

A Multifactorial Model of T Cell Expansion and Durable Clinical Benefit in Response to a PD-L1 Inhibitor

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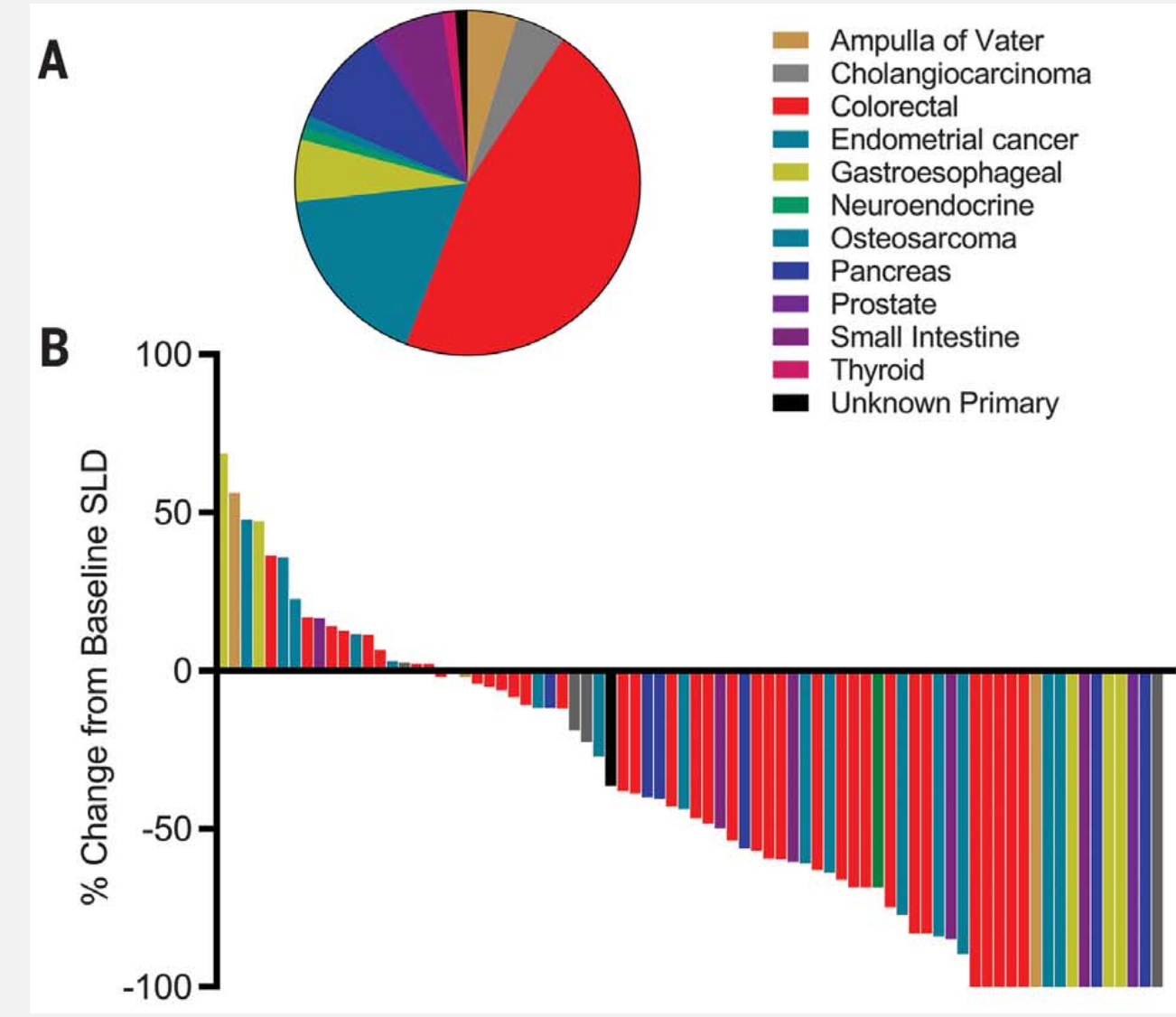
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Predicting response to checkpoint inhibitor immunotherapy

Promise of checkpoint inhibitors

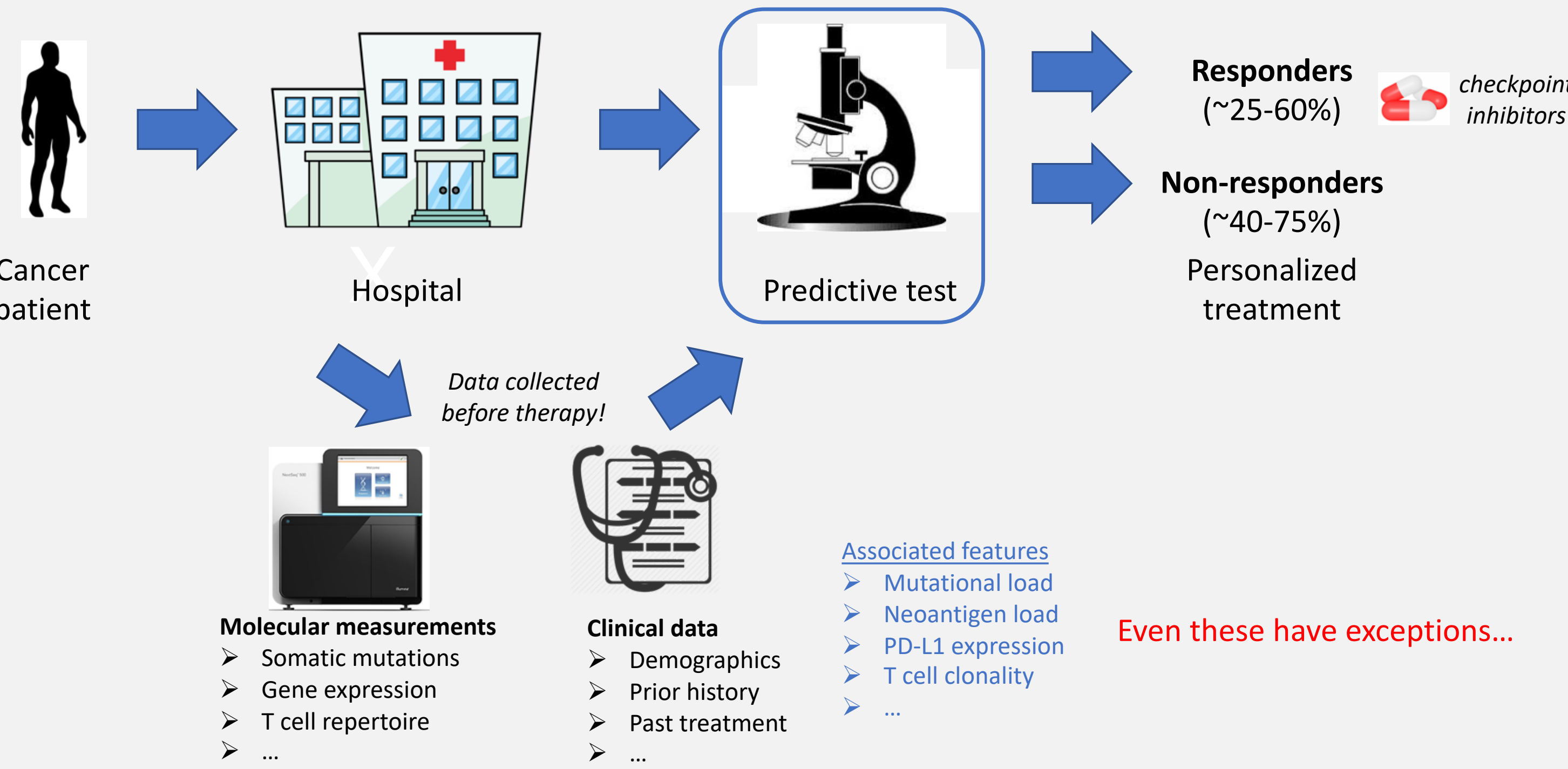
- Success in previously untreatable late-stage cancers
- Treatment scales across patients and cancer types



Mismatch-repair deficiency predicts response of solid tumors to PD-1 blockade
Le *et al.* (Science, 2017)

Biomarkers for response are a priority

- Efficacy
- Adverse effects
- Side effects



Our approach: predicting *molecular* response

Several key challenges

1. **Integrate** molecular and clinical features
2. **Select** appropriate outcome to predict
 - A. Clinical responses are noisy, impacted by confounding factors, and poor proxies for molecular response
3. **Assess** value for new patients from small datasets
 - A. Predictive accuracy
 - B. Feature importance

How to predict the response of immune system post-therapy?

Our approach: Predict T cell clones in the tumor (TILs) that expand in the blood

Advantages over predicting standard clinical outcomes (e.g., overall survival or durable clinical benefit)

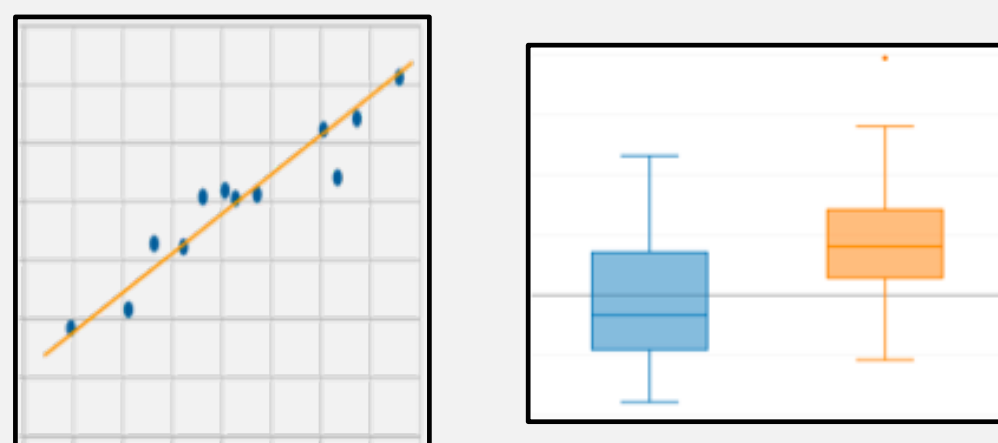
- More immediate outcome
- Finer-grained measurement

Molecular & clinical features

Age	71	68	46	66	78
Chemo regimens	3	1	1	2	1
Mutational load	540	198	23	14	412
PD-L1 expression	IC0	IC1	IC2	IC2	IC0
Unique TCRs	90,690	71,400	63,345	73,805	57,890

learn model

Outcomes



Predict outcomes Associate with response

Features of 21 bladder cancers

N=21 patients treated with PD-L1 inhibitor (*atezolizumab*)

Clinical	Tumor	Circulating
Prior intravesical Bacillus Calmette–Guérin (BCG)	Missense SNV count (mutation load)	Productive unique TCR count
Age	Expressed missense SNV count	Clonality (TCR)
Albumin < 4	Neoantigen count	Diversity (TCR)
Baseline neutrophil to lymphocyte ratio	Expressed neoantigen count	T cell fraction
Time since last chemotherapy (days)	Clonality (TCR)	Top clone frequency (%)
5-factor score	Diversity (TCR)	
Number of chemotherapy regimens	T cell fraction	

Public dataset from Snyder, et al. (*PLoS Medicine*, 2017)

A multifactorial model of immune response

Approach: Predict number of **expanded TIL clones** in blood three weeks after treatment as a function of **19 clinical, tumor, and circulating features**

Elastic Net: High-dimensional regression procedure designed to **reduce overfitting** and **be robust to irrelevant features** (*Hastie & Zhou, JRSS B, 2005*)

$$\hat{\beta} = \operatorname{argmin}_{\beta} ||y - X\beta||_2^2 + \lambda_1 ||\beta||_2^2 + \lambda_2 ||\beta||_1$$

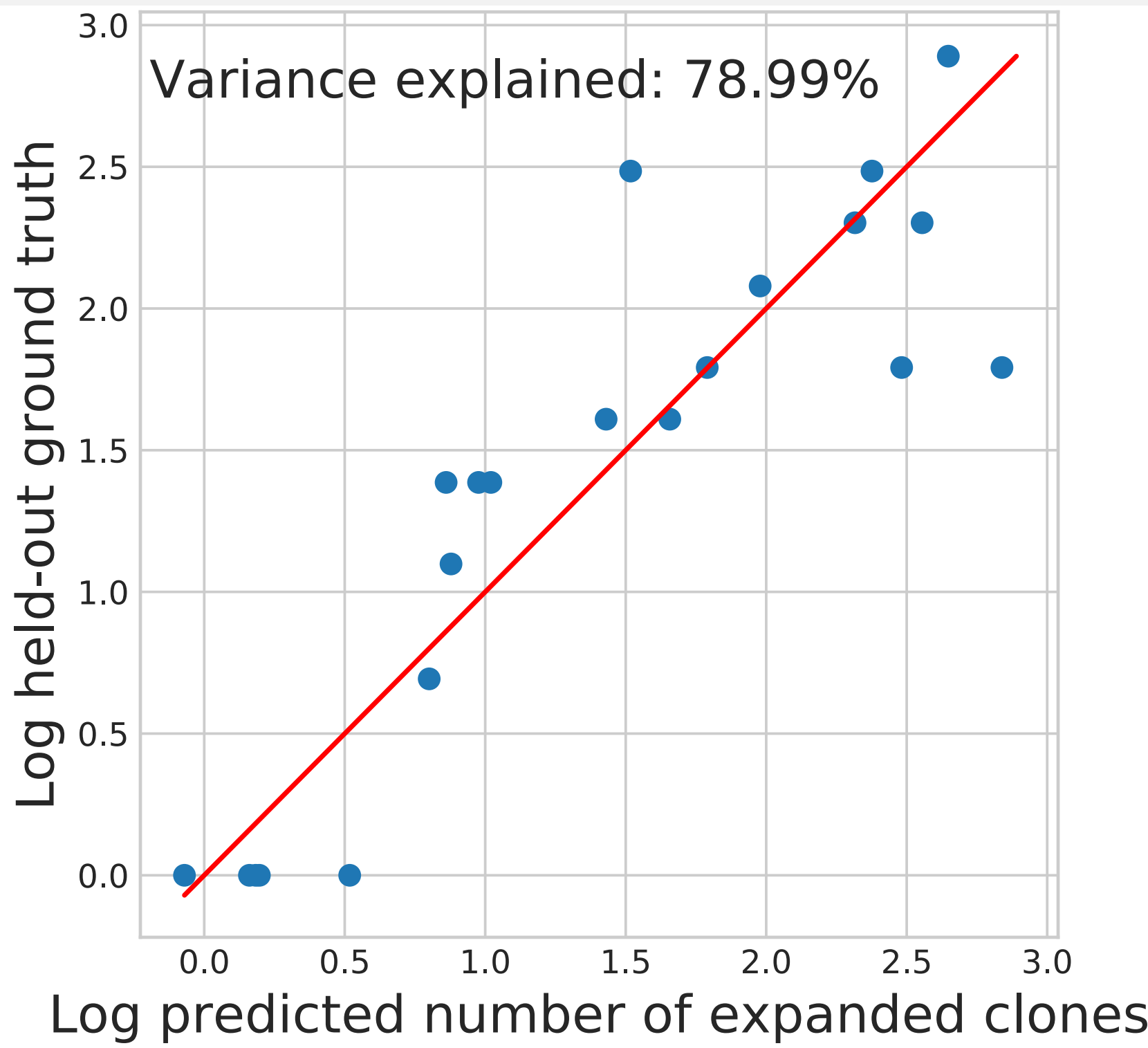
Cross-validation (CV): **Assess prediction accuracy on unseen patients** by repeating predictive analysis with each patient withheld

Permutation test: **Test whether cross-validation error significantly smaller** than error under permuted labeling ("Permutation Tests for Studying Classifier Performance," *Ojala & Garriga, JMLR, 2010*)

Our code: <https://github.com/lrg/multifactorial-immune-response>

Accurate prediction of TIL clone expansion

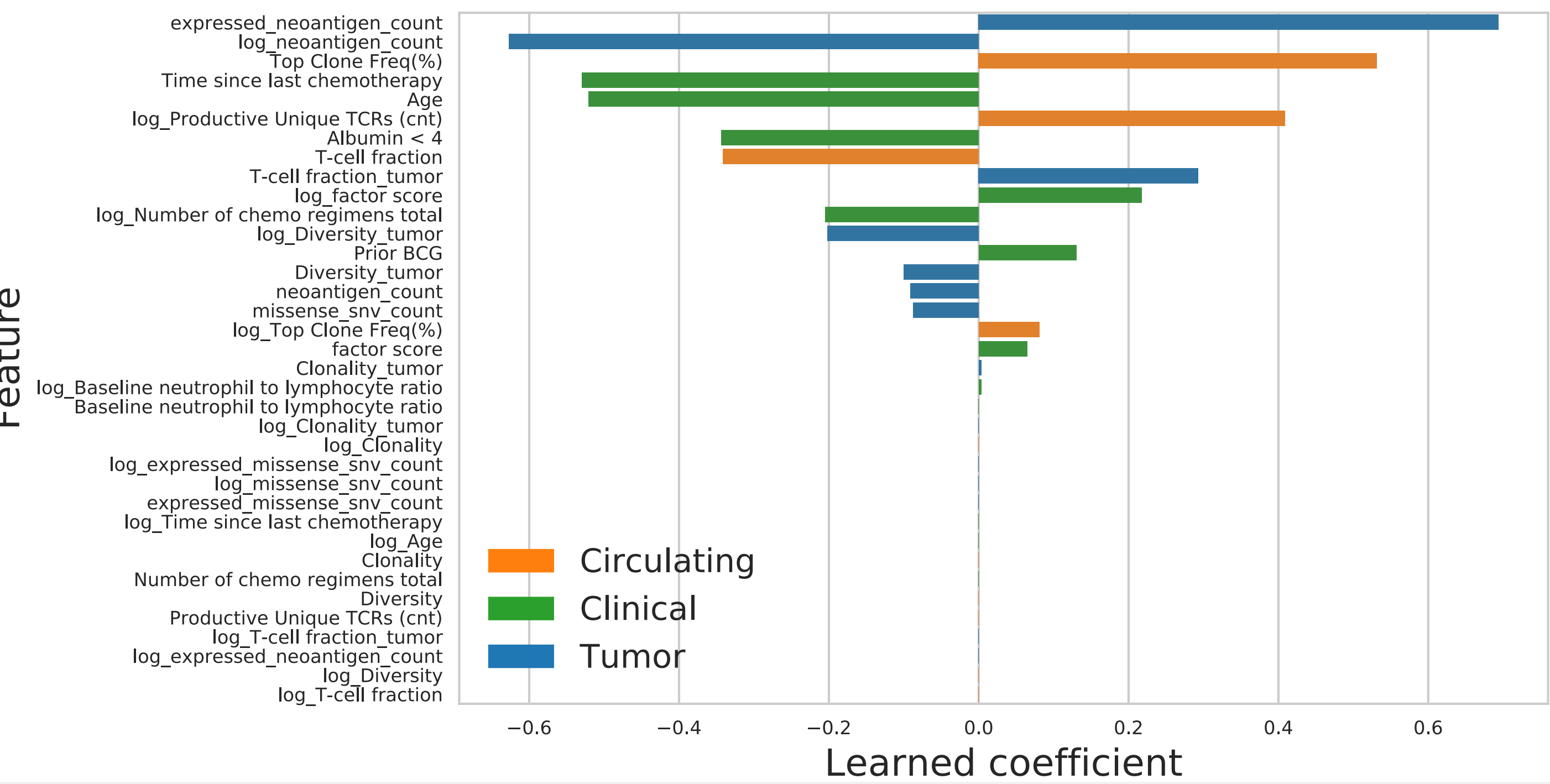
Learned model **explains 79% of variance in held-out patients**



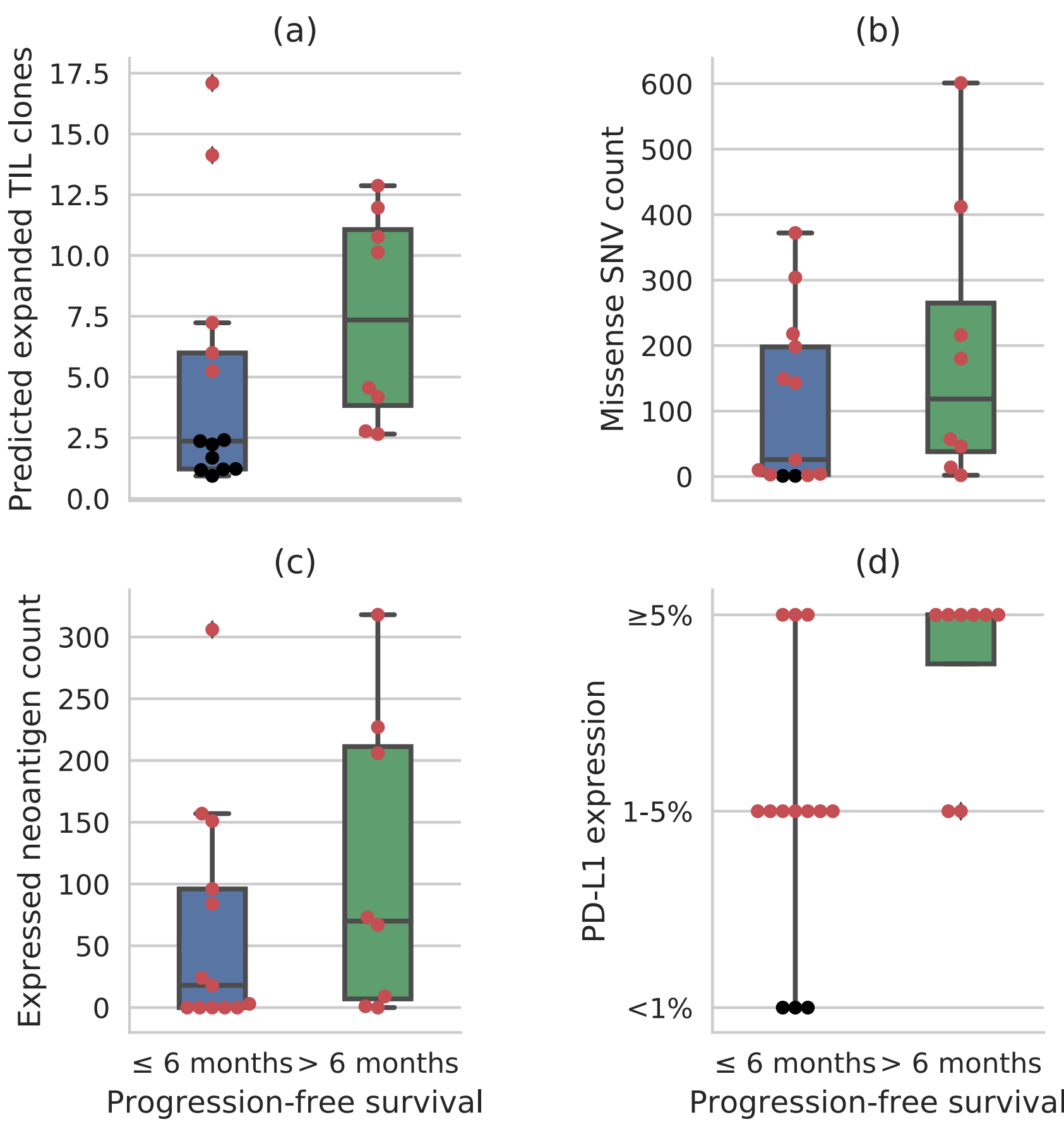
Clinical, tumor, and circulating feature importance

Clinical, tumor, and circulating features all contribute significantly

- **Only 26%** of variance explained if all tumor features excluded
- **Only 13%** of variance explained if all circulating features excluded
- **0%** of variance explained if all clinical features excluded



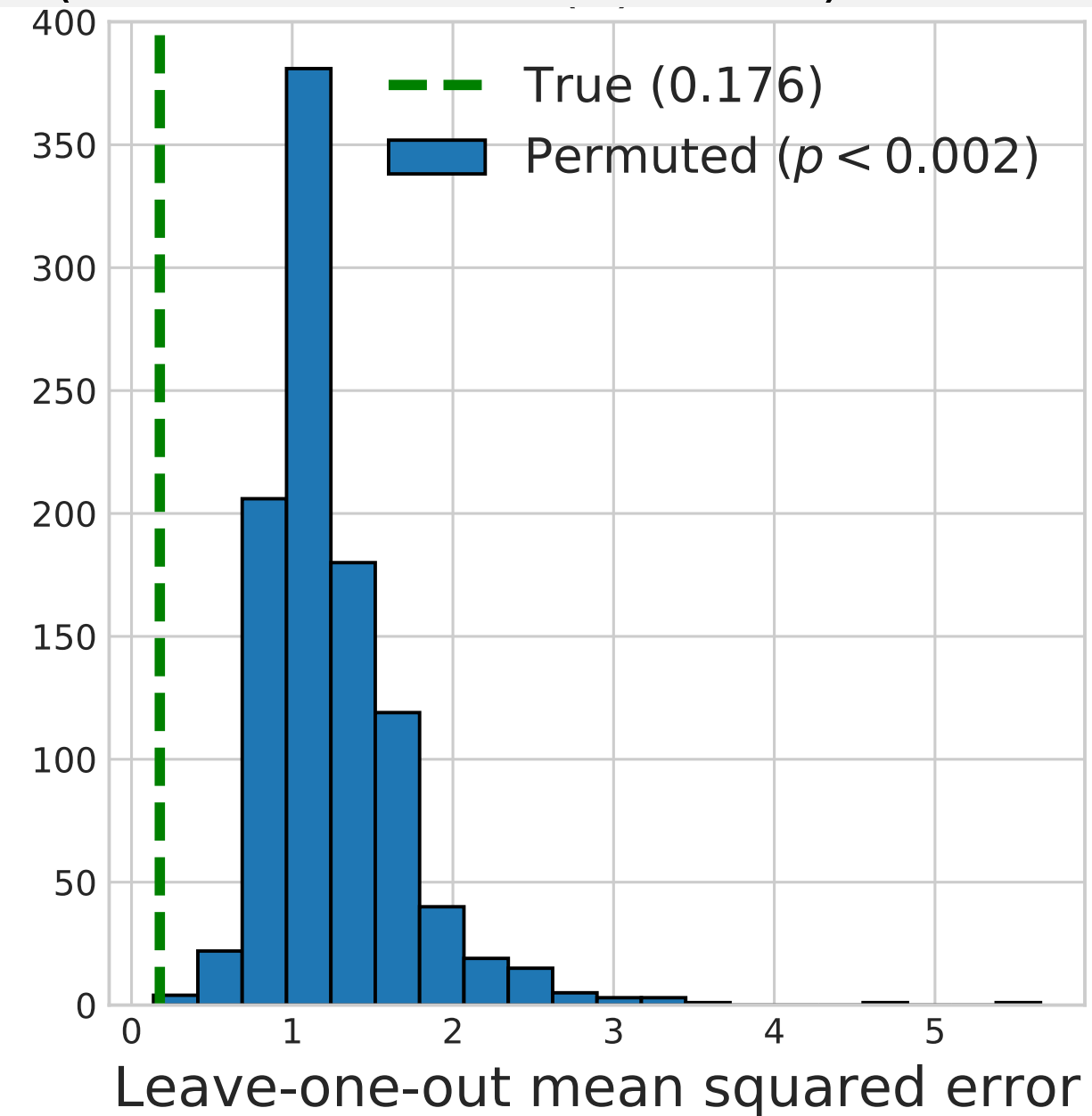
Implications for durable clinical benefit



Sanity-check test of CV error

Cross-validation error significantly worse when labels randomly permuted (1000 permutations)

- **Intuition:** CV error uninformative if comparable error achieved with wrong (permuted) labels
- Rules out perverse case where CV known to be unreliable (*Blum et al., COLT, 1999*)



Open questions

1. Can more predictive signal be extracted from previously completed studies of immunotherapy benefit?
2. What other cancer immunotherapy prediction problems could benefit from multifactorial modeling?
3. What value can patients or trial designers derive from accurate TIL expansion prediction?
4. What other "non-standard" phenotypes would be of value to predict?