Bioinformatic Identification of Potential Hub Genes in Muscle-Invasive Bladder Urothelial Carcinoma

Changgang Guo1,2,*, Ting Shao3,*, Dadong Wei1, Chunsheng Li1,2, Fengjun Liu1, Minghui Li1, Zhiming Gao1,2, and Guochang Bao1,2

Abstract
Despite aggressive treatment approaches, muscle-invasive bladder urothelial carcinoma (MIBC) patients still have a 50% chance of developing general incurable metastases. Therefore, there is an urgent need for candidate markers to enhance diagnosis and generate effective treatments for this disease. We evaluated four mRNA microarray datasets to find differences between non-MIBC (NMIBC) and MIBC tissues. Through a gene expression profile analysis via the Gene Expression Omnibus database, we identified 56 differentially expressed genes (DEGs). Enrichment analysis of gene ontology, Kyoto Encyclopedia of Genes and Genomes, and Reactome pathways revealed the interactions between these DEGs. Next, we established a protein-protein interaction network to determine the interrelationship between the DEGs and selected 10 hub genes accordingly. Bladder urothelial carcinoma (BLCA) patients with COL1A2, COL5A1, and COL5A2 alterations showed poor disease-free survival rates, while BLCA patients with COL1A1 and LUM alterations showed poor overall survival rates. Oncomine analysis of MIBC versus NMIBC tissues showed that COL1A1, COL5A2, COL1A2, and COL3A1 were consistently among the top 20 overexpressed genes in different studies. Using the TCGA portal, we noted that the high expression of each of the four genes led to shorter BLCA patient overall survival. It was evident that BLCA patients with an elevated high combined gene expression had significantly shorter overall survival and relapse-free survival than those with low combined gene expression using PROGgeneV2. Using Gene Expression Profiling Interactive Analysis, we noted that COL1A1, COL1A2, COL3A1, and COL5A2 were positively correlated with each other in BLCA. These genes are considered as clinically relevant genes, suggesting that they may play an important role in the carcinogenesis, development, invasion, and metastasis of MIBC. However, considering we adopted a bioinformatic approach, more research is crucial to confirm our results. Nonetheless, our findings may have important prospective clinical implementations.

Keywords
muscle-invasive bladder carcinoma, gene expression profile, COL1A1, COL1A2, COL3A1, COL5A2

Introduction
Bladder urothelial carcinoma (BLCA) is the most common cancer of the urinary system, with 420,000 new cases and 160,000 deaths each year. The incidence of BLCA increases with age and is higher in men than in women. Many risk factors play a role in the pathogenesis of bladder cancer, such as environmental factors, living factors, and diet factors. Bladder cancers appear in many forms, namely urothelial carcinoma, squamous carcinoma, and adenocarcinoma; among which urothelial carcinoma is the most common, accounting for more than 90% of bladder cancer cases. About 75% of newly diagnosed patients suffer from non-muscle invasive bladder cancer (NMIBC), while the
4. The typical response to MIBC is neoadjuvant chemotherapy; however, despite all efforts, there remains a 50% chance of developing general incurable metastatic diseases, especially in patients with advanced T-stage and lymph-node positive diseases at the time of operation. Although cytotoxic chemotherapy has always been the main treatment for advanced bladder cancer, there seems to be a lack of high-quality evidence to guide the treatment of rare subgroups that are usually excluded from research. In recent years, the combination of microarray technology and bioinformatic tools has led to the identification of new genes associated with cancer progression, diagnosis, and prognosis; bioinformatic analysis is a feasible and valuable approach that is not only helpful for better understanding the mechanism underpinning MIBC occurrence and development but also permits the identification of novel targets for MIBC diagnosis and prognosis. Therefore, we adopted this approach and effectively screened 56 differentially expressed genes (DEGs) from GEO, among which 10 hub genes were associated with MIBC progression and prognosis; we believe our findings may serve as potential biomarkers for the clinical setting.

### Materials and Methods

#### Microarray Data

Gene Expression Omnibus (GEO) is a public functional genomics data repository of data from high-throughput gene expression analyses, chips, and microarrays. Four gene expression profiles (GSE7476, GSE37317, GSE48075, and GSE120736) were downloaded from GEO. The four datasets contained both NMIBC and MIBC tissue samples (6 vs 3; 8 vs 10; 69 vs 73; and 78 vs 61, respectively) (Table 1). All data are freely available, and the current research does not involve any human or animal experiments.

#### Identification of DEGs

DEGs were screened in both NMIBC and MIBC samples using GEO2R (http://www.ncbi.nlm.nih.gov/geo/geo2r); GEO2R is an interactive web-based tool through which comparisons between at least two GEO datasets are performed to select DEGs across experimental conditions. We analyzed DEGs in our four datasets taking into account the following cut-off criteria: adjusted \( P < 0.05 \) and \(|\log FC| \geq 1\). The intersecting genes were checked using the Venn diagram web tool (http://bioinformatics.psb.ugent.be/webtools/Venn/).

#### Protein-Protein interaction (PPI) Network Establishment

To construct a PPI network, protein interactions were first analyzed using the Search Tool for the Retrieval of Interacting Genes (STRING, version 11.0; http://string-db.org). Analyzing functional protein interactions is a useful method for studying the mechanisms underpinning BLCA. We used the STRING database to construct a PPI network with a confidence score above 0.4 as the cut-off criterion. Then, we utilized Cytoscape (version 3.6.1), which is a software platform with an open bioinformatic source, to establish the visualization of the PPI network.

#### Kyoto Encyclopedia of Genes and Genomes (KEGG) and Reactome Pathway Enrichment Analysis and Gene Ontology (GO) Annotation Analysis of DEGs

GO is considered one of the main bioinformatic methods for analyzing gene biological process; the KEGG is a database that was established through high-throughput experimental techniques, which allows users to explore biological systems and advanced functions from a wide range of molecular datasets, while Reactome is simultaneously an archive of biological processes and a tool for discovering unexpected functional relationships between datasets, such as gene expression profiles or somatic mutation catalogs of tumor cells. In this study, GO annotation analysis, as well as KEGG and Reactome pathway enrichment analysis of DEGs were executed using STRING (version 11.0; http://string-db.org); adjusted \( P < 0.05 \) showed statistical significance.

#### Hub Gene Selection and Analysis

CytoHubba is a plug-in for Cytoscape that is used to explore the degree of each protein node. In this study, the genes with the top 10 highest degree values were considered as hub genes. The network comprising central genes and their co-expressed genes was analyzed by CytoHubba; the relationship between the gene expression levels of NMIBC patients and MIBC patients was previously analyzed by Sanchez-Carbayo et al. using the online Oncomine database (http://www.oncomine.com). Moreover, hierarchical clustering was performed to analyze the expression levels of these hub genes in different pathological stages of BLCA using the UCSC Cancer Genomics Browser (http://genome-cancer.ucsc.edu); the Kaplan–Meier estimator was utilized to analyze the effect of the hub genes on patient disease-free survival.

| Dataset ID | NMIBC | MIBC | Total |
|------------|-------|------|-------|
| GSE7476    | 6     | 3    | 9     |
| GSE37317   | 8     | 10   | 18    |
| GSE48075   | 69    | 73   | 142   |
| GSE120736  | 78    | 61   | 139   |

GEO: gene expression omnibus; MIBC: muscle invasive bladder cancer; NMIBC: non-MIBC.

### Statistics of the Four Microarray Databases Derived from the GEO Database.
survival and overall survival via the cBioPortal platform (http://www.cbioportal.org)\(^{19,20}\).

**Oncomine Database Analysis**

Oncomine is an online tool for discovering new biomarkers in a variety of tumor microarray databases (https://www.oncomine.org/). We utilized this tool with the following screening criteria: (1) Cancer Type: Bladder Urothelial Cancer; (2) Data Type: mRNA; and (3) Analysis Type: Bladder Cancer Histology Analysis. We compared the overexpressed genes between MIBC and NMIBC tissues across eight studies in the ONCOMINE database\(^ {17,21-27}\).

**Analysis of COL1A1, COL1A2, COL3A1, and COL5A2 using TCGAportal, PROGgeneV2, and Gene Expression Profiling Interactive Analysis (GEPIA)**

The TCGAportal (http://www.tcgaportal.org/)\(^ {28}\) is an online web tool used to retrieve *COL1A1, COL1A2, COL3A1,* and *COL5A2* expression data in different BLCA grades and stages, and the corresponding overall survival statuses (data from The Cancer Genome Atlas database, https://portal.gdc.cancer.gov/). We determined the prognostic relevance of a multiple-gene signature on overall and disease-free survival in various publicly available gene expression datasets using the PROGgeneV2-Pan Cancer Prognostics Database (http://genomics.jefferson.edu/proggene/)\(^ {29}\). We also analyzed the correlation between the four genes using the GEPIA database (http://gepia.cancer-pku.cn/index.html)\(^ {30}\).

**Results**

**Identification of DEGs in MIBC**

After acquiring four different gene expression profiles, we identified a large number of DEGs (1156 in GSE7476, 746 in GSE120736, 1093 in GSE48075, and 604 in GSE37317). We then conducted a Venn analysis to examine the intersection among the DEG profiles (Fig. 1A). Ultimately, a total of 56 common DEGs were evident. Notably, the expression

---

**Fig. 1.** Identification of DEGs: Venn diagram and PPI network analyses. (A) A Venn diagram, including the total DEGs, was designed and used to detect DEG profile intersections; a total of 56 DEGs were expressed in the four groups. (B) A protein–protein interaction network was constructed using the differentially expressed genes; purple nodes represent upregulated genes, and green nodes represent downregulated genes. (C) The network of hub genes and their co-expressed genes was analyzed using the CytoHubba tool; purple nodes represent the co-expressed genes, while the other colored nodes represent the hub genes: the red node represents the highest degree gene, and the yellow node represents the lowest degree gene. DEGs: differentially expressed genes; PPI: protein-protein interaction.
level of these DEGs varied between MIBC and NMIBC tissues, as 32 DEGs were significantly upregulated and 24 were significantly downregulated in the former.

**KEGG and Reactome Pathway Enrichment Analysis and GO Annotation Analysis of DEGs**

As shown in Table 2, the GO analysis demonstrated that the DEGs were involved in biological process (BP) variations that were markedly enriched in extracellular structure organization; changes in cell components (CC) were prominently enriched in fibrillar collagen trimers. Moreover, the most significantly enriched molecular function (MF) term was platelet-derived growth factor binding. Furthermore, KEGG and Reactome pathway analyses showed that the DEGs were mainly enriched in protein digestion and absorption and extracellular matrix organization, respectively (Table 2).

**Establishment of a PPI Network and Hub Gene Selection**

The PPI network of the 56 DEGs was established via STRING tools and Cytoscape; this network consisted of 71 edges and 37 nodes (Fig. 1B). The DEGs were then ranked according to their degree of interaction, and the top 10 were identified as hub genes (Table 3). We saw that collagen type I alpha 1 chain (COL1A1) was the most significant gene, with a degree = 12, followed by COL3A1, COL1A2, VCAN, Table 2. Significantly Enriched GO Terms, KEGG, and RCTM Pathways of DEGs.

| Category      | Term                        | Description                                      | Count in gene set | FDR      |
|---------------|-----------------------------|--------------------------------------------------|-------------------|----------|
| BP term       | GO:0043062                  | Extracellular structure organization              | 16                | 4.18E-12 |
| BP term       | GO:0030198                  | Extracellular matrix organization                 | 14                | 1.48E-10 |
| BP term       | GO:0030199                  | Collagen fibril organization                      | 8                 | 5.17E-10 |
| BP term       | GO:0071229                  | Cellular response to acid chemical                | 7                 | 8.90E-04 |
| BP term       | GO:0001568                  | Blood vessel development                          | 9                 | 2.90E-03 |
| CC term       | GO:0005583                  | Fibrillar collagen trimer                          | 6                 | 1.18E-09 |
| CC term       | GO:0044420                  | Extracellular matrix component                    | 7                 | 4.25E-08 |
| CC term       | GO:0044421                  | Extracellular region part                         | 18                | 1.41E-06 |
| CC term       | GO:0031012                  | Extracellular matrix                              | 9                 | 5.15E-06 |
| CC term       | GO:0062023                  | Collagen-containing extracellular matrix          | 7                 | 7.99E-06 |
| MF term       | GO:0048407                  | Platelet-derived growth factor binding            | 4                 | 2.78E-05 |
| MF term       | GO:0004497                  | Monooxygenase activity                            | 4                 | 7.50E-03 |
| MF term       | GO:0005201                  | Extracellular matrix structural constituent       | 4                 | 7.50E-03 |
| MF term       | GO:0016404                  | 15-Hydroxyprostaglandin dehydrogenase (NAD+) activity | 2          | 7.50E-03 |
| MF term       | GO:0043167                  | Ion binding                                       | 32                | 7.50E-03 |
| KEGG pathway  | hsa04974                    | Protein digestion and absorption                  | 5                 | 5.80E-04 |
| RCTM pathway  | HSA-1474244                 | Extracellular matrix organization                 | 12                | 1.59E-08 |
| RCTM pathway  | HSA-300178                  | ECM proteoglycans                                 | 8                 | 1.59E-08 |
| RCTM pathway  | HSA-300170                  | Syndecan interactions                             | 6                 | 3.17E-08 |
| RCTM pathway  | HSA-1442490                 | Collagen degradation                              | 7                 | 8.18E-08 |
| RCTM pathway  | HSA-216083                  | Integrin cell surface interactions               | 7                 | 3.53E-07 |

BP: biological process; CC: cellular component; DEGs: differentially expressed genes; GO: gene ontology; KEGG: Kyoto encyclopedia of genes and genomes; MF: molecular function; RCTM: Reactome.

**Table 3. Top 10 Hub Genes with Higher Degree.**

| No. | Gene symbol | Full name | Alias | Degree |
|-----|-------------|-----------|-------|--------|
| 1   | COL1A1      | Collagen type I alpha 1 chain | EDSTH1, EDSC, OI1, OI2, OI3, OI4 | 12     |
| 2   | COL1A2      | Collagen type I alpha 2 chain | EDSTH2, EDSCV, OI4 | 11     |
| 3   | COL3A1      | Collagen type III alpha 1 chain | EDSA4A, EDSC, EDSCV, PMGEDSV | 11     |
| 4   | VCAN        | Versican  | CSPG2, ERVR, GHAP, PG-M, WGN, WGN1 | 10     |
| 5   | COL5A2      | Collagen type V alpha 2 chain | EDSC, EDSCV | 9      |
| 6   | COL5A1      | Collagen type V alpha 1 chain | EDSC, EDSC1 | 8      |
| 7   | LUM         | Lumican   | LDC, SLR2D | 7      |
| 8   | TIMP2       | TIMP metallopeptidase inhibitor 2 | CSC-21 K, DDC8 | 6      |
| 9   | SULF1       | Sulfatase 1 | SULF-1 | 5      |
| 10  | TPX2        | TPX2 microtubule nucleation factor | C20orf1, C20orf2, DIL-2, DIL2, FLS353, GD: C20orf1, HCA519, HCTP4, REPP86, p100 | 5      |
COL5A2, COL5A1, LUM, TIMP2, SULF1, TPX2, respectively. All the hub genes were markedly upregulated in MIBC tissues.

**Analysis for Hub Genes**

In Table 3, we listed the symbols, full names, and aliases of the hub genes. A central gene network along with co-expressed genes was detected by CytoHubba (Fig. 1C). In the Sanchez-Carbayo dataset, the gene expression level of all hub genes, except for TPX2, was higher in MIBC patients than in NMIBC patients (Fig. 2). Notably, hierarchical clustering showed that every hub gene was expressed in BLCA (Fig. 3A); however, gene expression levels in BLCA patients with stage III or IV tumors were significantly higher than those in BLCA patients with stage II tumors, with the exception of TPX2 (Fig. 3B). Subsequently, the effect of these hub genes on the disease-free survival and overall survival rates was analyzed by Kaplan–Meier curve using the online cBioPortal platform. BLCA patients with COL1A2, COL5A2, COL1A1 alterations were associated with poor disease-free survival and overall survival rates was analyzed by Kaplan–Meier curve using the online cBioPortal platform. BLCA patients with COL1A2, COL5A2, COL5A2 alterations were associated with poor disease-free survival and overall survival rates was analyzed by Kaplan–Meier curve using the online cBioPortal platform. BLCA patients with COL1A1, COL5A2, COL5A2 alterations were associated with poor disease-free survival and overall survival rates was analyzed by Kaplan–Meier curve using the online cBioPortal platform. BLCA patients with COL1A1 and LUM alterations were correlated to poor overall survival (Fig. 4B).

**Oncomine Database Analysis**

The MIBC and NMIBC tissues were analyzed using the Oncomine database. Notably, COL1A1, COL5A2, COL1A2,
and COL3A1 were among the top 20 overexpressed genes of both tissues in different studies (Fig. 5). Our findings suggest that these genes are crucial in bladder cancer, revealing that they may play an important role in the carcinogenesis and/or development of MIBC.

**Analysis of COL1A1, COL1A2, COL3A1, and COL5A2 using TCGAportal, PROGgeneV2, and GEPIA**

The expressions of COL1A1, COL1A2, COL3A1, and COL5A2 were directly proportional to the clinical stage and grade of BLCA according to TCGAportal online analysis. Out of 276 BLCA patients, 138 patients were in the low-expression gene group and 138 patients were in the high-expression gene group. High expression of COL1A1, COL1A2, COL3A1, and COL5A2 genes with BLCA patients showed remarkably worse overall survival time using the TCGAportal online tool (Fig. 6). We analyzed the ability of our four-gene signature to predict survival via PROGgeneV2; the results of combined gene expression analyses from public data are presented in Fig. 7A, B. Analyzing the GSE13507 gene array dataset, from the study of Wun-Jae Kim et al., revealed that patients with high combined gene

---

**Fig. 3.** Expression levels of hub genes in different pathological stages of BLCA using the UCSC browser. (A) The hierarchical clustering of hub genes was constructed. BLCA samples were analyzed for gene upregulation (marked in red) and downregulation (marked in blue). (B) Gene expression levels in BLCA patients with stage III or IV tumors are significantly higher than those in BLCA patients with stage II tumors, except for TPX2. BLCA: bladder urothelial carcinoma.
expression had significantly shorter overall survival than those with low combined gene expression in bladder cancer\(^{31}\). Whereas, the microarray dataset (GSE31684) by Schmidt et al. indicated that high combined gene expression was associated with poor relapse-free survival in bladder cancer\(^{32}\). Furthermore, we analyzed the correlation between our four genes of interest using the GEPIA-BLCA dataset; we found that \(COL1A1\) was positively correlated with \(COL1A2\) \((R = 0.97, P < 0.01)\), \(COL3A1\) \((R = 0.97, P < 0.01)\), and \(COL5A2\) \((R = 0.91, P < 0.01)\), respectively; while \(COL1A2\) was positively correlated with \(COL3A1\) \((R = 0.97, P < 0.01)\) and \(COL5A2\) \((R = 0.88, P < 0.01)\), respectively.

**Fig. 4.** Disease-free survival and overall survival analyses of hub genes via the cBioPortal online platform. (A) BLCA patients with \(COL1A2\), \(COL5A1\), and \(COL5A2\) alterations were associated with shorter disease-free survival \((P < 0.05)\). (B) BLCA patients with \(COL1A1\) and \(LUM\) alterations were linked to shorter overall survival \((P < 0.05)\). BLCA: bladder urothelial carcinoma.
Bladder cancer is one of the most common cancers in the world, with about 430,000 newly diagnosed cases each year. Most bladder cancers are of epithelial origin, among which urothelial tumors account for about 90% of the cases. Traditionally, radical cystectomy has been the standard treatment for MIBC; however, this approach is associated with potential complications that might negatively affect the patient’s quality of life. Although early diagnosis and multimodal treatments grant promising results, metastatic diseases are usually incurable and have a frightening 5-year overall survival rate of just 15%. Metastasis and recurrence are the leading causes of death in bladder cancer patients, especially those with MIBC. Therefore, identifying novel biomarkers and further understanding the molecular mechanism of MIBC are crucial for enhanced diagnosis strategies and treatment options for MIBC.

In this study, we identified 56 DEGs that were common to both MIBC and NMIBC; however, 32 DEGs were upregulated and 24 were downregulated in MIBC compared with those in NMIBC. KEGG and Reactome pathway enrichment analyses, along with GO function annotation, suggested that these 56 DEGs were mainly enriched in “extracellular structure organization,” “fibrillar collagen trimer,” “platelet-derived growth factor binding,” “protein digestion and absorption,” and “extracellular matrix organization” terms. PPI network construction allowed the visualization of DEG interconnections, which we then used to determine 10 hub genes. Notably, all our hub genes were overexpressed in MIBC. In the Sanchez-Carbayo Bladder 2 dataset, we found that COL1A1, COL3A1, COL1A2, VCAN, COL5A2, COL5A1, LUM, TIMP2, and SULF1 were differentially expressed in NMIBC and MIBC, except for TFP2, which is consistent with previous findings available on the UCSC platform (http://genome-cancer.ucsc.edu) that compared stage II tumors with stage III or stage IV tumors. Kaplan–Meier estimation revealed that BLCA patients with COL1A2, COL5A1, and COL5A2 alterations were linked to poor disease-free survival, whereas, BLCA patients having COL1A1 and LUM alterations were associated with poor overall survival. While comparing MIBC and NMIBC tissues via the Oncomine database revealed that four genes, COL1A1, COL5A2, COL1A2, and COL3A1, were consistently among the top 20 overexpressed genes in different studies. Combining the results of our different approaches, COL1A1, COL1A2, COL3A1, and COL5A2 were considered to be clinically relevant genes that may play an important role in the carcinogenesis and/or development of MIBC. We further demonstrated that with the increase in disease severity (clinical stage and tumor grade), the expression of COL1A1, COL1A2, COL3A1, and COL5A2 genes also increased. Furthermore, high expression of COL1A1, COL1A2, COL3A1, and COL5A2 led to a shorter overall survival of BLCA patients, as per data retrieved from TCGAportal online tool. We also analyzed that patients with high combined gene expression had significantly shorter overall survival and relapse-free survival than those with low combined gene expression in BLCA by PROGgeneV2 online tool; this finding is even more relevant considering that COL1A1, COL1A2, COL3A1, and COL5A2 were positively correlated according to our GEPIA investigation.

COL1A1 gene encodes the pro-alpha1 chains of type I collagen; Wu et al. previously revealed that the expression of COL1A1 was increased in colorectal adenocarcinoma and that COL1A1 silencing significantly inhibited cell proliferation, migration, and invasion and promoted apoptosis. Furthermore, Ma et al. reported that the siRNA-mediated COL1A1 silencing (siCOL1A1) inhibited the clonality, motility, and invasiveness of hepatocellular carcinoma cells and abolished the Slug-dependent epithelial-to-mesenchymal transition and hepatocellular carcinoma stem gene signatures. Zhu et al. showed that the loss of BMP1 promotes the expression of COL1A1 and EMT progression in colon...
cancer\textsuperscript{38}, \textit{COL1A1} may be a novel prognostic biomarker and a promising therapeutic target for breast cancer\textsuperscript{39}, malignant astrocytoma\textsuperscript{40}, colorectal cancer\textsuperscript{41}, and gastric cancer\textsuperscript{42}. Rong et al. have shown that \textit{COL1A2}, which encodes the pro-alpha2 chain of type I collagen, has a higher expression in gastric cancer than in normal tissues\textsuperscript{43}. Additionally, Tamilzhalagan et al. reported that miR-25 can selectively target the \textit{COL1A2} gene. Interestingly, miR-25 silencing increased the expression of \textit{COL1A2} and inhibited the expression of E-cadherin, revealing the inhibitory effect of mir-25 on diffuse gastric cancer\textsuperscript{44}. Notably, \textit{COL1A2} was highly expressed in pancreatic cancer\textsuperscript{45}, sarcoma\textsuperscript{46}, and gastric cancer\textsuperscript{47} tissues and promoted cell proliferation, migration, and invasion, which provided insights for mechanistic studies and clinical trials. \textit{COL3A1} was found to be the main structural component of hollow organs such as the great vessels, the uterus, and the intestines. Other functions of type III collagen include the interaction with platelets in the
coagulation cascade and the signaling in wound-healing processes\textsuperscript{48}. The expression of $COL3A1$ is associated with shorter overall survival of patients with epithelial ovarian cancer of four major tissue types\textsuperscript{49}. Qiu et al. reported that abnormal expression of miR-29a/b in nasopharyngeal carcinoma tissue may regulate $COL3A1$ expression and contribute to migration and invasion\textsuperscript{50}, while Su et al. indicated that let-7d can at least partially inhibit growth, metastasis, and macrophage infiltration in renal cell carcinoma by targeting $COL3A1$\textsuperscript{51}. $COL5A2$, collagen type V alpha 2 chain,
encodes an alpha chain that composes one of the low abundance fibrillar collagens, and mutations in this gene are associated with the Ehlers–Danlos syndrome. Zeng et al. demonstrated that COL5A2 was related to the poor clinical outcome and survival rate of patients with bladder cancer, suggesting its potential as a biomarker for bladder cancer. Vargas et al. reported that compared with ductal carcinoma, the in situ expression of COL5A2 increased in invasive breast cancer and was associated with extracellular matrix remodeling, suggesting that COL5A2 is associated with tumor progression in breast cancer. Moreover, the expression of COL5A2 was related to the malignancy of colorectal cancer and tongue squamous cell carcinoma.

In addition to COL1A1, COL1A2, COL3A1, and COL5A2, we found six other hub genes related to MIBC—VCAN, COL5A1, LUM, TIMP2, SULF1, and TPX2. These hub genes may be considered as valuable markers for the diagnosis, treatment, and prognosis of MIBC tumors.

Conclusion
In this study, we performed a comprehensive bioinformatic analysis to identify hub genes that may be involved in the progression, carcinogenesis, and development of MIBC. We found 10 hub genes with this potential; in particular, COL1A1, COL1A2, COL3A1, and COL5A2 caught our attention. Our findings may provide a new perspective for future research on the molecular mechanism of MIBC. However, considering we adopted a bioinformatics approach, further biological experiments are imperative.

Ethical Approval
This study was approved by our institutional review board.

Statement of Human and Animal Rights
This article does not contain any studies with human or animal subjects.

Statement of Informed Consent
There are no human subjects in this article and informed consent is not applicable.

Declaration of Conflicting Interests
The author(s) declared no potential conflicts of interest with respect to the research, authorship, and/or publication of this article.

Funding
The author(s) received no financial support for the research, authorship, and/or publication of this article.

ORCID iD
Guochang Bao  https://orcid.org/0000-0001-9982-1391

References
1. Hurst C, Rosenberg J, Knowles M. SnapShot: bladder cancer. Cancer Cell. 2018;34(2):350–350.e1.
2. Cumberbatch MGK, Jubber I, Black PC, Esperto F, Figueroa JD, Kamat AM, Kiemeney L, Lotan Y, Pang K, Silverman DT, Znaor A, et al. Epidemiology of bladder cancer: a systematic review and contemporary update of risk factors in 2018. Eur Urol. 2018;74(6):784–795.
3. Zhu S, Yu W, Yang X, Wu C, Cheng F. Traditional classification and novel subtyping systems for bladder cancer. Front Oncol. 2020;10:102.
4. Smith AB, Porten S, Kim S, Willis D, Plimack ER, Hoffman-Censits J, Roth B, Cheng T, Tran M, Lee IL, Melquist J, et al. Identification of distinct basal and luminal subtypes of muscle-invasive bladder cancer with different sensitivities to frontline chemotherapy. Cancer Cell. 2014;25(2):152–165.
5. Song BN, Kim SK, Mun JY, Choi YD, Leem SH, Chu IS. Identification of an immunotherapy-responsive molecular subtype of bladder cancer. EBioMedicine. 2019;50:238–245.
6. Szklarczyk D, Morris JH, Cook H, Kuhn M, Wyder S,�ka M, Vezzi F, process association networks, made broadly accessible. Nucleic Acids Res. 2017;45(D1): D362–D368.
7. Smoot ME, Ono K, Ruscheinski J, Wang PL, Ideker T. Cytoscape 2.8: new features for data integration and network visualization. Bioinformatics. 2011;27(3):431–432.
8. Ashburner M, Ball CA, Blake JA, Botstein D, Butler H, Cherry JM, Davis AP, Dolinski K, Dwight SS, Eppig JT, Harris MA, et al. Gene ontology: tool for the unification of biology the gene ontology consortium. Nat Genet. 2000;25(1):25–29.
9. Kanehisa M. The KEGG database. Novartis Found Symp. 2002;247:91–101.
10. Fabregat A, Jupe S, Matthews L, Sidiroopoulos N, Gillespie M, Garapati P, Haw R, Jassal B, Körning C, May B, Milacic M, et al. The reactome pathway knowledgebase. Nucleic Acids Res. 2018;46(D1): D649–D655.
17. Sanchez-Carbayo M, Socci ND, Lozano J, Saint F, Cordon-Cardo C. Defining molecular profiles of poor outcome in patients with invasive bladder cancer using oligonucleotide microarrays. J Clin Oncol. 2006;24(5):778–789.

18. Kent WJ, Sugnet CW, Furey TS, Roskin KM, Pringle TH, Zahler AM, Haussler D. The human genome browser at UCSC. Genome Res. 2002;12(6):996–1006.

19. Gao J, Aksoy BA, Dogrusoz U, Dresdner G, Gross B, Sumer SO, Sun Y, Jacobsen A, Sinha R, Larsson E, Cerami E, et al. Integrative analysis of complex cancer genomics and clinical profiles using the BioPortal. Sci Signal. 2013;6(269):pl1.

20. Cerami E, Gao J, Dogrusoz U, Gross BE, Sumer SO, Aksoy BA, Jacobsen A, Byrne CJ, Heuer ML, Larsson E, Antipin Y, et al. The cBio cancer genomics portal: an open platform for exploring multidimensional cancer genomics data. Cancer Discov. 2012;2(5):401–404.

21. Blaveri E, Simko JP, Korkola JE, Brewer JL, Baehner F, Mehta K, Devries S, Koppie T, Pejavar S, Carroll P, Waldman FM. Bladder cancer outcome and subtype classification by gene expression. Clin Cancer Res. 2005;11(11):4044–4055.

22. Dyrsjkot L, Thykjaer T, Kruhoff M, Jensen JL, Marcussen N, Hamilton-Dutoit S, Wolf H, Orntoft TF. Identifying distinct classes of bladder cancer using microarrays. Nat Genet. 2003;33(1):90–96.

23. Dyrsjkot L, Kruhoff M, Thykjaer T, Marcussen N, Jensen JL, Moller K, Orntoft TF. Gene expression in the urinary bladder: a common carcinoma in situ gene expression signature exists disregarding histopathological classification. Cancer Res. 2004;64(11):4040–4048.

24. Dyrsjkot L, Zieger K, Real FX, Malats N, Carrato A, Hurst C, Kotwal S, Knowles M, Malmstrom PU, de la Torre M, Wester K, et al. Gene expression signatures predict outcome in non-muscle-invasive bladder carcinoma: a multicenter validation study. Clin Cancer Res. 2007;13(12):3545–3551.

25. Lee JS, Leem SH, Lee SY, Kim SC, Park ES, Kim SB, Kim SK, Kim YJ, Kim WJ, Chu IS. Expression signature of E2F1 and its associated genes predict superficial to invasive progression of bladder tumors. J Clin Oncol. 2010;28(16):2660–2667.

26. Modlich O, Prisack HB, Pitschke G, Ramp U, Ackermann R, Bojar H, Vogeli TA, Grimm MO. Identifying superficial, muscle-invasive, and metastasizing transitional cell carcinoma of the bladder: use of cDNA array analysis of gene expression profiles. Clin Cancer Res. 2004;10(10):3410–3421.

27. Stransky N, Valtott C, Reyal F, Bernard-Pierrot I, de Medina SG, Segreves R, de Rycke Y, Elvin P, Cassidy A, Spraggion G, Graham A, et al. Regional copy number-independent deregulation of transcription in cancer. Nat Genet. 2006;38(12):1386–1396.

28. Xu S, Feng Y, Zhao S. Proteins with evolutionarily hypervariable domains are associated with immune response and better survival of basal-like breast cancer patients. Comput Struct Biotechnol J. 2019;17:430–440.

29. Goswami CP, Nakshatri H. PROGgeneV2: enhancements on the existing database. BMC Cancer. 2014;14:970.

30. Tang Z, Li C, Kang B, Gao G, Li C, Zhang Z. GEPIA: a web server for cancer and normal gene expression profiling and interactive analyses. Nucleic Acids Res. 2017;45(W1):W98–W102.

31. Kim WJ, Kim EJ, Kim SK, Kim YJ, Ha YS, Jeong P, Kim MJ, Yun SJ, Lee KM, Moon SK, Lee SC, et al. Predictive value of progression-related gene classifier in primary non-muscle invasive bladder cancer. Mol Cancer. 2010;9:3.

32. Riester M, Taylor JM, Feifer A, Koppie T, Rosenberg JE, Downey RJ, Bochner BH, Michor F. Combination of a novel gene expression signature with a clinical nomogram improves the prediction of survival in high-risk bladder cancer. Clin Cancer Res. 2012;18(5):1323–1333.

33. Antoni S, Ferlay J, Soerjomataram I, Znaor A, Jamel A, Bray F. Bladder cancer incidence and mortality: a global overview and recent trends. Eur Urol. 2017;71(1):96–108.

34. Mason SJ, Downing A, Wright P, Hounsome L, Bottomley SE, Corner J, Richards M, Catto JW, Glaser AW. Health-related quality of life after treatment for bladder cancer in England. Br J Cancer. 2018;118(11):1518–1528.

35. Nadal R, Bellmunt J. Management of metastatic bladder cancer. Cancer Treat Rev. 2019;76:10–21.

36. Wu W, Yang Z, Long F, Luo L, Deng Q, Wu J, Ouyang S, Tang D. COL1A1 and MZB1 as the hub genes influenced the proliferation, invasion, migration and apoptosis of rectum adenocarcinoma cells by weighted correlation network analysis. Bioorg Chem. 2020;95:103457.

37. Ma HP, Chang HL, Bamodu OA, Yadav VK, Huang TY, Wu ATH, Yeh CT, Tsai SH, Lee WH. Collagen 1A1 (COL1A1) is a reliable biomarker and putative therapeutic target for hepatocellular carcinogenesis and metastasis. Cancers (Basel). 2019;11(6):786.

38. Zhu X, Luo X, Jiang S, Wang H. Bone Morphogenetic protein 1 targeting COL1A1 and COL1A2 to regulate the epithelial-mesenchymal transition process of colon cancer SW620 cells. J Nanosci Nanotechnol. 2020;20(3):1366–1374.

39. Liu J, Shen JX, Wu HT, Li XL, Wen XF, Du CW, Zhang GJ. Collagen 1A1 (COL1A1) promotes metastasis of breast cancer and is a potential therapeutic target. Discov Med. 2018;25(139):211–223.

40. Sun S, Wang Y, Wu Y, Gao Y, Li Q, Abdulrahman AA, Liu XF, Ji GQ, Gao J, Li L, Wan FP, et al. Identification of COL1A1 as an invasion-related gene in malignant astrocytoma. Int J Oncol. 2018;53(6):2542–2554.

41. Zhang Z, Wang Y, Zhang J, Zhong J, Yang R. COL1A1 promotes metastasis in colorectal cancer by regulating the WNT/PCP pathway. Mol Med Rep. 2018;17(4):5037–5042.

42. Shi Y, Duan Z, Zhang X, Zhang W, Wang G, Li F. Down-regulation of the let-7i facilitates gastric cancer invasion and metastasis by targeting COL1A1. Protein Cell. 2019;10(2):143–148.

43. Rong L, Huang W, Tian S, Chi X, Zhao P, Liu F. COL1A2 is a Novel Biomarker to improve clinical prediction in human gastric cancer: integrating bioinformatics and meta-analysis. Pathol Oncol Res. 2018;24(1):129–134.

44. Tamilzhagan S, Rathinam D, Ganesan K. Amplified 7q21-22 gene MCM7 and its intrinsic miR-25 suppress COL1A2
associated genes to sustain intestinal gastric cancer features. Mol Carcinog. 2017;56(6):1590–1602.

45. Wu J, Liu J, Wei X, Yu Q, Niu X, Tang S, Song L. A feature-based analysis identifies COL1A2 as a regulator in pancreatic cancer. J Enzyme Inhib Med Chem. 2019;34(1):420–428.

46. Omar R, Cooper A, Maranyane HM, Zerbini L, Prince S. COL1A2 is a TBX3 target that mediates its impact on fibrosarcoma and chondrosarcoma cell migration. Cancer Lett. 2019;459:227–239.

47. Ao R, Guan L, Wang Y, Wang JN. Silencing of COL1A2, COL6A3, and THBS2 inhibits gastric cancer cell proliferation, migration, and invasion while promoting apoptosis through the PI3k-Akt signaling pathway. J Cell Biochem. 2018;119(6):4420–4434.

48. Kuivaniemi H, Tromp G. Type III collagen (COL3A1): gene and protein structure, tissue distribution, and associated diseases. Gene. 2019;707:151–171.

49. Engqvist H, Parris TZ, Kovacs A, Nemes S, Werner Ronnerman E, De Lara S, Biermann J, Sundfeldt K, Karlsson P, Helou K. Immunohistochemical validation of COL3A1, GPR158 and PITHD1 as prognostic biomarkers in early-stage ovarian carcinomas. BMC Cancer. 2018;19(1):928.

50. Qiu F, Sun R, Deng N, Guo T, Cao Y, Yu Y, Wang X, Zou B, Zhang S, Jing T, Ling T, et al. miR-29a/b enhances cell migration and invasion in nasopharyngeal carcinoma progression by regulating SPARC and COL3A1 gene expression. PLoS One. 2015;10(3):e0120969.

51. Su B, Zhao W, Shi B, Zhang Z, Yu X, Xie F, Guo Z, Zhang X, Liu J, Shen Q, Wang J, et al. Let-7d suppresses growth, metastasis, and tumor macrophage infiltration in renal cell carcinoma by targeting COL3A1 and CCL7. Mol Cancer. 2014;13:206.

52. Park AC, Phan N, Massoudi D, Liu Z, Kernien JF, Adams SM, Davidson JM, Birk DE, Liu B, Greenspan DS. Deficits in Col5a2 expression result in novel skin and adipose abnormalities and predisposition to aortic aneurysms and dissections. Am J Pathol. 2017;187(10):2300–2311.

53. Zeng XT, Liu XP, Liu TZ, Wang XH. The clinical significance of COL5A2 in patients with bladder cancer: a retrospective analysis of bladder cancer gene expression data. Medicine (Baltimore). 2018;97(10):e0091.

54. Vargas AC, McCart Reed AE, Waddell N, Lane A, Reid LE, Smart CE, Cocciardi S, da Silva L, Song S, Chenevix-Trench G, Simpson PT, et al. Gene expression profiling of tumour epithelial and stromal compartments during breast cancer progression. Breast Cancer Res Treat 2012;135:153–165.

55. Fischer H, Stenling R, Rubio C, Lindblom A. Colorectal carcinogenesis is associated with stromal expression of COL11A1 and COL5A2. Carcinogenesis. 2001;22(6):875–878.

56. Chen IC, Tseng YK, Shu CW, Weng TJ, Liou HH, Yen LM, Hsieh IC, Wang CC, Wu PC, Shiu YE, Fu TY, et al. Differential clinical significance of COL5A1 and COL5A2 in tongue squamous cell carcinoma. J Oral Pathol Med. 2019;48(6):468–476.