### Original Article Association of KDR mutation with better clinical outcomes in pan-cancer for immune checkpoint inhibitors

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**Abstract:** Kinase insert domain receptor (KDR) activation is associated with the immunosuppressive microenvironment. However, the efficacy of immunotherapy in patients with *KDR* mutations is still unclear. To investigate the relationship between *KDR* gene mutations and the prognosis of pan-cancer, and whether immune checkpoint inhibitors (ICIs) may improve the prognosis of patients with *KDR* mutations, we analyzed public cohorts of pan-cancer immunotherapeutic patients including genomic and clinical data.Further analysis was performed on an internal validation data set including 67 non-small cell lung cancer. Through bioinformatics analysis, potential mechanism was studied in TCGA data. We found better responses to ICIs in patients with *KDR* mutation from pan-cancer public datasets (objective response rate [ORR], 45.0% vs 25.1%, *P*=0.0058; progression-free survival [PFS], P=0.039, HR=0.586, 95% CI 0.353-0.973) and validation cohort (overall survival (OS), P=0.05, HR=0.62; 95% CI, 0.38-1.00). Our NSCLC cohort verified the value of *KDR* mutation in predicting better clinical outcomes, including ORR (70.0% vs 22.81%, P=0.0057) and PFS (HR=0.158; 95% CI, 0.045-0.773, P=0.007). *KDR* mutation was associated with tumor mutation burden high, neoantigen burden and immune cellular activities. Meanwhile, *KDR* mutation was indicative of an immune-hot status, characterized by higher expression of PD-L1 and abundance of cytotoxic lymphocytes. *KDR* mutations may be potential positive predictors for pan-cancer received ICIs.

Keywords: KDR, pan-cancer, immune checkpoint inhibitors, biomarker, NSCLC

#### Introduction

Immune Checkpoint Inhibitors (ICIs) are transforming the landscape of treatment in cancers [1]. With ICIs initially approved for advanced melanoma (MM) [2], immunotherapy has become an important treatment for cancers, including non-small cell lung cancer (NSCLC), breast cancer (BRC), clear cell renal cell carcinoma (ccRCC), head and neck squamous cell carcinoma (HNSCC), etc [3-5].

However, ICIs are not effective for all patients. Patients with different molecular, histological, or genetic characteristics may have different response rates to ICIs. The response rate in monotherapy of ICIs is just 15%-20% [6-9]. Although some biomarkers were proven to be significant positive predictors of ICIs treatment, such as microsatellite instability (MSI), PD-L1, and tumor mutation burden (TMB), nearly half of patients could not benefit from ICIs even if the biomarker was positive [10-12]. The development of biomarkers lacks a breakthrough. The expression of PD-L1 is the most widely accepted predictor of immunotherapy. It has been included in the majority of clinical trials of immunotherapy as an important observation indicator. It is generally believed that patients with PD-L1 positive have a better response rate and prognosis during immunotherapy [13, 14]. There are still many technical limitations in PD-L1 expression detection [7, 15]. Cytoplasmic PD-L1 protein seems to have an uncertain

function in immunotherapy. The antibody used to detect of PD-L1 protein expression remains specific. Tumor specimens collected from bronchoscopy or percutaneous lung biopsy are not enough to get stable results. TMB is evolving as a potential biomarker of ICIs efficacy. It also has some limitations. Each tumor mutation was given equal weight, but not all mutations had the same efficacy to the ICIs. For example, RCC patients have promising efficacy with moderate TMB level [16, 17]. In addition, some genes mutations, such as NORTH 1/2/3 mutation, were associated with good ICIs efficacy, while JAK1/2 mutations, MDM2/4 amplification, and EGFR mutations were associated with poor efficacy [18-20]. As a biomarker, these results were not imperfect. Therefore, new biomarkers are urgently needed to screen more patients suitable for ICIs treatment.

Kinase insert domain receptor (*KDR*) encodes a key receptor that mediates tumor angiogenesis/metastasis switches [21]. Several studies have reported that VEGF-VEGFR2 axis activation is associated with the immunosuppressive microenvironment. Activation of VEGF-VEGFR2 could induce the aggregation of immature dendritic cells, bone marrow-derived suppressor cells and regulatory T cells (Tregs), and inhibit T lymphocyte migration [22, 23]. However, the clinical significance of *KDR* mutations for ICIs treated pan-cancer is unclear.

### Methods

### Clinical cohorts and TCGA cohort

Based on six published immunotherapy studies, survival and mutation dates were collected and integrated as discovery cohort for this study (<u>Supplementary Figure 1</u>) [24-29]. In the discovery cohort, samples were sequenced using whole-exome sequencing (WES) and the Memorial Sloan Kettering-Integrated Mutation Profiling of Actionable Cancer Targets (MSK-IMPACT) panel. Patients with a sample size of <10 (n=3), receiving other treatments (n=10), or unknown treatments (n=4) were excluded. 662 patients were finally included in the discovery cohort, including ccRCC (n=35), NSCLC (n=385), MM (n=205), HNSCC (n=10), and bladder cancer (BLCA) (n=27).

To verify the predictive power of *KDR* gene to ICIs treatment, we used an expanded pan-cancer cohort with overall survival (OS) and muta-

tional data as the validation cohort [30]. MSK-IMPACT panel were used for the samples from this cohort. After filtering, 1333 patients were included from eight cancer types: renal cell carcinoma (RCC) (n=146), NSCLC (n=303), MM (n=302), HNSSC (n=133), esophagogastric cancer (ESCA) (n=104), colorectal cancer (CRC) (n=106), BRC (n=39), and BLCA (n=200, Figure 1B). All the clinical and mutational dates of the discovery and the validation cohort were obtained from cBioPortal.

Sixty-seven patients from our hospital who received at least two cycles of immunotherapy (camrelizumab, nivolumab, pembrolizumab toripalimab, or tislelizumab) were included in our NSCLC cohort (**Figure 1C**). All patients had genomic profiling of tumor tissue before anti-PD-1 treatment (288-gene panel).

The clinical data, somatic mutation data, and RNA-seq data of the TCGA cohort were retrieved from the UCSC Xena data portal.

### Calculation of the TMB and NAL

The TMB was defined as the total number of non-synonymous mutations/exome size or normalized to the exonic coverage of MSK-IMPACT panels for the discovery or validation cohorts, respectively. The high TMB cutoff was the highest 20% of TMB in each cancer, as previously described [30]. NAL (neoantigen load) and TCR (T cell receptor) diversity scores were from pancancer immune landscape project [31].

### Immune infiltration estimation, leukocyte, lymphocyte and TIL fraction analyses

Cell-type identification was performed by estimating relative subgroup of RNA transcripts (CIBERSORT) based on an online method. Gene profiles was used to describe cellular composition of complex tissues [32]. Hematopoietic subsets were counted using mixed mRNA mixtures from TCGA database.

By analyzing Thorsson's data, we estimated TIL level using genomics and HE-stained in the TCGA pan-cancer cohort [31]. The leukocyte fraction data was derived from DNA methylation, and an aggregation of plasma cells, NK cells, CD8 T cells, gamma-delta T cells, Tregs, follicular helper T cells, CD4 T cells, and B cells estimated by CIBERSORT. TIL fraction analysis was based on the mappings of TILs for >5,000 HE-stained images on TCGA dataset [33].



**Figure 1.** *KDR*-MUT was associated with the clinical benefit to ICIs in the discovery cohort. A. Progression-free survival curves comparing the *KDR*-WT and *KDR*-MUT groups in patients treated with ICIs therapy from the discovery cohort. B. Overall survival curves comparing the *KDR*-WT and *KDR*-MUT groups in patients treated with ICIs therapy from the discovery cohort. C. Bar graph showing proportions of ORR in *KDR*-MUT and *KDR*-WT patients. D. Bar graph showing proportions of DCB in *KDR*-MUT and *KDR*-WT patients (Fisher's exact test). E. Overall survival curves comparing the *KDR*-WT and *KDR*-WT patients (Fisher's exact test). E. Overall survival curves comparing the *KDR*-WT and *KDR*-MUT groups in the TCGA cohort. F. Progression-free survival curves comparing the *KDR*-WT and *KDR*-MUT groups in the TCGA cohort.

### Assessment of immune signatures

Gene set enrichment analysis (GSEA) was used to associate the gene signature with *KDR*-MUT and *KDR*-WT with the "Cluster Profiler" R package. The normalized enrichment score is the main statistical data to test results of gene set enrichment. In addition, in order to study the relationship between the antitumor immunity and *KDR* mutation, we obtained 29 classical immune signatures from a previous study [34]. GSEA was used to quantitatively analyze enrichment levels of 29 immune signatures in each sample.

### Immunohistochemistry

The paraffin sections of NSCLC were taken from pathology department of our hospital. After dewaxing and rehydrating, the tumor sections were immersed into pH 6.0 citrate buffer for antigen extraction. Endogenous peroxidase was blocked with 3% H<sub>2</sub>O<sub>2</sub>, and 3% BSA was added to evenly cover the tissue at 37°C for 30 minutes, then blocking solution was gently removed. The primary antibody was added at 4°C overnight. Then, the HRP labeled secondary antibody from the corresponding species of the primary antibody was added and incubated at 37°C for 50 minutes. DAB staining, hematoxylin dying at 37°C for 3 minutes, dehydration and neutral resin sealing were performed. Anti-CD3 (Servicebio, GB11-1337), anti-CD31 (Servicebio, GB11063-1), anti-ICAM1 (Servicebio, GB11106) were used. Immunohistochemical staining was quantified by HALO image analysis software.

### Statistical analysis

All statistical analyses were conducted with R software (version 3.6.3). Fisher's exact test was used to assess the *KDR* status and the response. Log-rank test and Cox proportional hazards regression analysis were used to analyze the differences between *KDR*-MUT and *KDR*-WT in PFS and OS. The Cox proportional hazards model was used for subgroup survival

analysis and adjustments were made for available confounding. Interactions between the *KDR* status and age, sex, cancer type, TMB level, and drug class were assessed in the validation cohort. Statistical analysis of comparisons between two groups was conducted using the Wilcoxon test. *P* values were two-tailed, and <0.05 were considered statistically significant.

### Results

# Association between VEGF signaling pathway and ICIs efficacy in the discovery cohort

Considering that *KDR* is a gene in the VEGF signaling pathway, we further explored the relationships between genes in the VEGF pathway and the efficacy of immunotherapy. In the discovery cohort, eight genes involved in the VEGF signaling pathway were investigated, including *KDR*, *FLT1*, *FLT3*, *FLT4*, *VEGFC*, *VEGFA*, *VEGFD*, and *VEGFB*. Among these genes, *KDR* mutation was the only one that significantly prolonged PFS and OS (<u>Supplementary Figure 2A</u>, 2B). In addition, only *KDR* mutation was gathered in patients with ORR and DCB (<u>Supplementary</u> <u>Figure 2C</u>), indicating that *KDR* mutation may be the best biomarker for predicting the efficacy of immunotherapy.

Association between KDR-MUT and ICIs efficacy in the discovery cohort

Six publicly available studies involving 5 cancer types were consolidated into the discovery cohort. The basic characteristics were summarized in <u>Supplementary Table 1</u>. Fifty-one patients were harboring *KDR* mutations (*KDR*-MUT), accounting for 7.7% of the discovery cohort, and 611 patients were *KDR* wild-type (*KDR*-WT). 3 (5.9%) patients were confirmed harboring *p.S803Y* mutations and 3 (5.9%) patients with *p.R1032Q* mutations (<u>Supplementary Figure 3A</u>). We found that patients with *KDR*-MUT had longer PFS than patients with *KDR*-WT (median PFS: 48.91 vs 6.63 months, log-rank test P=0.003, multivariable-

adjusted P=0.039, multivariable-adjusted HR 0.586; 95% Cl, 0.353-0.973, Figure 1A). Superior OS was also observed in KDR-MUT group (median OS: not reached vs 24.08 months, log-rank test P=0.028, multivariableadjusted P=0.069, multivariable-adjusted HR= 0.595; 95% CI, 0.340-1.040, Figure 1B). After adjusted for sex, age, cancer types, drug class, and TMB level, the significant difference remained in PFS (multivariable-adjusted P= 0.039, multivariable-adjusted HR=0.586; 95% CI, 0.353-0.973), but only a numerically significant OS benefit (multivariable-adjusted P= 0.069, multivariable-adjusted HR=0.595; 95% CI, 0.340-1.040). According to RECIST 1.1, the overall response of 594 patients was evaluable, including 40 KDR-MUT patients and 554 KDR-WT patients. As expected, the ORR in patients with KDR-MUT was almost twice as higher as in patients with KDR-WT (45.0% vs 25.1%, P=0.0058, Figure 1C). As for DCB, 56% of patients with KDR-MUT were from ICIs treatment, while 41.8% of patients with KDR-WT were from ICIs (P=0.0522, Figure 1D).

Further, to assess the potential prognostic value of KDR mutations, we performed survival analyses based on KDR status in the TCGA database. There was no significant difference in OS between the KDR-WT and KDR-MUT patients treated with standard treatment (Figure 1E), and the same results across multiple cancer types were presented in Supplementary Figure 4. Although significantly worse PFS was observed in KDR-MUT patients (median PFS 49.249 vs 70.125 months, log-rank test P=0.04, HR=1.170; 95% CI, 1.01-1.36), it was no longer significant after adjusting for sex, age, and cancer types (multivariable-adjusted P=0.36, multivariable-adjusted HR=0.93; 95% CI, 0.80-1.08, Figure 1F). Taken together, KDR-MUT may potentially predict the efficacy and favorable clinical outcomes of ICIs treatment.

# Association between KDR-MUT and ICIs efficacy in the validation cohort

The basic characteristics of an expanded ICItreated cohort (n=1333) were summarized in <u>Supplementary Table 2</u>. There were 73 cases of *KDR*-MUT, including 39 MM, 11 BLCA, 9 NSCLC, 5 HNSCC, 3 ESCA, 3 RCC, 2 CRC, and 1 BRC, accounting for 5.5% of the population. In the validation cohort, 73 of 1333 patients were confirmed to have *KDR* mutations (<u>Sup-</u>

plementary Figure 2B). p.R1032Q (2.7%) was high frequency mutation site in *KDR* mutation, followed by p.P351S (2.7%), p.E469K (2.7%). Adjusted for confounding factors (sex, age, cancer types, drug class, and TMB level), KDR mutation remained an independent predictor for superior OS (median OS 46.0 vs 22.0 months, log-rank test P=0.0001, multivariable-adjusted P=0.05, multivariable-adjusted HR=0.62; 95% CI, 0.38-1.00, Figure 2A). Even compared with other oncogenes, KDR remained the most stable predictor for multiple cancers (Supplementary Table 3) [25, 26, 35-38]. In the stratification analysis of OS, KDR-MUT also had a survival advantage over KDR-WT in subgroups of age, sex, cancer type, TMB status, and drug class (Figure 2B, P>0.05).

Recently, TMB has been proven to be an effective immunotherapy biomarker [39]. Hence, to evaluate the predictive performances of TMB and KDR, patients were divided into four groups based on KDR status and TMB level. We found that *KDR*-MUT patients achieved the longest OS. Moreover, KDR mutation was associated with higher TMB (adjusted P=0.04, HR= 0.40; 95% Cl, 0.19-0.94, Figure 2C). We also explored the effect of the most frequent types of KDR mutations on TMB in discovery and validation cohort. As shown in Supplementary Figure 6, patients with specific KDR mutation sites had higher TMB than KDR-WT patients, but there was no difference between patients with different KDR mutation subtypes.

# Association between KDR-MUT and ICIs efficacy in NSCLC cohort

In our study, 67 NSCLC patients with genomic profiling were included (<u>Supplementary Table 4</u>). In NSCLC cohort, 54 of the 67 patients (80.6%) were male with a median age of 64-year-old (range, 32-80). The ORR was 29.8%. In the NSCLC cohort, the frequency of *KDR* mutations was 14.9%. And *p.R1032Q* mutation was not found in NSCLC with *KDR* mutation (<u>Supplementary Figure 3C</u>).

The genomic mutational landscape of 67 patients is displayed in **Figure 3A**. Consistent with other studies, the rate of *TP53* and *LRP-1B* mutations were higher in responders than non-responders [37, 40]. Besides these, *KDR* mutation enrichment was discovered in responders as well (**Figure 3B**). *KDR* mutation

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Figure 2. Validation of the predictive function of *KDR*-MUT. A. Overall survival curves comparing the *KDR*-WT and *KDR*-MUT groups in patients treated with ICIs therapy from the validation cohort. B. Stratification analysis of OS in the validation cohort. NSCLC, non-small cell lung cancer; SKCM, melanoma; HNSC, head and neck cancer; CRC, colorectal cancer; BLCA, bladder cancer; ESCA, esophagogastric cancer; RCC, renal cell carcinoma; BRCA, Breast invasive carcinoma. C. Overall survival curves comparing *KDR*-MUT&TMB high, *KDR*-MUT&TMB low, *KDR*-WT&TMB high, and *KDR*-WT&TMB low groups in the validation cohort.





Figure 3. KDR-MUT was associated with a better response to ICIs in the NSCLC cohort. A. Stacked plots show mutational burden (histogram, top); lines of treatment (histogram); mutations in TP53, DPYD, LRP1B, KDR, KRAS, ERBB2, SMARCA4, FAT1, ROS1, ERBB4, NF1, PIK3CA, PDGFRA, EGFR, ARID1A, FBXW7, STK11, PTEN, NOTCH1, and POLE (tile plot, middle); their mutational rates in patients having achieved objective response or progressive disease (histogram, right); and mutational marks (bottom). B. Scatter diagram displaying the mutational rate in patients having achieved objective response or progressive disease. KDR is emphasized in red. C. Histogram depicting proportions of ORR and DCR in KDR-MUT and KDR-WT patients. D. Progression-free survival curves in the patients with or without KDR-MUT.

was associated with higher ORR (70.0% vs 22.81%, P=0.0057, **Figure 3C**), longer PFS (median PFS 20 months vs 7 months, log-rank test P=0.007, HR=0.158; 95% CI, 0.045-0.773, **Figure 3D**) in NSCLC patients. These results further demonstrate the predictive function of *KDR* mutation to ICIs treatment.

### Association of KDR-MUT with enhanced immunogenicity and activated immune response

The mutational landscape of *KDR* and clinical characteristics were shown in **Figure 4A**, with MM patients (16.4%) having the highest levels of *KDR* mutations, followed by NSCLC (10.5%) and GBM (8.8%) (**Figure 4B**). Across all patients, the mutational frequencies of *KDR* was 5.0%. The somatic mutations of *KDR* were evenly distributed (**Figure 4C**, <u>Supplementary Figure 4A</u>), and there was no difference in PFS and OS with the common *KDR* mutation subtypes (<u>Supplementary Figure 7</u>).

To further explore the potential mechanisms associated with *KDR*-MUT in predicting the efficacy of ICIs. We first analyzed the correlation between *KDR* and TMB and neoantigen burden. As shown, *KDR* mutations were associated with higher TMB in the TCGA, discovery and validation cohorts, as well as significantly higher predicted neoantigens in the TCGA cohort (**Figure 5A**; <u>Supplementary Figure 3A</u>), suggesting that the *KDR* positive patients may have a higher immune reaction to tumor neoantigens.

GSEA revealed the significant enrichment of antigen processing and presentation pathway in the KDR-MUT group, along with other immune activation-related pathways, including T-cell receptor signaling, NK cell-mediated immunity, CD8+ T cell activation, and IFNgamma response (Figure 5B). In addition, KDR-MUT tumors had higher expression of MHC I- and II-associated antigen-presenting molecules than KDR-WT tumors (Figure 5C). Antigen-presenting cells (APCs) have long been known to present tumor-associated antigens to T cells, which elicited tumor-specific immune responses. Immune evasion associated with in antigen presentation to T lymphocytes defects is a common phenomenon and highlights the relevance of T cells in solid tumors rejection. These results attenuate this effect in KDR-MUT tumors, which may make tumor cells susceptible to ICIs.

Following antigen processing and presentation by APC, two signals are essential to activate the immune response against cancer cells: peptide-MHC complexes occupy TCRs and subsequent activate the costimulatory molecules [41]. The host immune system needs to maintain a diverse TCR repertoire to recognize multiple tumor neoantigens [42]. We found the TCR diversity and costimulatory molecules were significantly higher in *KDR*-MUT tumors than *KDR*-WT tumors (**Figure 5C** and **5D**). Taken together, these results suggested that *KDR*-MUT was associated with enhanced immunogenicity and activation of tumor neoantigens immune response.

Notably, we also found that angiogenesis and VEGFR signaling was down-regulated in KDR-MUT tumors than KDR-WT tumors across multiple cancer types (Figure 5B and 5E). VEGFR2 activation promotes endothelial cell proliferation, migration. Consistently, significantly decreased mRNA expression levels of vascular epithelial cell migration and proliferationrelated genes were observed (Supplementary Figure 5B). Abnormal angiogenesis was closely associated with immunosuppressive and is characterized by T cell infiltration overcoming higher interstitial fluid pressure, and down-regulated vasculature cell adhesion molecule-1 (VCAM-1) and intracellular cell adhesion molecule-1 (ICAM-1) to impair the T cell extravasation [43]. As shown in Figure 5E and 5F, vascular permeability and expression of ICAM1 were decreased and increased in KDR-MUT patients, respectively. Decreased tumor vessel density labeled by CD31 and increased expression of ICAM1 were further confirmed by immunohistochemistry (Figure 5G). These results implied that KDR-MUT was associated with the inhibition of angiogenesis, which could promote immune cell infiltration.

## KDR-MUT was indicative of an immune-hot status

Hot tumor microenvironment (TME) characterized by the presence of tumor-infiltrating lymphocytes and PD-L1 expression is associated with increased response to anti-PD-1/L1 monotherapy [44, 45]. By using CIBERSORT, we found that activated NK cells and CD8+ T cells were more abundant in *KDR*-MUT tumors. Besides, MO and M1 macrophages infiltration increased, while the M2 macrophages infiltra-



**Figure 4.** The pan-cancer mutational landscape of *KDR* in TCGA cohort. A. Association between *KDR* status and annotated clinical characteristics in TCGA cohort. (cancer type, sex, age, CNA, TMB, PFS, and OS were annotated. Samples were sorted by *KDR* status, while *KDR*-MUT and *KDR*-WT samples were separated by a gap. B. The proportion of *KDR*-MUT tumors identified for each cancer type with alteration frequency above 1%. C. Lollipop plot showing the loci distribution of mutations across the *KDR* altered patient cohorts from the TCGA database. Truncating mutations included nonsense, nonstop, splice site mutations, and frameshift insertion and deletion; Non truncating mutations included missense mutations and inframe insertion and deletion.

tion decreased (Figure 6A). To cross-examine the above results with different immune cells assessment methods, we next analyzed the expression profiles of 29 immune signatures. As shown in Figure 6B, KDR-MUT tumors have rich immune signatures and microenvironment cell populations, such as dendritic cells (DCs) and cytotoxic lymphocytes. Then, using three different methods, we further verified the abundance of immune cells in KDR-MUT tumors. First, the larger leukocyte fraction in KDR-MUT tumors was assessed based on DNA methylation arrays (Figure 6C). Next, we obtained a similar high TIL results in KDR-MUT tumors based on TIL score estimated from HE-stained slides (Figure 6D) [33], which was consistent with the results of lymphocyte fraction estimated by the CIBERSORT method (Figure 6E). Importantly, CD3 immunohistochemical results further confirmed the presence of tumor-infiltrating T cells in KDR-MUT tumors (Figure 6F).

Some chemokines (such as CXCL9, CXCL10, and CCL5) were also up-regulated in the *KDR*-MUT tumors, and these chemokines have been shown to attract CD8 T cells and DCs [46, 47]. Additionally, the expression of genes related with cytotoxic activity (such as *GZMA*, *PRF1*), the surrogate measures of cytotoxic T lymphocyte (CTL) activity, was higher in *KDR*-MUT tumors, indicating enhanced tumor-killing capacity. Moreover, PD-L1 (CD274) and CTLA4 were up-regulated in *KDR*-MUT tumors (**Figure 6G**). Both of these results demonstrate *KDR*-MUT was associated with a hot TME and enhanced the ICIs efficacy.

### Discussion

In this work, *KDR*-MUT were identified for the first time as a positive factor for better clinical benefit in pan-cancer patients treated with ICIs, particularly NSCLC. However, patients receiving standard care did not receive the clinical benefits of OS. In the exploratory analyses, higher TMB, higher neoantigen burden, and reduced expression of genes associated with *VEGFR* pathway activation may be potential

mechanism for predicting *KDR* mutation in pan-cancer (Supplementary Figure 8). GSEA revealed prominent enrichment of signatures related to antigen processing and presentation in patients with *KDR*-MUT. Using CIBERSORT, we found cytotoxic lymphocytes were more abundant in *KDR*-MUT tumors. GSEA showed that *KDR*-MUT tumors had rich immune signatures and microenvironment cell populations, further proving that *KDR*-MUT was associated with a hot tumor microenvironment. These results suggest that *KDR*-MUT may be a potentially positive predictor of pan-cancer patients treated with ICIs.

In recent years, more and more immunotherapy treatments has been applied to clinical practice. How to select patients who can benefit from immunotherapy has become a thorny clinical problem. Some specific mutations have been proved to influence the response to immunotherapy. It was reported that MMR, POLE, and POLD1 were related to the response to immunotherapy [27, 28, 37-40]. These genes seem to lack powerful clinical evidence. Our study demonstrated that KDR-MUT has the potential to become a biomarker for ICIs. We showed the association between KDR gene status and the responses of immunotherapy in three independent cohorts. Importantly, in our own NSCLC cohort, we excluded some confounding factors, such as the course of treatment which were not covered in the public cohort. It can reduce the statistical bias to some extent and make the result more reliable. KDR mutation is not so rare in tumors, accounting for about 5.0% of the pan-cancer cohort, with MM ranking first (16.4%) followed by NSCLC (10.5%) and GBM (8.8%). This means that KDR-MUT could serve more patients than other rare mutants. Unlike PD-L1 expression or TMB, KDR-MUT can be easily detected by next-generation sequencing. KDR-MUT is also included in most commercially available gene panels.

*KDR* is a type 2 vascular endothelial growth factor receptor that plays an important role in



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**Figure 5.** *KDR*-MUT was associated with enhanced tumor immunogenicity and anti-tumor immunity. A. Comparison of the TMB and NAL levels between the *KDR*-MUT and *KDR*-WT tumors in the TCGA cohort. B. Significantly enriched pathways with GSEA between *KDR*-MUT and *KDR*-WT tumors in the TCGA dataset. C. Comparison of the MHC molecules and co-stimulators expression levels between *KDR*-WT and *KDR*-MUT tumors in the TCGA cohort. D. Comparison of the expression of TCR richness between *KDR*-WT and *KDR*-WT and *KDR*-WT and *KDR*-WT and *KDR*-MUT tumors. E. Heatmap shows clustering of tumor types based on angiogenesis and APC correlated gene sets. The heatmap is colored by the normalized enrichment score of a gene set for a tumor type. F. Comparison of the adhesion molecules expression levels in *KDR*-MUT and *KDR*-WT groups. G. Quantitative immunohistochemistry (IHC) analysis of CD31 and ICAM1 protein expression in the *KDR*-MUT and *KDR*-WT groups (n=4/group). Representative micrographs of these samples are shown. Data represent the mean ± SD. (Mann-Whitney U test; ns, not significant; \*P<0.05, \*\*P<0.01, \*\*\*\*P<0.001).





**Figure 6.** *KDR*-MUT was indicative of an immune-hot status. A. Comparison of the 22 immune cells infiltration levels in *KDR*-MUT and *KDR*-WT tumors. CIBERSORT was used to calculate the infiltration degree of these immune cells. B. Volcano plots of 29 immune signatures estimated by the GSEA method. Immune signatures enriched in *KDR*-MUT tumors and *KDR*-WT tumors are marked in red and blue, respectively. C. Comparison of the leukocyte fractions between *KDR*-WT and *KDR*-MUT tumors. D. Comparison of the TIL regional fractions between *KDR*-WT and *KDR*-MUT tumors. E. Comparison of the lymphocyte fractions between *KDR*-WT and *KDR*-MUT tumors. F. Quantitative immunohistochemistry (IHC) analysis of CD3 protein expression in the *KDR*-MUT and *KDR*-WT groups (n=4/group). Representative micrographs of these samples are shown. Data represent the mean ± SD. G. Boxplot comparing the immune-related genes expression levels in *KDR*-MUT and *KDR*-WT groups (Mann-Whitney U test; ns, not significant; \*P<0.05, \*\*P<0.01, \*\*\*P<0.001.

the regulation of angiogenesis and vascular integrity. However, a possible functional association between PD-1/PD-L1 and VEGF/VEGFR has been reported. In patients with ccRCC and classical Hodgkin lymphoma addition, PD-L1 expression was associated with VEGF and microvessel density [48, 49]. Our study showed similar results, with significantly higher TMB values and PD-L1 expression associated with KDR-MUT compared to KDR-WT. It produced more neoantigens, which were processed by APCs and presented to T cells. Up-regulated MHC-related molecules and TCR diversity activated this process. Meanwhile, chemokines contributed to the invasion of CD8 T cells and DCs into the tumor tissues. CTL released GZMA and PRF1 and then enhanced tumor-killing capacity. This is one of the reasons why we believe that KDR nonsynonymous mutations are associated with good clinical outcomes in ICIs.

To our knowledge, this is the first time for exploring the relationship between KDR-MUT and ICIs in pan-cancer. Some limitations still existed in our study including these inherent to a retrospective design. The analysis was based on a universal carcinomatous public cohort that received WES or panel sequencing, which may have resulted in selection bias. However, our own independent cohort of NSCLC was included in the analysis as a strong complement. Due to the limited number of patients, the relationship between KDR-MUT and the efficacy of immunotherapy cannot be further explored, but the overall efficacy of KDR-MUT has been satisfactory. These results should be confirmed in large cohorts. In addition, the potential mechanisms by which KDR mutation enhance the efficacy of immunotherapy were explored only through bioinformatics analysis. Our conclusion is only that KDR mutation is related to the immune hot environment, but the specific mechanism how KDR-MUT caused the immune alterations is worthy of further experimental study.

In conclusion, our results suggested that *KDR*-MUT was associated with better PFS and OS in pan-cancer patients who received ICIs. *KDR*-MUT might be an important indicator of the immunogenetic landscape, and should be considered as a therapeutic biomarker for immunotherapy.

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### Disclosure of conflict of interest

#### None.

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### References

- [1] Johnson DB, Sullivan RJ and Menzies AM. Immune checkpoint inhibitors in challenging populations. Cancer 2017; 123: 1904-1911.
- [2] Billan S, Kaidar-Person O and Gil Z. Treatment after progression in the era of immunotherapy. Lancet Oncol 2020; 21: e463-e476.
- [3] Rassy EE, Khoury Abboud RM, Ibrahim N, Assi T, Aoun F and Kattan J. The current state of immune checkpoint inhibitors in the first-line treatment of renal cancer. Immunotherapy 2018; 10: 1047-1052.
- [4] Zhang Z, Xie T, Zhang X, Qi C, Shen L and Peng Z. Immune checkpoint inhibitors for treatment of advanced gastric or gastroesophageal junction cancer: current evidence and future perspectives. Chin J Cancer Res 2020; 32: 287-302.
- [5] Zam W and Ali L. Immune checkpoint inhibitors in the treatment of cancer. Curr Rev Clin Exp Pharmacol 2022; 17: 103-113.
- [6] Borghaei H, Paz-Ares L, Horn L, Spigel DR, Steins M, Ready NE, Chow LQ, Vokes EE, Felip E, Holgado E, Barlesi F, Kohlhäufl M, Arrieta O, Burgio MA, Fayette J, Lena H, Poddubskaya E, Gerber DE, Gettinger SN, Rudin CM, Rizvi N, Crinò L, Blumenschein GR Jr, Antonia SJ, Dorange C, Harbison CT, Graf Finckenstein F and Brahmer JR. Nivolumab versus docetaxel in advanced nonsquamous non-small-cell lung cancer. N Engl J Med 2015; 373: 1627-39.

- [7] Ikeda S, Goodman AM, Cohen PR, Jensen TJ, Ellison CK, Frampton G, Miller V, Patel SP and Kurzrock R. Metastatic basal cell carcinoma with amplification of PD-L1: exceptional response to anti-PD1 therapy. NPJ Genom Med 2016; 1: 16037.
- [8] Shitara K, Özgüroğlu M, Bang YJ, Di Bartolomeo M, Mandalà M, Ryu MH, Fornaro L, Olesiński T, Caglevic C, Chung HC, Muro K, Goekkurt E, Mansoor W, McDermott RS, Shacham-Shmueli E, Chen X, Mayo C, Kang SP, Ohtsu A and Fuchs CS; KEYNOTE-061 investigators. Pembrolizumab versus paclitaxel for previously treated, advanced gastric or gastrooesophageal junction cancer (KEYNOTE-061): a randomised, open-label, controlled, phase 3 trial. Lancet 2018; 392: 123-133.
- [9] Jardim DL, Goodman A, de Melo Gagliato D and Kurzrock R. The challenges of tumor mutational burden as an immunotherapy biomarker. Cancer Cell 2021; 39: 154-173.
- [10] Mok TSK, Wu YL, Kudaba I, Kowalski DM, Cho BC, Turna HZ, Castro G Jr, Srimuninnimit V, Laktionov KK, Bondarenko I, Kubota K, Lubiniecki GM, Zhang J, Kush D and Lopes G; KEYNOTE-042 Investigators. Pembrolizumab versus chemotherapy for previously untreated, PD-L1-expressing, locally advanced or metastatic non-small-cell lung cancer (KEY-NOTE-042): a randomised, open-label, controlled, phase 3 trial. Lancet 2019; 393: 1819-1830.
- [11] André T, Shiu KK, Kim TW, Jensen BV, Jensen LH, Punt C, Smith D, Garcia-Carbonero R, Benavides M, Gibbs P, de la Fouchardiere C, Rivera F, Elez E, Bendell J, Le DT, Yoshino T, Van Cutsem E, Yang P, Farooqui MZH, Marinello P and Diaz LA Jr; KEYNOTE-177 Investigators. Pembrolizumab in microsatellite-instability-high advanced colorectal cancer. N Engl J Med 2020; 383: 2207-2218.
- [12] Carbone DP, Reck M, Paz-Ares L, Creelan B, Horn L, Steins M, Felip E, van den Heuvel MM, Ciuleanu TE, Badin F, Ready N, Hiltermann TJN, Nair S, Juergens R, Peters S, Minenza E, Wrangle JM, Rodriguez-Abreu D, Borghaei H, Blumenschein GR Jr, Villaruz LC, Havel L, Krejci J, Corral Jaime J, Chang H, Geese WJ, Bhagavatheeswaran P, Chen AC and Socinski MA; CheckMate 026 Investigators. First-line nivolumab in stage IV or recurrent non-smallcell lung cancer. N Engl J Med 2017; 376: 2415-2426.
- [13] Reck M, Rodríguez-Abreu D, Robinson AG, Hui R, Csőszi T, Fülöp A, Gottfried M, Peled N, Tafreshi A, Cuffe S, O'Brien M, Rao S, Hotta K, Leiby MA, Lubiniecki GM, Shentu Y, Rangwala R and Brahmer JR; KEYNOTE-024 Investigators. Pembrolizumab versus chemotherapy for PD-L1-positive non-small-cell lung cancer. N Engl J Med 2016; 375: 1823-1833.

- [14] Balar AV, Castellano D, O'Donnell PH, Grivas P, Vuky J, Powles T, Plimack ER, Hahn NM, de Wit R, Pang L, Savage MJ, Perini RF, Keefe SM, Bajorin D and Bellmunt J. First-line pembrolizumab in cisplatin-ineligible patients with locally advanced and unresectable or metastatic urothelial cancer (KEYNOTE-052): a multicentre, single-arm, phase 2 study. Lancet Oncol 2017; 18: 1483-1492.
- [15] Zhou ZQ, Zhao JJ, Pan QZ, Chen CL, Liu Y, Tang Y, Zhu Q, Weng DS and Xia JC. PD-L1 expression is a predictive biomarker for CIK cellbased immunotherapy in postoperative patients with breast cancer. J Immunother Cancer 2019; 7: 228.
- [16] Gunjur A. Nivolumab plus ipilimumab in advanced renal-cell carcinoma. Lancet Oncol 2018; 19: e232.
- [17] Motzer RJ, Escudier B, McDermott DF, George S, Hammers HJ, Srinivas S, Tykodi SS, Sosman JA, Procopio G, Plimack ER, Castellano D, Choueiri TK, Gurney H, Donskov F, Bono P, Wagstaff J, Gauler TC, Ueda T, Tomita Y, Schutz FA, Kollmannsberger C, Larkin J, Ravaud A, Simon JS, Xu LA, Waxman IM and Sharma P; CheckMate 025 Investigators. Nivolumab versus everolimus in advanced renal-cell carcinoma. N Engl J Med 2015; 373: 1803-13.
- [18] Zhang K, Hong X, Song Z, Xu Y, Li C, Wang G, Zhang Y, Zhao X, Zhao Z, Zhao J, Huang M, Huang D, Qi C, Gao C, Cai S, Gu F, Hu Y, Xu C, Wang W, Lou Z, Zhang Y and Liu L. Identification of deleterious NOTCH mutation as novel predictor to efficacious immunotherapy in NSCLC. Clin Cancer Res 2020; 26: 3649-3661.
- [19] Shin DS, Zaretsky JM, Escuin-Ordinas H, Garcia-Diaz A, Hu-Lieskovan S, Kalbasi A, Grasso CS, Hugo W, Sandoval S, Torrejon DY, Palaskas N, Rodriguez GA, Parisi G, Azhdam A, Chmielowski B, Cherry G, Seja E, Berent-Maoz B, Shintaku IP, Le DT, Pardoll DM, Diaz LA Jr, Tumeh PC, Graeber TG, Lo RS, Comin-Anduix B and Ribas A. Primary resistance to PD-1 blockade mediated by JAK1/2 mutations. Cancer Discov 2017; 7: 188-201.
- [20] Kato S, Goodman A, Walavalkar V, Barkauskas DA, Sharabi A and Kurzrock R. Hyperprogressors after immunotherapy: analysis of genomic alterations associated with accelerated growth rate. Clin Cancer Res 2017; 23: 4242-4250.
- [21] Peng FW, Liu DK, Zhang QW, Xu YG and Shi L. VEGFR-2 inhibitors and the therapeutic applications thereof: a patent review (2012-2016). Expert Opin Ther Pat 2017; 27: 987-1004.
- [22] Voron T, Colussi O, Marcheteau E, Pernot S, Nizard M, Pointet AL, Latreche S, Bergaya S, Benhamouda N, Tanchot C, Stockmann C, Combe P, Berger A, Zinzindohoue F, Yagita H, Tartour E, Taieb J and Terme M. VEGF-A modu-

lates expression of inhibitory checkpoints on CD8+ T cells in tumors. J Exp Med 2015; 212: 139-48.

- [23] Terme M, Pernot S, Marcheteau E, Sandoval F, Benhamouda N, Colussi O, Dubreuil O, Carpentier AF, Tartour E and Taieb J. VEGFA-VEGFR pathway blockade inhibits tumor-induced regulatory T-cell proliferation in colorectal cancer. Cancer Res 2013; 73: 539-49.
- [24] Miao D, Margolis CA, Vokes NI, Liu D, Taylor-Weiner A, Wankowicz SM, Adeegbe D, Keliher D, Schilling B, Tracy A, Manos M, Chau NG, Hanna GJ, Polak P, Rodig SJ, Signoretti S, Sholl LM, Engelman JA, Getz G, Jänne PA, Haddad RI, Choueiri TK, Barbie DA, Haq R, Awad MM, Schadendorf D, Hodi FS, Bellmunt J, Wong KK, Hammerman P and Van Allen EM. Genomic correlates of response to immune checkpoint blockade in microsatellite-stable solid tumors. Nat Genet 2018; 50: 1271-1281.
- [25] Miao D, Margolis CA, Gao W, Voss MH, Li W, Martini DJ, Norton C, Bossé D, Wankowicz SM, Cullen D, Horak C, Wind-Rotolo M, Tracy A, Giannakis M, Hodi FS, Drake CG, Ball MW, Allaf ME, Snyder A, Hellmann MD, Ho T, Motzer RJ, Signoretti S, Kaelin WG Jr, Choueiri TK and Van Allen EM. Genomic correlates of response to immune checkpoint therapies in clear cell renal cell carcinoma. Science 2018; 359: 801-806.
- [26] Rizvi NA, Hellmann MD, Snyder A, Kvistborg P, Makarov V, Havel JJ, Lee W, Yuan J, Wong P, Ho TS, Miller ML, Rekhtman N, Moreira AL, Ibrahim F, Bruggeman C, Gasmi B, Zappasodi R, Maeda Y, Sander C, Garon EB, Merghoub T, Wolchok JD, Schumacher TN and Chan TA. Cancer immunology. Mutational landscape determines sensitivity to PD-1 blockade in nonsmall cell lung cancer. Science 2015; 348: 124-8.
- [27] Hellmann MD, Nathanson T, Rizvi H, Creelan BC, Sanchez-Vega F, Ahuja A, Ni A, Novik JB, Mangarin LMB, Abu-Akeel M, Liu C, Sauter JL, Rekhtman N, Chang E, Callahan MK, Chaft JE, Voss MH, Tenet M, Li XM, Covello K, Renninger A, Vitazka P, Geese WJ, Borghaei H, Rudin CM, Antonia SJ, Swanton C, Hammerbacher J, Merghoub T, McGranahan N, Snyder A and Wolchok JD. Genomic features of response to combination immunotherapy in patients with advanced non-small-cell lung cancer. Cancer Cell 2018; 33: 843-852.
- [28] Snyder A, Makarov V, Merghoub T, Yuan J, Zaretsky JM, Desrichard A, Walsh LA, Postow MA, Wong P, Ho TS, Hollmann TJ, Bruggeman C, Kannan K, Li Y, Elipenahli C, Liu C, Harbison CT, Wang L, Ribas A, Wolchok JD and Chan TA. Genetic basis for clinical response to CTLA-4

blockade in melanoma. N Engl J Med 2014; 371: 2189-2199.

- [29] Rizvi H, Sanchez-Vega F, La K, Chatila W, Jonsson P, Halpenny D, Plodkowski A, Long N, Sauter JL, Rekhtman N, Hollmann T, Schalper KA, Gainor JF, Shen R, Ni A, Arbour KC, Merghoub T, Wolchok J, Snyder A, Chaft JE, Kris MG, Rudin CM, Socci ND, Berger MF, Taylor BS, Zehir A, Solit DB, Arcila ME, Ladanyi M, Riely GJ, Schultz N and Hellmann MD. Molecular determinants of response to anti-programmed cell death (PD)-1 and anti-programmed death-ligand 1 (PD-L1) blockade in patients with nonsmall-cell lung cancer profiled with targeted next-generation sequencing. J Clin Oncol 2018; 36: 633-641.
- [30] Samstein RM, Lee CH, Shoushtari AN, Hellmann MD, Shen R, Janjigian YY, Barron DA, Zehir A, Jordan EJ, Omuro A, Kaley TJ, Kendall SM, Motzer RJ, Hakimi AA, Voss MH, Russo P, Rosenberg J, Iyer G, Bochner BH, Bajorin DF, Al-Ahmadie HA, Chaft JE, Rudin CM, Riely GJ, Baxi S, Ho AL, Wong RJ, Pfister DG, Wolchok JD. Barker CA. Gutin PH. Brennan CW. Tabar V. Mellinghoff IK, DeAngelis LM, Ariyan CE, Lee N, Tap WD, Gounder MM, D'Angelo SP, Saltz L, Stadler ZK, Scher HI, Baselga J, Razavi P, Klebanoff CA, Yaeger R, Segal NH, Ku GY, DeMatteo RP, Ladanyi M, Rizvi NA, Berger MF, Riaz N, Solit DB, Chan TA and Morris LGT. Tumor mutational load predicts survival after immunotherapy across multiple cancer types. Nat Genet 2019; 51: 202-206.
- [31] Thorsson V, Gibbs DL, Brown SD, Wolf D, Bortone DS, Ou Yang TH, Porta-Pardo E, Gao GF, Plaisier CL, Eddy JA, Ziv E, Culhane AC, Paull EO, Sivakumar IKA, Gentles AJ, Malhotra R, Farshidfar F, Colaprico A, Parker JS, Mose LE, Vo NS, Liu J, Liu Y, Rader J, Dhankani V, Reynolds SM, Bowlby R, Califano A, Cherniack AD, Anastassiou D, Bedognetti D, Mokrab Y, Newman AM, Rao A, Chen K, Krasnitz A, Hu H, Malta TM, Noushmehr H, Pedamallu CS, Bullman S, Ojesina AI, Lamb A, Zhou W, Shen H, Choueiri TK, Weinstein JN, Guinney J, Saltz J, Holt RA, Rabkin CS; Cancer Genome Atlas Research Network, Lazar AJ, Serody JS, Demicco EG, Disis ML, Vincent BG and Shmulevich I. The immune landscape of cancer. Immunity 2018; 48: 812-830.
- [32] Newman AM, Liu CL, Green MR, Gentles AJ, Feng W, Xu Y, Hoang CD, Diehn M and Alizadeh AA. Robust enumeration of cell subsets from tissue expression profiles. Nat Methods 2015; 12: 453-7.
- [33] Saltz J, Gupta R, Hou L, Kurc T, Singh P, Nguyen V, Samaras D, Shroyer KR, Zhao T, Batiste R, Van Arnam J; Cancer Genome Atlas Research

Network, Shmulevich I, Rao AUK, Lazar AJ, Sharma A and Thorsson V. Spatial organization and molecular correlation of tumor-infiltrating lymphocytes using deep learning on pathology images. Cell Rep 2018; 23: 181-193.

- [34] He Y, Jiang Z, Chen C and Wang X. Classification of triple-negative breast cancers based on immunogenomic profiling. J Exp Clin Cancer Res 2018; 37: 327.
- [35] Haradhvala NJ, Kim J, Maruvka YE, Polak P, Rosebrock D, Livitz D, Hess JM, Leshchiner I, Kamburov A, Mouw KW, Lawrence MS and Getz G. Distinct mutational signatures characterize concurrent loss of polymerase proofreading and mismatch repair. Nat Commun 2018; 9: 1746.
- [36] Mehnert JM, Panda A, Zhong H, Hirshfield K, Damare S, Lane K, Sokol L, Stein MN, Rodriguez-Rodriquez L, Kaufman HL, Ali S, Ross JS, Pavlick DC, Bhanot G, White EP, DiPaola RS, Lovell A, Cheng J and Ganesan S. Immune activation and response to pembrolizumab in POLE-mutant endometrial cancer. J Clin Invest 2016; 126: 2334-40.
- [37] Dong ZY, Zhong WZ, Zhang XC, Su J, Xie Z, Liu SY, Tu HY, Chen HJ, Sun YL, Zhou Q, Yang JJ, Yang XN, Lin JX, Yan HH, Zhai HR, Yan LX, Liao RQ, Wu SP and Wu YL. Potential predictive value of TP53 and KRAS mutation status for response to PD-1 blockade immunotherapy in lung adenocarcinoma. Clin Cancer Res 2017; 23: 3012-3024.
- [38] Wu YM, Cieślik M, Lonigro RJ, Vats P, Reimers MA, Cao X, Ning Y, Wang L, Kunju LP, de Sarkar N, Heath EI, Chou J, Feng FY, Nelson PS, de Bono JS, Zou W, Montgomery B, Alva A, Robinson DR and Chinnaiyan AM; PCF/SU2C International Prostate Cancer Dream Team. Inactivation of CDK12 delineates a distinct immunogenic class of advanced prostate cancer. Cell 2018; 173: 1770-1782.
- [39] Cui Y, Chen H, Xi R, Cui H, Zhao Y, Xu E, Yan T, Lu X, Huang F, Kong P, Li Y, Zhu X, Wang J, Zhu W, Wang J, Ma Y, Zhou Y, Guo S, Zhang L, Liu Y, Wang B, Xi Y, Sun R, Yu X, Zhai Y, Wang F, Yang J, Yang B, Cheng C, Liu J, Song B, Li H, Wang Y, Zhang Y, Cheng X, Zhan Q, Li Y and Liu Z. Whole-genome sequencing of 508 patients identifies key molecular features associated with poor prognosis in esophageal squamous cell carcinoma. Cell Res 2020; 30: 902-913.
- [40] Chen H, Chong W, Wu Q, Yao Y, Mao M and Wang X. Association of LRP1B mutation with tumor mutation burden and outcomes in melanoma and non-small cell lung cancer patients treated with immune check-point blockades. Front Immunol 2019; 10: 1113.

- [41] Pahari S, Negi S, Aqdas M, Arnett E, Schlesinger LS and Agrewala JN. Induction of autophagy through CLEC4E in combination with TLR4: an innovative strategy to restrict the survival of Mycobacterium tuberculosis. Autophagy 2020; 16: 1021-1043.
- [42] Riaz N, Havel JJ, Makarov V, Desrichard A, Urba WJ, Sims JS, Hodi FS, Martín-Algarra S, Mandal R, Sharfman WH, Bhatia S, Hwu WJ, Gajewski TF, Slingluff CL Jr, Chowell D, Kendall SM, Chang H, Shah R, Kuo F, Morris LGT, Sidhom JW, Schneck JP, Horak CE, Weinhold N and Chan TA. Tumor and microenvironment evolution during immunotherapy with nivolumab. Cell 2017; 171: 934-949.
- [43] Teng MW, Ngiow SF, Ribas A and Smyth MJ. Classifying cancers based on T-cell infiltration and PD-L1. Cancer Res 2015; 75: 2139-45.
- [44] Galon J and Bruni D. Approaches to treat immune hot, altered and cold tumours with combination immunotherapies. Nat Rev Drug Discov 2019; 18: 197-218.
- [45] Hegde PS, Karanikas V and Evers S. The where, the when, and the how of immune monitoring for cancer immunotherapies in the era of checkpoint inhibition. Clin Cancer Res 2016; 22: 1865-74.
- [46] Harlin H, Meng Y, Peterson AC, Zha Y, Tretiakova M, Slingluff C, McKee M and Gajewski TF. Chemokine expression in melanoma metastases associated with CD8+ T-cell recruitment. Cancer Res 2009; 69: 3077-85.
- [47] Bergamaschi C, Pandit H, Nagy BA, Stellas D, Jensen SM, Bear J, Cam M, Valentin A, Fox BA, Felber BK and Pavlakis GN. Heterodimeric IL-15 delays tumor growth and promotes intratumoral CTL and dendritic cell accumulation by a cytokine network involving XCL1, IFN-γ, CXCL9 and CXCL10. J Immunother Cancer 2020; 8: e000599.
- [48] Shin SJ, Jeon YK, Kim PJ, Cho YM, Koh J, Chung DH and Go H. Clinicopathologic analysis of PD-L1 and PD-L2 expression in renal cell carcinoma: association with oncogenic proteins status. Ann Surg Oncol 2016; 23: 694-702.
- [49] Koh YW, Han JH, Yoon DH, Suh C and Huh J. PD-L1 expression correlates with VEGF and microvessel density in patients with uniformly treated classical Hodgkin lymphoma. Ann Hematol 2017; 96: 1883-1890.



**Supplementary Figure 1.** Flowchart of the study design. A. Consolidation of the discovery cohort from six published studies. B. Consolidation of the validation cohort. C. Inclusion criteria of NSCLC cohort.



**Supplementary Figure 2.** Survival analysis of eight genes involved in the VEGF signaling pathway. A. PFS of the eight genes involved in the VEGF signaling pathway in the discovery cohort. B. OS of the eight genes involved in the VEGF signaling pathway in the discovery cohort. C. Associations between gene mutation and clinical responses (ORR and DCB). Both dashed lines indicated P=0.05 regarding DCB and ORR, respectively (two-tailed Fisher's exact test).

Characteristics	No (%)	KDR status [No. (%)] <sup>a</sup>		
Characteristics	NO. (%)	KDR-MUT	KDR-WT	
No. of patients	662	51 (7.7)	611 (93.2)	
Median age, years (range)	63 (18-92)	64 (36-86)	63 (18-92)	
Age				
<60	211	15 (7.1)	196 (92.9)	
≥60	451	36 (8.0)	415 (92.0)	
Gender				
Male	370 (55.9)	32 (8.6)	338 (91.4)	
Female	292 (44.1)	19 (6.5)	273 (93.5)	
Cancer type				
BLCA	27 (4.1)	0(0)	27 (100)	
HNSC	10 (1.5)	0 (0)	10 (100)	
MM	205 (31.0)	31 (15.1)	174 (84.9)	
RCC	35 (5.3)	1 (2.9)	34 (97.1)	
NSCLC	385 (58.1)	19 (4.9)	366 (95.1)	
Drug				
PD-(L)1	345 (52.1)	12 (3.5)	333 (96.5)	
CTLA-4	199 (30.1)	31 (15.6)	168 (84.4)	
PD-(L)1+CTLA-4	118 (17.8)	8 (6.8)	110 (93.2)	
RECIST				
CR	26 (3.9)	4 (15.4)	22 (84.6)	
PR	131 (19.8)	14 (10.7)	117 (89.3)	
SD	182 (27.5)	11 (6.0)	171 (94.0)	
PD	255 (38.5)	11 (4.3)	244 (95.7)	
NE <sup>b</sup>	68 (10.3)	11 (16.2)	57 (83.8)	
Durable clinical benefit				
DCB	277 (41.8)	28 (10.1)	249 (89.9)	
NDB	368 (55.6)	22 (6.0)	346 (94.0)	
NE°	17 (2.6)	1 (5.9)	16 (94.1)	

Supplementary Table 1. Patie	nt characteristics in the discovery	y cohort stratified by KDR status
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<sup>a</sup>Indicated percentage of *KDR*-MUT or *KDR*-WT patients in a given category (i.e. specific gender, specific age group). <sup>b</sup>Sixty-eight patients with RECIST not evaluable due to missing data. <sup>c</sup>Seventeen patients with durable clinical benefit not evaluable. Abbreviations: BLCA, Bladder cancer; CR, Complete response; CTLA-4, Cytotoxic T-cell lymphocyte-4; DCB, Durable clinical benefit; HNSC, Head and neck cancer; MM, Melanoma; NDB, No durable benefit; NE, Not evaluable; NSCLC, Non-small cell lung cancer; PD, Progressive disease; PD-(L)1, Programmed cell death-1 or programmed death-ligand 1; PR, Partial response; RCC, Renal cell carcinoma; SD, Stable disease.



**Supplementary Figure 3.** Pie charts of patients with *KDR* mutations. Pie charts showing the proportions of *KDR* mutation (*KDR*-MUT) and wild-type (*KDR*-WT) tumors in the discovery cohort (A), validation cohort (B) and NSCLC cohort (C).



**Supplementary Figure 4.** Survival analysis of cancer subgroups in the TCGA cohort. A. PFS of cancer subgroups in the TCGA cohort. B. OS of cancer subgroups in the TCGA cohort.

	NL- (0/)	KDR status [No. (%)] <sup>a</sup>		
Characteristics	NO. (%)	KDR-MUT	KDR-WT	
No. of patients	1333 (100)	73 (5.5)	1260 (94.5)	
Median age, years (range)	62 (16-90)	66 (31-87)	62 (16-90)	
Age				
<60	497 (37.3)	24 (4.8)	473 (95.2)	
≥60	836 (62.7)	49 (5.9)	787 (94.1)	
Gender				
Male	844 (63.3)	51 (6.0)	793 (94.0)	
Female	489 (36.7)	22 (4.5)	467 (95.5)	
Cancer type				
BLCA	200 (15.0)	11 (5.5)	189 (94.5)	
BRCA	39 (2.9)	1 (2.6)	38 (97.4)	
CRC	106 (8.0)	2 (1.9)	104 (98.1)	
ESCA	104 (7.8)	3 (2.9)	101 (97.1)	
HNSC	133 (10.0)	5 (3.8)	128 (96.2)	
MM	302 (22.7)	39 (12.9)	263 (87.1)	
NSCLC	303 (22.7)	9 (3.0)	294 (97.0)	
RCC	146 (11.0)	3 (2.1)	143 (97.9)	
Drug				
PD-(L)1	1021 (76.6)	45 (4.4)	976 (95.6)	
CTLA-4	89 (6.7)	7 (7.9)	82 (92.1)	
PD-(L)1+CTLA-4	223 (16.7)	21 (9.4)	202 (90.6)	

Supplementary Table 2. Patien	characteristics in the validation	cohort stratified by KDR status
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<sup>a</sup>Indicated percentage of *KDR*-MUT or *KDR*-WT patients in a given category (i.e. specific gender, specific age group). Abbreviations: BLCA, Bladder cancer; BRCA, Breast cancer; CR, Complete response; CRC, Colorectal cancer; EAC, Esophagogastric adenocarcinoma; CTLA-4, Cytotoxic T-cell lymphocyte-4; DCB, Durable clinical benefit; HNSC, Head and neck cancer; MM, Melanoma; NDB, No durable benefit; NE, Not evaluable; NSCLC, Non-small cell lung cancer; PD, Progressive disease; PD-(L)1, Programmed cell death-1 or programmed death-ligand 1; PR, Partial response; RCC, Renal cell carcinoma; SD, stable disease.

### KDR mutation as a novel biomarker for ICI

Cana	Fixed offecto	HR (95% CI)							
Gene	Fixed effects	BLCA	CRC	ESCA	HNSC	MM	NSCLC	RCC	BRCA
KDR	0.31 (0.17-0.53)	0.14 (0.02-1.07)	0.04 (0-587.32)	0.04 (0-47.13)	0.30 (0.04-2.19)	0.49 (0.24-1.02)	0.48 (0.15-1.52)	0.04 (0-124.47)	0.04 (0-3912.12)
MLH1	0.80 (0.37-1.71)	0.47 (0.06-3.45)	0.66 (0.15-2.74)	0.04 (0.00-96.96)	10.15 (1.32-77.61)	1.43 (0.45-4.54)	0.04 (0.00-516.91)	0.04 (0.00-13081.77)	(-)#
MSH2	0.75 (0.38-1.46)	0.37 (0.05-2.66)	0.31 (0.04-2.25)	1.83 (0.43-7.67)	1.08 (0.26-4.44)	0.31 (0.04-2.23)	0.39 (0.09-1.59)	(-)#	(-)#
MSH6	0.45 (0.21-0.96)	0.04 (0.00-3.45)	0.23 (0.03-1.70)	(-)#	10.15 (1.32-77.61)	0.22 (0.03-1.62)	0.76 (0.24-2.38)	1.14 (0.15-8.30)	0.04 (0.00-20.74)
PMS2	0.75 (0.33-1.70)	1.13 (0.15-8.16)	5.21 (1.22-22.11)	0.04 (0.00-96.96)	0.33 (0.04-2.39)	0.04 (0.00-19.58)	5.08 (1.24-20.76)	(-)#	(-)#
POLE	0.69 (0.45-1.05)	0.60 (0.24-1.49)	0.13 (0.01-0.98)	0.55 (0.16-1.84)	0.86 (0.27-2.76)	0.82 (0.36-1.88)	0.52 (0.23-1.18)	(-)#	0.04 (0-71.72)
POLD1	0.93 (0.54-1.58)	0.87 (0.35-2.15)	0.84 (0.25-2.76)	0.35 (0.04-2.57)	0.04 (0-856.64)	2.06 (0.75-5.62)	0.65 (0.09-4.69)	0.04 (0-281.99)	(-)#
TP53	1.45 (1.23-1.70)	0.62 (0.40-0.96)	2.27 (1.16-4.41)	1.41 (0.69-2.86)	1.34 (0.83-2.17)	0.80 (0.49-1.28)	1.19 (0.87-1.62)	0.81 (0.36-1.80)	2.00 (0.82-4.88)
KRAS	1.08 (0.86-1.35)	0.30 (0.04-2.22)	1.03 (0.56-1.90)	0.32 (0.07-1.38)	0.26 (0.03-1.94)	0.55 (0.13-2.26)	0.86 (0.63-1.18)	0.04 (0-66534.63)	(-)#
CDK12	1.00 (0.6-1.65)	0.74 (0.23-2.36)	0.65 (0.19-2.12)	0.04 (0-18852.06)	1.98 (0.48-8.15)	1.14 (0.41-3.10)	1.04 (0.38-2.82)	(-)#	0.04 (0-361.70)

Supplementary Table 3. The estimated hazard ratios of reported gene mutations as predictive markers in the validation cohort

"Indicated no specific gene mutations in the given cancer type.



**Supplementary Figure 5.** *KDR*-MUT was associated with enhanced tumor immunogenicity and anti-angiogenesis. A. Comparison of tumor mutational burden between the *KDR*-MUT and *KDR*-WT tumors in the discovery cohort, validation cohort, and NSCLC cohort (Mann-Whitney U test; ns, not significant; B. Comparison of expression levels of angiogenesis-related genes in *KDR*-MUT and *KDR*-WT groups (\*P<0.05, \*\*P<0.01, \*\*\*P<0.001, \*\*\*\*P<0.0001).



Supplementary Figure 6. Comparison of TMB levels in patients with different mutation sites of KDR.

Supplementary Table 4. Patient characteristics in the NSCLC cohort stratified by KDR status					
Characteristics	N (0()	KDR status [No. (%)] <sup>a</sup>			
Characteristics	NO. (%)	KDR-MUT	KDR-WT		
No. of patients	67	10 (14.9)	57 (85.1)		
Median age, years (range)	64 (32-80)	59 (52-78)	64 (32-80)		
Age					
<60	21 (31.3)	5 (23.8)	16 (76.2)		
≥60	46 (68.7)	5 (10.9)	41 (89.1)		
Gender					
Male	54 (80.6)	9 (16.7)	45 (83.3)		
Female	13 (19.4)	1(7.7)	12 (92.3)		
Histology					
LUAD	43 (64.2)	6 (14.0)	37 (86.0)		
LUSC	21 (31.3)	4 (19.0)	17 (81.0)		
other	3 (4.5)	0(0)	3 (100)		
Drug					
PD-1	67 (100)	10 (14.9)	57 (85.1)		
RECIST					
PR	20 (29.8)	7 (35)	13 (65)		
SD	32 (47.8)	3 (9.4)	29 (90.6)		
PD	15 (22.4)	O (O)	15 (100)		
Lines of treatment					
First	30 (44.8)	6 (20)	24 (80)		
Second	13 (19.4)	2 (15.4)	11 (84.6)		
Third or subsequent	24 (35.8)	2 (8.3)	22 (91.7)		
Stage					
I	3 (4.5)	2 (66.7)	1 (33.3)		
П	3 (4.5)	1 (33.3)	2 (66.7)		
III	15 (22.4)	1(6.7)	14 (93.3)		
IV	46 (68.6)	6 (13.0)	40 (87.0)		
Smoking history					
Yes	27 (40.3)	7 (25.9)	20 (74.1)		
No	40 (59.7)	3 (7.5)	37 (92.5)		

Supplementary Table 4.	Patient characteristics	in the NSCLC	cohort stratified by	KDR status
Supplementary lable 4.			contractined by	NDN Statuc

<sup>a</sup>Indicated percentage of *KDR*-MUT or *KDR*-WT patients in a given category (i.e. specific gender, specific age group). Abbreviations: LUAD, Lung adenocarcinoma; LUSC, Lung squamous cell carcinoma; Other, NSCLC included pulmonary sarcomatoid carcinoma and unclassified NSCLC; PD, Progressive disease; PD-1, Programmed cell death-1; PR, Partial response; SD, Stable disease.



Supplementary Figure 7. Survival curves comparing the different KDR mutation in patients treated with ICIs therapy.



**Supplementary Figure 8.** Potential mechanisms associated with *KDR*-MUT in predicting the efficacy of ICIs. Red color words represent an increased activation; Blue color words represent a decreased activation. *KDR*-MUT tumors produced more neoantigens, which were processed and presented by APCs to T cells. Upregulated MHC-related molecules and TCR diversity activated this process. In addition, *KDR*-MUT was related to anti-angiogenesis, and then promoted the T cells to extravasate from the tumor vasculature. Meanwhile, chemokines (CXCL9, CXCL10, CCL5) contributed to the infiltration of CD8 T cells and dendritic cells into the tumor tissues. CTL (cytotoxic T lymphocyte) released GZMA and PRF1 and then enhanced tumor-killing capacity. Finally, *KDR*-MUT tumors expressed higher PD-L1, which was associated with immunotherapeutic benefit.