


Research Article

Multi-Omics Analyses Revealed Transcriptional Regulators associated with Immune Checkpoint Inhibitor Treatment in Advanced Bladder Cancer

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Simple Summary

Immune checkpoint inhibitor (ICI) therapy shifted the paradigm for advanced urothelial carcinoma (mUC) treatment, however, majority mUC patients present non-durable clinical benefit (non-DCB). Functional genomic analysis revealed key gene mutations and resistant gene expression in non-DCB group. Transcriptional reprogram reshapes the tumor microenvironment contributing to sensitivity of ICI treatment in mUC. This multi-omics study leverages our understanding of the molecular mechanisms of the intrinsic-or induced immune evasion, helps defining biomarkers for stratifying subgroup of patients for effective treatment.

Abstract

Background: Urothelial Bladder Cancer (UBC) is one of the most lethal cancers worldwide, the 5-year survival rate remains poor with platinum-based chemotherapy regimens as the standard of cancer treatment protocol. Recent FDA approval of a programmed death ligand-1 (PD-L1) inhibitor, atezolizumab, in advanced UBC patients is changing the therapeutic landscape. Although the response to anti-PD-L1 is correlated to PD-L1 expression and tumor mutation burden, the molecule determinants of responsiveness or non-responsiveness to Immune Checkpoint Inhibitor (ICI) is largely unknown.

Methods: R package maftools was used for genomic characterization and differential mutational analysis. EdgeR and DysRegSig algorithm were used for differential gene expression and dysregulator analysis. ConsensusTME algorithm was used for deconvolution of cell types within tumor microenvironment from bulk RNAseq data.

Result: A published immunotherapy cohort with whole exome sequencing, RNAseq and clinic outcome data for 29 metastatic urothelial cancer patients was used, paralleled with The Cancer Genome Atlas (TCGA) Bladder Cancer cohort, *GSE78220 cohort* and MSKCC-bladder cancer cohort. Genomic mutational profiling, mutational signature, a panel genes in antigen presentation and interferon signaling in bladder cancer were delineated with potential correlation with Durable Clinic Benefit (DCB) or non-DCB of PD-L1 inhibitor treatment. Characterized immune-responsive or resistant associated genes showed differentially expressed between DCB group and non-DCB group. Furthermore, transcriptional signature and transcriptional regulators between DCB and non-DCB were identified from transcriptomic data.

Conclusion: Our exploratory analyses provide multidimensional view of complexity of molecular determinants of immune responsiveness and suggest the influences of transcriptional reprogram in immune checkpoint blockade therapy.

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Whole exome sequencing for 391 bladder cancer samples were obtained from The Cancer Genome Atlas (TCGA), non-synonymous mutation was retained for analysis. The total number of non-synonymous mutations in each patient is heterogeneous with median number of 173.5 (Figure 1A). Somatic point mutations have two types of DNA substitution: transitions (A ↔ G and C ↔ T) and transversions (A ↔ C, G ↔ T, A ↔ T, C ↔ G). Transition mutation is higher than transversion mutation, and C > T transition and C > G transversion are higher than other type of mutations in average (p < 0.05) and in each individual case (Figure 1B, C). We then analyzed the mutational signature against known mutational signatures [16], three predominant signatures were defined that highly correlated with APOBEC signature (signature 2, 13), aging signature (signature 1) and ultra-hypermutators with error-prone polymerase POLE somatic mutation (Signature 10). In addition, microsatellite unstable signature (Signature 6) and defect DNA mismatch repair signature (Signature 20) are also high in bladder cancer (Figure 1D). These broad mutation processes in bladder cancers corresponded with the hyper mutation rate in bladder cancer genome and correlated to the subpopulation with high PD-L1 VCD8A expression [17]. Recently, Snyder et al [18] systematically analyzed multi-omic data from subset patients who receiving immune checkpoint inhibitor, atezolizumab,

in a clinical trial cohort and demonstrated the complex nature of immune response to ICI. We take advantage of whole exome sequencing data from this subset cohort to compare the genomic difference between patients with durable clinic benefit (DCB, response lasting more than 6 months), and patients without durable clinic benefit (non-DCB). We observed that average mutation variants are higher in DCB group than non-DCB group (unpaired two sample t-test, p < 0.05) (Supplementary Figure 1), this is consistent with higher mutation burden in responders relative to non-responders [14]. Interestingly, the distribution of mutation types (T > G, T > A, T > C, C > T, C > G, C > A) in DCB and non-DCB groups are similar. We obtained top 20 frequent mutated cancer genes in bladder cancer from COSMIC (supplementary Figure 2), we did not observe COSMIC cancer mutated genes enriched in either group (Supplementary Figure 2), however, the top 20 mutated gene profiling displayed difference between DCB and non-DCB group, such as *DSPP*, *FAM186A* and *NBPF10* mutations are only in non-DCB group (Supplementary Figure 3A and 3B). Moreover, pathways analysis showed increased mutated genes in each oncogenic pathways in non-DCB group, such as RTK-RAS, WNT, PI3K, MYC, TGF-beta and NRF2 pathways (Supplementary Figure 3C and 3D). These pathway activations may contribute to the resistance to ICI treatment. We also examined the MSKCC-bladder cancer

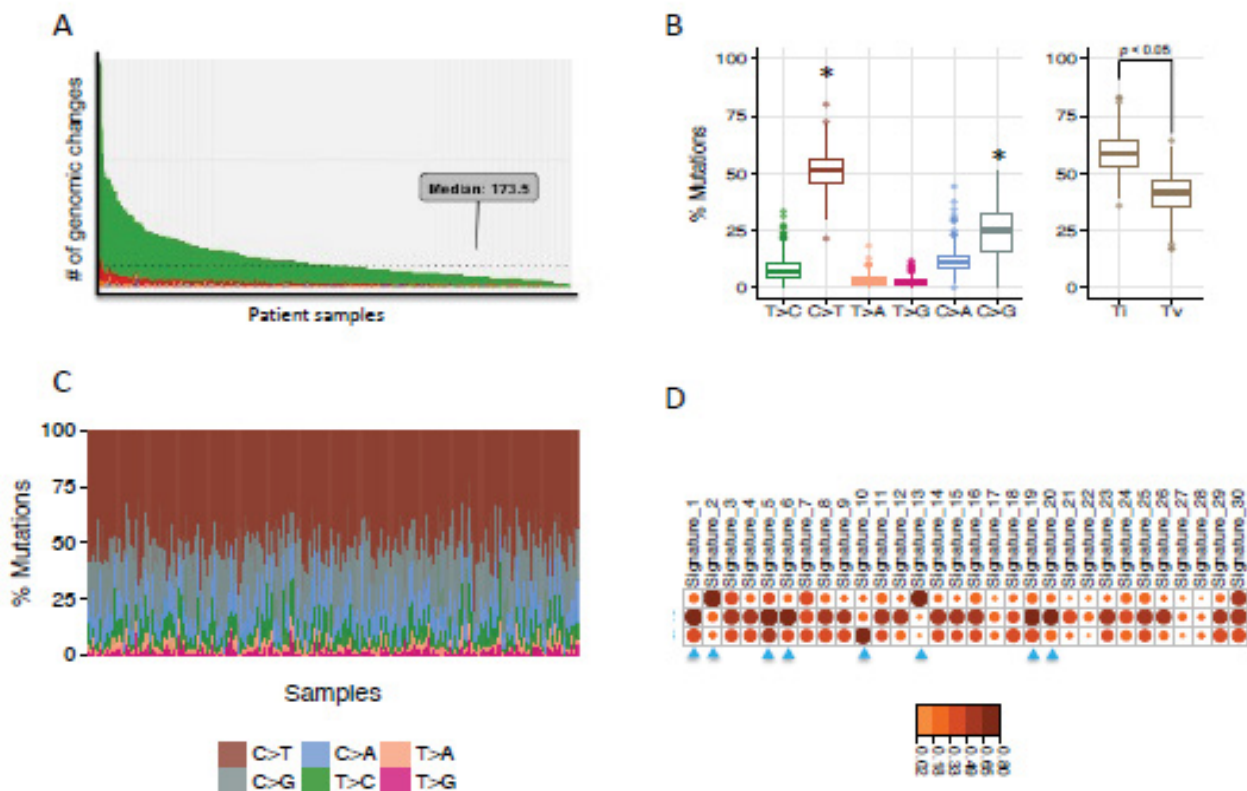


Figure 1: Mutational landscape of urothelial bladder cancer. Whole exom sequence data for TCGA-BLCA were used to investigate the tumor mutational burden (A), mutation type distribution (B, C) and somatic mutational signatures (D). Ti: Transitions, Tv: Transversions.

cohort including clinical and targeted genomic sequencing data [13], the PIK3CA mutation significantly higher in deceased patient group relative to living patient group after ICI treatment (Supplementary Figure 4), suggesting tumors harboring PI3K/Akt pathway activation might have short clinic benefit from ICI therapy.

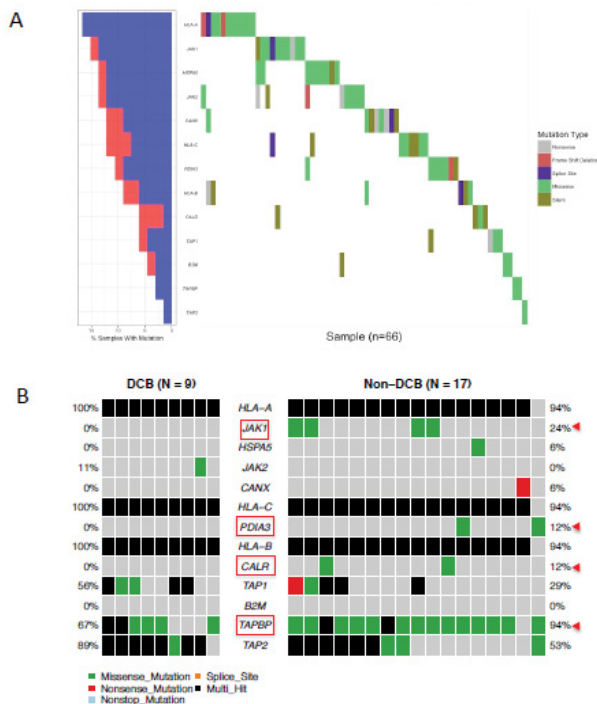
Interferon-Receptor Signaling and Antigen Presentation Pathways Defects in Bladder Cancer

Immune checkpoint blockade therapy on melanoma, lung cancer and colorectal cancer provided underlying mechanisms of response, resistance to PD-1 or CTLA-4 inhibitors. From genomic level, JAK1, JAK2 and B2M mutations have been found associated with acquired resistance to anti-PD-1 therapy in melanoma [9]. Positive selection for HLA and antigen-processing machinery mutations in tumors with TILs may have implications for potential immunotherapy resistance [12]. Castro et al reported that B2M mutation (occurring relatively early event in tumors) and HLA mutations were highly enriched in patients with microsatellite instability. In addition, these mutations had higher levels of immune infiltration by natural killer and CD8+ T cells and associated

with higher levels of cytotoxicity [19]. From TCGA bladder cancer patients (n=396) with whole exome sequence data, we found 66 (~17%) patients harboured mutations associated with either MHC binding molecules (HLA gene A, B, C), antigen-processing machinery (APM) pathway or interferon receptor signalling. Interestingly, these mutations exhibit mutually exclusive mutation pattern, indicating the non-redundant roles of these genes in immune response regulation (Figure 2A). We then explored these gene mutations in the ICI-treated cohort. We found that *JAK1*, *PDIA3*, *CALR*, *TAP1*, *TAPBP* have higher mutation frequency in non-DCB group than in DCB group, although there is no significant statistical difference due to a small sample size (Figure 2B). Taken together, genetic alterations tend to increase in non-DCB group from ICI treatment relative to DCB group.

Putative Immune Checkpoint Sensitivity Gene Expressions in Bladder Cancer

From transcriptional level, if a subset of genes such as granzyme A (*GZMA*), perforin (*PRF1*, CD8 T cell cytolytic score), *PDCD1LG2* (PD-L2), and *CTLA4* were highly expressed in the pretreatment melanoma tumors, these patients exhibited clinical benefit from CTLA-4 antibodies treatment [20]. On the other hand, immune suppressive surface receptors including *PDCD1* (PD-1), *LAG3*, *HAVCR2* (Tim-3), *CD160* and *CD244* as well as transcription factors such as *EOMES*, *PRDM1* (Blimp-1), and *TBX21* (T-BET) were identified from T cell exhaustion, which is reminiscent of non-response to ICI treatment [21]. We were wondering how these gene expressions changed in ICI-treated bladder cancer patients. From bulk RNAseq data, we observed that *CD8A*, *CD8B*, *PDCD1(PD-1)*, *PRF1*, *GZMA*, *EOMES*, and *TBX21* highly expressed in DCB group than in non-DCB group ($p < 0.2$, unpaired two sample t-test) (Figure 3A). Hierarchical



cluster of the correlation of these genes showed different patterns between two groups (Supplementary Figure 5). We then estimated the relative abundance of different cell types in the tumor microenvironment (TME) using bulk tumor RNAseq data with advanced consensusTME algorithm [22]. In ICI-treated cohort, among 19 different cell types within TME, the abundances of mast cells and fibroblast cells are marginally higher in non-DCB than in DCB group ($p = 0.082$ and $p=0.16$, respectively. Supplementary Figure 6). Tumor-infiltrating mast cells have recently reported to associate with resistance to anti-PD-1 therapy in humanized melanoma mouse model [23]. Tumor-associated fibroblasts could activate transforming growth factor β signalling resulting in restricting T-cell infiltration and restrain anti-tumor immunity in urothelial cancer [14]. Because the numbers of tumor-infiltrating CD8+ T cells (TILs) increase overall survival and immune checkpoint inhibitor efficacy. From RNA-seq data, the expression of CD8A (one component of the CD8 dimer) has been used as a surrogate for TIL levels [24]. Within 390 bladder cancer patients with clinical information and RNA-seq data, bladder cancer with high CD8A expression showed significantly longer overall survival time than that with lower

CD8A expression (log-rank test, $p < 0.05$). (Figure 3B). We observed similar trend of higher expression of CD8A and higher CD8+ T cell infiltration in DCB group compared to non-DCB group. These positive findings support the TIL cells and antigen recognition by T cells in DCB group.

Differentially Expressed Genes and Transcription Regulators between DCB and non-DCB

Although the ICI-treated advanced bladder cancer cohort has been analyzed by Snyder et al, but the study mainly focused on genomic alteration and T cell receptor functions [18]. We took advantage of RNAseq from pre-treated samples to examine the differentially expressed genes (DEGs) and master regulators between patients with DCB and non-DCB. Figure 4A showed 33 DEGs (absolute $\log_2FC > 1.5$ and false discovery rate < 0.1). Among them, the expressions of 9 genes (27.8%) are down-regulated in DCB and the expressions of 24 genes (72.2%) are up-regulated in DCB. These genes involved in multiple GO biological processes such as oxidative stress, metabolism and cell-cell adhesion (Figure 4B). Interestingly, there are several non-coding transcripts in DEGs, their biological functions remain elusive. In order

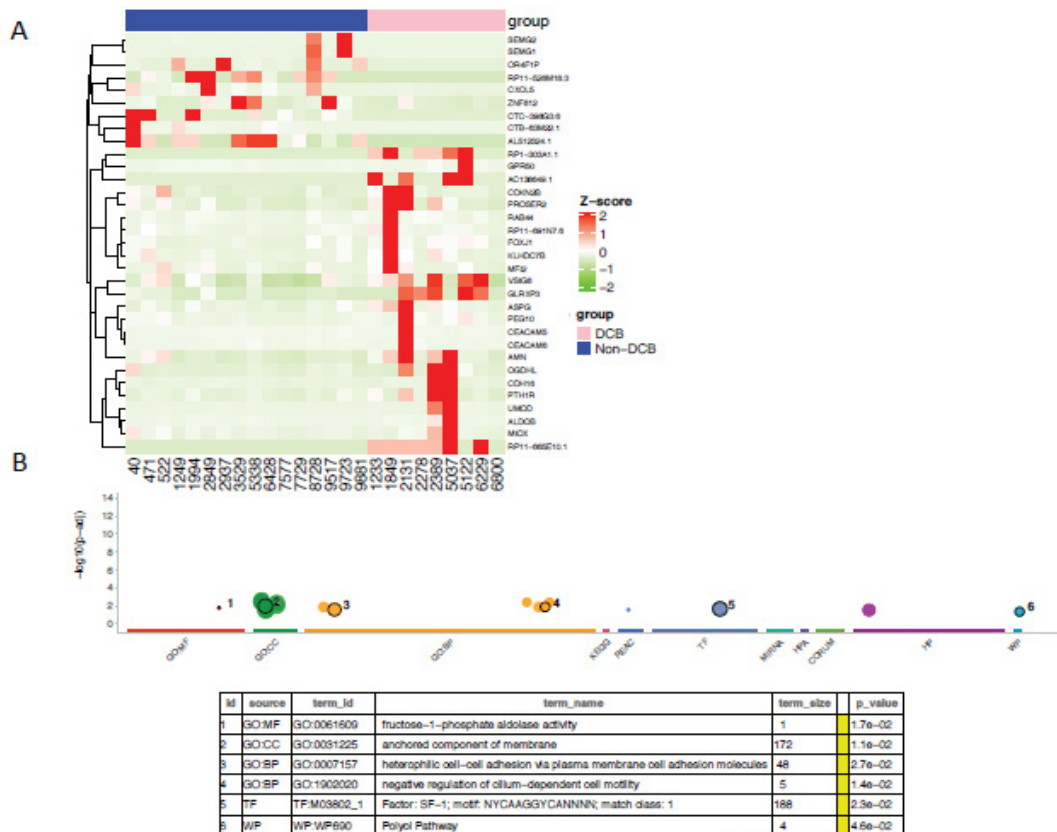


Figure 4: Differentially expressed genes between bladder cancer patients presenting durable clinic benefit (DCB) and non-DCB after PD-L1 inhibitor treatment. (A) Bulk RNAseq was derived from pretreatment tumor samples, the differentially expressed genes were identified based on $|\log_2(FC)| \geq 1.5$, and $FDR < 0.05$, and visualized in heatmap. (B) The GO terms of differentially expression genes were analyzed with goprofiler with multiple pathway databases. The selected representative GO terms were listed in the table below the plot. P-value for the enrichment < 0.05 .

to deeply understand the transcriptional regulation, we used DysRegSig algorithm, a machine learning-based gene dysregulation analysis [25], to explore the gene expression dysregulation between DCB and non-DCB patients, there are 184 significantly dysregulated transcription factor (TF) – targeted genes pairs (Supplementary Table 1). The dysregulated target genes are predominately enriched in immune-related signalling such as TNFs, STAT5-IL2-T cell activation, TNFR2 non-canonical NF-KB pathway, IL-18 signalling and cytokine-cytokine receptor interaction (Supplementary Table 2). Further gene regulatory network analysis revealed 10 top-ranked master transcriptional regulators (Supplementary Table 3). One of them is PATZ1-WNK4 with regulatory intensity 3.107, which is highly correlated to DCB phenotype (Figure 5A). Transcription factor PATZ1 negatively regulates the development of FOXP3 regulatory T cells [26], this may contribute to patients with favourable clinic benefit when receiving ICI treatment. We verified PATZ1-WNK4 high expression pattern in responder group from an independent ICI-treated melanoma cohort [8]. Other TF-target pairs such as TCF7L2-HOXA7, ETV6-ZNF277, and CIC-SGSM2 showed stronger regulatory intensities in DCB group than in non-DCB group (Figure 5B-D). All these TF-targets regulations tend to promote cancer cells or its microenvironments responding to ICI treatment in DCB group, the molecular mechanisms remain to be further investigated.

Discussion

Remarkable clinical efficacy, durable response and low toxicity of immune checkpoint blockade treatment have been observed in various malignancies including UBC [5, 6, 27]. Anti-PD-L1 for advanced bladder cancer could reach 43.3% response rate in early small clinical trial cohort [27], it reached >10% response rate in later large clinical trial cohort [5, 6]. However, a large proportion of patients failed to respond to checkpoint inhibitors, therefore, it is crucial to identify biomarker(s) to stratify or predict responders to achieve better clinical outcome. The molecular determinants of responsiveness to PD-1/PD-L1 and CTLA-4 inhibitors appear to be heterogeneous and complex. We take advantage of bladder cancer cohort with PD-L1 inhibitor treatment to analyze the molecular determinants of immunosensitivity or resistance in bladder cancer. First, urothelial bladder cancer is a genomic disorder with high mutation load, consistent with TCGA-bladder cancer cohort. Three dominant mutation signatures have been identified such as APOBEC, POLE signature and aging signature. The PD-L1 inhibitor (Atezolizumab) treated patients with advanced bladder cancer (n=29) exhibit limited recurrent gene mutations correlated to DCB group or non-DCB group from WES, for example, although relatively higher tumor mutation burden in DCB [14], we found several non-DCB group unique gene mutations (*DSPP*, *FAM186A* and *NBPF10*) are

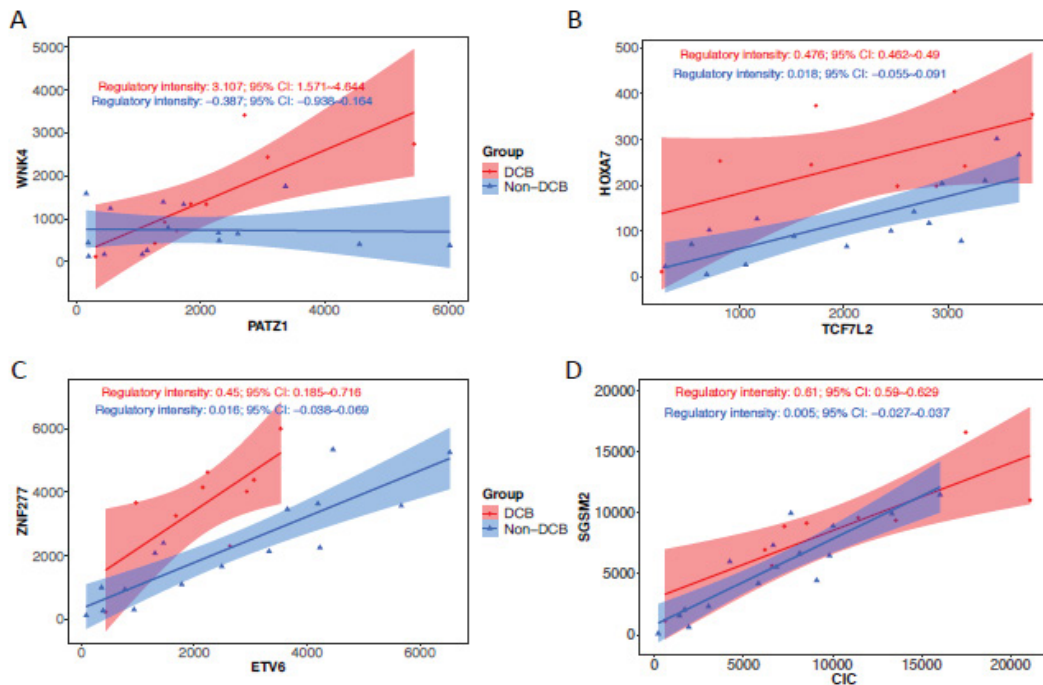


Figure 5: Representative dysregulators and targets in immune PD-L1 inhibitor treated bladder cancer. RNAseq derived from advanced bladder cancer patients with durable clinic benefit (DCB) group and non-DCB group was subjected to systematically identification of dysregulation events associated with treatment benefit using DysRegSig algorithm. The top one ranked transcription factor (PATZ1) and target (MNK4) and other transcription factor targets (TCF7L2-HOXA7, ETV6-ZNF277, CIC-SGSM2) were highly correlated in DCB group compared to non-DCB group. The regulatory intensity and 95% confidence interval was shown in each group.

present when compared to DCB group, the functions of these gene mutations are unknown. From the pathway analysis, we observed mutated genes in each pathway increased in non-DCB group compared to DCB group (Supplementary Figure 3), orthogonal MSKCC-bladder cancer cohort showed *PIK3CA* mutation significantly higher in deceased patient group compared to living patient group after receiving ICI treatment. This suggesting the activation of these pathways (RTK-RAS, PI3K, MYC, TGF-beta, NRF2, Hippo) collectively contribute to unfavorable clinical outcome of ICI-treatment. Second, we found 17% bladder cancer patients from TCGA-BLCA cohort harbouring mutations involving MHC molecules, antigen processing machinery or interferon-receptor signalling pathway. These mutations exhibited mutually exclusive pattern. In the ICI-treated cohort, we observed that *JAK1*, *TAPBP* mutation frequency is relative higher in non-DCB group than DCB group (Figure 2). Third, a panel of nineteen immuno-responsive genes found in other types of cancers differentially expressed between DCB group and non-DCB group of bladder cancer patients who received ICI treatment, such as *CD8A*, *CD8B*, *GZMA* (granzyme A), transcription factor *EOMES*, *TBX21* marginally higher in DCB relative to non-DCB, these gene expressions indicate functional activities of immunity (Figure 3A). Moreover, patients with high CD8A expression demonstrated longer survival time in TCGA bladder cancer cohort (Figure 3B). Finally, we identified differentially expressed genes between DCB group and non-DCB group. These transcriptional signature is able to discriminate ICI-treated clinic outcomes (Figure 4). Ten significant transcriptional regulators were further characterized from the RNAseq, a group of transcription factor-target (e.g., *PATZ1-WNK4*, *TCF7L2-HOXA7*, *ETV6-ZNF277*, and *CIC-SGSM2*) demonstrated higher regulatory intensities in DCB group relative to non-DCB group, which reflects transcriptional reprogramming in tumor cells and or within tumor microenvironment that influence the consequence of immune checkpoint blockade therapy. Transcriptional signatures are potential predictors for responding to antagonists of PD-1. For instance, up-regulation of mesenchymal transition genes (*AXL*, *ROR2*, *WNT5A*, *LOXL2*, *TWIST2*, *TAGLN*, *FAP*), immunosuppressive genes (*IL10*, *VEGFA*, *VEGFC*) and monocyte and macrophage chemotactic genes (*CCL2*, *CCL7*, *CCL8*, *CCL13*) preferentially in non-responding tumors [8], indicating patients with these expression signature most likely do not respond to anti-PD-1 treatment. These findings highlight the complexity of interplay between cancer cells and the immune system which will need further elucidation in urothelial bladder cancer. The limitation of this analysis is that the TCGA-BLCA cohort did not have PD-1 or PD-L1 inhibitor treatment information, and the ICI-treated metastatic bladder cancer cohort contains small samples, particularly patients with durable clinic benefit group. Most the difference between

DCB and non-DCB group are marginal. Nevertheless, comprehensive analyses of this valuable cohort with multi-omics and clinic data provide genomic and transcriptional insight into potential molecular mechanisms for ICI-treated durable clinical benefit. The findings remain to be confirmed in a large data in the future.

Supplementary Materials

The following supporting information can be downloaded at: www.mdpi.com/

Authors' Contributions

Conceived by WC, X.F.X, investigated by FX, DF, X.Q.S, JH, data collection and analysis with ZW, Y.H.C. J.P.D, T.L.Z., reviewed and edited by J.P.G.

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Ethics Approval and Consent to Participate

NA.

Availability of Data and Materials

The datasets analyzed during the current study are public available from weblinks: Somatic mutations shared by bladder cancer patients in COSMIC database (<https://cancer.sanger.ac.uk/cosmic/browse/tissue>). TCGA-BLCA whole exome mutation annotation file was downloaded from firehose broad institute (<http://firebrowse.org/?cohort=BLCA>). TCGA-BLCA RNAseq data was downloaded from UCSC Xena browser ([https://xenabrowser.net/datapages/?cohort=TCGA%20Bladder%20Cancer%20\(BLCA\)&removeHub=https%3A%2F%2Fxena.treehouse.gi.ucsc.edu%3A443](https://xenabrowser.net/datapages/?cohort=TCGA%20Bladder%20Cancer%20(BLCA)&removeHub=https%3A%2F%2Fxena.treehouse.gi.ucsc.edu%3A443)). MSKCC-bladder cancer cohort was downloaded from cbioport website

(https://www.cbioportal.org/study/summary?id=tmbskcc_2018). whole exome sequence, bulk RNAseq and clinical information derived from 29 locally advanced or metastatic bladder cancer patients (Memorial Sloan Kettering Cancer Center) were downloaded from <http://doi.org/10.5281/zenodo.546110> and <https://github.com/hammerlab/multi-omicurothelialanti-pd1>, respectively. GSE78220 RNAseq data was downloaded from GEO website.

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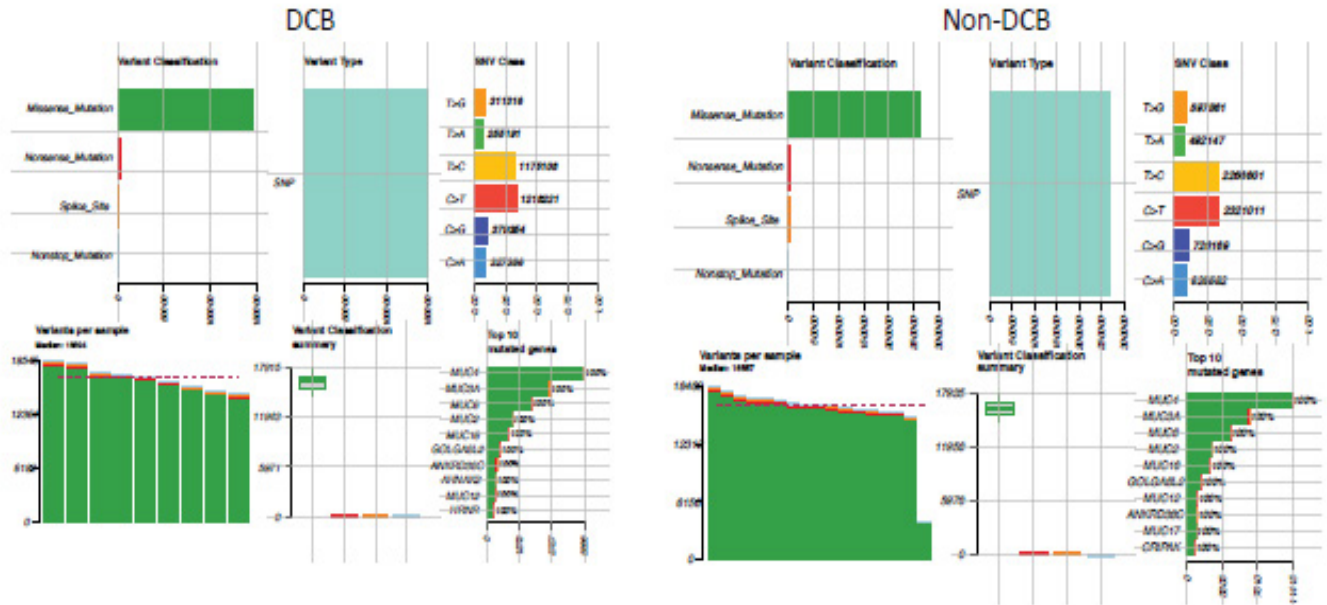
Conflict of Interests

The authors declare no conflict of interest.

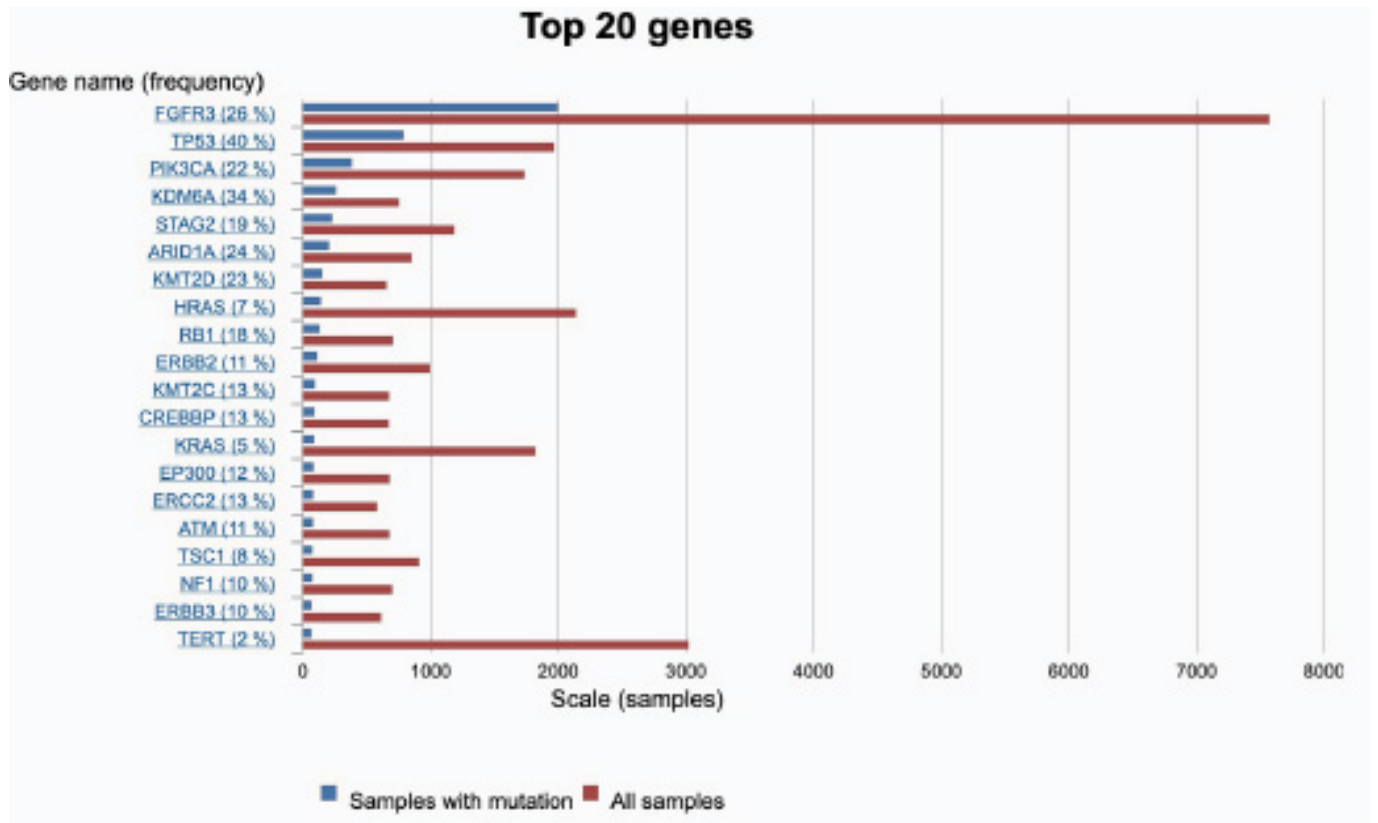
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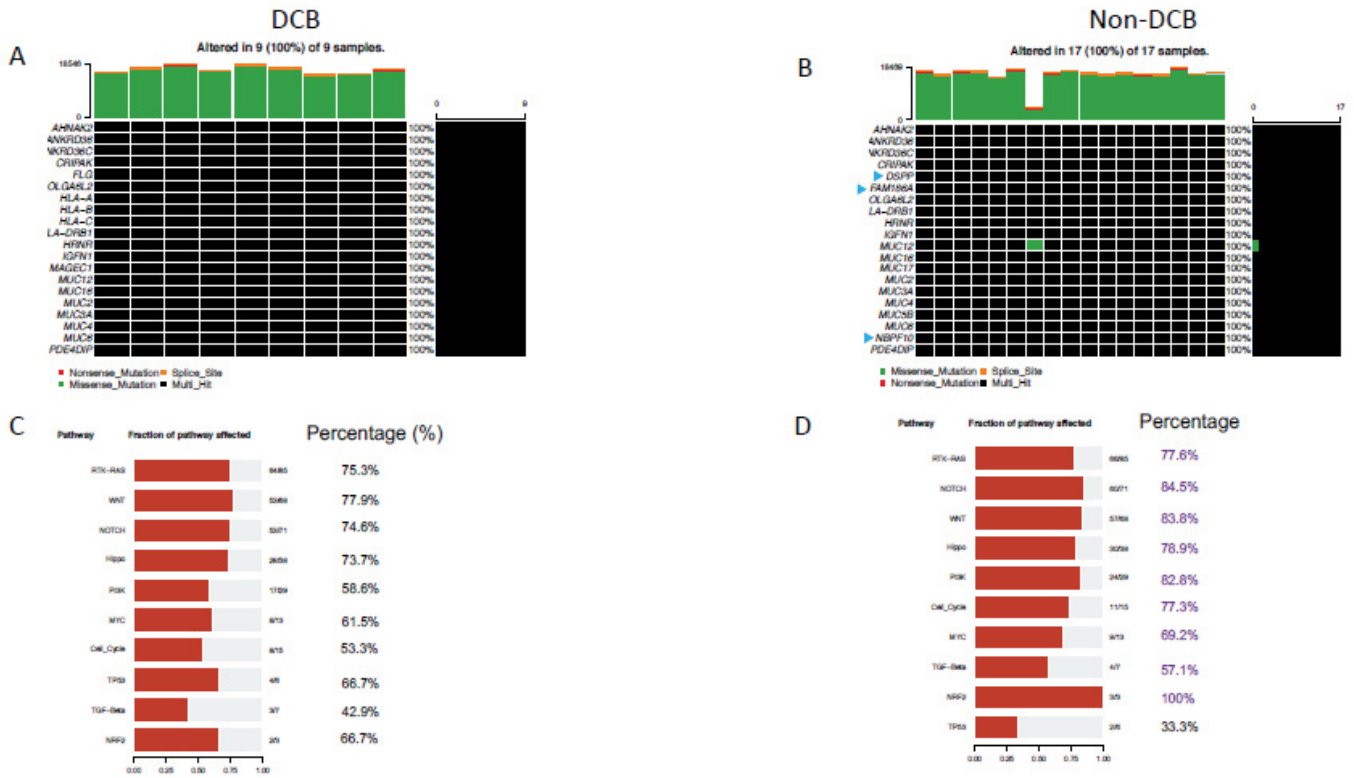
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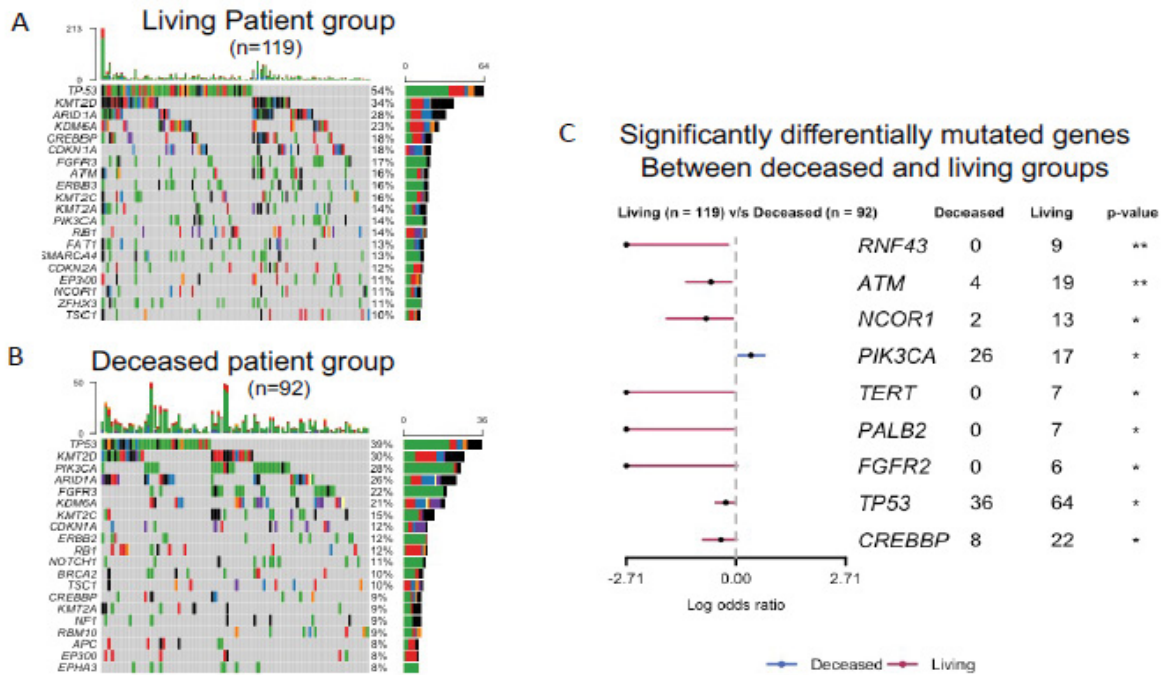
Supplementary Figure 1: Mutational landscape of pre-treatment tumor samples from advanced bladder cancer patients with durable clinical benefit (DCB) or without DCB after receiving immune checkpoint PD-L1 inhibitor (atezolizumab) therapy.



Supplementary Figure 2: The mutational frequency of top 20 cancer genes in bladder cancer derived from COSMIC. Red bar: all bladder cancer samples were tested; blue bar: samples with gene mutations.



Supplementary Figure 3: Concurrent gene mutations and involved pathways in pre-treatment tumor samples from advanced bladder cancer patients with durable clinical benefit (DCB) or without DCB after receiving immune checkpoint PD-L1 inhibitor (atezolizumab) therapy. (A, B) Oncoprint of gene mutations in DCB, non-DCB group. (C, D) Pathway enrichment from non-synonymous mutated genes in DCB or non-DCB group.



Supplementary Figure 4: Mutational landscape (A, B) and differentially mutated genes (C) between patient living and deceased group after immune checkpoint inhibitor treatment. MSKCC-bladder cancer cohort with clinical and targeted exome sequencing data were obtained from cBioportal website and maftools algorithm was used for genomic analysis.

Supplementary Table 1: Dysregulation of Transcription factor-target gene pairs in immune checkpoint inhibitor PD-L1 treated advanced bladder cancer between durable clinic benefit (DCB) group and non-DCB group

TF	Target	unb.coef.1	low.lim.1	up.lim.1	P.val.1	unb.coef.2	low.lim.2	up.lim.2	P.val.2	de.logFC
ZNF764	ABCC6	0.72	0.16	1.28	0.09	-0.38	-0.71	-0.05	0.08	541.48
TFAP4	ACSM5	0.42	0.41	0.43	0	-0.05	-0.2	0.09	0.53	170.29
ZNF33B	ACSM5	0.19	0.18	0.2	0	0.05	-0.04	0.14	0.36	170.29
ZFP2	ACSM5	1.05	1.04	1.06	0	0.11	0.01	0.21	0.07	170.29
ZNF554	ADCY1	0.12	0.1	0.15	0	-0.01	-0.08	0.06	0.75	422.31
ZNF530	ADCY1	0.32	0.29	0.35	0	0	-0.08	0.07	0.87	422.31
HOXA6	ADCY1	0.5	0.47	0.53	0	0.03	-0.04	0.1	0.31	422.31
ZNF764	AGXT2	0.65	0.46	0.84	0	0	-0.14	0.13	0.95	144.8
POU5F1	AGXT2	0.27	0.07	0.46	0.02	-0.15	-0.34	0.03	0.18	144.8
EN1	ALDH8A1	0.57	0.32	0.83	0	0	-0.12	0.12	1	146.9
HOXD9	ALDH8A1	0.41	0.08	0.73	0	-0.13	-0.27	0.01	0.06	146.9
SCRT2	AMY1C	0.4	0.18	0.63	0	-0.15	-0.47	0.16	0.43	124.32
ZNF324B	ARFRP1	0.23	0.11	0.36	0	0	-0.03	0.04	0.42	709.58
ZNF574	ARFRP1	0.29	0.16	0.42	0	0	-0.03	0.04	0.39	709.58
MAFG	ARFRP1	0.52	0.39	0.65	0	0.11	0.07	0.14	0	709.58
ZNF263	ATP6V0A4	0.34	-0.02	0.7	0	-0.22	-0.4	-0.04	0.04	239.32
PURA	BNIP1	0.89	0.82	0.96	0	-0.06	-0.21	0.08	0.36	77.97
STAT1	BPY2B	0.97	0.84	1.11	0	0.07	-0.37	0.51	0.79	16.88
ZNF213	C1orf167	0.73	0.52	0.93	0	-0.12	-0.3	0.05	0.23	241.73
ZNF263	C6orf136	0.39	0.25	0.54	0	0	-0.03	0.03	0.8	444.98
MAZ	C6orf136	0.61	0.46	0.76	0	0	-0.03	0.03	0.92	444.98
ZNF444	CAMK2N2	0.94	0.76	1.12	0	0	-0.33	0.33	0.99	96.89
FOXL1	CCDC179	0.88	0.6	1.17	0	-0.02	-0.46	0.41	0.93	30.91
TLX2	CCR5	0.47	0.47	0.48	0	0.12	-0.07	0.3	0.3	159.36
ZSCAN29	CCR5	0.19	0.18	0.2	0	0	-0.1	0.11	0.97	159.36
ZNF418	CCR5	0.25	0.24	0.26	0	0.02	-0.08	0.13	0.72	159.36
NR1I2	CCR5	0.23	0.22	0.24	0	-0.01	-0.14	0.12	0.91	159.36
RELB	CD8A	0.91	0.87	0.95	0	0.48	0.16	0.8	0.01	144.98
FOXJ2	CHST13	-0.08	-0.21	0.06	0.2	-0.76	-0.96	-0.56	0	48.72
FEV	CIB3	0.68	0.41	0.95	0.01	0	-0.38	0.38	1	86.9
ZBTB4	CLDND2	0.66	0.44	0.88	0	-0.11	-0.13	-0.1	0	52.87
SPIB	CLEC10A	0.82	0.71	0.93	0	0.01	-0.02	0.03	0.73	273.06
SPI1	CLEC10A	0.18	0.08	0.27	0	-0.02	-0.05	0.01	0.12	273.06
ZNF17	CLEC10A	0.01	-0.09	0.12	0.72	-0.12	-0.15	-0.09	0	273.06
HIVEP3	CLIC5	0.72	0.51	0.93	0	0.11	-0.25	0.47	0.5	580.51
HOXC8	COL11A2	0.29	0.25	0.33	0	0.1	-0.03	0.24	0.22	491.05
ZNF30	COL11A2	0.42	0.37	0.46	0	0.13	-0.01	0.27	0.13	491.05
ZNF18	COL11A2	0.45	0.41	0.49	0	0.02	-0.11	0.15	0.8	491.05
SPIC	COLEC10	0.54	0.5	0.57	0	-0.1	-0.41	0.21	0.59	50.78
NFIC	CYP21A2	0.61	0.42	0.81	0	0.1	-0.01	0.21	0.12	107.19
NFIA	CYP21A2	0.27	0.08	0.46	0.02	-0.09	-0.24	0.06	0.31	107.19
ALX3	DENND1C	0.43	0.37	0.5	0	0.02	-0.2	0.25	0.7	2753.22
HIC2	DENND1C	0.54	0.48	0.6	0	0.08	-0.13	0.29	0.35	2753.22
HOXC4	DMRT2	0.63	0.46	0.81	0	-0.2	-0.45	0.05	0.19	185.24

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ZNF324B	DRD5	0.47	0.25	0.68	0	-0.75	-1.71	0.21	0.2	164.51
ZNF785	EMID1	0.95	0.83	1.06	0	0.04	-0.25	0.33	0.6	293.63
KLF8	FAM78B	0.67	0.56	0.79	0	0.19	0.16	0.22	0	104.57
ZSCAN22	FAM78B	0.01	-0.1	0.13	0.85	-0.59	-0.62	-0.56	0	104.57
ZNF283	FCRL6	0.01	-0.03	0.04	0.35	-0.11	-0.18	-0.03	0	237.38
NHLH1	FCRL6	0.81	0.77	0.85	0	0.06	-0.03	0.15	0.21	237.38
ZNF423	FOXL1	0.91	0.53	1.28	0	-0.28	-0.52	-0.05	0.05	170.89
HIC2	FRMD1	0.69	0.51	0.87	0	0.07	-0.25	0.39	0.71	254.57
ZNF71	GABRB2	0.8	0.18	1.42	0.09	-0.32	-0.72	0.08	0.22	99.53
KLF12	GABRG3	0.81	0.74	0.87	0	-1.68	-3.59	0.23	0.15	66.83
ZNF740	GALNT16	0.71	0.58	0.85	0	-0.11	-0.54	0.31	0.66	270.08
ZNF263	GCGR	0.75	0.63	0.87	0	-0.27	-0.78	0.24	0.38	132.51
TFE3	GIMAP1	0.41	0.35	0.47	0	-0.37	-0.64	-0.11	0.02	430.81
ZNF774	GIMAP6	0.29	0.25	0.32	0	-0.08	-0.12	-0.04	0	332.49
ETV2	GIMAP6	0.27	0.23	0.3	0	-0.02	-0.05	0.02	0.02	332.49
ZNF75A	GIMAP6	0.44	0.4	0.47	0	0.01	-0.03	0.05	0.3	332.49
ZNF785	GIMAP6	0.27	0.24	0.3	0	0	-0.04	0.03	0.6	332.49
GLIS2	GIPR	0.61	0.43	0.79	0	0.13	-0.13	0.38	0.12	162.22
ZSCAN22	GJA3	0.51	0.49	0.53	0	-0.06	-0.18	0.06	0.03	33.47
ZNF764	GJA3	0.18	0.16	0.19	0	-0.01	-0.13	0.12	0.79	33.47
FOX11	GJA3	0.48	0.46	0.49	0	0.04	-0.07	0.15	0.32	33.47
ZNF224	GNG4	0.46	0.17	0.75	0	0.01	-0.12	0.13	0.93	75.27
FOX12	GOLGA8J	1.5	0.42	2.58	0.07	-0.2	-0.5	0.11	0.31	419.13
GLI4	GRID2IP	0.26	0.06	0.46	0	-0.01	-0.08	0.05	0.38	564.11
ESRRA	GRID2IP	0.1	-0.12	0.31	0.05	-0.23	-0.29	-0.16	0	564.11
STAT1	GRIN3B	0.36	0.28	0.43	0	0.03	-0.11	0.18	0.19	618.29
TBX21	GRIN3B	0.32	0.24	0.4	0	-0.05	-0.16	0.07	0.16	618.29
ERF	HOXA10	0	-0.01	0.02	0.16	-0.18	-0.25	-0.11	0	279.37
HOXA9	HOXA10	0.78	0.76	0.8	0	0.03	-0.05	0.1	0.49	279.37
ETV7	HOXA10	0.01	-0.01	0.02	0.09	-0.4	-0.46	-0.33	0	279.37
BHLHE40	HOXA7	0.12	0.08	0.16	0	0	-0.05	0.06	0.48	133.01
GFI1	HOXA7	0.4	0.36	0.45	0	0.02	-0.03	0.07	0.02	133.01
KLF14	HOXA7	0.53	0.49	0.57	0	0	-0.05	0.05	0.81	133.01
TCF7L2	HOXA7	0.38	0.35	0.42	0	0	-0.05	0.05	0.99	133.01
MNT	HOXA7	0.1	0.06	0.14	0	0	-0.05	0.06	0.47	133.01
MXD4	HOXA9	0.66	0.5	0.83	0	-0.5	-1.16	0.15	0.21	259.31
TFCP2L1	HPD	0.78	0.47	1.08	0	0.01	-0.13	0.14	0.93	202.34
ZNF263	HS6ST3	0.68	0.62	0.74	0	-0.91	-1.79	-0.04	0.09	346.44
ZBTB22	HSD3B1	0.86	0.45	1.27	0.03	0.04	-0.19	0.28	0.77	95.4
ZNF76	HUNK	0.72	0.65	0.78	0	-0.03	-0.15	0.1	0.39	329.81
ZNF281	HUNK	0.24	0.18	0.31	0	-0.01	-0.13	0.11	0.77	329.81
ESRRB	HYAL1	0.8	0.66	0.95	0	0.13	0.05	0.21	0	330.13
TP53	IGFBP2	0.64	0.47	0.81	0	-0.52	-1.22	0.18	0.22	609.33
ZSCAN22	IL17RB	0.31	0.28	0.33	0	0.08	0.01	0.15	0	547.83
HNF4A	IL17RB	0.28	0.27	0.3	0	0.12	0.07	0.17	0	547.83
ZNF619	IL17RB	0.55	0.52	0.57	0	0.01	-0.07	0.09	0.33	547.83
STAT1	IL2RB	1.08	0.9	1.26	0	0.09	-0.16	0.33	0.03	559.25

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TP53	INCA1	0.43	0.3	0.56	0	0	-0.03	0.03	0.71	443.87
ETV6	INCA1	0.33	0.21	0.45	0	0	-0.03	0.03	0.57	443.87
MTF1	INCA1	-0.04	-0.16	0.09	0.33	-0.35	-0.38	-0.32	0	443.87
ZSCAN5A	INPP5J	0.76	0.67	0.85	0	-0.26	-0.56	0.04	0.15	397.03
CREB3L4	ITGAE	0.68	0.42	0.95	0	-1.73	-3.21	-0.26	0.05	387.33
MBNL2	KATNAL2	0.52	0.47	0.57	0	0.14	0.04	0.25	0	217.84
ZNF84	KATNAL2	0	-0.05	0.06	0.88	-0.21	-0.31	-0.11	0	217.84
ZIC4	KCNK9	0.47	0.39	0.55	0	-0.07	-0.51	0.38	0.72	32.43
KLF3	KHK	0.28	0.15	0.41	0	0	-0.09	0.08	0.88	379.17
ZNF419	KLHDC9	0.46	0.43	0.5	0	0.01	-0.06	0.07	0.75	160.84
ZNF554	KLHDC9	0.56	0.52	0.59	0	0.05	-0.01	0.11	0.02	160.84
MYCL	KNG1	0.85	0.45	1.25	0.03	0.07	-0.16	0.31	0.61	247.15
ZNF444	LBX2	0.25	0.23	0.26	0	-0.01	-0.05	0.04	0.47	357.93
MZF1	LBX2	1	0.99	1.01	0	0	-0.05	0.04	0.61	357.93
HIC2	LBX2	0.14	0.13	0.15	0	-0.02	-0.06	0.02	0.12	357.93
ZNF574	LYPD5	0.84	0.68	1.01	0	-0.57	-1.61	0.47	0.37	97.14
KLF4	MGLL	0.29	-0.01	0.58	0.1	-0.49	-0.79	-0.19	0.01	386.5
ZNF574	MGLL	0.49	0.3	0.68	0	-0.43	-0.85	-0.01	0.09	386.5
ZSCAN22	MPPED1	2.17	0.9	3.44	0.05	-0.55	-1.66	0.56	0.43	42.99
RREB1	MSH5-SAPCD1	0.34	0.23	0.45	0	0.03	-0.08	0.13	0.1	104.53
ETV7	NKG7	0.56	0.52	0.59	0	0.04	-0.1	0.17	0.24	471.26
ZNF282	NKG7	0.43	0.4	0.47	0	-0.01	-0.13	0.1	0.82	471.26
ZNF555	NKX2-1	1.12	0.86	1.38	0	0.09	-0.6	0.77	0.83	174.3
ZBTB7A	NPBWR1	0.84	0.47	1.2	0.01	-0.13	-0.42	0.15	0.46	92.97
ZNF675	NPR3	0.64	0.53	0.75	0	-0.03	-0.5	0.44	0.91	349.53
ZBTB22	NRL	0.13	0.09	0.17	0	-0.01	-0.06	0.05	0.72	84.19
MESP1	NRL	0.13	0.09	0.17	0	0.02	-0.04	0.08	0.28	84.19
ZNF768	NRL	0.48	0.44	0.52	0	0	-0.06	0.06	0.9	84.19
POU5F1	NRL	0.4	0.36	0.43	0	-0.01	-0.07	0.05	0.58	84.19
ZNF85	OPRK1	1.17	1.07	1.28	0	-0.28	-0.7	0.15	0.31	307.76
PROP1	OR11L1	0.33	0.15	0.52	0.03	-0.38	-0.88	0.13	0.24	73.21
FERD3L	OR11L1	1.77	1.17	2.36	0	-0.04	-0.48	0.39	0.87	73.21
ZNF177	OR2AG2	0.76	0.53	0.98	0	-0.29	-0.77	0.19	0.32	73.11
ZNF232	OR51B5	0.97	0.95	0.99	0	-0.31	-0.5	-0.12	0.01	97.78
ZNF774	OR51E2	0.33	0.3	0.36	0	-0.01	-0.09	0.07	0.74	33.6
ZFP69	OR51E2	1.07	1.03	1.11	0	0.01	-0.07	0.08	0.83	33.6
ZNF7	OR8B12	0.36	0.17	0.56	0	0.01	-0.13	0.15	0.86	113.26
ZNF614	OR8B12	0.78	0.59	0.97	0	0.03	-0.11	0.17	0.69	113.26
ZNF334	OR8B3	0.98	0.84	1.11	0	0.14	-0.29	0.58	0.6	241.06
ZNF3	OR8D1	0.94	0.9	0.99	0	-0.07	-0.45	0.32	0.78	221.93
TCF24	OR8K3	1.04	0.44	1.63	0.05	-1.43	-2.15	-0.72	0.01	31.9
ZNF740	PCLO	0.75	0.62	0.87	0	0.05	-0.09	0.19	0.48	2473.74
HIC2	PCSK1N	0.78	0.68	0.88	0	-0.02	-0.07	0.04	0.02	543.24
FOXI2	PKHD1	0.74	0.4	1.09	0	0.02	-0.13	0.16	0.86	3625.56
ZSCAN22	PODXL	0.47	0.25	0.69	0	0	-0.01	0.02	0.17	1075.02
SP2	PODXL	0.31	0.09	0.54	0	0	-0.02	0.02	0.21	1075.02
ZNF398	PRAMEF20	0.71	0.49	0.92	0	-0.02	-0.15	0.12	0.84	93.63

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DRGX	PRAMEF20	0.48	0.02	0.94	0.09	-0.16	-0.28	-0.04	0.03	93.63
GMEB2	PROSER2	0.69	0.53	0.85	0	0.04	-0.2	0.28	0.77	1745.19
ZSCAN30	PTCH1	0.91	0.83	0.99	0	0.21	-0.21	0.64	0.41	1102.54
ZNF202	RBAK-RBAKDN	0.38	0.31	0.45	0	-0.03	-0.18	0.12	0.44	74.31
ZNF789	RBAK-RBAKDN	0.5	0.43	0.57	0	-0.02	-0.17	0.13	0.63	74.31
HOXB2	REM2	0.64	0.41	0.87	0	0.1	-0.13	0.33	0.36	60.6
XBP1	RGP1	0.56	0.31	0.81	0	0.02	0.01	0.04	0	1086.75
ZNF30	RGP1	0.07	-0.18	0.32	0.28	-0.2	-0.22	-0.19	0	1086.75
ZNF263	RGP1	0.29	0.04	0.55	0	0	-0.02	0.01	0.35	1086.75
HEY1	RMI2	0.87	0.68	1.05	0	0.34	0.29	0.39	0	70.65
ZNF71	RNF5	0.22	0.19	0.26	0	0.01	-0.05	0.07	0.39	345.54
ZNF324B	RNF5	0.15	0.12	0.18	0	0.01	-0.04	0.07	0.32	345.54
ZNF785	RNF5	0.32	0.28	0.35	0	0.02	-0.04	0.08	0.18	345.54
ZKSCAN3	RNF5	0.23	0.2	0.27	0	0	-0.06	0.05	0.8	345.54
E2F7	RNF5	0.26	0.22	0.3	0	-0.02	-0.08	0.04	0.18	345.54
TFAP2A	RPH3AL	0.89	0.81	0.97	0	-0.01	-0.34	0.33	0.97	641.74
ZNF552	SCIMP	0.04	-0.01	0.08	0.09	-0.11	-0.19	-0.03	0	206.38
PRDM4	SCIMP	0.29	0.24	0.33	0	0	-0.08	0.08	0.9	206.38
ZNF708	SCIMP	0.4	0.36	0.45	0	0	-0.08	0.08	0.93	206.38
IRF2	SERPINA6	0.35	0.18	0.52	0	-0.14	-0.37	0.08	0.29	65.83
POU4F3	SERPINA6	0.45	0.27	0.62	0	-0.01	-0.19	0.17	0.95	65.83
ATOH8	SGPP2	0.37	0.27	0.48	0	0.02	-0.11	0.15	0.53	395.19
ETV6	SGPP2	0.66	0.53	0.78	0	-0.09	-0.21	0.04	0.02	395.19
CTCFL	SGPP2	0.33	0.24	0.41	0	-0.5	-0.62	-0.38	0	395.19
ATF6B	SGSM2	0.56	0.32	0.79	0	0	-0.02	0.03	0.41	3235.36
CIC	SGSM2	0.3	0.09	0.51	0	0	-0.02	0.03	0.55	3235.36
ZNF7	SIAH3	0.49	0.33	0.65	0	-0.33	-0.64	-0.02	0.07	60.91
PLAG1	SLC16A13	1.16	1.1	1.22	0	0.35	-0.15	0.86	0.25	508.92
KLF16	SLC16A13	0.99	0.94	1.05	0	0.15	-0.49	0.79	0.7	508.92
KLF17	SLC25A47	0.58	0.48	0.67	0	0.21	-0.01	0.42	0.11	52.39
ZNF324B	SLC26A1	0.35	0.31	0.39	0	0	-0.01	0.02	0.09	3526.65
ZNF341	SLC26A1	0.31	0.27	0.36	0	0	-0.01	0.02	0.1	3526.65
ZNF692	SLC26A1	0.33	0.29	0.37	0	0.07	0.05	0.08	0	3526.65
ZNF792	SLC27A6	0.72	0.69	0.76	0	0.04	-0.01	0.08	0.17	99.26
ZNF614	SLC27A6	0.15	0.11	0.18	0	-0.01	-0.05	0.03	0.76	99.26
ZNF860	SLC27A6	0.11	0.08	0.15	0	-0.01	-0.06	0.03	0.67	99.26
ZSCAN22	SLC9A3	0.8	0.15	1.44	0.04	-0.33	-0.79	0.12	0.23	3449.62
BPTF	SNTB1	0.43	0.41	0.44	0	0.05	-0.17	0.26	0.13	528.13
ZFP64	SNTB1	0.32	0.3	0.33	0	0.02	-0.17	0.21	0.52	528.13
KLF3	SOX30	0.82	0.58	1.06	0	0.21	-0.09	0.5	0.27	132.94
ZNF141	SPAG8	0.46	0.34	0.59	0	0.03	-0.07	0.13	0.12	103.97
PPARD	SPAG8	0.6	0.48	0.72	0	0.04	-0.05	0.14	0.03	103.97
CEBPA	SPATS1	0.62	-0.02	1.26	0.11	-0.57	-1.07	-0.08	0.06	61.74
ZNF574	SPPL2C	0.81	0.51	1.11	0.01	-0.27	-0.95	0.41	0.53	50.69
ZNF394	STOML1	0.03	-0.1	0.17	0.47	-0.26	-0.28	-0.24	0	305.06
ZBTB48	STOML1	0.05	-0.09	0.18	0.31	-0.44	-0.46	-0.42	0	305.06
ZNF263	STOML1	0.65	0.51	0.78	0	0.32	0.3	0.35	0	305.06

Citation: Feng Xu, Zuheng Wang, Dian Fu, Xiuquan Shi, Jie Huang, Yuhao Chen, Jianping Da, Tingling Zhang, Jingping Ge, Xiaofeng Xu, Wen Cheng. Multi-Omics Analyses Revealed Transcriptional Regulators associated with Immune Checkpoint Inhibitor Treatment in Advanced Bladder Cancer. Journal of Biotechnology and Biomedicine. 6 (2023): 49-66.

FOXA3	TCEAL5	0.99	0.87	1.11	0	-0.05	-0.64	0.54	0.89	107.81
ZNF487	THEM5	0.01	-0.11	0.13	0.76	-0.41	-0.7	-0.12	0.02	881.73
ZNF93	THEM5	0.41	0.29	0.54	0	-0.23	-0.5	0.05	0.17	881.73
ZNF610	TMEM107	0.51	0.47	0.54	0	0	-0.02	0.03	0.63	177.42
ZNF524	TMEM107	0.38	0.34	0.42	0	0.19	0.17	0.22	0	177.42
RFX5	TMEM107	0.43	0.39	0.47	0	0.01	-0.02	0.03	0.31	177.42
RBPJL	TMEM176A	0.47	0.16	0.78	0.01	-0.34	-0.6	-0.08	0.03	1285.38
NFIB	TMEM176B	0.39	0.16	0.63	0	0.04	-0.03	0.11	0.21	1399.49
RBPJL	TMEM176B	0.56	0.31	0.8	0	-0.23	-0.29	-0.17	0	1399.49
ZNF785	TMEM213	0.48	0.27	0.7	0	-0.4	-0.55	-0.26	0	102.97
ZNF524	TMEM256-PLSCR3	0.68	0.56	0.8	0	0.05	-0.16	0.26	0.45	340.76
ZNF276	TNFRSF11B	0.58	0.46	0.69	0	-0.2	-0.57	0.17	0.37	214.79
ZNF776	TNFRSF11B	0.03	-0.09	0.16	0.11	-1.55	-1.9	-1.2	0	214.79
GLI4	TREX2	0.83	0.8	0.86	0	0.03	-0.17	0.24	0.44	163.83
HOXB2	TREX2	0.45	0.42	0.49	0	0.07	-0.14	0.27	0.14	163.83
DMRT3	TRIM42	0.82	0.72	0.91	0	0.02	-0.1	0.13	0.41	87.25
PAX3	TRIM42	0.27	0.16	0.37	0	0.02	-0.09	0.14	0.25	87.25
ZIC4	TRIM71	0.45	0.43	0.46	0	0.17	0.01	0.33	0.01	142.93
ZNF692	TRIM71	0.43	0.42	0.45	0	-0.04	-0.22	0.14	0.62	142.93
TFAP2C	TSPAN10	0.88	0.74	1.01	0	0.54	0.36	0.72	0	209.56
ETV4	TSSK6	0.29	0.1	0.49	0	0.01	-0.03	0.05	0.1	112.44
ELK1	TSSK6	0.3	0.1	0.49	0	0.01	-0.03	0.05	0.08	112.44
IRX3	TSSK6	0.49	0.3	0.68	0	0.01	-0.03	0.04	0.21	112.44
VSX2	UCP1	0.78	0.64	0.91	0	0	-0.03	0.03	0.71	38.42
ZNF550	UCP1	0.04	-0.09	0.17	0.44	-0.35	-0.38	-0.32	0	38.42
MYNN	UCP1	0.01	-0.11	0.14	0.75	-0.29	-0.32	-0.26	0	38.42
HNF1A	UGT2A3	0.8	0.7	0.9	0	0.11	-0.28	0.49	0.65	183.73
TFAP2C	USP43	0.74	0.69	0.79	0	-0.1	-0.34	0.14	0.5	756.62
RFX5	VAMP2	0.8	0.77	0.82	0	0.05	0.01	0.1	0	514.01
KLF7	VAMP2	0.57	0.54	0.6	0	0.01	-0.04	0.05	0.52	514.01
MECP2	VAMP2	0.3	0.27	0.33	0	0	-0.05	0.04	0.69	514.01
ZNF343	VAMP2	0.18	0.16	0.21	0	-0.01	-0.06	0.04	0.29	514.01
MLX	VEGF	0.86	0.66	1.07	0	0.01	-0.03	0.05	0.43	191.12
TCFL5	VEGF	1.47	1.26	1.69	0	0.01	-0.03	0.05	0.36	191.12
ZNF610	VEGF	0.37	0.16	0.58	0	-0.01	-0.05	0.03	0.4	191.12
ZNF485	VEGF	0.8	0.52	1.08	0	0	-0.04	0.04	0.97	191.12
VAX2	VSIG8	0.54	0.34	0.74	0	-0.03	-0.08	0.02	0.03	749.53
CDX2	VSIG8	0.32	0.12	0.51	0	-0.1	-0.16	-0.05	0	749.53
PATZ1	WNK4	0.66	0.56	0.76	0	-0.07	-0.37	0.22	0.68	754.94
MEIS3	ZMYND10	0.73	0.61	0.85	0	0.18	0.02	0.34	0.06	256.22
T	ZMYND10	0.46	0.26	0.66	0	-0.01	-0.21	0.2	0.97	256.22
ZNF785	ZNF277	0.4	0.22	0.58	0	0	-0.03	0.02	0.82	1408.62
KLF3	ZNF277	0.03	-0.15	0.21	0.54	-0.35	-0.38	-0.32	0	1408.62
ELK1	ZNF277	0.36	0.18	0.54	0	0	-0.02	0.03	0.57	1408.62
ETV6	ZNF277	0.37	0.18	0.55	0	0.02	-0.01	0.05	0	1408.62

Supplementary Table 2: Gene ontology pathway analysis.

pathway	source	p-value	q-value	external_id	members_input_overlap	members_input_overlap_geneids	size	effective size
Proximal tubule transport	Wikipathways	0.002	0.133	WP4917	SLC34A1; ATP6V0A4; SLC22A2	6582; 6569; 50617	57	57
Norepinephrine Neurotransmitter Release Cycle	Reactome	0.003	0.133	R-HSA-181430	VAMP2; SLC22A2	6844; 6582	18	18
SLC-mediated transmembrane transport	Reactome	0.006	0.133	R-HSA-425407	SLC34A1; SLC26A1; SLC27A6; SLC39A5; SLC22A2	283375; 28965; 10861; 6569; 6582	243	243
Surfactant metabolism	Reactome	0.007	0.133	R-HSA-5683826	SLC34A1; SFTPD	6569; 6441	26	26
TNFs bind their physiological receptors	Reactome	0.008	0.133	R-HSA-5669034	FASLG; TNFRSF11B	4982; 356	29	29
IL2 signaling events mediated by STAT5	PID	0.009	0.133	il2_stat5pathway	IL2RB; FASLG	356; 3560	30	30
Transport of small molecules	Reactome	0.01	0.133	R-HSA-382551	ATP6V0A4; RNF5; AMN; SLC34A1; SLC26A1; SLC22A2; SLC27A6; SLC39A5	6048; 6569; 28965; 10861; 50617; 283375; 6582; 81693	641	641
SNARE interactions in vesicular transport - Homo sapiens (human)	KEGG	0.01	0.133	path:hsa04130	VAMP2; BNIP1	662; 6844	33	33
Nicotine addiction - Homo sapiens (human)	KEGG	0.015	0.156	path:hsa05033	GABRG3; GRIN3B	2567; 116444	40	40
il-2 receptor beta chain in t cell activation	BioCarta	0.021	0.156	il2rbpathway	IL2RB; FASLG	356; 3560	48	48
Nephrotic syndrome	Wikipathways	0.021	0.156	WP4758	PODXL; NPHS2	5420; 7827	48	48
Neuroactive ligand-receptor interaction - Homo sapiens (human)	KEGG	0.022	0.156	path:hsa04080	GABRG3; OPRK1; GRIN3B; OPRL1; DRD5	116444; 2567; 1816; 4986; 4987	341	341
Retrograde transport at the Trans-Golgi-Network	Reactome	0.022	0.156	R-HSA-6811440	ARFRP1; RGP1	9827; 10139	49	49
TNFR2 non-canonical NF-kB pathway	Reactome	0.023	0.156	R-HSA-5668541	FASLG; TNFRSF11B	4982; 356	50	50
Neurotransmitter release cycle	Reactome	0.023	0.156	R-HSA-112310	VAMP2; SLC22A2	6582; 6844	50	50
keratinocyte differentiation	BioCarta	0.026	0.163	keratinocy tepathway	HOXA7; FASLG	3204; 356	53	53
Hepatitis C and Hepatocellular Carcinoma	Wikipathways	0.028	0.166	WP3646	PODXL; FASLG	5420; 356	56	56
Transmission across Chemical Synapses	Reactome	0.029	0.166	R-HSA-112315	GABRG3; VAMP2; GRIN3B; SLC22A2	6844; 2567; 116444; 6582	249	249

IL12-mediated signaling events	PID	0.035	0.189	il12_2pathway	IL2RB; FASLG	356; 3560	63	63
IL-18 signaling pathway	Wikipathways	0.039	0.194	WP4754	FASLG; STOML1; TNFRSF11B; SNTB1	6641; 4982; 9399; 356	274	274
Downstream signaling in naïve CD8+ T cells	PID	0.04	0.194	cd8tcrdownstreampathway	IL2RB; FASLG	356; 3560	68	68
Hedgehog	INOH	0.045	0.194	None	CDKN2B; PTCH1	5727; 1030	72	72
Primary focal segmental glomerulosclerosis (FSGS)	Wikipathways	0.046	0.194	WP2572	PODXL; NPHS2	7827; 5420	73	73
PPAR signaling pathway - Homo sapiens (human)	KEGG	0.047	0.194	path:hsa03320	SLC27A6; SCD5	28965; 79966	76	74
Peptide GPCRs	Wikipathways	0.048	0.194	WP24	OPRK1; OPRL1	4987; 4986	75	75
Cytokine-cytokine receptor interaction - Homo sapiens (human)	KEGG	0.049	0.194	path:hsa04060	IL2RB; FASLG; IL17RB; TNFRSF11B	4982; 55540; 356; 3560	295	295

Supplementary Table 3: Top 10 ranked master transcriptional regulators between durable clinic benefit (DCB) group and non-DCB group treated with immune checkpoint inhibitor (Atezolizumab) in advanced bladder cancer.

Rank	Gene	Dysregulation degree	Main functions in cancer	PMID
1	PATZ1	3876.64541	Inhibits proliferation and induces apoptosis	35509848
2	HIC2	1772.141003	a transcription repressor,	36650953
3	FO XK1	1739.29726	promoting the malignant behavior	32175400
4	ZNF324B	1685.109337	unknown	
5	PAX7	1607.067305	transcription factor of the forkhead family (FKHR) are associated with alveolar rhabdomyosarcomas.	9973247
6	ZNF785	1593.57106	unknown	
7	ELK1	1444.439546	transcription factor, promotes cancer progression	34966781
8	ERF	1254.483741	an ETS transcriptional repressor, acts as a tumor-suppressor gene	28515055
9	ZNF341	1184.604744	a transcription factor, low expression is related to radiosensitivity	34504527
10	PAX2	1145.368368	induces formation of vascular-like structures, promotes cancer progression	35601066