

1 **Identification of a novel inflamed tumor microenvironment signature as a predictive**
2 **biomarker of bacillus Calmette-Guerin immunotherapy in non-muscle invasive bladder**
3 **cancer**

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5 Jeffrey S. Damrauer^{a#}, Kyle R. Roell^{a,b#}, Markia A. Smith^{a,c#}, Xuezheng Sun^{a,b}, Erin L. Kirk^{a,b},
6 Katherine A. Hoadley^{a,d}, Halei C. Benefield^b, Gopakumar Iyer^{e,f}, David B. Solit^{e,f,g}, Matthew I.
7 Milowsky^{a,h}, William Y. Kim^{a,h}, Matthew E. Nielsenⁱ, Sara E. Wobker^{a,i,j}, Guido Dalbagni^{k,l}, Hikmat
8 A. Al-Ahmadie^m, Andrew F. Olshan^{a,b}, Bernard H. Bochner^{k,l}, Helena Furbergⁿ, Melissa A.
9 Troester^{a,e,j^}, Eugene J. Pietzak^{k,l^*}

10

- 11 a. University of North Carolina Lineberger Comprehensive Cancer Center
12 b. Department of Epidemiology, University of North Carolina at Chapel Hill
13 c. Department of Pathology and Laboratory Medicine, University of North Carolina at
14 Chapel Hill
15 d. Department of Genetics, University of North Carolina at Chapel Hill
16 e. Department of Medicine (Genitourinary Oncology Service), Memorial Sloan Kettering
17 Cancer Center
18 f. Department of Medicine, Weill Cornell Medicine
19 g. Human Oncology and Pathogenesis Program, Memorial Sloan Kettering Cancer Center
20 h. Department of Medicine, University of North Carolina at Chapel Hill
21 i. Department of Urology, University of North Carolina at Chapel Hill
22 j. Department of Pathology and Laboratory Medicine, University of North Carolina at
23 Chapel Hill
24 k. Department of Surgery (Urology Service), Memorial Sloan Kettering Cancer Center
25 l. Department of Urology, Weill Cornell Medicine
26 m. Department of Pathology, Memorial Sloan Kettering Cancer Center
27 n. Department of Epidemiology and Biostatistics, Memorial Sloan Kettering Cancer Center

28

29 # Authors Contributed Equally

30 ^ Co-senior Authors

31

32 ***Corresponding author:**

33 Eugene Pietzak, MD

34 Sidney Kimmel Center for Prostate and Urologic Cancers

35 Memorial Sloan Kettering Cancer Center
36 353 East 68th Street, New York, NY 10065
37 +1 646 422 4781
38 pietzake@mskcc.org

39

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83 **Translational Relevance Statement:** We performed RNA-based profiling by NanoString
84 nCounter on non-muscle invasive bladder cancer (NMIBC) clinical specimens and found that a
85 novel expression signature of an inflamed tumor microenvironment (TME), but not molecular
86 subtyping, was associated with improved recurrence-free survival after bacillus Calmette-Guerin
87 (BCG) immunotherapy. We further demonstrate that immune checkpoint gene expression was
88 not associated with higher recurrence rates after BCG. These findings were externally validated
89 in a large RNAseq dataset of NMIBC suggesting our immune signature could be a robust
90 predictive biomarker for BCG response and that an immunologically “cold” TME is a mechanism
91 of resistance to BCG. Our results also raise concerns about treatment strategies combining
92 BCG and immune checkpoint blockade in NMIBC and instead support approaches focused on
93 modulating the TME. Our integrated transcriptomic and panel sequencing found FGFR3
94 overexpression and mutations to be associated with an “cold” TME, further supporting
95 investigations into FGFR inhibitors for NMIBC.

96

97 **ABSTRACT**

98 **Purpose:** Improved risk stratification and predictive biomarkers of treatment response are
99 needed for non-muscle-invasive bladder cancer (NMIBC). Here we assessed the clinical utility
100 of targeted RNA and DNA molecular profiling in NMIBC.

101 **Experimental Design:** Gene expression in NMIBC samples was profiled by NanoString
102 nCounter, a RNA quantification platform, from two independent cohorts (n = 28, n = 50);
103 targeted panel sequencing was performed in a subgroup (n = 50). Gene signatures were
104 externally validated using two RNAseq datasets of NMIBC tumors (n = 438, n=73). Established
105 molecular subtype classifiers and novel gene expression signatures were assessed for
106 associations with clinicopathologic characteristics, somatic tumor mutations, and treatment
107 outcomes.

108 **Results:** Molecular subtypes distinguished between low-grade Ta tumors with FGFR3
109 mutations and overexpression (UROMOL-class 1) and tumors with more aggressive
110 clinicopathologic characteristics (UROMOL-classes 2 and 3), which were significantly enriched
111 with TERT promoter mutations. However, UROMOL subclasses were not associated with
112 recurrence after bacillus Calmette-Guerin (BCG) immunotherapy in two independent cohorts. In
113 contrast, a novel expression signature of an inflamed tumor microenvironment (TME) was
114 associated with improved recurrence-free survival after BCG. Expression of immune checkpoint
115 genes (PD-L1/PD-1/CTLA-4) was associated with an inflamed TME, but not with higher
116 recurrence rates after BCG. FGFR3 mutations and overexpression were both associated with
117 low immune signatures.

118 **Conclusions:** Assessment of the immune TME, rather than molecular subtypes, is a promising
119 predictive biomarker of BCG response. Modulating the TME in an immunologically “cold” tumor

120 warrants further investigation. Integrated transcriptomic and exome sequencing should improve
121 treatment selection in NMIBC.

122 **INTRODUCTION**

123 Current risk stratification for non-muscle-invasive bladder cancer (NMIBC) is inadequate and
124 relies on clinicopathologic features, with limited ability to accurately predict which tumors are
125 most likely to recur and/or progress to muscle-invasive disease(1). Hence, patients with NMIBC
126 undergo intensive surveillance with frequent, invasive cystoscopies, rendering bladder cancer
127 among the most expensive cancers to manage(2). Thus, there is a critical need for prognostic
128 and predictive biomarkers in NMIBC. Most urgently needed are molecular biomarkers that are
129 predictive for response to immunotherapies. Intravesical bacillus Calmette-Guérin (BCG)
130 immunotherapy has been a standard of care in NMIBC for over 40 years, but there exists no
131 biomarker to prioritize patients for optimal treatment in this era of recurrent BCG shortages. The
132 anti-PD-1 immune checkpoint inhibitor pembrolizumab was recently FDA-approved for BCG-
133 unresponsive NMIBC based on a single-arm phase II trial demonstrating an initial response rate
134 of 40% at 3 months. However, only 20% of patients remained disease-free at 12 months(3) and
135 PD-L1 expression by immunohistochemical (IHC) staining was not associated with response,
136 leaving little guidance on who is likely to benefit(3).

137
138 A potential approach to risk-stratify NMIBC patients and predict treatment responses is
139 transcriptomic profiling. This strategy has led to the identification and validation of luminal and
140 basal-like molecular subtypes of bladder cancer, similar to the luminal and basal-like breast
141 cancer subtypes(4,5). For muscle-invasive bladder cancer (MIBC), multiple retrospective
142 studies indicate that molecular subtyping is prognostic and predictive of response to
143 chemotherapy and immunotherapy, and a consensus molecular classification has recently been
144 established(4-6). In contrast, while early unsupervised microarray analyses of gene expression
145 had focused on NMIBC, there have been only a few recent gene expression efforts in NMIBC
146 using more contemporary expression profiling platforms(7-9). The largest of these was
147 UROMOL, a prospective multicenter European study that established three major molecular

148 subtypes of NMIBC(7). However, the 460 NMIBC specimens analyzed represent only 34% of
149 the initial 1,372 fresh-frozen specimens obtained, after applying strict criteria on RNA quality
150 and carcinoma cell percentage. Inclusion of only the highest-quality tumor tissue increases
151 confidence in the resulting biologic insights but can limit the clinical translation of these findings.
152 Most NMIBC specimens available for clinical testing are archival, formalin-fixed, paraffin-
153 embedded (FFPE) tissues, from which RNA can be difficult to extract and is generally of poor
154 quality(10). Thus, a clinically feasible method of multiplex gene expression profiling in NMIBC
155 must overcome the limited amount of available tissue, low cellularity, and poor RNA quality
156 inherent in the majority of these specimens.

157

158 RNA counting methods, such as nCounter (NanoString Technologies, Inc), that do not require
159 enzymatic reactions allow for FFPE samples to be used in expression profiling(11), but the
160 feasibility and value of these approaches have not been robustly assessed in NMIBC. Thus, we
161 sought to evaluate gene expression signatures determined by RNA counting to identify
162 associations with clinicopathologic characteristics, assess prognostic and predictive
163 significance, and evaluate the added value of gene expression data over established DNA
164 mutation sequencing methods in NMIBC.

165

166

167 **MATERIALS AND METHODS**

168 *Patient samples*

169 Clinical NMIBC specimens were collected from two sites, University of North Carolina at Chapel
170 Hill (UNC) and Memorial Sloan Kettering Cancer Center (MSK). All specimens from UNC and
171 MSK were archival FFPE NMIBC tissue samples procured by transurethral resection of bladder
172 tumor (TURBT) from treatment naïve, newly diagnosed patients and were reviewed by a
173 genitourinary pathologist (SW, HAA) to confirm grade, stage, and urothelial histology.

174 Specimens in the UNC cohort were collected through the institutional review board-approved
175 UNC Health Registry/Cancer Survivor Cohort Study, which prospectively ascertained persons
176 with newly diagnosed with cancer seen in the UNC Hospital system. Enrolled patients gave
177 informed consent for use of biospecimens, clinical data, and questionnaire data for approved
178 research. Specimens in the MSK cohort were collected through an institutional review board-
179 approved sequencing effort, in which tumor specimens and matched germline DNA were
180 profiled by targeted panel sequencing using a panel of 341 (later updated to 410) cancer-
181 associated genes within a CLIA-certified laboratory as previously described by Pietzak et al and
182 available on cbiportal.org (http://www.cbiportal.org/study/summary?id=blca_nmibc_2017)(12).
183 We included all tumors with available FFPE for RNA extraction and gene expression analysis
184 from Pietzak et al. Treatment and management was at the discretion of the treating urologic
185 oncologist. Patients treated with BCG immunotherapy received 6 weekly full doses of TICE
186 BCG with only 8% receiving additional maintenance BCG. Patients were then followed every 3
187 months with cystoscopy and urine cytology for the first year, then every 3–6 months. All HGT1
188 tumors had re-staging TURBTs with confirmation of uninvolved detrusor muscle at initial
189 diagnosis. Recurrence was defined as histologically proven cancer on biopsy or TURBT.

190

191

192

193 *Expression analysis*

194 RNA extraction and gene expression methods have been previously published(13). Briefly, RNA
195 was isolated from a 1-mm FFPE core or two 10- μ m unstained FFPE slides using the Qiagen
196 RNeasy FFPE Kit and protocol (cat. # 73504). RNA was quantitated using a ThermoScientific
197 NanoDrop 2000 Spectrophotometer (cat. # ND-2000) and Agilent 4200 TapeStation. After
198 excluding samples with low concentration or low percentages of RNA molecules >300
199 nucleotides long, the remaining 90% of samples were processed by the UNC Translational

200 Genomics Laboratory using the NanoString nCounter platform. Samples were run on a custom
201 codeset that included gene sets for the Hedegaard classifier(7), for p53 pathway defects(14),
202 and for 52 immune-related genes(15). Samples were randomized to batch and two Stratagene
203 Universal Human Reference RNA samples were included to assess batch variability. Batch
204 variability was low, with correlations between reference standards exceeding 0.98.

205
206 After counting RNA, QC procedures eliminated samples with low tumor gene expression. Four
207 steps were used to identify such samples. (1) Expression values below the limit of detection,
208 defined as the average of a sample's negative probe mean, was set to zero. (2) Correlations
209 between 6 housekeeping genes (*ACTB*, *GUSB*, *KU70/XRCC6*, *NAGA*, *PGK1*, *RPS10*) were
210 evaluated across samples. Pairwise correlations were all above 80%, with an overall mean
211 correlation of 87.2%, and therefore all 6 genes were retained in the dataset. (3) The relationship
212 between housekeeping gene missingness and endogenous gene missingness was assessed,
213 where missingness was defined as having an expression below the limit of detection.
214 Missingness in endogenous genes was correlated with housekeeping gene missingness.
215 Samples with greater than half of the endogenous genes marked as missing were removed
216 from further analysis. Principal component analysis (PCA) was then used to identify outliers. (4)
217 Finally, we used the R package Remove Unwanted Variation (RUV) to normalize the data(16).
218 Dimensions of unwanted variation (k) from 1 to 5 were assessed and k=1 was selected based
219 on sample PCA plots. Upon normalization, nearest-neighbors averaging imputation was applied
220 to all samples via the impute R package(16). Samples from MSK and UNC were assessed for
221 quality concurrently. When duplicate assays both passed QC, these were averaged to produce
222 a single expression vector for each patient.

223

224 *Gene expression signatures*

225 Genes for the BASE47 gene signature were as originally described by Damrauer et al(4). Gene
226 expression data were clustered and Prediction Analysis of Microarrays (PAM) was used to
227 select a parsimonious set of genes that represented the three UROMOL subclasses described
228 by Hedegaard et al.(7) For UROMOL subclasses, a total of 117 genes were included in the
229 classifier, though the same clusters were recapitulated with 110 genes. This smaller gene set
230 was used to cluster the samples. Classification by the p53 pathway defect expression signature
231 originally described in breast cancer(14) and recently shown to be associated with *TP53* status
232 in MIBC(17) was according to published methods(14,17). Finally, we developed a panel of
233 immune genes based on work previously described by Bindea et al.(15). As NMIBC specimens
234 either had high expression of all genes in the panel or low expression of all genes, the immune
235 score for a tumor was defined as the average of the counts of genes comprising the immune
236 signature panel. Tumors were grouped into tertiles by ranking tumors by gene scores and
237 identifying the 33rd and 66th percentile thresholds adapted from Miller et. al.(18). For
238 recurrence analyses, the top two tertiles were combined into a “high” immune signature score
239 and compared to the bottom tertile, considered a “low” immune score. Proliferation was
240 determined from the median expression of proliferation-related genes in Parker et. al.(19). Java
241 TreeView was used to visualize the data according to *TP53* status and immune score(11).

242

243 *Comparison of RNA-based classes with DNA mutations*

244 Targeted panel sequencing of MSK samples was performed using the MSK-IMPACT
245 assay(12,20,21). Coding and promoter mutations and indels, excluding silent mutations, were
246 considered. Mutation calls were not assessed for samples that failed MSK-IMPACT QC.
247 Specific genes were curated from Pietzak et al. based on genes mutated at >10% in
248 NMIBC(12). Missense mutations and amplifications in known oncogenes were deemed
249 significant if the variant was recurrent or if existing literature reported it as a functionally

250 validated activating alteration. Alterations in tumor suppressor and DNA damage repair genes
251 were deemed deleterious if truncating mutations (nonsense, frameshift indels), recurrent
252 missense mutations, or homozygous deletions. Recurrent missense mutations were defined as
253 those reported in either the Catalogue of Somatic Mutations in Cancer (COSMIC) or in the
254 cBioPortal for Cancer Genomics more than 10 times. Missense mutations reported by the MSK-
255 IMPACT bioinformatics pipeline, but not meeting our definition for recurrent alterations, were
256 reported as “novel missense mutations”/“missense mutations of unknown significance” and not
257 included in statistical analyses of clinicopathologic and recurrence associations.

258

259 *Validation Cohorts*

260 UROMOL Cohort: Publicly available RNAseq data from the UROMOL Project (Prediction of
261 bladder cancer disease course using risk scores that combine molecular and clinical risk
262 factors) were downloaded from the supplemental material section from Hedegaard et al(7). and
263 from www.medrxiv.org (<https://www.medrxiv.org/content/10.1101/2020.06.19.20054809v1>) to
264 determine BCG treatment history and recurrence outcomes. Only 19% of the UROMOL cohort
265 received BCG and maintenance BCG was infrequently given(7).

266

267 Northwestern Cohort: Publicly available RNAseq data from GSE154261 was downloaded from
268 Robertson et al(9). This cohort comprised of 73 primary T1 tumors treated at Northwestern
269 University with all patients receiving induction BCG, 64% receiving some maintenance BCG,
270 and 84% having a restaging TURBT before BCG(9).

271

272

273 *Statistical analyses*

274 Clustering of gene expression data was performed in R using the heatmap.plus R package(11).
275 Pearson distance measures and complete linkage were used to determine clusters. Java

276 TreeView was used to visualize the clustered data and generate heatmaps(11). The
277 compareGroups R package was used to generate descriptive tables and to determine the
278 statistical significance of associations among variables(11). Continuous variables were
279 assessed using t-tests or ANOVA, where appropriate, and categorical variables were assessed
280 using chi-squared tests or Fisher's exact tests. Cox regression modeling was used to determine
281 the association between gene signatures and recurrence after BCG. The Kaplan-Meier method
282 and log-rank test were used for estimations of recurrence-free survival (RFS).

283

284

285 **RESULTS**

286 *Patient and tumor characteristics*

287 RNA was extracted from tumor samples of 41 UNC patients and 68 MSK patients, of which 28
288 samples (68%) and 50 samples (74%), respectively, met QC criteria. In general, samples with
289 low (<30%) tumor cellularity more often failed QC (**Supplemental Table 1, Supplemental**
290 **Figure 1**). Patient demographics and clinicopathologic characteristics for profiled samples are
291 described in **Table 1**. Recurrence outcomes were available only for the MSK cohort.
292 Clinicopathologic characteristics were not associated with recurrence risk in this cohort
293 (**Supplemental Table 2**), consistent with our prior study(12). One patient was treated with
294 immediate cystectomy, so their tumor was analyzed but they were excluded from recurrence
295 analyses. Median follow-up among the MSK cohort was 39 months (IQR 13–66.2), with
296 recurrence occurring in 23 patients. Median follow-up among the 38 patients (76%) receiving
297 BCG immunotherapy was 38.2 months (IQR 10.7–66.6), with recurrence occurring in 18
298 patients.

299

300 *Molecular subtyping by RNA counting*

301 We first sought to determine molecular subtype using the BASE47 classifier developed to
302 separate MIBC into luminal and basal-like subtypes(4). We found limited discriminatory capacity
303 as only 4 NMIBC tumors were classified as basal-like, whereas the remainder were luminal. We
304 therefore focused our subsequent analyses on a gene expression classifier derived from the
305 large UROMOL cohort of NMIBC samples(7). Our combined NMIBC sample cohort was well-
306 distributed across the 3 previously described UROMOL subclasses (**Figure 1**) and clinical
307 characteristics differed across the groups (**Table 2**). UROMOL class 1, previously described as
308 luminal-like/well-differentiated, was enriched in low-grade Ta tumors that had greater tumor
309 cellularity due to their predominantly papillary architecture. As expected, these tumors had
310 significantly higher FGFR3 gene expression levels in both the UNC and MSK cohorts
311 (**Supplemental Figure 2**)(7). Consistent with UROMOL class 2 tumors being the most
312 aggressive subtype, they had the highest proliferation expression scores and expression of the
313 proto-oncogene forkhead box M1 (FOXM1) in both cohorts (**Supplemental Figure 3**).
314 UROMOL class 3, previously described as being an early basal-like subtype, was enriched in
315 high-grade T1 tumors.

316

317 There were very few progression events within the MSK cohort, preventing formal analysis, but
318 all 3 patients with progression to secondary MIBC had UROMOL class 2 tumors, consistent with
319 the original study(7). There was no correlation between UROMOL classes and RFS within the
320 entire MSK cohort or the subset of 38 patients treated with BCG (**Figure 2A–B**). To rule out the
321 possibility that this lack of association was the result of small sample size, we analyzed publicly
322 available data from the UROMOL study and found that class 2 was associated with worse RFS
323 in the UROMOL study (n=438), but subtype was not associated with recurrence rates in the
324 subset of patients treated with BCG (n=83) (**Figure 2C–D**). These results suggest that the
325 UROMOL classifier is prognostic of outcome but not predictive of BCG response.

326

327 *Immune gene expression and signatures*

328 As UROMOL classes were not predictive of recurrence following BCG treatment among either
329 the MSK or UROMOL cohorts, we sought to evaluate other RNA-based signatures as potential
330 predictive biomarkers of clinical benefit from BCG immunotherapy. Previous biomarker studies
331 have identified higher pretreatment T cell infiltration and an inflamed tumor microenvironment
332 (TME) as associated with improved response to systemic immunotherapies in various cancer
333 types, including bladder cancer(15,22,23). We evaluated the expression of known immune-
334 related genes to determine the degree of immune cell infiltration and inflammation within the
335 TME(15). NMIBC tumors could be broadly classified as having either low expression or higher
336 expression across all immune genes studied, with no groups emerging with more complex
337 patterns (i.e., innate immune response only) (**Figure 3A**). Therefore, the median expression of
338 all immune genes was used to calculate an immune score. Tumors were grouped into tertiles by
339 ranking tumors by gene scores and identifying the 33rd and 66th percentile thresholds adapted
340 from Miller et al.(18). These scores varied among UROMOL classes, with basal-like UROMOL
341 class 3 tumors having the highest expression of immune-related genes, whereas class 1 tumors
342 had the lowest (**Supplemental Figure 4**). Tumors with high immune scores (top two tertiles)
343 were associated with improved RFS in the MSK cohort (hazard ratio [HR]=0.33, 95%
344 confidence interval [CI] 0.14–0.78, $p=0.01$) as well as among the subset treated with BCG
345 (HR=0.23, 95% CI 0.09–0.59, $p=0.002$) (**Figure 3B–C, Supplemental Table 2**). We also sought
346 to externally validate the immune score using the UROMOL cohort (**Supplemental Figure 5**)
347 and found that higher immune scores were associated with improved RFS among the 83-patient
348 UROMOL subgroup treated with BCG (HR=0.5, 95% CI 0.24–1.00, $p=0.05$), but not among the
349 entire 438-patient UROMOL cohort (HR=0.96, 95% CI 0.70–1.33, $p=0.84$) (**Figure 3D–E**). No
350 statistically significant difference was seen for high immune score with improved RFS in the
351 Northwestern HGT1 cohort (HR=0.7, 95% CI 0.52-3.6 $p=0.4$) (**Supplemental Figure 6**).

352

353 We next sought to determine which individual immune-related genes were associated with
354 favorable outcomes following treatment with BCG. Correlates of remaining free from high-grade
355 recurrences >2 years after BCG treatment included high expression of HLA class I
356 histocompatibility antigens, granzyme B (GZMB), and inducible T cell co-stimulator (ICOS).
357 Surprisingly, we found high expression of the immune checkpoint genes PD-L1 and LAG-3 to
358 also be associated with remaining recurrence free after BCG (**Figure 4A**).

359

360 As this finding contradicts a prevailing hypothesis that immune checkpoint expression leads to
361 BCG treatment failure, and there is growing interest in anti-PD-L1/PD-1 immunotherapy for
362 patients with NMIBC, we further assessed the significance of expression of targetable immune
363 checkpoints (PD-L1, PD-1, and CTLA-4) within our cohort. Expression of these immune
364 inhibitory molecules was highest in UROMOL class 3 tumors and lowest in class 1 tumors
365 (**Supplemental Figure 7**). Higher expression of PD-L1 ($p=0.058$), PD-1 ($p<0.01$), and CTLA-
366 4($p<0.01$) in pretreatment NMIBC specimens was associated with lower recurrence rates after
367 BCG immunotherapy in our cohort, suggesting that these molecules correspond to increased
368 immune infiltration and are not associated with resistance to BCG (**Figure 4B, Supplemental**
369 **Figure 7**). Examining pretreatment expression levels of these immune checkpoint genes in the
370 83-patient BCG-treated UROMOL cohort found no association with RFS in BCG-treated
371 patients (**Figure 4C, Supplemental Figure 7**). Furthermore, expression of PD-L1, PD-1, and
372 CTLA-4 in pre-treatment specimens from the Northwestern HGT1 cohort were also not
373 associated with RFS after BCG (**Supplemental Figure 7**).

374

375 *Association between gene expression-based signatures and somatic DNA mutations*

376 As the UROMOL study inferred somatic mutations from RNA expression data alone(7), which
377 can lead to false-positive deleterious mutation calls(24), we evaluated correlations between
378 UROMOL subclasses and somatic DNA mutations using tumor-normal target panel data
379 available for the MSK cohort. *FGFR3* mutations were enriched among luminal class 1 tumors,
380 while infrequently altered in basal-like class 3 tumors (**Figure 1, Table 3**). Alterations in *ERBB2*,
381 a gene known to be mutually exclusive with *FGFR3*(12), were absent in class 1 tumors but seen
382 in 20% of class 2 tumors and 36% of class 3 tumors (p=0.02) (**Table 3**). Mutations in the
383 chromatin remodeling gene *TET2* were enriched in basal-like class 3 tumors (**Table 3**). Somatic
384 mutations in the *TERT* promoter were highly enriched in UROMOL class 2 tumors, found to be
385 present in all 20 samples (100%) (**Figure 1**). UROMOL class 2 tumors also had the highest
386 fraction of the genome altered (**Supplemental Figure 8**).

387
388 Alterations in *TP53* and cell cycle genes are among the most common genetic events in MIBC
389 and high-risk NMIBC, yet biomarker investigations into p53 status to guide clinical management
390 have reported conflicting results(12,25). We explored a mutant p53-pathway gene signature,
391 developed and validated in breast cancer(14), within our NMIBC cohort. This signature was
392 recently shown to be associated with *TP53* status in MIBC(17), but has little overlap with the
393 “p53-like” gene expression signature identified by Choi et al. in MIBC(6) (**Supplemental Figure**
394 **9**). The mutant p53-pathway gene signature was highest in class 2 tumors, followed by class 3
395 tumors; these two subclasses had the most *TP53* mutations (**Figure 1, Supplemental Figure**
396 **9**). RNA-based p53 pathway assessment detected defects in the p53 pathway more than *TP53*
397 DNA mutations alone and had fair correlation with DNA alterations in an expanded set of
398 commonly altered cell cycle pathway genes (*TP53, MDM2, RB1, CDKN2A, CDKN1A, CCND1*)
399 (**Supplemental Figure 10**). Neither p53 mutant-like gene expression nor *TP53* DNA alterations
400 were associated with recurrence in the entire MSK cohort or those treated with BCG.
401 Additionally, mutant p53-pathway gene signature tumors had high immune scores, whereas

402 p53-wild-type-like tumors had intermediate immune scores (**Supplemental Figure 10**). Higher
403 mutant p53-pathway gene signatures also correlated with increased expression of proliferation
404 and FOXM1 (**Supplemental Figure 11**).

405
406 Finally, as the pretreatment TME inferred by immune score was associated with BCG response,
407 we sought to determine whether there was a correlation between genomic alterations in
408 individual genes and immune score. We found that *FGFR3* mutations by targeted panel
409 sequencing and *FGFR3* overexpression by RNA profiling were both associated with lower
410 immune scores (**Supplemental Table 3, Supplemental Figure 2**). *FGFR3* mutations remained
411 significantly associated with low immune score even after adjusting for multiple comparisons.
412 Conversely, *ERBB2* alterations were enriched in immune-score-high tumors (38% v. 0 v. 0,
413 $p=0.001$) (**Supplemental Table 3**).

414

415 **DISCUSSION**

416 There are several barriers that must be overcome before molecular biomarkers identified within
417 the context of retrospective clinical studies can be successfully translated to clinical practice.
418 These include the limited quantity and often poor quality of tissue available for many patients
419 with NMIBC. Here, we show that gene expression profiling by RNA counting is feasible using
420 archival FFPE NMIBC clinical samples, despite relatively limited tumor tissue and low tumor
421 cellularity.

422

423 To build upon prior work with the NMIBC-specific UROMOL gene expression classifier(7), we
424 combined targeted panel sequencing of tumor-normal pairs along with transcriptomic profiling
425 and observed that *TERT* promoter mutations were significantly enriched in UROMOL class 2
426 (100%) and class 3 (82%) tumors. This novel observation demonstrates the potential molecular
427 insights that can be obtained by integrated RNA and DNA molecular profiling and supports the

428 development of urine-based screening and surveillance biomarkers that detect *TERT* promoter
429 mutations in urinary cell free DNA to identify aggressive bladder tumors earlier(26). We also
430 found UROMOL molecular subtypes correlated with expected clinicopathologic characteristics
431 but were not associated with BCG response. UROMOL subtypes can be recapitulated within
432 bladder cancer cell lines, suggesting that these molecular subtypes are tumor cell “intrinsic”
433 signatures(7). We hypothesize that molecular subtypes based on tumor cell phenotype,
434 independent of signals from infiltrating stromal and immune cells, provide important biologic
435 insights but are not likely to be robust predictive biomarkers for response to immunotherapy
436 such as BCG.

437
438 To develop a better predictive biomarker for BCG response, we assessed immune cell
439 infiltration and inflammation within the TME using a novel gene expression signature. We
440 observed that a high immune score was associated with improved RFS after BCG
441 immunotherapy in the MSK and UROMOL cohorts, demonstrating the importance of the
442 pretreatment TME in determining BCG response, consistent with the work of several other
443 groups(27,28). While BCG is a live, attenuated bacteria known to provoke an influx of innate
444 and adaptive immune cells within the bladder wall, the recruitment of these immune cells
445 appears unable to overcome an existing immunologically “cold” TME. Notably, no statistical
446 difference in RFS after BCG was seen within the Northwestern cohort for the immune score.
447 This may be due to it being a HGT1 only cohort compared to the MSK and UROMOL cohorts
448 which also included LGTa and HGTa tumors. It is also possible that the addition of maintenance
449 BCG may attenuate the negative effects of a “cold” pre-treatment TME, as 64% of the
450 Northwestern cohort received maintenance BCG compared to only a few patients within the
451 MSK and UROMOL cohorts. Maintenance BCG is recommended in multiple guidelines as it
452 reduces the risk of recurrence by 19% at 5 years compared to induction BCG alone(29,30), but
453 many patients do not receive it due to the persistent global BCG shortage and concerns over

454 treatment-related toxicity(29-31). If prospectively validated, the immune score may provide a
455 rational, biomarker-driven approach to selecting which patients would most benefit from
456 induction and maintenance BCG. This could be invaluable in helping to alleviate the global BCG
457 shortage.

458

459 Further investigation into the causes of a cold TME in NMIBC is needed, but our study suggests
460 that combination strategies that modulate the TME to promote antitumor immune cell
461 recruitment hold promise in NMIBC. Our finding that mutation and overexpression of *FGFR3* are
462 associated with a low immune score in NMIBC is consistent with data from studies of MIBC,
463 upper tract urothelial carcinoma, and mouse models of bladder cancer(23,32,33), providing
464 further support for investigations into the potential immunomodulating benefits of *FGFR3*-
465 targeted therapies. Interestingly, our data also suggest that *FGFR3* and *ERBB2* are not only
466 mutually exclusive in NMIBC but result in contrasting differences in the TME. Both *FGFR3* and
467 *ERBB2* are attractive “targets,” given the prevalence of these genomic alterations in patients
468 with NMIBC. As more targeted therapies with demonstrable activity in metastatic urothelial
469 cancer are evaluated in clinical trials for patients with NMIBC, these differences in the TME will
470 likely become even more relevant.

471

472 Our analysis also found that high gene expression of PD-L1, PD-1, and CTLA-4 was not
473 associated with worse BCG response in either the MSK, UROMOL, or Northwestern cohorts.
474 While this may seem contrary to the prevailing theory that PD-1/PD-L1 expression is a
475 mechanism of resistance to BCG(34,35), others have reported high PD-L1 gene expression to
476 be associated with favorable outcomes with intravesical therapy in T1 NMIBC(36). These
477 observations might be from differences in pre-analytical preparations or post-translational
478 modifications between gene level and protein level expression of PD-1/PD-L1(37). However,
479 IHC-based studies assessing the role of PD-1/PD-L1 expression in NMIBC by comparing BCG

480 “responders” to “non-responders” have had conflicting results, varying considerably depending
481 on the antibody and cut-off used, and studies supporting PD-1/PD-L1 as a mechanism of
482 resistance to BCG generally do not adjust for differences in important clinical factors that are
483 known to affect both BCG response and PD-L1 expression levels, such as concomitant CIS(34-
484 36,38). The first and most highly cited study suggesting PD-L1 expression may be a mechanism
485 of immune evasion to BCG was by Inman et al., where 11 of 16 patients with recurrent tumors
486 had associated post-BCG granulomata with strong PD-L1 expression by IHC staining(35).
487 However, the authors did not assess PD-L1 expression in post-BCG granulomata from patients
488 without recurrence, which is particularly relevant as post-BCG granulomata are associated with
489 a favorable response to BCG(39) and comprised of macrophages that often express PD-L1
490 even in healthy tissue(40). Pretreatment specimens from the prospective NMIBC cohort within
491 Inman et al. showed no difference in recurrence rates after BCG by PD-L1 expression (PD-L1+
492 30% [3/10] vs. PD-L1- 38% [13/34], p=0.2)(35). Taken together with our data, PD-1/PD-L1
493 expression is unlikely to mediate resistance to BCG immunotherapy in treatment-naïve NMIBC,
494 and high expression of PD-L1, PD-1, or CTLA-4 should not preclude treatment with BCG. Our
495 results raise concerns about ongoing trials combining BCG and immune checkpoint blockade,
496 as they indicate that an immunologically “cold” TME may be a shared mechanism of resistance
497 to both BCG and checkpoint inhibitors(23).

498
499 As p53 defects by IHC staining has been one of the most heavily investigated biomarkers in
500 NMIBC with numerous conflicting reports of significance(25), we sought to investigate the
501 additive value of DNA and RNA molecular profiling for p53-pathway alterations in NMIBC. While
502 we found overlap between RNA-based p53 pathway assessment and genomic alterations in cell
503 cycle pathway genes, this does not appear to be clinically relevant for recurrence after BCG.
504 Whether defects in the p53-pathway are associated with progression to secondary MIBC
505 remains unclear, but warrants further investigation(41).

506 Limitations of our study include the use of targeted sequencing panels rather than whole
507 exome/transcriptomic sequencing, and that clinical outcomes were available only for the MSK
508 cohort. These limitations may have caused us to miss subtle but important associations with
509 gene expression signatures and precluded our ability to perform robust multivariable analyses.
510 Additional, large prospective validation that account for differences in tumor cell purity is
511 required and ongoing at our center and elsewhere.

512

513 In sum, the current analysis demonstrates the feasibility of performing RNA-based subtyping on
514 clinical NMIBC specimens, even those with relatively low tumor cellularity. Gene expression
515 signatures can provide novel biologically and clinically relevant information on the TME (i.e.,
516 immune score) and tumor-intrinsic properties (i.e., mutant p53 pathway signature) that are
517 additive and complementary to analysis of genomic DNA. The integration of transcriptomic
518 tumor profiling with exome sequencing is a promising approach to improve risk stratification and
519 treatment selection for patients with NMIBC.

520

521

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525

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- 526
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- 665
- 666

667 **Table 1. Patient and tumor characteristics.** Data are presented as mean (SD) or n (%).

	All n = 78	UNC n = 28	MSK n = 50	MSK BCG-treated n = 38
Age (years)	65.0 (12.5)	63.9 (12.1)	65.6 (12.7)	66.8 (11.4)
Male sex	59 (75.6%)	22 (78.6%)	37 (74.0%)	29 (76.3%)
Stage and grade¹				
T1 high-grade	35 (46.1%)	18 (64.3%)	17 (34.0%)	14 (36.8%)
Ta high-grade	25 (32.0%)	0 (0.0%)	25 (50.0%)	23 (60.5%)
Ta low-grade	18 (21.1%)	10 (35.7%)	8 (16.0%)	1 (2.6%)
Carcinoma in situ (CIS)	22 (28.2%)	8 (28.5%)	14 (28.0%)	14 (36.8%)
Tumor cellularity¹				
0-40	33 (43.4%)	13 (46.4%)	18 (36.0%)	17 (44.7%)
50-90	43 (56.7%)	11 (39.3%)	32 (64.0%)	21 (55.3%)
UROMOL subclass				
1/ Luminal	26 (33.3%)	7 (25.0%)	19 (38.0%)	10 (26.3%)
2/ CIS-like	28 (35.9%)	8 (27.6%)	20 (40.0%)	17 (44.7%)
3/ Early basal-like	24 (30.85%)	13 (46.4%)	11 (22.0%)	11 (30.0%)
Immune score²				
Low	20 (25.6%)	6 (21.4%)	14 (28.0%)	11 (30.0%)
Medium	20 (25.6%)	5 (17.9%)	15 (30.0%)	8 (21.0%)
High	38 (48.7%)	17 (60.7%)	21 (42.0%)	19 (50.0%)
Treatment				
BCG			38 (76%)	38 (100%)
Mitomycin			5 (10%)	0
TURBT only			6 (12%)	0
Cystectomy			1 (2%)	0
Tumor size ≥3 cm			26 (52%)	18 (47.4%)
Multiple tumors			19 (38%)	18 (47.4%)

668

669 ¹Missing values omitted

670 ²Grouped into tertiles based on score rank BCG = bacillus Calmette-Guérin; CIS = carcinoma in

671 situ; MSK = Memorial Sloan Kettering Cancer Center; SD = standard deviation; TURBT =

672 transurethral resection of the bladder tumor; UNC = University of North Carolina.

673

674 **Table 2. Descriptive statistics by UROMOL signature classification.** Data are presented as
 675 mean (SD) or n (%).

	1/ Luminal n = 26	2/ Luminal CIS-like n = 28	3/ Early basal-like n = 24	p
Institution/cohort				0.08
UNC	7 (26.9%)	8 (28.6%)	13 (54.2%)	
MSK	19 (73.1%)	20 (71.4%)	11 (45.8%)	
Age (years)	61.6 (14.9)	65.3 (11.2)	68.3 (10.3)	0.159
Male sex	20 (76.9%)	20 (71.4%)	19 (79.2%)	0.797
Stage				0.001
T1	4 (15.4%)	15 (53.6%)	16 (66.7%)	
Ta	22 (84.6%)	13 (46.4%)	8 (33.3%)	
Grade¹				< 0.001
High	12 (46.2%)	27 (96.4%)	21 (87.5%)	
Low	14 (53.8%)	0 (0.00%)	2 (8.33%)	
Carcinoma in situ	8 (30.8%)	2 (7.14%)	7 (29.2%)	0.063
Tumor cellularity¹				0.004
0-40	6 (23.1%)	11 (40.7%)	16 (69.6%)	
50-90	20 (76.9%)	16 (59.3%)	7 (30.4%)	

676 ¹ Missing values omitted

677 CIS = carcinoma in situ; MSK = Memorial Sloan Kettering Cancer Center; SD = standard
 678 deviation; UNC = University of North Carolina.

679

680 **Table 3. Gene alterations across UROMOL subclasses.** Bold indicates statistical
 681 significance.

Gene	1/ Luminal (n=19)	2/ Luminal CIS-like (n=20)	3/ Early basal-like (n=11)	p=
FGFR3	17 (89%)	10 (50%)	2 (18%)	<0.001
KRAS	1	2	0	0.8
HRAS	1	0	2	0.1
ERBB2	0	4 (25%)	4 (36%)	0.02
ERBB3	2	1	1	0.8
PIK3CA	5	3	4	0.4
TSC1	1	4	1	0.4
NF1	0	4	1	0.1
TP53	2	5	4	0.3
RB1	0	1	1	0.7
MDM2	1	3	0	0.5
CCND1	0	2	1	0.4
CDKN2A	3	7	2	0.3
CDKN1A	3	3	1	1
STAG2	4	4	2	1
KDM6A	9	6	7	0.2
ARID1A	2	5	2	0.6
KMT2A	2	1	0	0.6
KMT2C	2	3	0	0.5
KMT2D	6	5	1	0.4
EP300	4	1	2	0.3
CREBBP	1	5	1	0.2
TERT	12 (63%)	20 (100%)	9 (82%)	0.007
ERCC2	2 (11%)	2 (10%)	5 (45%)	0.044
ATM	0	4	2	0.1
BRCA1	0	2	0	0.3
BRCA2	0	1	1	0.7
TET2	0	1 (5%)	3 (27%)	0.03
MGA	1	3	0	0.5
BCOR	0	0	1	0.2
BRD4	0	0	0	1
DNAJB1	0	0	1	0.2
IDH1	0	0	0	1
JAK3	0	0	1	1
ARID1B	0	1	2	0.2
BAP1	1	3	2	0.6
PBRM1	2	2	0	0.6
NFE2L2	0	1	0	1
TBX3	1	0	0	0.6
NCOR1	1	0	1	0.5

682
 683 CIS = carcinoma in situ
 684

685 **Figure Legends**

686

687 **Figure 1. Heatmap of patient and tumor genetic characteristics according to cohort and**
688 **UROMOL subclass.**

689

690 **Figure 2. Recurrence-free survival (RFS) stratified by UROMOL subclass. A,** entire MSK
691 cohort; **B,** MSK patients treated with BCG; **C,** entire UROMOL cohort; **D,** UROMOL patients
692 treated with BCG.

693

694 **Figure 3. RNA expression-based immune score differs among UROMOL subclasses and**
695 **is associated with recurrence-free survival (RFS) after BCG. A,** Average expression counts
696 for markers of various immune cell types for each patient. **B-E,** RFS by immune score in the **B,**
697 entire MSK cohort; **C,** MSK patients treated with BCG; **D,** entire UROMOL cohort, **E,** UROMOL
698 patients treated with BCG.

699

700 **Figure 4. Immune correlates of recurrence following BCG treatment. A,** Volcano plot
701 demonstrating correlation of expression of individual immune-related genes with probability of
702 remaining free of a high-grade recurrence beyond 24 months after BCG (only genes statistically
703 significant after Benjamini-Hochberg false discovery correction are in red and labeled). **B-C,**
704 RFS by PD-L1 (CD274) expression among BCG-treated (**B**) MSK patients and (**C**) patients in
705 the UROMOL cohort.

706

707

708

709

Figure 1

Memorial Sloan Kettering

University of North Carolina

UROMOL

Class 1

Class 2

Class 3

Class 1

Class 2

Class 3

Gender

Stage

Grade

CIS

Gender

Male

Female

Stage

Ta

T1

Grade

Low

High

Missing

CIS

No

Yes

p53 Pathway Score

Mutant - Like

Wild Type - Like

Mutation

Mutant

Wild Type

Missing

p53 Pathway Score

TP53

FGFR3

TERT promoter

PIK3CA

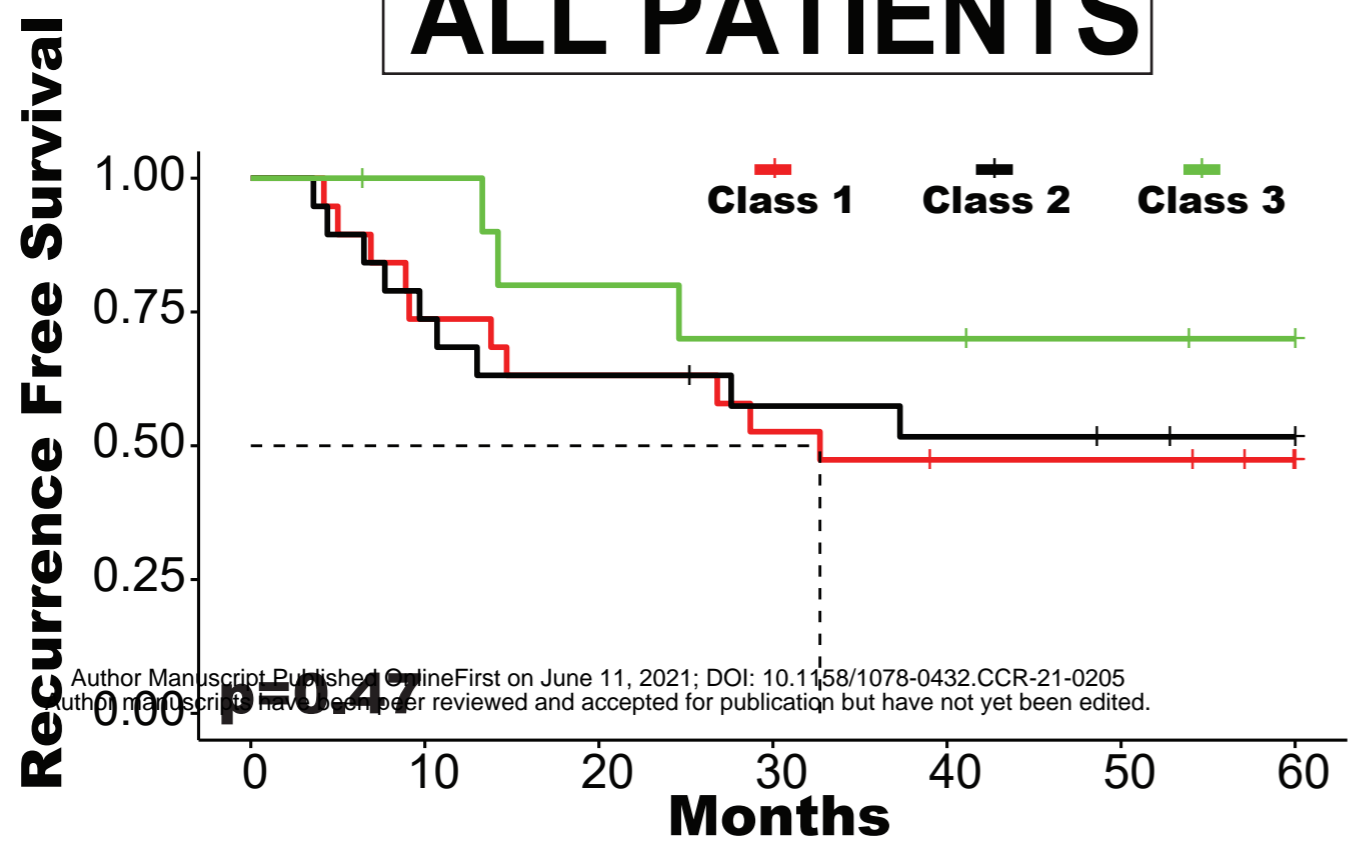
CDKN2A

STAG2

KDM6A

MSK COHORT

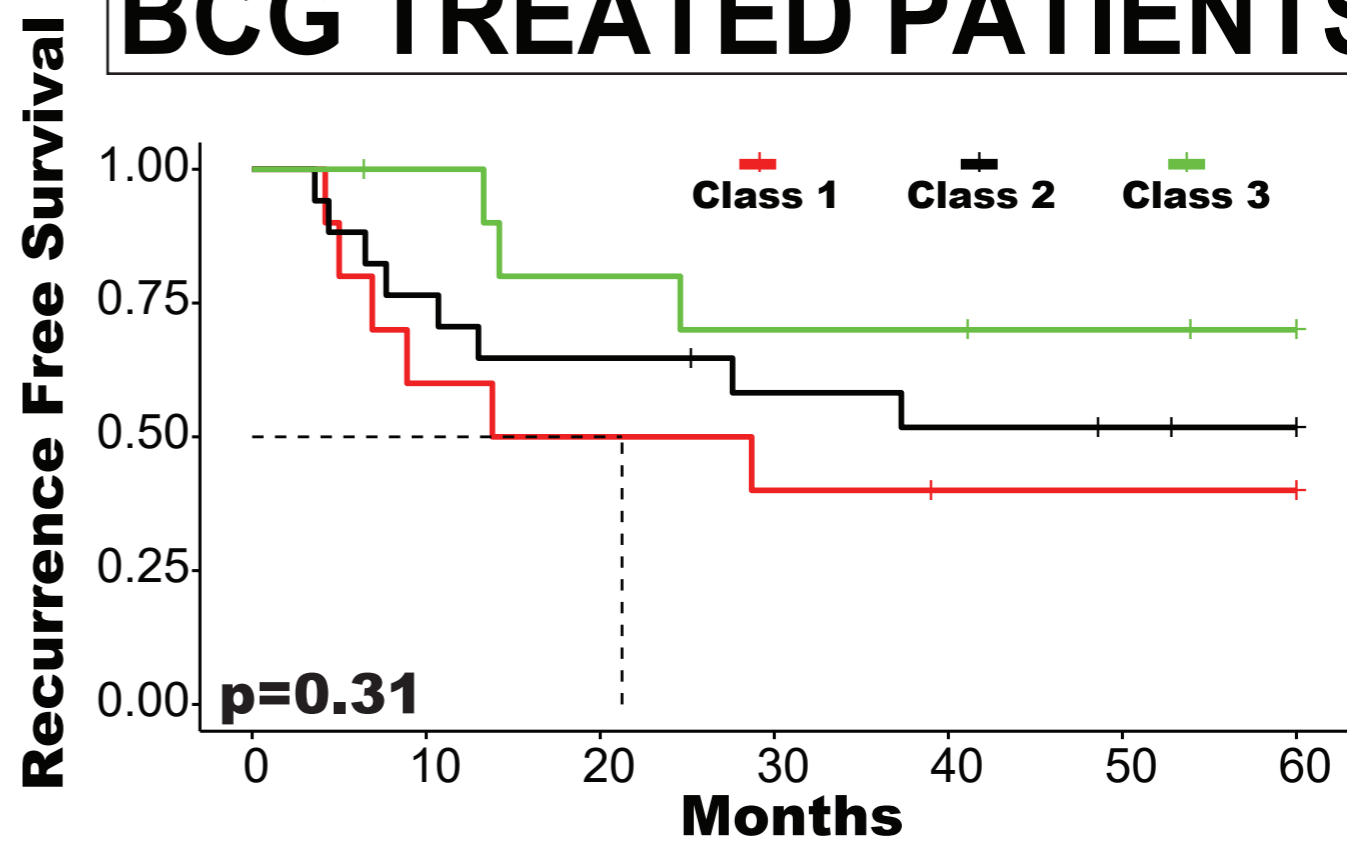
ALL PATIENTS



Class 1	19	14	12	10	8	8	5
Class 2	19	14	12	10	9	8	7
Class 3	11	10	8	7	7	6	5

2A

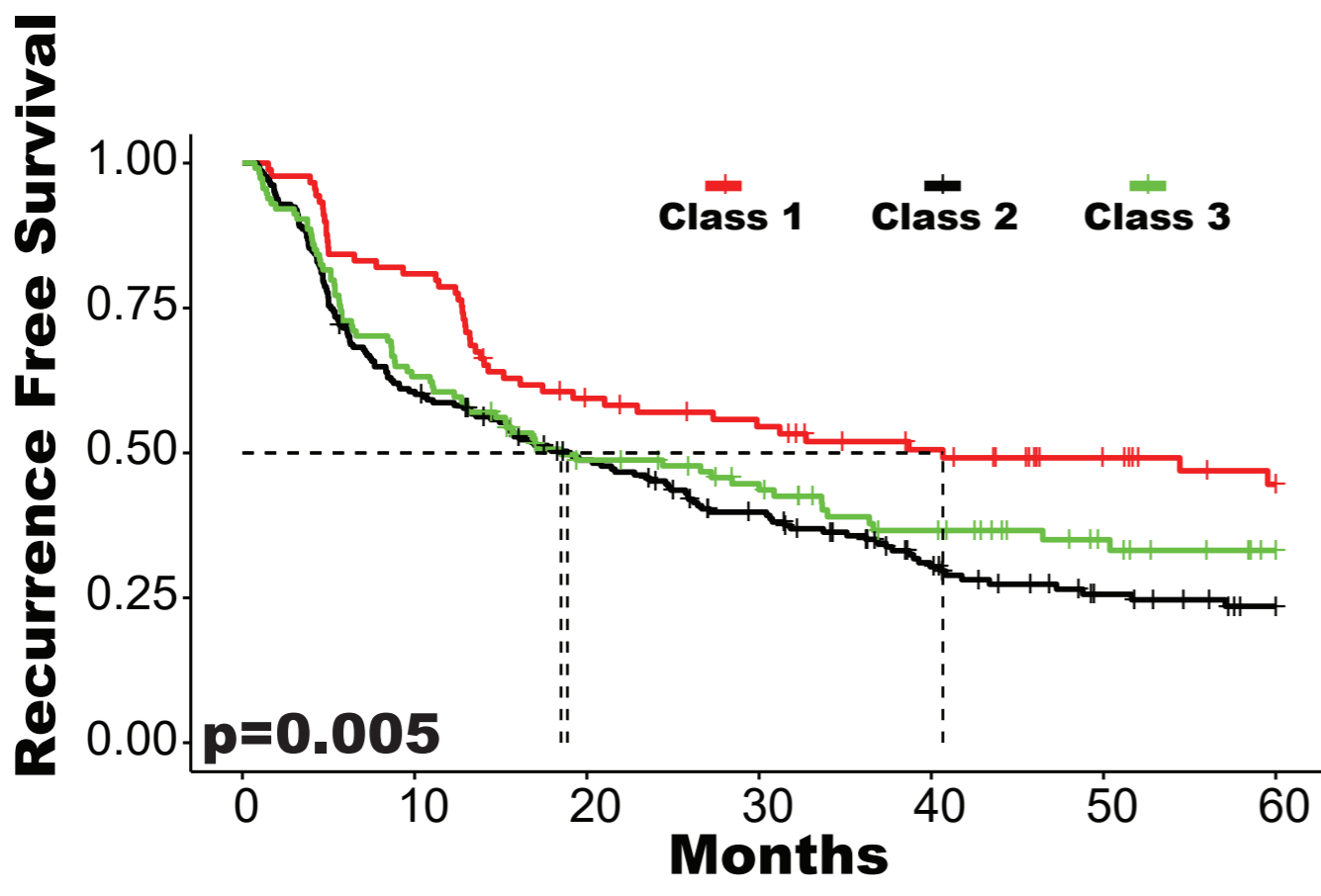
BCG TREATED PATIENTS



Class 1	10	6	5	4	3	3	3
Class 2	17	13	11	9	8	7	6
Class 3	11	10	8	7	7	6	5

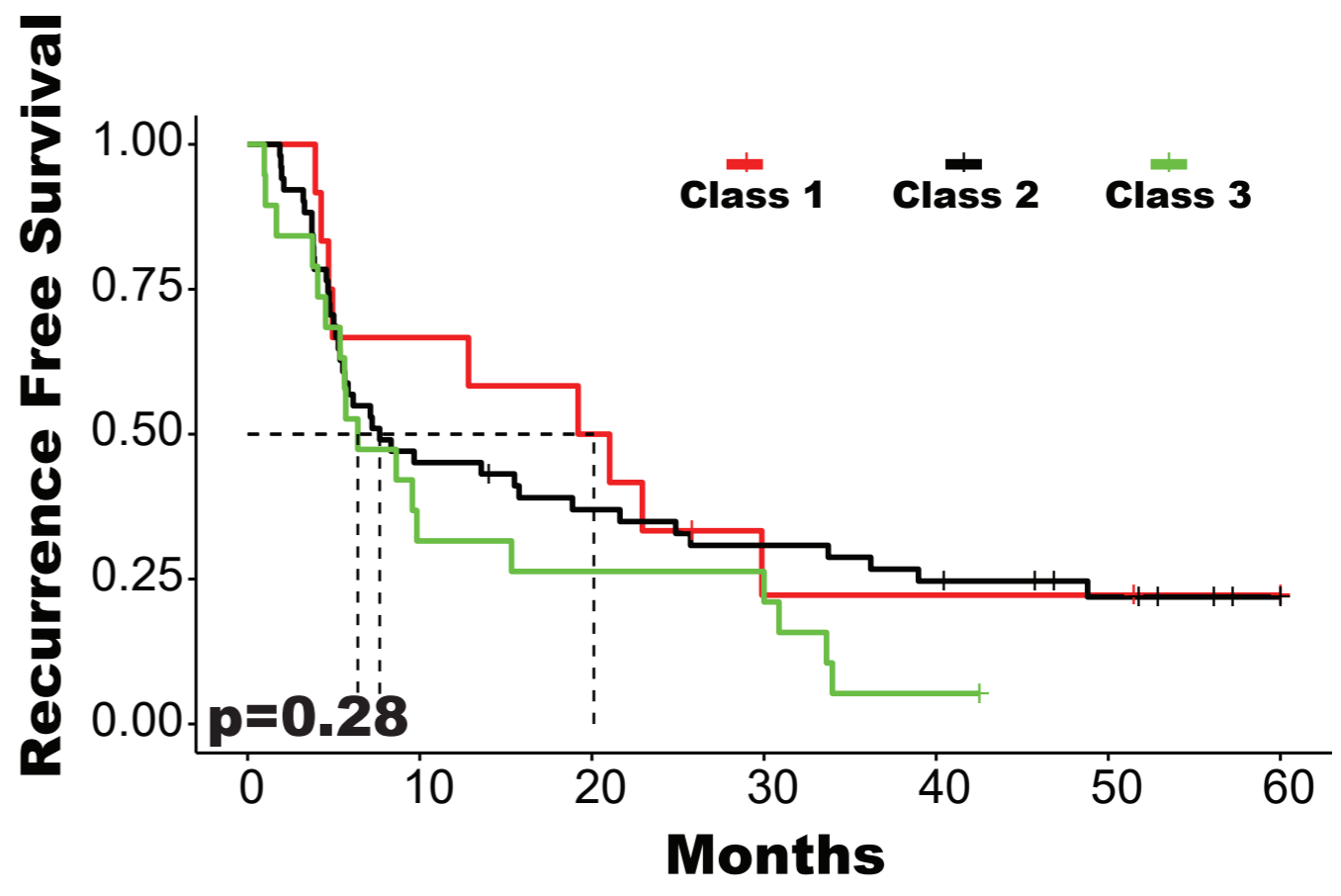
2B

UROMOL COHORT



Class 1	89	72	50	44	36	26	19
Class 2	211	127	95	70	44	27	18
Class 3	114	72	51	42	30	19	11

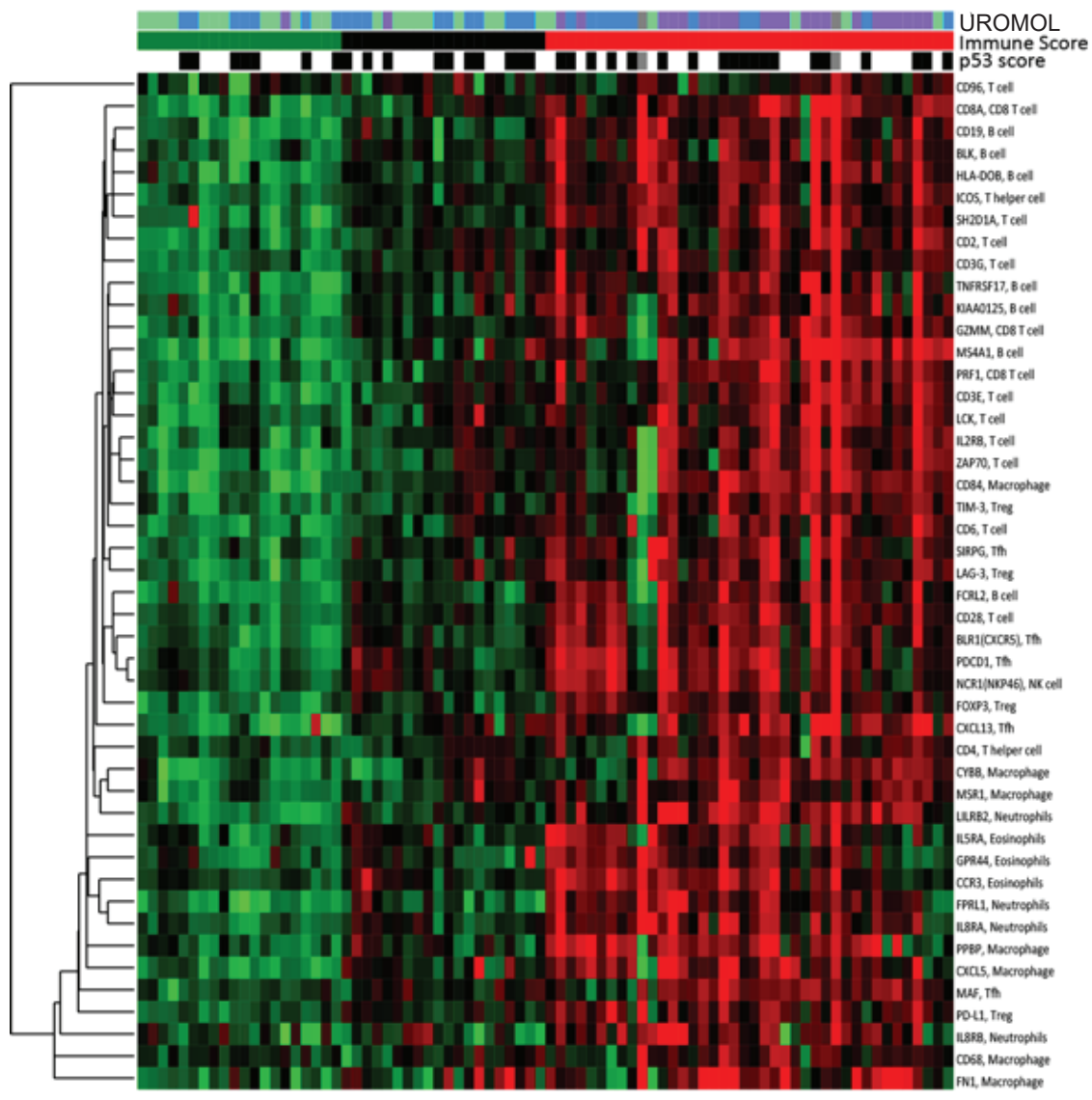
2C



Class 1	12	8	6	2	2	2	1
Class 2	51	23	18	15	12	8	4
Class 3	19	6	5	5	1	0	0

2D

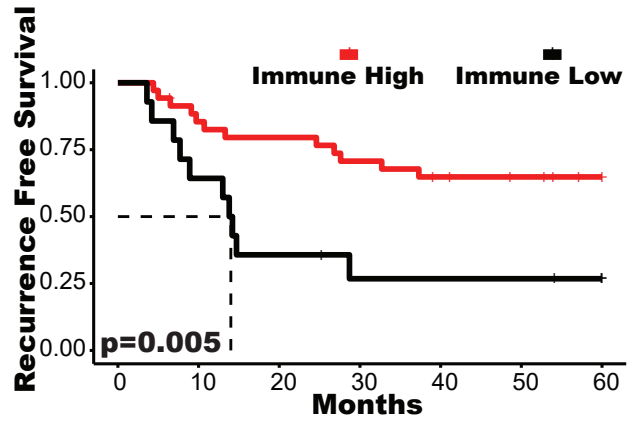
Figure 3



3A

ALL PATIENTS

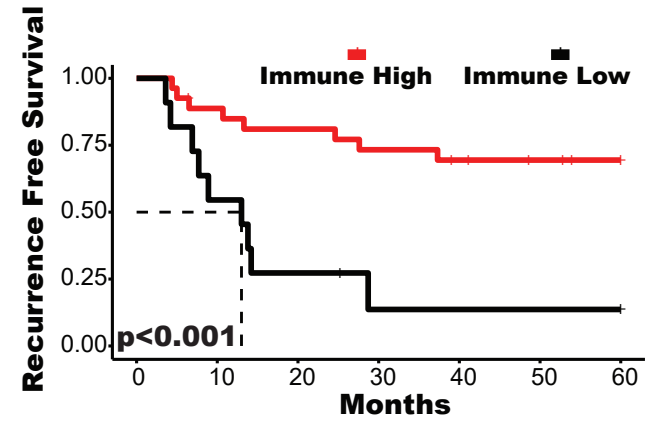
MSK COHORT



Immune High	35	29	27	24	21	19	16
Immune Low	14	9	5	3	3	3	1

3B

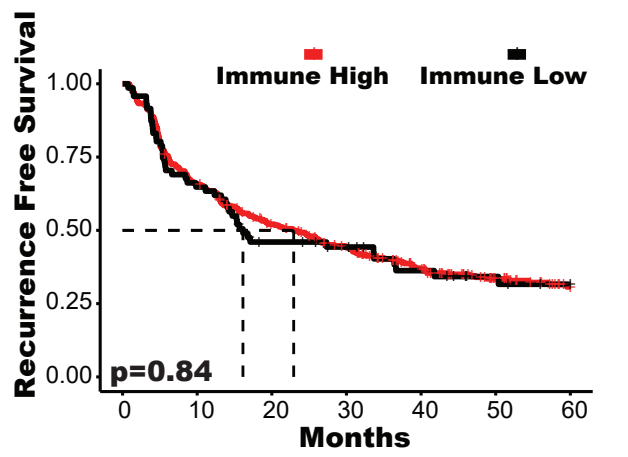
BCG TREATED PATIENTS



Immune High	27	23	21	19	17	15	13
Immune Low	11	6	3	1	1	1	1

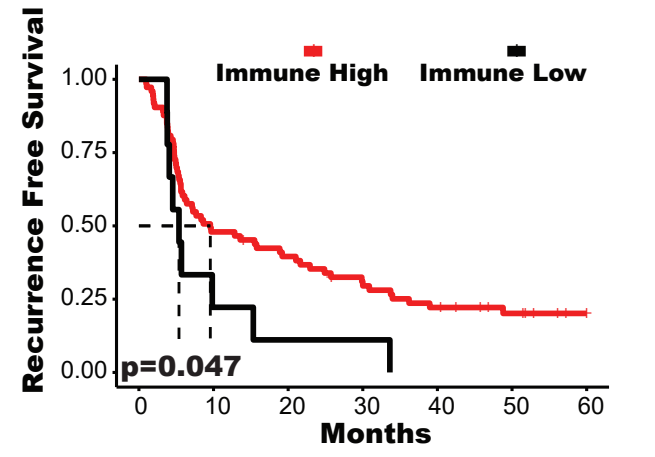
3C

UROMOL COHORT



Immune High	343	225	167	131	92	59	38
Immune Low	71	46	29	25	18	13	10

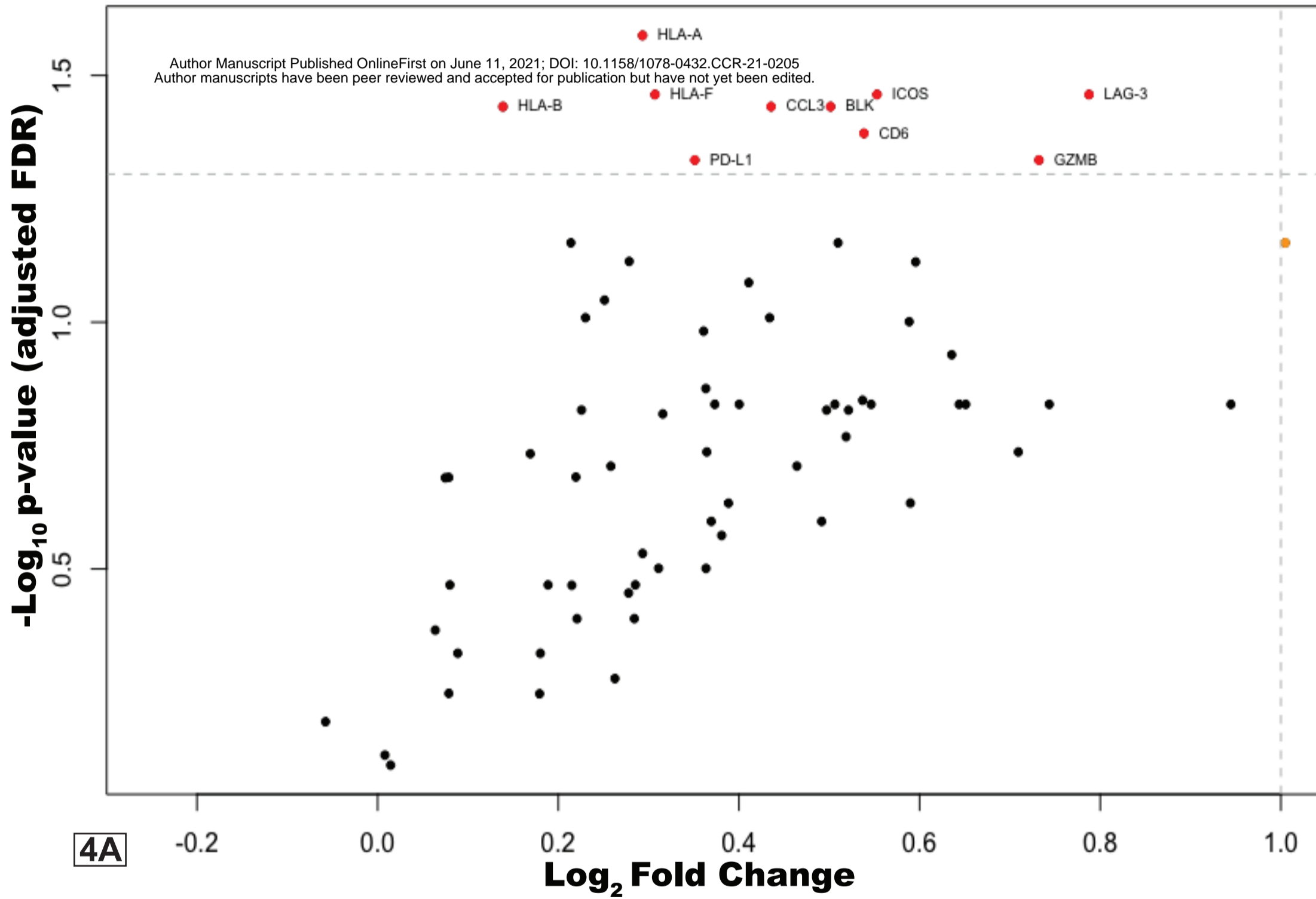
3D



Immune High	73	35	28	21	15	10	5
Immune Low	9	2	1	1	0	0	0

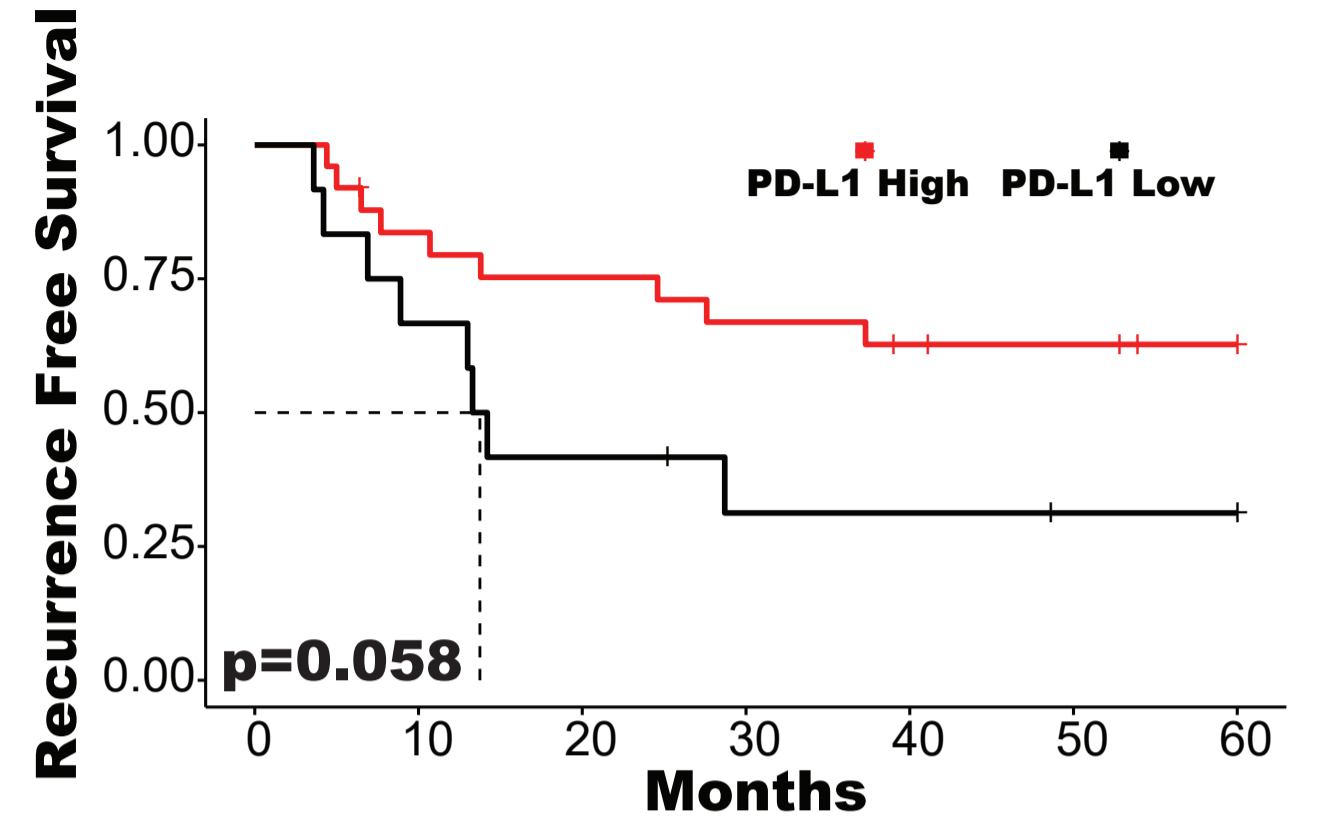
3E

Immune Gene Expression Associated with >2 Year RFS after BCG



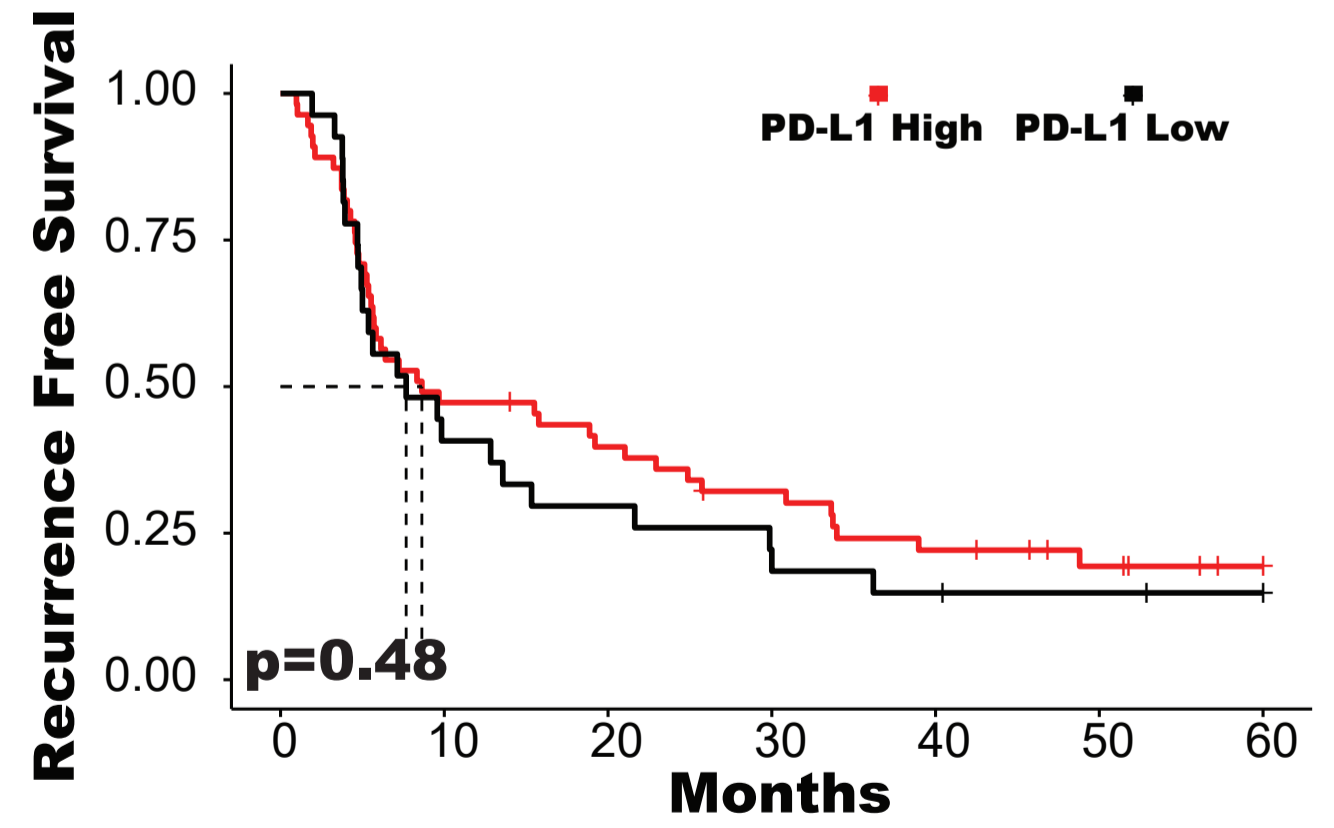
MSK COHORT

BCG TREATED PATIENTS



PD-L1 High	25	20	18	16	14	13	11
PD-L1 Low	12	8	5	3	3	2	2

UROMOL COHORT



PD-L1 High	55	26	21	16	11	7	3
PD-L1 Low	27	11	8	6	4	3	2