peptides in conditioned media. To identify the mutant peptides associated with prostate cancer, MS/MS spectra were searched using CMPD (downloaded from http://cgbc.cgu.edu.tw/cmmpd/) that includes 24,189 prostate cancer-related mutant proteins (Figure 1C). Although possibilities of detecting mutant peptides are higher when a large number of databases are used, the number of false positives also tends to increase in such cases. For this reason, only the prostate cancer-related database was downloaded for use.

Initially, 42 peptides on 36 mutant proteins were identified, and verified manually following a process described in a previous study (24). Eventually, we identified six possible mutant peptides related to prostate cancer (Table I), of which, three were not quantified, and one had no significantly different secretion across the two prostate cancer cell lines. However, we confirmed that the two peptides, DTEEEDFHVDQATTVK from alpha-1-antitrypsin and SCCSCCPVDCAK from metallothionein-1E, were greatly increased in DU145 compared to those in PC3 cells. Unusually, BAG family molecular chaperone regulator 3 had no mutant sites. Nevertheless, we assumed that GQVAAAAAAQPPASHGPER is generated by trypsin cleavage, because cysteine (C) in front of the sequence was mutated to arginine (R).

Discussion
Prostate cancer (PCa) is the most common non-cutaneous malignancy in men. In particular, most patients with prostate cancer progress to metastasis in different parts (2, 3). The lack of biomarkers to replace PSA and dearth of understanding of the factors and mechanisms, involved in metastasis, pose the biggest challenges in this regard. To resolve this big issue, we designed MS-based secretory proteomics using metastatic PCA cell lines, DU145 and PC3. Among the various PCa cells, DU145 and PC3 cells lack PSA, 5α-reductase, and androgen receptor (AR), hence are related to androgen-independent (AI) prostate cancers. Both cell lines were representatives of metastatic PCa, but the metastatic tissues are different; DU145 and PC3 were newly isolated from brain and bone metastasis, respectively, in a patient with prostate cancer. PC3 is known to be a more aggressive prostate cancer cell than DU145 (25-27). Also, recent study showed that invasion of PC3 was enhanced by LNCaP co-culture, whereas DU-145 was inhibited (28). Hence, two different AI prostate cancer cells, DU145 and PC3, have unlike properties.

We investigated the potential metastatic and/or aggressive factors in secretory proteins from different prostate cancer cell lines by an MS-based proteomics approach. Particularly, TMT was used for quantitative proteomic analysis. TMT, an isobaric chemical labeling method, is a powerful and convenient tool to estimate the relative abundance of proteins. In particular, isobaric tagging strategy with MS-based proteins is suitable for low-abundance proteins, such as the ones secreted in CM, due to high signal from native peptides (29). We implemented quality check for the proteins, both from CM and from MS-based proteomics with TMT labeling, via western blot and scatter plot, respectively. Our results also showed that β-actin, but not α-tubulin, exists in proteins from CM (Figure 1B). However, recent research has shown that β-actin may be secreted into the cell culture medium through an exosome-independent mechanism that is regulated by immune pathways (30). Pearson correlation coefficient between reporter ion intensities, examined to obtain the quantitative accuracy of MS-based proteomics, showed significance with an R value of 0.8 or higher (Figure 2A). Therefore, these results indicate that our proteomics results do have experimental significance.

We obtained 211 (37.6%) differentially expressed secretory proteins (DEP), out of the total 561 of quantified proteins across DU145 and PC3 cells. In contrast to our proteomics results, which indicated that secretory proteins were considerably different between DU145 and PC3 cells, a previous study on cell-surface protein profiles of prostate cancer cell lines reported that expression of cell surface proteins (as seen by cluster of differentiation (CD)) is comparable in DU145 and PC because of similar biological
Figure 3. GO function (A), and protein domain and KEGG pathway (B) enrichment analysis of highly secreted proteins in DU145 and PC3 cells.
Figure 4. Protein-Protein Interaction network of highly secreted proteins in DU145 (A) and PC3 (B).
behavior and gene expression (31). Among the secretory proteins, serpin B3 and LamC1 were used to verify the quantitative proteomics results using western blot; there seems to be a correlation between the results (Figure 2C). Serpin B3 appeared in both cell extract and CM of DU145 and PC3 cells. Serpin B3 is well-known in the positive regulation of cell migration and proliferation, hence implicating a role in invasiveness (32). Moreover, previous studies had indicated that high expression of serpin B3 could be utilized as a favorable, predictive, and prognostic biomarker in diverse forms of cancer, such as squamous cell carcinoma (SCC), breast cancer, and hepatocellular carcinoma (33-35). Our recent study also showed that Serpin B3 is over-secreted in PC3 cells, though not in normal prostate cell lines such as RWPE-1 (36).

Proteomics results were supplemented by bioinformatics analysis using DAVID and STRING for further understanding of the biological characteristics of DEP. Functional annotation enrichment, using DAVID, suggests that laminin and/or laminin-associated proteins have more potential biological functions in DU145 than in PC3 cells (Figure 3A). Furthermore, results of Protein-Protein interaction (PPI), via STRING, demonstrated that Laminin subunit gamma-1 (LAMC1) could be connected to the proteins highly secreted by DU145 cells, except for those known to be vastly expressed in the brain, like the Amyloid beta A4 protein (AAP) (Figure 4A). For this reason, we focused on LAMC1 which is specifically secreted from DU145, rather than PC3, but not expressed in cell extract.

LAMC1 acts as an oncogene located either in extracellular region or secreted. Indeed, invading tumor cells attach to laminin and increase their metastatic potential (37). Thus, LAMC1 was seen to significantly increase tumor cell invasion in hepatocellular carcinoma (38), and overexpressed in grade III meningioma compared to that in grade I (39). In a most recent study, LAMC1 gene was up-regulated in uterine carcinomas, revealing clinically aggressive phenotypes (40). A previous study already revealed that silencing of LAMC1 by miR-29 significantly inhibited cell migration and invasion in cancer cells. Conversely, up-regulated LamC1 in PC cells affects cell proliferation, migration, and invasion in PCa cell lines (41). Another report concerning LAMC1 at PCa indicated LAMC1 as direct target of miR-22 (related to apoptosis and cell migration) (42). However, previous LAMC1 studies were all conducted in cell extracts, not on secreted LAMC1. Therefore, the present study confirmed that LAMC1 was secreted at a higher level in CM than in CE, more so in DU145 than in PC3. Hence, highly secreted LAMC1 might be suggested as a novel biomarker of predictive and prognostic metastases in PCa.

In this study, we used proteogenomics that integrates proteomics and genomics to discover a novel metastatic mutation associated with prostate cancer (Figure 1C). One of the major causes of cancer is the accumulated mutations in key proteins, which modulate cell growth, differentiation, and death (43). Mutations that change the protein sequence significantly affect protein stability and function. Indeed, proteins having PCa-associated mutations include BRCA1, BRCA2, and HOXB13, that may cause a higher risk of aggressive prostate cancer (44, 45). The HOXB13 protein, with G84E mutation, is associated with an increased risk of prostate cancer and other malignancies (46). Thus, the mutant proteins and/or sites enable us to look for new hallmarks of cancer (47). For this reason, we hypothesized that the secreted mutant proteins could be potent as new biomarkers with metastatic and/or distinguishable factors between PC3 and DU145 cells through a secretome analysis.

MS/MS results, re-analyzed using CMPD including large-scale coding variants for cancer, have provided a collection of mutant protein sequences that correspond to large-scale
genome data from NCI and TCGA. As a result, we identified six mutant secretory peptides in prostate cancer using MS-based proteomics combined with CMPD. The false-positive and non-mutant peptides were carefully verified and manually removed. Hence, only three out of six mutant peptides were quantified, of which two were found predominantly increased in DU145 (Table I). Researchers assume that due to gene aberrations, many protein sequence get altered, thereby increasing the proteasomal degradation of the incorrectly folded protein, subsequently, generating low abundance of mutant protein (48). Therefore, it follows that more aggressive PCa in bone metastatic PC3 cell lines would have a lower level of mutant protein than in DU145. Unfortunately, the six identified mutant peptides, classified under single nucleotide polymorphism (SNP), contained missense mutation that is not reported to be implicated in the disease. Recently, millions of SNPs involved with diseases, but biological mechanisms and functions are still not understood (49, 50). Nevertheless, genome-wide association studies (GWAS) have been recently reported the 76 SNPs associated with prostate cancer (51). Other recent studies have shown that patient with CASP9 polymorphism (rs1052576) were related with increasing prostate cancer risk and the common SPOP mutant, F133V variant, increases invasion in prostate cancer cells (52, 53). Although the details regarding the increased preference of the six mutation sites in DU145 is not yet clear, the mutant peptides in PCa secrete are expected to provide an ideal opportunity for metastatic biomarker evaluation, using the evolving proteomics technologies such as selective (SRM) and Parallel reaction monitoring (PRM) (54, 55). Hence, lately some of papers were already indicated that Ras-mutant peptides as cancer-specific biomarkers identified and quantified in clinical specimens by MS-based proteomics approach as well as proteogenomics revealed that TP53 mutation related to signal phosphorylation of MASTL and EEF2K in breast cancer (56, 57). Furthermore, mutant peptides can be used for tumor-associated antigens to apply to cancer peptide vaccines that utilize the clinical immunotherapies (58, 59).

In summary, we performed the quantitative proteomic analysis relevant to different metastatic sites associated with prostate cancer. In total, 598 proteins were identified, among which 561 were quantified based on the proteomic analysis. Among the quantified proteins, 128 were up-regulated and 83 were down-regulated in DU145 compared to PC3 cells. Based on our current results, LAMC1 was found to be highly over-expressed and specifically secreted in PCa cell lines; preferentially in DU145, indicating its potential as a biomarker to compare DU145 with PC3. The six newly-identified mutations, besides being potent as specific markers for prostate cancer, are especially expected to be markers to distinguish the site of metastasis. In this regard, future studies should aim at exploring how LAMC1 is hyper-secreted and distinguishes between DU145 and PC3 prostate cancer cell lines. The identified mutant peptides should be confirmed as novel prostate cancer markers that can be identified from patient samples by SRM or PRM methods. In conclusion, we anticipate that the identified metastatic factors, LAMC1 and the six mutant peptides, could be used as biomarkers in the prognosis as well as new drug targets of metastatic prostate cancer.

**Conflicts of Interest**

The Authors declare that they have no competing interests.

**Acknowledgements**

This work was supported by the National Research Foundation of Korea (NRF) grant, funded by the Korean government (MSIP) (No.2015R1A2A2A01004286 and 2018R1D1A1A02043591).

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