The role of neoantigen in immune checkpoint blockade therapy

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Abstract
Immune checkpoint inhibitor induces tumor rejection by activated host immune system. The anti-tumor immune response consists of capture, presentation, recognition of neoantigen, as well as subsequent killing of tumor cell. Due to the interdependence among this series of stepwise events, neoantigen profoundly influences the efficacy of anti-immune checkpoint therapy. Moreover, the neoantigen-specific T cell reactivity is the cornerstone of multiple immunotherapies. In fact, several strategies targeting neoantigen have been attempted for synergetic effect with immune checkpoint inhibitor. Increasing neoantigen presentation to immune system by oncolytic virus, radiotherapy, or cancer vaccine is feasible to enhance neoantigen-specific T cell reactivity in theory. However, some obstacles have not been overcome in practice such as dynamic variation of neoantigen landscape, identification of potential neoantigen, maintenance of high T cell titer post vaccination. In addition, adoptive T cell transfer is another approach to enhance neoantigen-specific T cell reactivity, especially for patients with severe immunosuppression. In this review, we highlighted the advancements of neoantigen and innovative explorations of utilization of neoantigen repertoire in immune checkpoint blockade therapy.

Keywords: Neoantigen, PD-1/PD-L1, CTLA-4, Immunotherapy, Cancer vaccine, Adoptive T cell transfer

Introduction
It is well established that non-self-antigen generated by tumor somatic mutation confers tumor immunogenicity and induces anti-tumor immune response [1]. Cancer immunotherapies such as chimeric antigen receptor T cell, bispecific antibody, cancer vaccine, and immune checkpoint inhibitor, eradicate tumor cell by enhancing host cancer-specific immune reactivity [2–5]. In 2013, Cancer-Immunity cycle theory was established to describe anti-tumor immune response (Fig. 1) [6]. In the cycle, mutation-derived neoantigen is released by cancer cell and initiates the anti-tumor immune response. Then the neoantigen is captured and presented by professional antigen presentation cell (APC) which induces the priming and activation of neoantigen-specific T cell in peripheral immune organ. Peripheral activated T effector cell traffics to and infiltrates into tumor bed. Following recognition of neoantigen, tumor cell is killed by tumor infiltrating lymphocyte (TIL). During cancer immune evasion, one or more steps are undermined [6].

Strategies such as enhancing release and presentation of neoantigen, increasing neoantigen-specific T cell abundance, or blocking immune checkpoint, impact different steps in Cancer-Immunity cycle. Among these strategies, immune checkpoint blockade attracts intensive attention for potent and durable tumor control [7]. Cytotoxic T lymphocyte-associated antigen-4 (CTLA-4) antibody primarily blocks inhibitory signaling of T cell priming and activation while programmed cell death protein 1/programmed cell death ligand 1 (PD-1/PD-L1) antibody primarily recovers attenuated anti-tumor immune response in tumor bed [6]. However, due to the interdependence of multiple steps of cancer-immunity cycle, the efficacy of immune checkpoint blockade treatment is substantially affected by neoantigen presentation and recognition [6]. Therefore, it is presumed that neoantigen is a predictive biomarker and synergistic treatment target for immune checkpoint inhibitor.
Fig. 1 Cancer-Immunity cycle and neoantigen presentation. Cancer-immunity cycle: neoantigen released by dead cancer cell initiates the anti-tumor immune response. Then the neoantigen is captured and presented by antigen presentation cell (APC) which induces the priming and activation of neoantigen-specific T cell in peripheral immune organ. Peripheral activated T effector cell traffics to and infiltrates into tumor bed. Following recognition of neoantigen, tumor cell is killed by tumor infiltrating lymphocyte (TIL). Neoantigen presentation: in the proteasome of tumor cell, mutant protein derived from somatic mutation is degenerated into peptide and then transported to endoplasmic reticulum. The peptide binds to major histocompatibility complex I (MHC-I) binding site by transporter associated with antigen processing (TAP). Simultaneously, the assembled peptide-MHC-I complex is transported to membrane of tumor cell. Cytotoxic T cell could recognize peptide-MHC-I complex and kill the tumor cell.
Predictive value of neoantigen in immune checkpoint blockade therapy

Tumor somatic mutation

Somatic mutation participates in tumor initiation and progression. In the meanwhile, tumor mutation landscape influences immune surveillance and evasion. In the proteasome of tumor cell, mutant protein derived from somatic mutation is degenerated into peptide and then transported to endoplasmic reticulum. The peptide binds to major histocompatibility complex I (MHC-I) binding site. Simultaneously, the assembled peptide-MHC-I complex is transported to membrane of tumor cell (Fig. 1). Cytotoxic T cell could recognize peptide-MHC-I complex and kill the tumor cell [8].

Neoantigen derives from tumor somatic mutation, thus tumor mutation burden (TMB) is considered as the surrogate of neoantigen burden and the predictive biomarker of checkpoint blockade therapy [9–11]. Effective tumor regression induced by immune checkpoint blockade therapy is commonly observed in some specific cancer types [12]. A main factor contributing to different treatment effect among various type of cancers is TMB [13]. Based on data from TCGA, tumor mutation burden analysis across multiple cancers was conducted (Fig. 2). The result showed that cancer types which are approved for immune checkpoint inhibitor therapy such as melanoma, bladder cancer, and head and neck cancer tend to harbor high TMB. As an exception, three types of renal cell cancer have relative low TMB, but they response to nivolumab well. The sensitivity of kidney cell cancer to PD-L1 antibody is attributed to high-frequency indel variation-derived frameshift mutation. It is found that the probability of frameshift mutation generating neoantigen is eight times higher than non-synonymous single nucleotide variation [13].

The treatment effect not only varies with tumor type, but also positively relates to TMB in patients with the same type of tumor [14, 15]. Rizvi et al. investigated relationship between efficacy of PD-1 blockade and TMB in non-small cell lung cancer (NSCLC). The result showed that patients with high TMB showed a significant advantage in progression-free survival (PFS) than low TMB group (Hazard Ratio = 0.19, log-rank P = 0.0004) [14]. Subsequently, Carbone et al. further confirmed the correlation between TMB and efficacy of nivolumab in stage IV or recurrent NSCLC patients [16].

For some tumors, especially gastrointestinal tumor, accumulated somatic mutations closely relate to mismatch repair deficiency (dMMR) [5]. Mismatch repair system participates in correcting base substitution, insertion, and deletion in the process of DNA replication [17]. It is generally thought that high microsatellite instability (MSI-H)/dMMR heralds the clinic benefit
from immune checkpoint inhibitors [15, 18–20]. Le et al. evaluated the relationship between dMMR and efficacy of PD-1 inhibitor across 12 tumor types. Objective radiographic response rate was up to 53% in patients with dMMR, indirectly indicating the predictive role of TMB in immune checkpoint blockade treatment [15].

**Predictive value of neoantigen in immune checkpoint inhibitor**

Though high TMB and neoantigen burden contribute to high response rate to immune checkpoint inhibitor generally, the two biomarkers could not fully determine treatment effect. For example, in NSCLC patients undergoing PD-1 or PD-1/CTLA-4 inhibitor treatment, it was detected that both TMB and candidate neoantigen level of post-progression tumor tissue were higher compared to pre-immunotherapy tumor tissue [21]. By analysis of landscape of tumor somatic mutation, it was observed that some mutant genes encoding neoantigens were eliminated while some mutations not encoding neoantigen were gained after resistance to anti-immune checkpoint therapy [21]. Besides, the eliminated neoantigen was found to have higher MHC binding affinity compared with retained and gained neoantigen [21]. Further exploration revealed that eliminated mutations harboring altered domain related MHC binding and T cell receptor (TCR) binding [21]. By autologous T cell culture, the eliminated neoantigen could induce clonal T cell expansion successfully, indicating that the loss of these antigen might relate to immune escape and resistance to immunotherapy [21].

It is notable that as a widely adopted biomarker for patient selection and efficacy prediction prior to immune checkpoint blockade therapy, TMB is not a perfect surrogate of immunogenic neoantigen. Actually, anti-tumor immune response is initiated by the recognition of neoantigen-MHC complex rather than mutated gene [22]. In the “lottery of neoantigen formation”, the production of neoantigen related with clinic benefit is a probabilistic event. High TMB could elevate the probability of production of immunogenic neoantigen but could not guarantee the occurrence of neoantigen-specific response [1, 23]. Snyder et al. noticed that some melanoma patients failed to response to CTLA-4 inhibitor even harboring high TMB. Further investigation showed that patients benefiting from CTLA-4 inhibitor tented to have neoantigens containing the same tetrapeptide, which was absent in the patients resistant to CTLA-4 inhibitor [23]. Moreover, therapeutic benefit-related neoantigens were found to resemble epitopes from pathogen which had strong immunogenicity [23]. Intriguingly, the homology between tumor neoantigen and pathogen epitope was speculated to relate with cross-reactivity of immunity [23]. Some observations showed that antigen from microbiota influenced the efficacy of immune checkpoint inhibitor [24–27].

**Prediction of neoantigen**

For further application of neoantigen as a predictive biomarker or a therapeutic target for checkpoint blockade therapy, the prediction of candidate neoantigen is indispensable. The common prediction algorithm consists of three parts: (A) identification of mutated protein; (B) identification of MHC typing; (C) prediction of the binding affinity between MHC and neo-peptides (Fig. 3) [28, 29].

Next generation sequencing (NGS) provides an available access to obtain somatic mutation data in the whole genome. Based on public databases of gene, transcription, and protein sequence, annotation software interprets the data from NGS to expression status of protein [30, 31]. In terms of identification of MHC typing, as the most polymorphic region, MHC typing is usually obtained by sequencing data which could be further interpreted by well-annotated sequences in public databases [32–35].

In the process of antigen recognition, only 0.5% peptides could bind to MHC, which is the most selective step [36]. Therefore, MHC binding affinity is the core parameter for most prediction algorithms of candidate neoantigen [37]. In most cases, the researchers tend to focus on binding affinity of MHC-I molecule. The primary reason

![Fig. 3 Prediction of candidate neoantigen. The common prediction algorithm consists of three parts: a identification of mutated protein; b identification of major histocompatibility complex (MHC) typing; c prediction of the binding affinity between MHC and neo-peptides.](image-url)
is that MHC-I molecule directly relates with neoantigen presentation on tumor cell [38, 39]. Besides, peptides binding to MHC-I molecules are usually distributed in narrow length, mainly in 8 and 11 amino acids while MHC-II molecules could bind to peptides which are distributed in broader length (from 11 to 30 amino acids) [28, 40]. It is more available to utilize MHC-I molecule for neoantigen prediction in silico. Valuation of affinity is mainly conducted by two methods: 3D structure-based prediction or sequence-based method [28]. Propelled by NGS, sequence-based methods have been developing, from previously verified peptide binders of a specific MHC allele to nonlinear methods [41]. However, due to the limited amount of detected MHC allele, machine learning-based method was created to overcome the problem. Beyond MHC allele have been detected, trained neuron network could calculate the affinity of given peptide-MHC complex by mimicking MHC sequence from peptide-binding residues, called pan-specific tool [42].

Though multiple tools have been created (Table 1), high false positive rate of neoantigen prediction could not be ignored. Some steps of neoantigen processing prior to MHC binding including cleavage in proteasome and transportation to endoplasmic reticulum influence the effect of neoantigen presentation [43]. Therefore, comprehensive algorithm involving affinity of peptide-MHC complex, cleavage in proteasome, and transporter associated with antigen processing (TAP) shows better performance than analysis with single approach [44, 45]. Some prediction algorithms such as NetTepi even contain parameters including binding stability and T cell propensity [46].

Due to decreased false positive rate benefiting from improved algorithm, it is more available to analyze the shared features among the candidate neoantigens for potential therapeutic target [47]. Kim et al. designed a novel algorithm consisting of peptide-MHC affinity and other immunogenicity-related parameters for candidate neoantigen prediction [36]. Notably, just 16 out of 3760 candidate neoantigens were detected in more than one sample. Besides, analysis of 1867 known neoantigens revealed that most neoantigens were homological with pathogens such as Mycobacterium tuberculosis, Trypanosoma cruzi, Vaccinia virus, Human herpes virus, and Hepatitis C virus [36].

**Therapeutic application of neoantigen in immune checkpoint blockade therapy**

**Increased release of neoantigen to immune system**

Restricted antigen presentation renders some neoantigens hidden from immune surveillance [48]. However, these neoantigens could be released post cancer cell death. Released neoantigen has a probability to induce the emergency of new T cell clones which could further circulate and kill other cancer cells [48]. Interventions such as oncolytic virus and radiotherapy effectively eliminate local cancer cells that might influence the immune surveillance status systemically [48]. Combination therapy of these interventions and immune checkpoint blockade showed more potent tumor eradication than mono-therapy of immune checkpoint inhibitor [1].

**Immune checkpoint inhibitor combined with oncolytic virus**

Distinguished from normal cell, tumor cell tends to have an undermined capability to counteract virus infection [49]. Therefore, oncolytic virus preferentially infects cancer cell and induces lysis of cancer cell subsequently. By deleting or silencing gene which is essential to virus

| Algorithm                | Brief description of algorithm                                                                 | Refs.   |
|--------------------------|-------------------------------------------------------------------------------------------------|---------|
| NetMHC-3.0               | Artificial neural network-based algorithm for prediction of binding affinity between MHC-I and peptides of length 8–11 | [89]    |
| NetMHCpan-3.0            | Machine-learning model-based algorithm for prediction of binding affinity between MHC-I and peptides, a pan-specific version | [90]    |
| NetMHCcons              | Comprehensive algorithm for prediction of binding affinity between MHC-I and peptides          | [41]    |
| NetMHCstab              | Artificial neural network-based algorithm for prediction of stability of peptide-MHC-I complex | [91]    |
| PickPocket              | Position-specific scoring matrix-based algorithm for prediction of binding affinity between MHC-I and peptides | [92]    |
| FRED2                   | Epitope prediction for neoantigen                                                             | [93]    |
| NetCTL-1.2              | Comprehensive prediction algorithm containing proteasome cleavage, TAP transport, and MHC-I binding affinity | [94]    |
| NetCTLpan               | The pan-specific version of NetCTL                                                             | [95]    |
| NetTepi                 | Integrated prediction algorithm containing binding affinity, stability of peptide-MHC-I complex, and T cell propensity | [46]    |
| pVAC-Seq                | Identification of neoantigen by tumor mutation and expression data                            | [96]    |
| EpitoolKit              | Prediction of MHC-I typing and T cell epitope                                                  | [97]    |
| WAPP                    | Comprehensive prediction algorithm containing proteasome cleavage, TAP transport, and MHC-I binding affinity | [98]    |

MHC major histocompatibility complex, TAP transporter associated with antigen processing
replication in normal cell, engineered oncolytic virus is
designed for higher tumor specificity [50]. In addition to
direct cancer cell lysis effect, oncolytic virus induces local
and systemic tumor specific immune response which
might influence tumor surveillance [51, 52]. After infec-
tion by oncolytic virus, cancer cell upregulates the pro-
duction of reactive oxygen species and some cytokines
(e.g. interleukin-2 and interferon-γ) to counteract the
infection [48]. Subsequently activated innate immune
response and magnified adaptive immune response pro-
mote recognition of tumor associated antigen and neo-
antigen [48]. Characteristics of oncolytic virus including
tumor cell killing and immune stimulatory effect are
favorable to immune checkpoint blockade therapy
[53–55].

Woller et al. found the synthetic effect of oncolytic virus
with immune checkpoint inhibitor therapy [52]. In mice
model bearing disseminated CMT64 tumor, researchers
observed that neither PD-1 inhibitor nor oncolytic virus
alone could inhibit tumor progression [52]. However,
the combination therapy effectively inhibit the growth of
primary and metastatic lesions [52]. Systemic resistance
to PD-1 inhibitor was overcome by local oncolytic virus
infection, which was primarily attributed to broadened
neoantigen spectrum and elevated inflammation mag-
nitude [52]. Similarly, Zamarin et al. observed that the
intratumoral oncolytic virus treatment combined with
PD-1 inhibitor resulted in regression of primary and dis-
tant tumor lesions, suggesting the potential application
of oncolytic virus for overcoming immune checkpoint
inhibitor resistance [56].

Immune checkpoint inhibitor combined with radiotherapy
Radiotherapy substantially influences tumor immuno-
genicity and tumor microenvironment [57, 58]. Actually,
radiotherapy-induced tumor cell death releases neoanti-
gen which is equal to a tumor vaccine in situ [59]. Con-
current neoantigen specific immune response is activated
accompanied with abundant T cell infiltration [60]. The
conversion of tumor immune microenvironment from
“cold” to “hot” synergizes with immune checkpoint inhib-
itors, enhancing anti-tumor response by different steps of
Cancer-Immunity cycle [61, 62].

Aboudaram et al. interrogated the treatment effect of
concurrent radiotherapy and PD-1 inhibitor (pembroliz-
ubum or nivolumab) [63]. It was showed that metastatic
melanoma patients receiving PD-1 inhibitor and concur-
rent radiotherapy had higher objective response rate than
patients receiving PD-1 inhibitor monotherapy (64.7% vs.
33.3%, $P=0.02$) [63]. In addition, in the NSCLC
patients, Fiorica et al. observed the similar phenomenon
[64]. The results suggested that nivolumab combined
concurrent radiotherapy had advantage over nivolumab
monotherapy in prognosis (1-year overall survival rate:
57.8% vs 27.4%, $P=0.043$; 1-year progression-free sur-
vival rate: 57.8% vs 20.6%, $P=0.040$) [64].

Apart from increased release of neoantigen, radio-
therapy bi-directionally influences the infiltration of
immune cells. On the one hand, radiotherapy alters the
expression of adhesion molecules on tumor vascula-
ture endothelial cells (e.g. upregulated selectins vascu-
lar cell adhesion molecule-1, and intercellular adhesion
molecules 1), which are helpful to recruitment and
infiltration of immune cell [65]. On the other hand,
radiotherapy-induced cytokines including transforming
growth factor-β (TGF-β) participate in remodeling of
extracellular matrix. TGF-β could promote the produc-
tion of extracellular matrix protein which impedes the
infiltration of immune cell into tumor bed [65]. Notably,
the inhibitory tumor immune microenvironment could
be counteracted by ICI treatment. It was observed that
the α-CTLA-4 treatment modulated cytokine milieu
including the upregulated interleukin-2 and downregu-
lated TGF-β, indicating the synthetic effect of ICI treat-
ment and radiotherapy [65].

Neoantigen-based cancer vaccine
Compared with defined molecular neoantigen vaccine,
neoantigen release induced by interventions such as
oncolytic virus and radiotherapy is unpredictable. Neo-
antigen released by tumor cell death is diluted by larger
amount of non-mutant peptides, decreasing the probabil-
ity of neoantigen presentation and recognition [1]. Thus,
neoantigen vaccine containing candidate neoantigen
would be a more effective synthetic therapy for immune
checkpoint inhibitor [66]. As a monotherapy, neoantigen
vaccine is not sufficient for tumor control primarily due
to inhibitory tumor immune microenvironment. How-
ever, the combination therapy with immune checkpoint
inhibitor overcomes the obstacle and shows promising
application prospect [47, 67].

Neoantigen cancer vaccine-induced tumor rejection
In the immune response induced by neoantigen vaccine,
neoantigen is mainly recognized by cross-presentation
[47]. Following professional APC recruited to the vac-
cination site, activated APC migrates to drain lymph
node and induces the activation of T cell [47]. For the
design of neoantigen vaccine, selection of neoantigen
which could be effectively presented and recognized is a
technical challenge. In addition to immune-stimulating
components, different adjuvants and vaccine platforms
influence efficacy of vaccine as well [66, 68, 69]. Actu-
ally, platforms including tumor cell, DNA, RNA, viral,
bacteria, dendritic cell, and peptide/protein have been

applied in vaccine design [66]. Among vaccine platforms mentioned above, viral based platform vaccine shows the potentiality to induce potent and durable T cell response [47]. Moreover, tumor could not be eliminated completely in a short time, so long term tumor control needs both prime vaccination and multiple boost vaccinations to maintain T cell activity.

Based on mice model bearing MCA-induced sarcomas, Gubin et al. conducted a trial to explore the effect of neoantigen vaccination on tumor control [70]. By comprehensive analysis containing Stabilized Matrix Method algorithm, the Artificial Neural Network algorithm, and the NetMHCpan algorithm, the affinity and stability of peptide-MHC I complex were calculated among all non-synonymous mutations [70]. Subsequently, candidate neoantigens were filtered by the following standards: A: screening out the neoantigen poorly processed in proteasome; B: eliminating neoantigens with lower binding affinity to MHC I than their corresponding wild type (WT) peptides. Finally, two predominant H-2 K^b restricted candidate epitopes were identified: an A506T mutation in Asparagine-linked glycosylation 8 (mAlg8) and a G1254V mutation in Laminin alpha subunit 4 (mLama4) [70]. To verify the role of mAlg8 and mLama4 in anti-PD-1 treatment, in the mice model experiencing anti-PD-1-induced sarcomas (d42m1-T3) rejection, researchers isolated CD8^+ T cell from spleen. Intriguingly, the isolated T cell could be stimulated to produce interferon-γ by mAlg8 and mLama4 rather than other antigens [70]. Moreover, selected reaction monitoring showed that mLama4 and mAlg8 were the only candidate epitopes with strong binding affinity in the H-2K^b eluate [70]. The observation that mAlg8 or mLama4 specific CD8^+ T cell increased temporally after anti-PD-1 treatment and mounted to peak value just before tumor regression further indicated the role of these neo-epitopes [70]. Given the results mentioned above, researchers designed a cancer vaccine mainly consisting of mAlg8 or mLama4. The vaccine led to potent tumor elimination in mice model compared with the control mice [70].

**Immune checkpoint inhibitor combined with neoantigen cancer vaccine**

On the one hand, some failures to response to neoantigen vaccines are primarily attributed to suppressive tumor microenvironment. Emerging immune modulators including anti-PD-1/PD-L1 antibody, anti-CTLA-4 antibody, and anti-T cell immunoglobulin and mucin domain-containing protein-3 (Tim-3) antibody could resolve the problem [71, 72]. Sahin et al. conducted a study to explore the treatment effect of RNA platform-based neoantigen cancer vaccine [73]. One out of three melanoma patients receiving vaccine experienced relapse and distant metastasis. However, by subsequent pembrolizumab treatment, the patient showed a complete response [73]. Compared with the reported complete response rate (below 10%), this treatment effect is satisfactory [73]. Further investigation revealed that neoantigen specific T cell was PD-1^+^, and the expression abundance of PD-L1 in tumor tissue was upregulated, suggesting the suppressive immune microenvironment induced by neoantigen cancer vaccine [73]. Presumably due to blockaded inhibitory immune regulation, the combination therapy showed more robust tumor control effect [73]. Simultaneously, Ott et al. investigated the efficacy of neoantigen cancer vaccine targeting up to 20 predicted neo-epitopes [74]. It was showed that 2 out of 6 melanoma patients experienced tumor relapse [74]. Similarly to the phenomenon mentioned above, both 2 recurrent melanoma patients had a complete tumor rejection after pembrolizumab treatment, which further verified the feasibility of combination therapy [74].

On the other hand, frequently-occurring adaptive resistance during immune checkpoint inhibitor is related with variation of neoantigen repertoire [67]. Due to heterogeneity of tumor, part of mutations are shared by all tumor cells while the others are exclusively expressed by subpopulations [75]. Under survival selective pressure, subpopulations sensitive to immune checkpoint inhibitor are eliminated. In the meanwhile, subpopulations resistant to immune checkpoint inhibitor have an advantage in proliferation [76]. As a result, loss of immunologic epitopes results in alternative subpopulation constitution and resistance to treatment, called immuneediting [77, 78]. However, the resistance could be overcome by neoantigen cancer vaccine, because immune-stimulating component of vaccine could be manipulated depending on dynamic variation of neoantigen spectrum during tumor evolution [67]. Carreno et al. conducted a study to investigate the influence of neoantigen cancer vaccine on neoantigen-specific T cell receptor repertoire [79]. The study recruited 3 melanoma patients which had been treated with ipilimumab [79]. Each patient received DC platform-based neoantigen cancer vaccine which containing 7 identified neoantigens [79]. Before and after vaccination, researchers collected peripheral blood sample and estimated the immune response to supposed neoantigens [79]. Immune monitoring showed that T cell response targeting these neoantigens was enhanced. Moreover, compared with pre-vaccination, vaccination induced T cell response to 2 additional neoantigens per patient [79]. Subsequently, composition and abundance of neoantigen-specific T cell was analyzed. In the purified CD8^+ T cell isolated from peripheral blood, researchers found that after vaccination, the frequency of existing
neoantigen-specific TCRβ clonotypes were increased accompanied with additional clonotypes for all each neoantigen [79]. The results showed that both TCRβ clonotypes targeting predominant and sub-predominant neoantigens were elevated after vaccination, suggesting the broadened spectrum of T cell response [79]. Two patients recruited in the study were resistant to ipilimumab and had recurrent tumor lesions. By intervention of neoantigen cancer vaccine, effective anti-tumor immune response was rebuilt [79].

**Adaptive T cell transfer**

The efficacy of immune checkpoint inhibitor directly depends on activity of neoantigen specific T cell [80]. Therefore, manipulating composition and abundance of T cell would be another approach to enhance treatment effect of immune checkpoint inhibitor [80]. In preclinical trials, adoptive T cell transfer targeting tumor specific mutations showed potent anti-tumor activity [81, 82]. In 2014, Tran et al. conducted a study to explore the treatment effect of adoptive T cell transfer in a patient with metastatic cholangiocarcinoma [83]. After identification of specific T cell clone targeting tumor specific mutation (mutation of erbb2 interacting protein, called ERBB2IP), autologous TIL was stimulated by interleukin-2 for proliferation and enhanced activity [83]. Subsequently, total 42.4 billion TILs were transferred to the patient which contained nearly 10 billion ERBB2IP mutation specific CD4+ T cells [83]. Prior to adoptive T cell transfer, the patient had received multiple chemotherapy regimens and tumor had metastasized to liver and lung [83]. Following the T cell injection, all lesions in liver and lung showed regression and reached a maximum reduction up to 30% [83]. Besides, followed by recurrent lesion in lung, the patient achieved a disease stabilization for 13 months [83]. Researchers further investigated whether the tumor rejection was attributed to ERBB2IP mutation specific CD4+ T cell. The patient with refractory tumor after treatment received a second adoptive T cell transfer which contained more than 95% ERBB2IP mutation specific CD4+ T cells [83]. Unexpectedly, the tumor was eliminated more quickly and potently than first T cell injection, and the tumor was observed obvious regression as early as following 1st month [83]. Apart from selection and expansion from TIL, neoantigen specific T cell could also be obtained from TCR-engineered T cell [84]. Commonly, neoantigen specific T cell is prepared by transferring genetic material which could encode corresponding TCR or synthetic chimeric antigen receptor [84–86].

One of determinants for efficacy of immune checkpoint inhibitor is pre-existing tumor specific T cell, so it is feasible to combine adoptive T cell transfer and immune checkpoint blockade especially for patient with severely inhibited immunity [5, 84]. It was noticed that combination of anti-immune checkpoint and T cell transfer targeting tumor associated antigen induced tumor eradication successfully [87]. Limited by complicated procedure for obtaining neoantigen-specific T cell, study verifying the effect of combination therapy of neoantigen-specific T cell transfer and ICI is unavailable. However, it has been verified that the effect of TIL transfer could be boosted by immune checkpoint blockade [88].

In theory, combination therapy of immune checkpoint inhibitor and neoantigen specific adoptive T cell transfer is a reasonable strategy in the absence of pre-existing tumor specific T cell, but the actual efficacy should be investigated further.

**Conclusion**

Mutation is a double-edged sword for tumor growth. It contributes to tumorigenesis and progression, but in the meanwhile, the mutation could be recognized by host immunity and lead to tumor elimination. It has been confirmed that neoantigen specific T cell activity is the main determinant of immunotherapy. Thus, strategies targeting neoantigen receive intensive attention for the synthetic effect with other immunotherapies such as immune checkpoint inhibitor. Based on NGSs and public databases, multiple algorithms were established and optimized for neoantigen prediction in silico, which further propelled development of neoantigen cancer vaccine and T cell transfer. In terms of the importance of pre-existing anti-tumor immune response for immune checkpoint inhibitor, we believe personalized neoantigen-based treatment would be a promising synthetic strategy.

**Abbreviations**

Alg8: Asparagine-linked glycosylation 8; APC: antigen presentation cell; CTLA-4: cytotoxic T lymphocyte-associated antigen-4; ERBB2IP: erbb2 interacting protein; HLA: human leukocyte antigen; Lamin: Laminin alpha subunit 4; MHC: major histocompatibility complex; MMR: mismatch repair; MSI: microsatellite instability; NGS: next generation sequencing; NSCLC: non-small cell lung cancer; PD-1: programmed cell death protein 1; PD-L1: programmed cell death ligand 1; PFS: progression-free survival; TAP: transporter associated with antigen processing; TCR: T cell receptor; TGF-β: transforming growth factor-β; TIL: tumor infiltrating lymphocyte; Tim-3: T cell immunoglobulin and mucin domain-containing protein-3; TMB: tumor mutation burden; WT: wild type.

**Authors’ contributions**

MY performed the selection of literature, drafted the manuscript, and prepared the figures. SQ, WZ, and SY collected the related references and participated in discussion. KW and QC designed this review and revised the manuscript. All authors contributed to this manuscript. All authors read and approved the final manuscript.

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Competition interests
The authors declare that they have no competing interests.

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