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# Beyond PD-L1: novel PD-1 biomarkers identified by driving T cell dysfunction in vitro

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## Agenus VISION Platform

Integrated discovery & development platform drives concepts to clinic



Figure 1: Representation of VISION platform as an integrated tool to recapitulate tumor immune interactions allowing perturbations to interrogate concepts and accelerate them to clinic.

### T cell dysfunction system drives CD8+ T cells to dysfunction through chronic cancer antigen exposure



Figure 2: Method for driving T cell dysfunction in vitro. T cell preparation: PBMCs were isolated from three healthy donors on a Ficoll gradient and rested overnight. CD8+ T cells were isolated on magnetic beads, stimulated with CD3/CD28, transduced with NY-ESO-1 lentivirus, and expanded. TCR expression was confirmed by flow cytometry. Cancer cell preparation: U251 MG cell lines were transduced with lentivirus encoding a fusion protein of b2-microgobulin, HLA2-A2 and NY-ESO-1 or MART-1 peptide and selected on blasticidin. Antigen expression was confirmed with an NFAT reporter system in Jurkat T cells. For co-culture: A fixed number of irradiated U251 MG cells were cultured for 24h. CD8+ T cells were added to achieve the desired cancer: T cell ratio. The co-culture was monitored every 24h. If cancer cells were depleted, CD8+ T cells were transferred to a fresh well of irradiated U251 MG cells and the remainder were collected for downstream analyses. If cancer cells were not depleted, media was changed and reassessed the following day. Top Panel) Cytotoxicity kinetics as monitored by live cell fluorescence microscopy. Middle Panel) Secreted effector molecules measured by Luminex Bead Array, normalized to the maximum signal per molecule. Bottom Panel) Transcriptomic landscape of T cell dysfunction across antigen exposures.

VISION PLATFORM PROVIDES CLINICAL INSIGHTS VISION T cell dysfunction signature correlate with Objective Response Rate in anti-PD1/PDL1 treatment



Figure 3: (Left Panel) Average response rate to anti-PD1 therapies. (Right Panel) Accuracy of predicting response to anti-PD1 on a per patient basis using either PD-L1 IHC or Agenus' VISION machine learning models using melanoma (Riaz et. Al and Hugo et. al) and cervical (Agenus C-550 and C-700) patients treated with anti-PD1.

VISION deep-learning model identifies anti-PD1 responders in melanoma and cervical cancer



Figure 4: Development of a deep learning model to differentiate responders from non-responders in a population of anti-PD1 treated melanoma patients. Melanoma was selected for initial training and validation due to the availability of public, clinically-annotated tumor RNA-seq datasets from large numbers of patients (Hugo et al. (n=25) and Riaz et al. (n=43)). Feature selection comprised of Agenus VISION T cell state signatures combined with differentially expressed genes between the two groups. Data partition for training and validation was followed by modelling and performance evaluation.

### VISION T cell dysfunction signature correlates with Objective Response Rate in anti-PD1/PDL1 treatment across indications



high patients (fPD1), C) CD8+ T-Cell Abundance, D) VISION T cell dysfunction signature, and E) Bivariate Tumor Mutation Burden and VISION T cell dysfunction signature.

F) Prediction of ORR for TCGA molecular subtypes using the bivariate model. TCGA indications were split into previously defined molecular subtypes (e.g. for BRCA - HER2, LumA, LumB, and basal) to validate correlations seen at the indication-level.

Identifying Responders and non-responders



PDL<sup>2</sup>

Accuracy

87% Agenus Model Melanoma

Accuracy

86% Agenus Model Cervical Accuracy



Multi Layer Perceptron (MLP) Model

### VISION PLATFORM DEFINES RATIONAL COMBINATIONS Single cell transcriptomics of T cells during early dysfunction identifies T cell subsets that respond to anti-PD1 Expression of dysfunction markers on the activated PD1<sup>high</sup> subset



Figure 6: A) Heatmap showing mean expression of select activation and dysfunction markers based on single cell RNA-sequencing across 14 T cell clusters that are present during the early dysfunction state. (B) Violin plot depicting single cell expression levels for select dysfunction markers within the activated PD1high subset

### Combination of PD-1 and TIGIT blockade enhanced T cell cytotoxicity of tumor cells relative to monotherapies



Figure 7: A) Tumor killing capacity of anti-PD1 and co-blockade B) Tumor killing capacity of PD1 and anti-TIM3 co-blockade.

- antibodies in the clinic.

### **References:**

- Riaz et. AL., Tumor and Microenvironment Evolution during Immunotherapy with Nivolumab.
- Anti-PD-1 Therapy in Metastatic Melanoma. Cell. 2016 Mar 24;165(1):35-44.
- burden. j. immunotherapy cancer 6, 32 (2018).

Normalized Expression

### Conclusions

• Agenus' VISION platform combines deep in vitro profiling and Al-based approaches to predict clinical outcomes, plus rational targets & combinations.

• We defined a predictive biomarker signature that outperforms standard PD-L1 IHC. • We identified a potential mechanism underlying the effective combination of anti-PD1 and anti-TIGIT

Hugo et. al., Genomic and Transcriptomic Features of Response to

• Morrison, C., Pabla, S., Conroy, J.M. et al. Predicting response to checkpoint inhibitors in melanoma beyond PD-L1 and mutational

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