Quantitative Spatial Profiling of PD-1/PD-L1 Interaction and HLA-DR/IDO-1 Predicts Improved Outcomes of Anti-PD-1 Therapies in Metastatic Melanoma

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Abstract

Purpose: PD-1/L1 axis–directed therapies produce clinical responses in a subset of patients; therefore, biomarkers of response are needed. We hypothesized that quantifying key immunosuppression mechanisms within the tumor microenvironment by multiparameter algorithms would identify strong predictors of anti–PD-1 response.

Experimental Design: Pretreatment tumor biopsies from 166 patients treated with anti–PD-1 across 10 academic cancer centers were fluorescently stained with multiple markers in discovery (n = 24) and validation (n = 142) cohorts. Biomarker-positive cells and their colocalization were spatially profiled in pathologist-selected tumor regions using novel Automated Quantitative Analysis algorithms. Selected biomarker signatures, PD-1/PD-L1 interaction score, and IDO-1/HLA-DR coexpression were evaluated for anti–PD-1 treatment outcomes.

Results: In the discovery cohort, PD-1/PD-L1 interaction score and/or IDO-1/HLA-DR coexpression was strongly associated with anti–PD-1 response (P = 0.0005). In contrast, individual biomarkers (PD-1, PD-L1, IDO-1, HLA-DR) were not associated with response or survival. This finding was replicated in an independent validation cohort: patients with high PD-1/PD-L1 and/or IDO-1/HLA-DR were more likely to respond (P = 0.0096). These patients also experienced significantly improved progression-free survival (HR = 0.36; P = 0.0004) and overall survival (HR = 0.39; P = 0.0011). In the combined cohort, 80% of patients exhibiting higher levels of PD-1/PD-L1 interaction scores and IDO-1/HLA-DR responded to PD-1 blockers (P = 0.000004). In contrast, PD-L1 expression was not predictive of survival.


Introduction

PD-1/L1 axis–directed therapies produce profound clinical responses, but only in a subset of patients (1). Although PD-L1 immunohistochemistry tests are currently utilized to select patients for anti–PD-1 therapy, multiple studies have demonstrated imperfect correlation with clinical response (2). Therefore, novel biomarker signatures encompassing the biological complexity of functional antitumor immune responses are being pursued using next-generation technologies (3, 4). In this study, we quantified two complementary immunosuppression signatures hypothesized to correlate with anti–PD-1 response using novel digital pathology algorithms. (i) density of PD-1–positive cells in close proximity to PD-L1–positive cells, namely, PD-1/PD-L1 interaction score as a measure of adaptive immune resistance, in contrast to PD-L1 alone (5); (ii) colocalization of HLA-DR and IDO-1 which are known to be upregulated by IFNγ in immune cell-rich tumors (6, 7). Our principal hypothesis was that adaptive immune resistance...
Translational Relevance

PD-1 and PD-L1 inhibitors are currently transforming the treatment landscape of many solid tumors. However, reliable biomarkers to predict response have remained elusive. The digital pathology approach described here identified unique biomarker signatures strongly associated with anti–PD-1 outcomes in a real world cohort of 166 melanoma patients from 10 different cancer centers. Patients positive for biomarker signatures (PD-1/PD-L1 and HLA-DR/IDO) had a high likelihood of response (80%) and a 3-fold improvement in overall survival (HR, 0.31). This digital pathology approach provides a biologically relevant profile of the tumor microenvironment and may be clinically useful for selecting patients for anti–PD-1 therapies. The unique advantages of this approach include its objective, quantitative nature, minimal tissue requirements (2 unstained slides), and high predictive capacity. Further, we speculate that this test could be used for promising novel combinations, including inhibitors of IDO-1 and LAG-3 (a cognate ligand of HLA-DR).

Discovery cohort. We used tumor samples from patients with metastatic melanoma (n = 24) treated with single-agent anti–PD-1 therapy ( pembrolizumab or nivolumab) at Vanderbilt University Medical Center between 2009 and 2015. All retrospective specimens were obtained from formalin-fixed, paraffin-embedded (FFPE) biopsy or resection tissue within the 12 months prior to the start of anti–PD-1 therapy. Response status was determined using RECIST criteria.

Independent validation cohort. We used tumor samples from patients with metastatic melanoma (n = 142) treated with single-agent anti–PD-1 ( pembrolizumab or nivolumab) at ten different cancer centers: Yale-New Haven Health system, Moffitt Cancer Center, University of Alabama Birmingham, Mayo Clinic Jacksonville, Mayo Clinic Phoenix, University of California San Francisco, University of Virginia, Medical University of South Carolina, Cleveland Clinic, and Washington University in St. Louis, between 2010 and 2016. All retrospective specimens were obtained from FFPE tissue collected prior to the start of anti–PD-1 therapy, with the vast majority within the 12 months prior to therapy initiation. Response status was determined by each institution using RECIST v1.1 criteria; when not available, irRECIST or immune-related response criteria were permitted (10–12). For the purpose of this study, “responders” were defined as patients experiencing complete response, partial response, or stable disease (SD) greater than 6 months, and “non-responders” as patients with progressive disease (PD) as best response, or SD for less than 6 months (13). Clinical outcomes were blinded to the operators performing specimen testing and image analyses in both the discovery and validation cohorts. Clinicopathologic characteristics for the entire cohort are listed in Table 1.

Control (standard-of-care) cohort. We obtained tumor samples from patients with metastatic melanoma (n = 44) who received standard-of-care treatments prior to approval of single-agent anti–PD-1 ( pembrolizumab or nivolumab) or were not eligible for anti–PD-1 therapies at University of Alabama, Birmingham, and Vanderbilt University Medical Center between July 2006 and March 2014. All retrospective specimens were obtained from FFPE biopsy. Clinicopathologic characteristics were collected from medical record review and are listed in Supplementary Table S1.

QIF staining, imaging, and analysis

Immunofluorescence staining. FFPE tissue or cell line samples were dewaxed, rehydrated through a series of xylene to alcohol washes to distilled water. Heat-induced antigen retrieval was then performed using the NsGen Decloaking Chamber (Biocare) in Diva buffer (Biocare) and transferred to Tris-buffered saline. All subsequent staining steps were performed at room temperature. Endogenous peroxidase was blocked using Peroxidized 1 (Biocare) followed by incubation with a protein-blocking solution (Background Sniper, Biocare) to reduce nonspecific antibody staining. Slides were stained 1:200 for 1 hour with mouse anti–PD-1 (NAT105, Biocare) and incubated with neat EnVision+ HRP Mouse (Agilent) for 30 minutes, and PD-1 staining was detected using 1:200 dilution of TSA-Cy3.5 (Perkin Elmer) for 10 minutes. Any residual HRP was then quenched using 2 × 10 minute washes of fresh 100 mmol/L benzylidine with 50 mmol/L hydrogen peroxide before staining with 3.6 μg/mL rabbit anti–PD-L1 (E1L3N, Cell Signaling Technology), incubated with a cocktail of neat EnVision+ HRP Rabbit

Materials and Methods

Patient specimens and study design

This study was conducted in accordance with U.S. Common Rule where the studies were performed after approval by each local Institutional Review Board with appropriate written consent from the subjects (wherever necessary) prior to specimen testing.
Table 1. Patient baseline characteristics from discovery and validation studies

<table>
<thead>
<tr>
<th>Characteristic</th>
<th>Discovery (n = 24)</th>
<th>Validation (n = 142)</th>
<th>p*</th>
</tr>
</thead>
<tbody>
<tr>
<td>Age, years</td>
<td>Mean (SD)</td>
<td>57.3 (13.9)</td>
<td>6.4(14.3)</td>
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<tr>
<td></td>
<td>Female</td>
<td>9 (38)</td>
<td>54 (38)</td>
</tr>
<tr>
<td></td>
<td>Male</td>
<td>15 (62)</td>
<td>68 (62)</td>
</tr>
<tr>
<td>BRAF mutation status, N (%)</td>
<td>Mutant</td>
<td>9 (38)</td>
<td>47 (33)</td>
</tr>
<tr>
<td></td>
<td>WT</td>
<td>15 (62)</td>
<td>63 (44)</td>
</tr>
<tr>
<td></td>
<td>Unknown</td>
<td>0</td>
<td>32 (23)</td>
</tr>
<tr>
<td>M staging of metastasis, N (%)</td>
<td>Lymph node</td>
<td>3 (12)</td>
<td>20 (14)</td>
</tr>
<tr>
<td></td>
<td>Skin</td>
<td>5 (21)</td>
<td>42 (30)</td>
</tr>
<tr>
<td></td>
<td>Otherb</td>
<td>16 (67)</td>
<td>66 (46)</td>
</tr>
<tr>
<td></td>
<td>Unknown</td>
<td>0</td>
<td>14 (10)</td>
</tr>
<tr>
<td>Location, N (%)</td>
<td>No</td>
<td>11 (46)</td>
<td>65 (46)</td>
</tr>
<tr>
<td></td>
<td>Yes</td>
<td>13 (54)</td>
<td>77 (54)</td>
</tr>
<tr>
<td>Previous systemic therapies, N (%)</td>
<td>0</td>
<td>3 (12)</td>
<td>53 (37)</td>
</tr>
<tr>
<td></td>
<td>1</td>
<td>10 (42)</td>
<td>49 (35)</td>
</tr>
<tr>
<td></td>
<td>2</td>
<td>10 (42)</td>
<td>17 (12)</td>
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<td></td>
<td>≥3</td>
<td>1 (4)</td>
<td>23 (16)</td>
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<tr>
<td>LDH level, N (%)</td>
<td>Normal</td>
<td>14 (58)</td>
<td>89 (63)</td>
</tr>
<tr>
<td></td>
<td>Elevated</td>
<td>10 (42)</td>
<td>31 (22)</td>
</tr>
<tr>
<td></td>
<td>Unknown</td>
<td>0</td>
<td>22 (15)</td>
</tr>
<tr>
<td>Anti-PD-1 response, N (%)</td>
<td>No</td>
<td>11 (46)</td>
<td>64 (45)</td>
</tr>
<tr>
<td></td>
<td>Yes</td>
<td>13 (54)</td>
<td>78 (55)</td>
</tr>
</tbody>
</table>

*pFrom t test for comparison of means, and χ² test for comparison of counts, across the discovery and validation samples.

bOther includes Adrenal, Bone, Brain, Breast, Colon, Esophagus, Intestine, Liver, Lung, Pancreas, Parotid, Sinus, Spleen, Trachea, Viscera, and Vulva.

Antibody validation. The specificity of the PD-L1 (E133N) antibody was verified on cell lines with known levels of PD-L1 expression, namely the positively expressing Karpas 299 and PD-L1–negative Ramos (RA 1) cell lines (Supplementary Fig. S1A and S1B). In addition, this PD-L1 clone has been validated previously (14) and shown to have high concordance with PD-L1 antibody clones, 22C3 and 28-8 (15, 16), which are approved as companion or complementary diagnostics with PD-1–targeted therapies associated with this study cohort. PD-1 (NAT105) antibody specificity was verified using CHO-K1 parental (PD-1 negative) and CHO-K1 cells transfected to express PD-1 (Supplementary Fig. S1C and S1D). HEK293 cells were negative for both IDO-1 (Supplementary Fig. S1E) and HLA-DR expression (Supplementary Fig. S1G). SKOV3 cells were previously identified as IDO-1 positive (Supplementary Fig. S1F), and A375 cells were identified as HLA-DR positive (Supplementary Fig. S1H).

Expression of all proteins was observed in the expected cellular localization, membranous for PD-L1, PD-1, and HLA-DR, and cytoplasmic for IDO-1. Quantitation of PD-1% and PD-L1% positive cellular expression by AQUA analysis in the CHO-K1/CHO-K1/PD-1 and Karpas 299/Ramos cell mixtures was compared with results obtained by flow cytometry and demonstrated very tight correlations with a Spearman rho 1.0 and 0.98 for PD-1 and PD-L1, respectively (Supplementary Fig. S1I and S1J).

Sample imaging. Fluorescence images were acquired on the Vectra 2 Intelligent Slide Analysis System (PerkinElmer; ref. 17). First, slides were imaged at 4x magnification based on DAPI, and an automated inForm software algorithm (v. 2.0) was used to identify tissue areas of the slides. The tissue areas were then imaged at 4x magnification for channels associated with DAPI (blue), FITC (green), and Cy5 (red) for the PD-1/PD-L1–stained slides or DAPI (blue), Cy3 (green), and Cy5 (red) for the IDO-1/HLA-DR–stained slides to create RGB images. These images were processed by automated inForm software enrichment algorithm to identify possible 20x high-power fields of view according to the highest Cy5 expression (PD-1/PD-L1) or Cy3 and Cy5 expression (IDO-1/HLA-DR). The top 40 fields of view were imaged at 20x magnification across DAPI, FITC, Texas Red, and Cy5 wavelengths (PD-1/PD-L1) or DAPI, FITC, Cy3, and Cy5 wavelengths (IDO-1/HLA-DR). All computer preselected images were reviewed by a board-certified pathologist for acceptability. Images lacking tumor cells, containing mainly necrotic cells, or abnormal fluorescence signals not expected with antibody localization or biology were rejected. Accepted images were processed using AQUA-duct (PerkinElmer), wherein each fluorophore was spectrally unmixed into individual channels and saved as a separate file for analysis.

Image analyses by AQUA technology. This technology has been widely validated in clinical settings for objective quantitation of biomarkers in tissues (9, 18–22). Briefly, the pathologist-accepted image files were subjected to AQUAAnalysys (v. 3.1.2.4) via a fully automated process. DAPI signal within each accepted image was used to identify cell nuclei and then dilated to the approximate size of an entire cell. Melanoma tumor marker S100 (Fig. 1A) detected with 488 dye was used to create a binary mask of all tumor area within that image (Fig. 1B). Overlap between this mask and the mask of all cells created a new mask for all tumor cells. Similarly, absence of the tumor cell marker in combination with the mask of all nuclei created a new mask for all nonnontumors.

(15–16)
Image analysis by AQUA Technology. Representative pseudo-colored images are shown for S100 (A), PD-L1 (C), PD-1 (E), IDO-1 (I), CD11b (K), and HLA-DR (K, M, O). Binary masks for each cell phenotype of interest were generated in AQUA analysis as shown for tumor cells (B), PD-L1 (D), and PD-1 (F). To identify the proportion of PD-1 cells (G) in close proximity to PD-L1 expression, an interaction mask was generated by expanding the PD-L1 mask (H). An example for phenotyping MDSC is represented by CD11b⁺IDO-1⁺HLA-DR⁻ cells (I), TAMs by CD11b⁺IDO-1⁺HLA-DR⁺ cells (L), and suppressive cells by IDO-1⁺HLA-DR⁺ (N) or S100⁺IDO-1⁺HLA-DR⁺ (P). S100⁺HLA-DR⁺IDO-1⁺ cells were identified as non-myeloid suppressive cells (O-P). Image resolution 1 pixel = 0.5 micron.
cells. Each Cy3.5 image (Fig. 1E) was used to create a binary mask for PD-1+ cells and overlapped with the mask of all nontumor cells to create a binary mask of all nontumor cells that are PD-1+ (Fig. 1F). The binary mask of all PD-L1+ cells was dilated to create an interaction mask encompassing the nearest neighbor cells (e.g., cells with PD-1). This interaction mask was combined with the binary mask of all PD-1+ nontumor cells to create an interaction compartment of the PD-1+ cells in close proximity to the PD-L1+ cells such that PD-1 is likely interacting with PD-L1 (Fig. 1G and H). The interaction algorithm was verified in PD-1+ and PD-L1+ cell line mixtures, wherein the total area from all accepted fields for the interaction compartment and the total area of the nontumor cells were calculated and multiplied by a factor of 10,000 to create a whole number representing an interaction score for each specimen, and this score increased with increasing proportions of PD-1+ cells (Supplementary Fig. S2). In accordance with the paradigm proposed by Bronte and colleagues (23) and based on the immunosuppressive function of IDO-1 in tumor and myeloid cells (24–26), the CD11b+/HLA-DR-/IDO-1+ signature was used to designate myeloid-derived suppressor cells (MDSC, Fig. 1I and J); CD11b+/HLA-DR+/IDO-1+ cells were described as tumor-associated macrophages (TAM, Fig. 1K and L); and HLA-DR+/IDO-1+ or SI100+/HLA-DR+/IDO-1+ cells were identified as nonmyeloid suppressive cells (Fig. 1O and P). Reproducibility of the AQUA quantitation for PD-1 and PD-L1 was evaluated on multiple staining runs performed on independent days using the YTMA322 control array. Quantitation for both markers was found to be highly reproducible using linear regression with $r^2 = 0.97$ and 0.98, respectively (Supplementary Fig. S3A and S3B). Reproducibility of the AQUA quantitation for HLA-DR and IDO-1 was evaluated using the MDSC-TMA-containing cores from cell lines, SKOV3, HEK293, and A375, treated with 0.1 μg/mL IFN, A375 treated with 0.05 pg/μL IFN, and three MDS patient samples. Both markers were highly reproducible using linear regression with $r^2 = 0.98$ and 0.97, respectively (Supplementary Fig. S3C and S3D).

**Cell lines.** Karpas299, Ramos (RA 1), HEK293, SKOV3, and A375 cell lines were purchased from the ATCC and maintained according to the manufacturer's instructions. The CHO-K1/PD-1 stable and parental CHO-K1 cell lines were purchased from GenScript and maintained according to the manufacturer's instructions. CHO-K1/PD-1 and CHO-K1 cells or Karpas299 and Ramos cells were then mixed across a range of positive/negative ratios to generate cell mixes with approximately 1%, 5%, 10%, 15%, 25%, 50%, or 100% of cells expressing PD-1 or PD-L1. Prior to mixing, dead cells were removed via centrifugation through Ficoll. Approximately 1 million cells were set aside for flow cytometry analysis with either the anti-human PD-1 BV421 antibody (MH4 clone, BD Biosciences) or anti-human PD-L1 APC antibody (MH1 clone, BD Biosciences). The remaining cells, approximately 30 million per mixture, were then fixed in 10% neutral-buffered formalin for a minimum of 6 hours, then transferred to 70% ethanol, and pelletted into 1.5 mL microcentrifuge tubes. All remaining ethanol was removed, and the cells were resuspended in 2% low melt agarose. The cell pellet was then processed through a series of ethanol gradients to xylene before embedding into paraffin for multiplexed fluorescence immunohistochemistry analysis. FFPE samples of Karpas/Ramos cell mixtures, normal tonsil, and placenta were also used to construct a 42-spot tissue microarray (YTMA322), used as a control for the PD-1/PD-L1 assay.

**Statistical analysis**

Youden index was used to identify optimal cut points for the PD-1/PD-L1 interaction score and IDO-1/HLA-DR tests from the discovery cohort. In particular, cutoff values for the PD-1/PD-L1 interaction score ranging from 900 to 1,400 and cutoff values of IDO-1 "HLA-DR" ranging from 2% to 8% were investigated. Values of 900 and 5%, respectively, maximized Youden index, so these cutoff values were used to define individuals with positive and negative combined tests: positives have PD-1/PD-L1 $\geq 900$ or IDO-1 "HLA-DR" $\geq 5$%; negative results have PD-1/PD-L1 < 900 and IDO-1 "HLA-DR" < 5%. The ability of the dual biomarker signatures to predict response and survival was then evaluated in the discovery and validation samples. The primary analytic framework used logistic regression, the Kaplan–Meier method, and Cox proportional hazards models of time to events. Specifically, logistic regression was used to evaluate the potential association between the dual test and response to treatment. Progression-free survival (PFS) and overall survival (OS) following therapy were assessed using Kaplan–Meier curves for the different dual test groups. The HR of PFS or OS was determined using Cox proportional hazards regression, contrasting patients with a high versus low dual test score (i.e., using the above dichotomized cutoff). For the logistic Cox and regression models, the potential confounding by metastatic stage, lactate dehydrogenase (LDH), age at receiving immunotherapy, and previous therapies was evaluated. The ability of individual markers (e.g., PD-L1) to predict survival was also evaluated. All analyses were performed using R (https://www.r-project.org/).

**Results**

**Identification of candidate biomarkers to predict response to anti–PD-1 therapy**

Based on the hypothesis that both PD-1 and PD-L1 would need to be expressed in close proximity and in sufficient quantity for patients to demonstrate a response to anti–PD-1 axis therapies, we first evaluated the ability of our PD-1/PD-L1 interaction score assessment to differentiate patients who responded versus those who did not in a discovery cohort of 24 patients. In support of this hypothesis, the PD-1/PD-L1 interaction score was able to significantly differentiate between responders and nonresponders (Fig. 2A) with a sensitivity of 0.69 and specificity of 0.82 (Table 2), whereas PD-1 or PD-L1 alone did not in this cohort (Fig. 2B and C). We also tested the FDA-approved PD-L1 22C3 chromogenic immunohistochemistry assay on this same cohort (performed by an independent laboratory), and this assay was also unable to distinguish responders from nonresponders (Fig. 2D).

We previously observed a correlation between major histocompatibility complex-II (HLA-DR) expression on tumor cells and anti–PD-1 response in patients with melanoma (8). However, HLA-DR is also robustly expressed by tumor-infiltrating myeloid cells (CD11b+), which are not known to contribute to this association. HLA-DR can be either constitutively expressed by melanoma and other tumor cells, or can be induced by IFNγ (8). Therefore, in the discovery cohort, we assessed additional combinations of HLA-DR, IDO-1 (an additional IFNγ-responsive
marker), and CD11b expression (to evaluate the myeloid compartment) to test the ability of these markers to predict patient response to PD-1 blockers. We observed significantly higher levels of IDO-1+/HLA-DR+ cells in patients who responded to anti–PD-1 therapy (Fig. 2E) that was independent of myeloid (CD11b) expression (Fig. 2F) with a sensitivity of 0.85 and specificity of 0.91 (Table 2), whereas individual biomarkers (Fig. 2G and H) or additional combinations were unable to differentiate responders.
Table 2. Correlation of novel biomarker signatures identified by quantitative digital pathology algorithms with anti-PD-1 response

<table>
<thead>
<tr>
<th>Measure</th>
<th>Discovery (n = 24)</th>
<th>Validation (n = 142)</th>
<th>Combined (n = 166)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>Y</td>
<td>N</td>
<td>P</td>
</tr>
<tr>
<td>PD-L1 (TC) b</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>≥5%</td>
<td>7</td>
<td>4</td>
<td>0.67</td>
</tr>
<tr>
<td>&lt;5%</td>
<td>5</td>
<td>5</td>
<td></td>
</tr>
<tr>
<td>PD-1/PD-L1 interaction score</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>≥900</td>
<td>9</td>
<td>2</td>
<td>0.02</td>
</tr>
<tr>
<td>&lt;900</td>
<td>4</td>
<td>9</td>
<td></td>
</tr>
<tr>
<td>IDO-1/HLA-DR</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>≥5%</td>
<td>11</td>
<td>1</td>
<td>0.0006</td>
</tr>
<tr>
<td>&lt;5%</td>
<td>2</td>
<td>10</td>
<td></td>
</tr>
<tr>
<td>Biomarker signature</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Single or dual positive c</td>
<td>12</td>
<td>2</td>
<td>0.0005</td>
</tr>
<tr>
<td>Negative</td>
<td>1</td>
<td>9</td>
<td></td>
</tr>
</tbody>
</table>

Note: From $\chi^2$ test, comparing biomarker signature-positive with -negative groups. Bold numbers indicate statistical significance.

bPD-L1 expression could not be assessed on a total of 4 patients (discovery cohort = 3 and validation cohort = 1) due to the absence of tumor-specific (S100) staining.

cSingle positivity is defined by the presence of either PD-1/PD-L1 interaction score or IDO-1/HLA-DR coexpression, and dual positivity is by presence of both biomarker signatures above cut point levels in the tumor sample.

from nonresponders (Supplementary Table S2). Representative images from responder and nonresponder patients for the PD-1/ PD-L1 and IDO-1/HLA-DR tests are shown in Fig. 2I–L.

To further verify the role of an IDO-1/HLA-DR c cellular phenotype in predicting response to anti–PD-1 therapy, we performed ex vivo culture of previously established patient-derived xenograft (PDX) tissue sections from a patient with melanoma with a profound clinical response and another patient with disease progression despite anti–PD-1 therapy (8). When these tumor samples were treated with IFNγ, we observed a significant induction and coexpression of HLA-DR and IDO-1 in the patient who had responded to anti–PD-1 therapy but not in the patient with progressive disease, demonstrating heightened sensitivity of the IDO-1/HLA-DR pathway to IFNγ in tumor tissue from the extraordinary responder (Supplementary Fig. S4A–S4D). In contrast, there was an absence of robust change in PD-L1 expression in either PDX model. These findings are largely replicative of immunoblotting data previously described by our group (8). These translational studies suggest IDO-1 and HLA-DR may more accurately reflect pre-existing antitumor immunity and inflammation, hence, more accurately reflect effective anti–PD-1 responses ($P = 0.0006$) than PD-L1 expression alone ($P > 0.05$).

From these two promising biomarker signatures, the maximized Youden index of 0.74 determined from the discovery cohort identified the optimal cut point for the PD-1/PD-L1 interaction score as ≥900 and IDO-1/HLA-DR ≥ 5% positivity (Fig. 2M). We then evaluated whether these two biomarker signatures were redundant or complementary. As summarized in Table 2, we observed the highest response rates in the group of patients whose tumors expressed either PD-1/PD-L1 interaction score and/or IDO-1/HLA-DR levels above their respective cut points; likewise, the lowest response rates were observed in the biomarker signature-negative population ($P = 0.0006$). The combination of high PD-1/PD-L1 interaction score and/or IDO-1/HLA-DR levels above their respective cut points; likewise, the lowest response rates were observed in the biomarker signature-negative population ($P = 0.0006$). The combination of high PD-1/PD-L1 interaction score and/or IDO-1/HLA-DR levels above their respective cut points; likewise, the lowest response rates were observed in the biomarker signature-negative population ($P = 0.0006$). The combination of high PD-1/PD-L1 interaction score and/or IDO-1/HLA-DR levels above their respective cut points; likewise, the lowest response rates were observed in the biomarker signature-negative population ($P = 0.0006$). The combination of high PD-1/PD-L1 interaction score and/or IDO-1/HLA-DR levels above their respective cut points; likewise, the lowest response rates were observed in the biomarker signature-negative population ($P = 0.0006$). The combination of high PD-1/PD-L1 interaction score and/or IDO-1/HLA-DR levels above their respective cut points; likewise, the lowest response rates were observed in the biomarker signature-negative population ($P = 0.0006$).
signatures \( (P = 0.0002; \text{Table 2}; \text{Supplementary Fig. S5}). \) We captured the highest proportion of responders (70.5%) by including patients that had either positive PD-1/L1 or IDO-1/HLA-DR scores. Smaller proportions of responders were identified if patients were stratified by a single signature: PD-1/PD-L1 interaction alone (51.3%) or IDO-1/HLA-DR (60.3%) alone (Table 2). More importantly, patients with PD-1/PD-L1 and/or IDO-1/HLA-DR above the prespecified cut point also experienced a significant improvement in PFS \( (HR = 0.36; P = 0.0004) \) and OS \( (HR = 0.39; P = 0.0011) \) as shown in Fig. 3A and B, whereas survival based on the presence of individual biomarker signatures was less profound (Supplementary Fig. S6). Furthermore, in multivariate analyses, survival predictions were not confounded by metastatic stage or LDH levels (see Supplementary Table S3). In contrast, tumor PD-L1 expression alone could not meaningfully \( (P > 0.1) \) segregate patients with improved PFS or OS regardless of the cut point.

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**Figure 3.**

Anti–PD-1-treated patients expressing novel biomarker signatures (PD-1/PD-L1 interaction and/or IDO-1/HLA-DR coexpression) experience significantly improved PFS (A) and OS (B) in contrast to PD-L1 alone in TC (C and D) at 5% cut point.
chosen, analyzed at 5% (Fig. 3C and D), 1%, and 25% (Supplementary Fig. S7). Likewise, total PD-L1 expression within tumor and immune cells produced survival curves nearly identical to tumor PD-L1 expression alone (data not shown, identical to PD-L1 TC). In addition, when viewed across all patients in discovery and validation cohorts (combined, n = 166), we found that patients whose tumors exhibited both PD-1/L1 interaction scores and IDO-1/HLA-DR biomarker signatures above cut point levels have the highest likelihood of response (80%, P = 0.000004) and superior OS (P = 0.0018; Fig. 4A), which is further illustrated by the highest sensitivity and specificity associated with the dual-biomarker test positive group (Table 2).

We then assessed whether this signature was associated with treatment-specific effect related to anti–PD-1 clinical benefit compared with a more generally positive prognosis. Thus, we evaluated survival relationships of these same biomarker signatures in a control cohort of patients with American Joint Committee on Cancer stage III–IV melanoma (n = 44), who had not received anti–PD-1 therapies (Supplementary Table S1). No statistically significant difference in survival was observed among biomarker signature–positive and –negative populations (see Supplementary Fig. S8).

Discussion
We described identification of immune biomarker signatures that correlate strongly with clinical outcomes to PD-1 blockade. The clinical utility of these predictive signatures was carefully validated in one of the largest cohorts of patients with metastatic melanoma treated at ten different institutions. More importantly, these assays apply minimal exclusion criteria and only require limited tumor specimens (2 unstained slides). The higher predictive value of these assays is likely due to the unique operating principles underlying the digital pathology approach, which involves use of spatial information and signal enrichment of rare cellular phenotypes within the tumor microenvironment and their conversion into actionable clinical signatures (9, 17, 19, 21, 22). This is exemplified by the precise determination of PD-1 and PD-L1 spatial interactions and accurate quantitation of their colocalization density, thus directly addressing the hypothesis that the therapeutic mechanism of PD-1 blockers requires both PD-1 and PD-L1 spatial interactions and accurate quantitation of their colocalization density, thus directly addressing the hypothesis that the therapeutic mechanism of PD-1 blockers requires both PD-1 and PD-L1 expression in sufficient quantity and spatial proximity within tumor tissue (5). This study along with multiple independent studies further reinforces the imperfect correlation between PD-L1 expression alone and clinical outcomes (1, 2).

The desire to explore alternate predictive pathways became imperative when we observed patients who progressed on anti–PD-1 therapy despite high PD-L1 expression. We have previously discovered a link between HLA-DR expression and response to anti–PD-1 (8) and thus hypothesized that this IFNγ-responsive MHC class II molecule could augment the signature's predictive capacity and identify other populations of immunogenic cells. HLA-DR has been reported in wide range of cancers, including breast cancer (27, 28), colon cancer (29), and Hodgkin lymphoma (30). Notably, MHC-II was shown to correlate with response to anti–PD-1 in complementar y fashion to PD-L1 expression in Hodgkin lymphoma (30). IDO-1 is also upregulated by IFNγ on tumor and other infiltrating cells in the tumor microenvironment (7, 25), and elevated gene expression correlated with anti–PD-1 response (4); thus incorporation of these two potentially overlapping IFNγ-responsive markers can enhance sensitivity. Accordingly, we observed a cellular phenotype coexpressing HLA-DR and IDO-1, but not the myeloid marker CD11b, which showed exceptionally strong correlation with clinical response. Importantly, 3 of 13 responders in the discovery cohort with low PD-1/PD-L1 interaction scores exhibited high levels of HLA-DR and IDO-1 (Fig. 2M), suggesting an alternate mechanism of immunosuppression, potentially mediated by engagement of HLA-DR on tumor cells with cognate ligands.

Figure 4.
Patients exhibiting both adaptive resistance mechanisms, namely, the presence of PD-1/PD-L1 interaction and IDO-1/HLA-DR coexpression (dual), experience further improvement in OS compared with presence of either mechanism alone (A). Adaptive resistance mechanisms observed in this study population are depicted with supportive imagery of a patient tumor (B).
on tumor-infiltrating lymphocytes, e.g., T-cell receptor (31) or LAG3 (32, 33). Interestingly, analyses of both pretreatment patient tumors and PDX models suggest that IDO-1/HLA-DR coexpression showed a better correlation with anti-PD-1 clinical outcomes than PD-L1. We are currently exploring the mechanisms of this finding and evaluating the predictive utility of these signatures in additional tumor indications for which PD-1 blockers were approved. Based on observations that PD-L1, HLA-DR, and IDO-1 are expressed along the adaptive immune resistance continuum mediated by IFNγ (5, 6, 34), we speculate that tumors may have evolved to express two complementary mechanisms of immunosuppression, one mediated by PD-1/PD-L1 interaction and a second via HLA-DR-cognate T-cell receptor or LAG3 engagement as modeled in Fig 4B.

Developing biomarkers to accurately predict response to anti-PD-1/PD-L1 therapy is a critical and unmet need. Other candidate markers involving similar immunosuppressive pathways have been identified by alternate methodologies (3, 4, 35–37). We believe that the biomarker signatures described herein compare quite favorably with previous studies from the standpoint of predictive capacity (80% response rate in double-positive patients), logistical feasibility (2 FFPE slides), objective quantitation (automated scoring system), and visual verification. Intriguingly, excellent clinical outcomes have also been reported with agents inhibiting PD-1, IDO-1, and PD-1/LAG-3 (a cognate ligand of HLA-DR), suggesting that these biomarker signatures may be relevant for enriching patients into promising next-generation combination immunotherapy trials (6, 24).

In conclusion, we report here discovery of a highly predictive, nonsubjective tool for administering anti-PD-1 monotherapies, whose clinical performance was verified in patients with melanoma treated across ten different cancer centers akin to a real-world setting. This approach utilizes minimal number of tumor sections and provides spatial information, serving as a powerful potential diagnostic tool that could fit well into a routine pathologist’s workflow and is easily interpretable by the clinical community.

Disclosure of Potential Conflicts of Interest
D.B. Johnson is a consultant/advisory board member for Array, Bristol-Myers Squibb, and Merck, and reports receiving commercial research grants from Bristol-Myers Squibb, and Incyte. D.L. Rimm is a consultant/advisory board member for Merck, and reports receiving commercial research grants from Novartis. R.W. Joseph is a consultant/advisory board member for Bristol-Myers Squibb, Exelixis, Incyte, Insys, Merck and Novartis. R.M. Conry reports receiving speakers bureau honoraria from Amgen, Bristol-Myers Squibb, Merck and Novartis. L.F. Hernandez-Aya reports receiving speakers bureau honoraria from Regeneron and is a consultant/advisory board member for Novartis. J.S. Witte is a consultant/advisory board member for Navigate Biopharma Services. S.B. McKee reports receiving speakers bureau honoraria from Amgen. J. Lameh has ownership interests in Novartis. J. Markowitz is a consultant/advisory board member for NewLink Genetics, and reports receiving commercial research grants from Navigate Biopharma Services. J.M. Balko reports receiving commercial research grants from Genentech and Incyte Pharma. No potential conflicts of interest were disclosed by the other authors.

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Acknowledgments
This study is supported by Navigate BioPharma Services, Inc., a Novartis Subsidiary. D.B. Johnson is funded by NC/IH K23 CA24726.

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Received March 7, 2018; revised June 1, 2018; accepted July 11, 2018; published first July 18, 2018.


Clinical Cancer Research

Quantitative Spatial Profiling of PD-1/PD-L1 Interaction and HLA-DR/IDO-1 Predicts Improved Outcomes of Anti–PD-1 Therapies in Metastatic Melanoma

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Access the most recent version of this article at:
doi:10.1158/1078-0432.CCR-18-0309

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