Data-driven design of targeted gene panels for estimating immunotherapy biomarkers

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Abstract

- Exome-wide biomarkers such as tumour mutation burden (TMB) are useful predictors of response to immunotherapy.
- While whole-exome sequencing directly measures TMB, its cost prevents it from being standard-of-care.
- We develop a data-driven framework both for selecting targeted gene panels and for using them to intelligently estimate immunotherapy biomarkers.
- To do this, we utilise an exome-wide generative model of mutation, whose structure can be chosen to reflect biological assumptions.

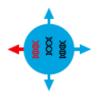
Outline

- Biological/Clinical Background
 - Cancer and immunotherapy
 - Exome-wide biomarkers
 - Targeted gene panels
- ② Generative/predictive modelling framework
 - Generative models: what and why?
 - From generative models to predictive models
- 3 Application to non-small cell lung cancer

Cancer is a disease of the genome

Cancer arises when DNA in cells changes (mutates) [1].





(a) Non-cancer cell: Normal DNA.

(b) Cancer cell: Mutated DNA.

Immunotherapy enables the immune system

Immunotherapy 'releases the brakes' on the immune system in order for it to attack tumours [2].



- (a) Low-damage cell: **unrecognisable** to immune system.
- (b) High-damage cell: **recognisable** to the immune system.

However, the immune system can only attack tumours that it recognises!

Tumour mutation burden stratifies patients for immunotherapy response

As a simple proxy for likelihood of immune response, we can use **tumour mutation burden**: the total count of non-synonymous mutations throughout the tumour exome [3, 4].

Advantages:

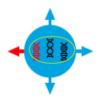
- Calculated from DNA sequencing only (compatible with liquid biopsy [5]).
- Relevant across cancer types.

Disadvantages:

- Requires Whole-Exome Sequencing (WES) to measure directly.
- Fails to incorporate other features of the immune environment.

Targeted panels make genomic biomarkers viable

Rather than the entire exome/genome, targeted gene panels sequence a **subset** of genes.





- (a) Whole-exome sequencing: **all genes** sequenced.
- (b) Targeted panel sequencing: **subset** of genes sequenced.

Problem Statement

Aims:

- Select a discrete set of genes to form a panel (of pre-specified cost).
- Estimate TMB from panel sequencing data.

Mutations and biomarkers: some notation

We'll use i, g, and s to refer to:

- **1** A sample i (ranging from i = 1 to i = N),
- ② A gene g (belonging to some index set G),
- **3** A variant type *s* (for example synonymous/non-synonymous).

Finally, we write:

 M_{igs} , to refer to a mutation count of a particular sample, gene, and mutation type.

We then can define the TMB of the *i*th sample via:

$$T_i = \sum_{\substack{g \in G \ s: \ ext{non-synonymous}}} M_{igs}.$$

What is a generative model of mutation?

A generative model of mutation attempts to capture the **underlying distribution** of variants across the entire exome/genome.

A simple example (Poisson):

$$M_{igs} \sim \text{Poisson}(\lambda_{gs})$$
 (1)

Why use generative models of mutation?

- Utilise known biology to inform the structure of the model.
- (Sometimes) make interpretable inferences.
- Generate trust that we've thought about it!

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• Utilise known biology to inform the structure of the model.

Choice of distribution, exome organisation, parameter space.

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Parameter estimates, post-model inference.

Generate trust that we've thought about it!

Generative models to predictive models

Aims:

- Select a discrete set of genes to form a panel (of pre-specified cost).
- ② A mechanism for estimating TMB from panel sequencing data.

Proposed mechanism: Linear estimator.

$$\hat{T} = \sum_{g,s} w_{gs} M_{igs} \tag{2}$$

Optimisation procedure:

$$\hat{w} := \arg\min_{w} \{ \mathbb{E}[(\hat{T}_i - T_i)^2] + \lambda |w| \}, \tag{3}$$

where λ is a penalty specifying panel cost, and the expectation $\mathbb E$ is over the distribution learned by our generative model.

NSCLC: Generative Model Structure

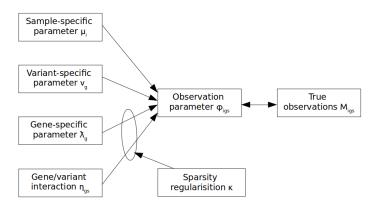


Figure: Generative model overview, as applied to NSCLC dataset [6].

NSCLC: Poisson-based generative model fit



Figure: Generative model parameters. [7]

NSCLC: Poisson-based generative model fit

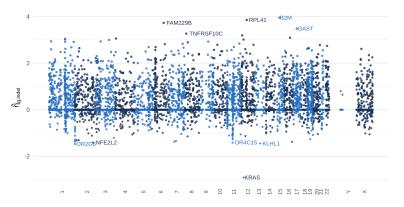


Figure: Generative model parameters (indel-specific) [7].

NSCLC: Estimation performance

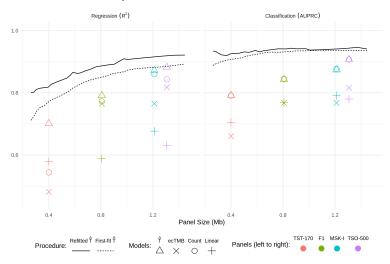


Figure: Predictive model performance: comparison with commercial panels and other predictive methods [7, 8].

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