The Search for Genetic Answers

Identifying candidate variants in families with unsolved rare genetic diseases

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Overview

1. Rare Genetic Diseases

- Rare genetic diseases collectively affect 1 in 16 people worldwide.(1)
- Diagnosis of a patient's causal genetic variant is essential for disease management but only 50% of patients are diagnosed following initial genetic evaluation.(2,3)

2. Pre-mRNA Splicing

- Normal protein production relies on correct mRNA processing including pre-mRNA splicing, a process primarily regulated by intronic sequences.
- Variants which disrupt these sequences are estimated to cause 15-60% of genetic diseases, but are underdiagnosed by current diagnostics.(4,5)

3. Approach

- Massively parallel sequencing (MPS) has improved diagnosis by an additional 10-40%.(6,7)
- This study combined whole exome (WES), genome (WGS), and RNA sequencing (RNAseq) data, to try identify candidate causal variants in 5 undiagnosed families.

Hypothesis

Causal genetic variants which disrupt pre-mRNA splicing and/or occur in novel disease genes are enriched in cohorts of undiagnosed families with rare genetic diseases.

Aim

Create a bioinformatics workflow using massively parallel sequencing data and in-silico tools to identify candidate pathogenic variants, especially those disrupting pre-mRNA splicing.

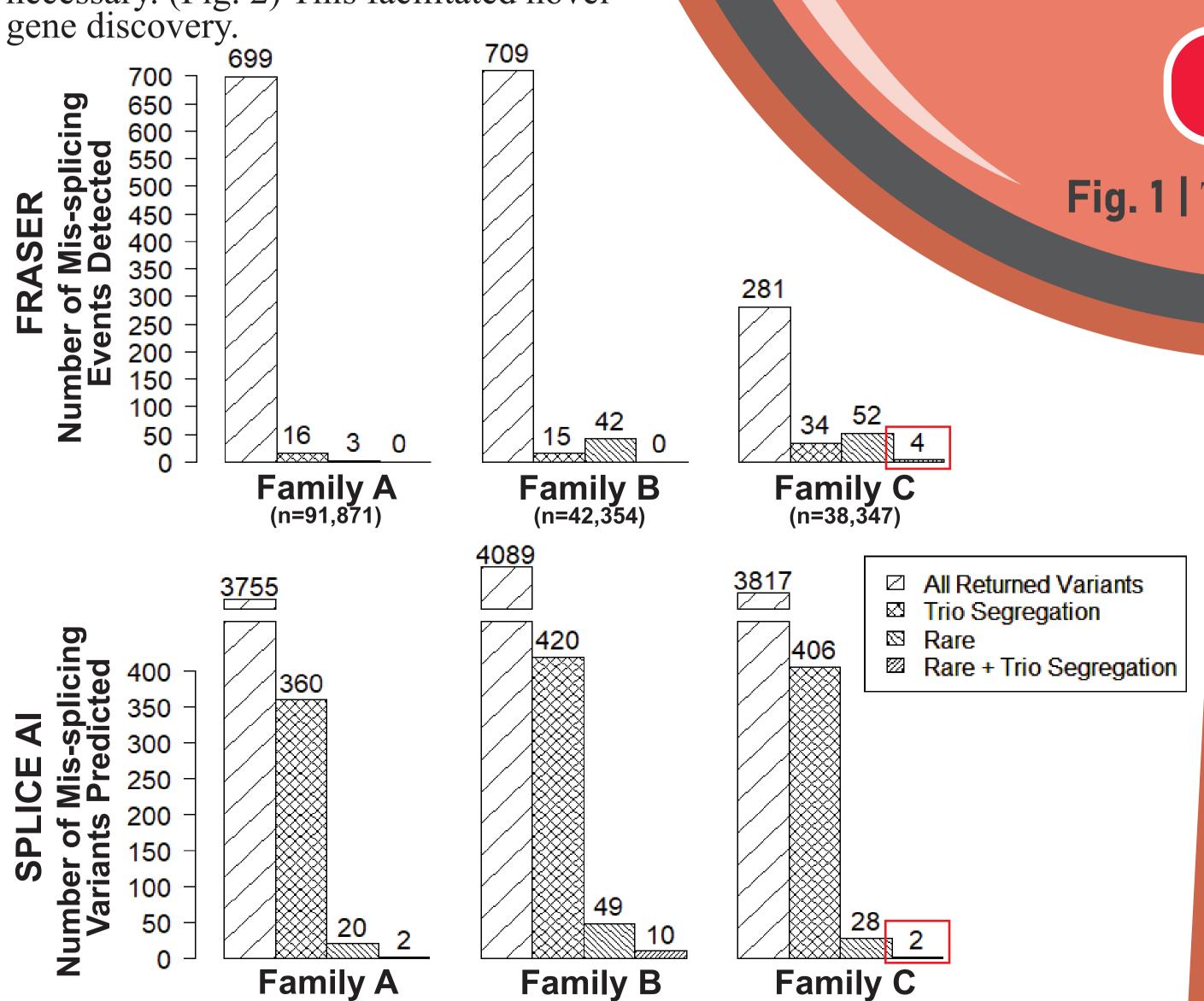
Results

The workflow utilised three MPS technologies, three computational tools, and three filtering parameters (Fig. 1).

For the three families with WGS and RNAseq, we focussed on mis-splicing variants via two separate approaches:

- 1. WGS + SpliceAI variant mis-splicing prediction (8)
- 2. RNAseq + FRASER mis-splicing event detection (9)

Both approaches returned candidate variants using low thresholds on tool score and variant population frequency.(10) The power of the tools and trio variant segregation was such that filtering by gene lists was not necessary. (Fig. 2) This facilitated novel



(n=1,133,269)

WGS RNAseq **WES** n=3 n=3 n=5 **FRASER** Splice Al Segr **Exonic Variants** Mis-splice Detection Mis-splice Prediction

Trio

Trio

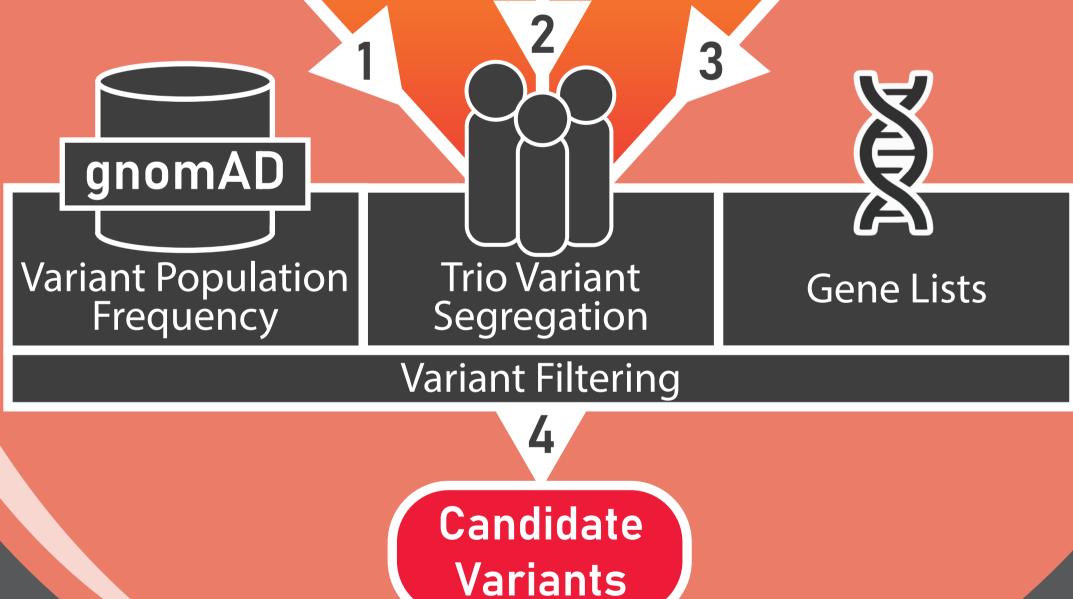


Fig. 1 | The final workflow

CDK5RAP3

This mis-splicing analysis workflow revealed I candidate variant in a novel disease gene, CDK5RAP3, in Family C (Fig. 3).

This gene is not associated with human disease but is embryonic lethal in mouse knock-out models.

Additionally, several disease genes in the same biochemical pathway as *CDK5RAP3* cause a distinct phenotype similar to the patient's in this family.

Further in-vitro analysis revealed that the variant segregated with disease in the family and thus, this intronic variant is likely responsible for the rare disease.

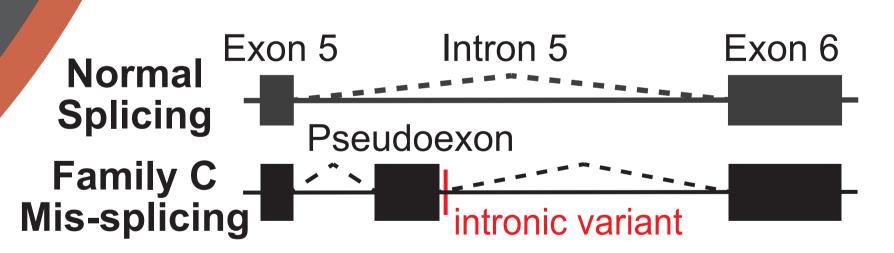


Fig. 3 | An intronic variant in Family C in CDK5RAP3 disrupts normal splicing by activating a pseudoexon

Conclusions & Future Directions

By leveraging two powerful tools to identify mis-splicing at both the DNA and RNA level, this workflow has provided a genetic answer for 1 of 5 analysed families with unsolved rare genetic disease. This result supports our hypothesis.

Further validation with both solved and unsolved cases will be essential to understand the workflow's limitations. However, even in its current form, this approach is likely to increase diagnostic rates of families with unsolved rare genetic diseases.

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(n=1,205,172)

Fig. 2 The final workflow returns candidate mis-splicing variants without

gene list filtering. Red boxes highlight *CDK5RAP3* variant in Family C

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(n=1,121,428)