

Identifying genetic markers that modify risk of carbamazepine-induced Stevens-Johnson Syndrome

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Introduction

- Carbamazepine (CBZ) is a highly effective medication for people with epilepsy.
- Steven Johnson Syndrome (SJS,) a severe disorder of the skin, is a rare but life-threatening adverse reaction to CBZ treatment.

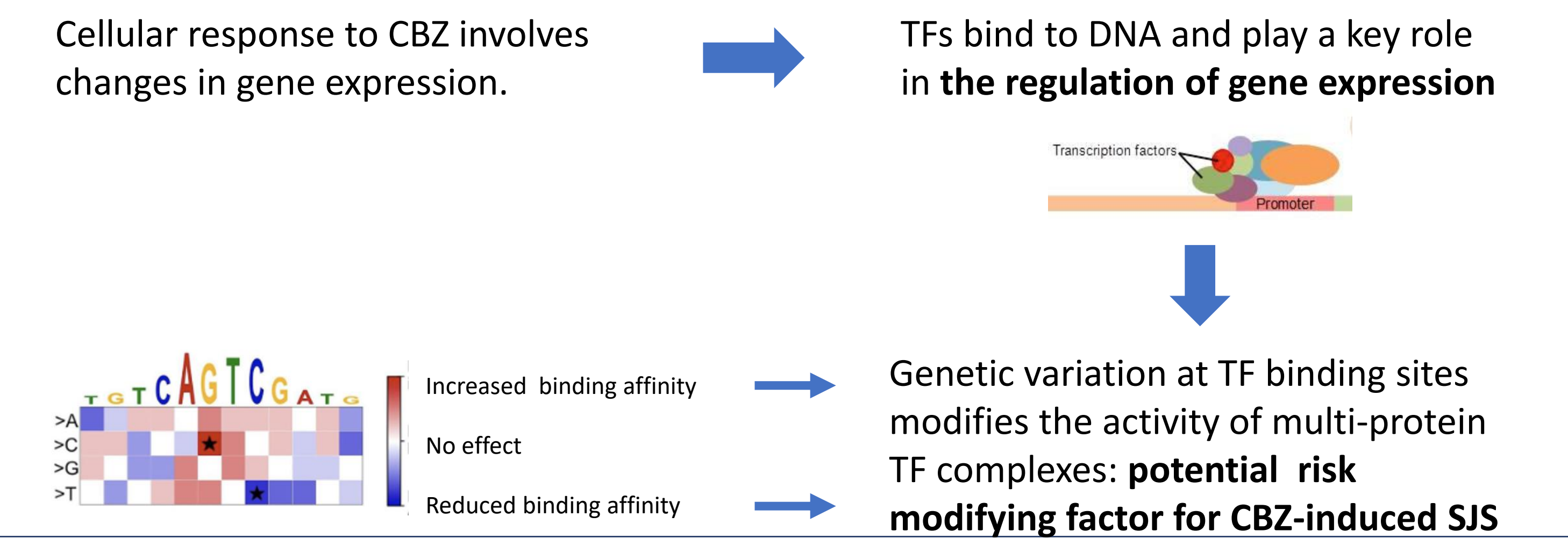


- In Han Chinese HLA-B*15:02 is associated with risk but is neither necessary nor sufficient so other genetic variations likely modify risk.

Objective: Identify additional genetic risk markers

Hypothesis

Hypothesis: Genetic variations in regions of the DNA where transcription factors (TF) bind modify risk of CBZ-induced SJS



Methods

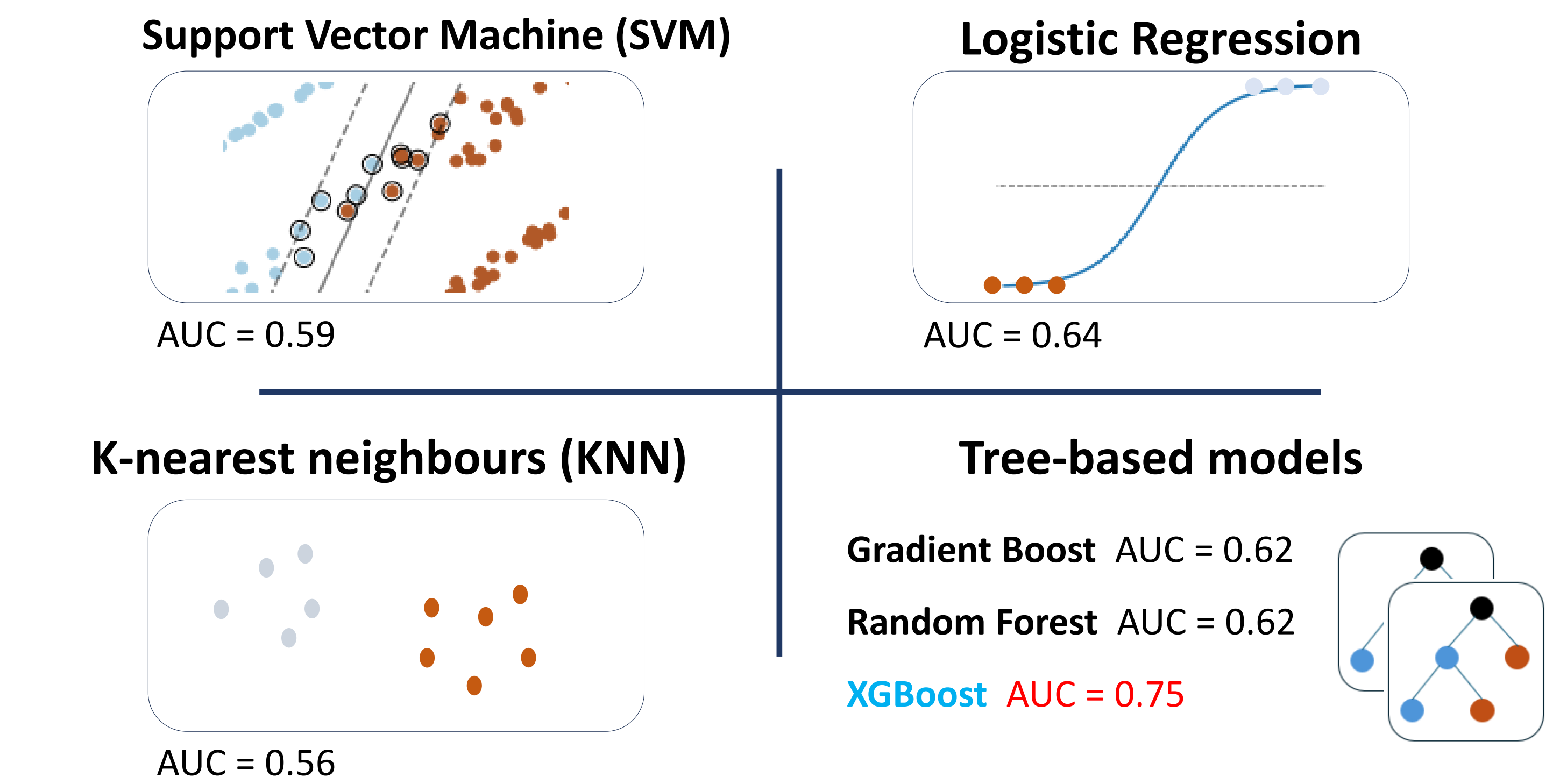
Step 1: Evaluate predictive performance of machine learning (ML) models

- Existing whole genome data obtained from 117 individuals with CBZ-induced SJS and 81 CBZ-tolerant.
- No one model is best in all situations. It depends on the underlying structure in the data
- Applied 6 ML models with 5-fold-cross validation to find the model with the best AUC for the known risk region (HLA-B region in chromosome 6)

Step 2: Use the best model to systematically look for risk in regions of the DNA where TFs bind

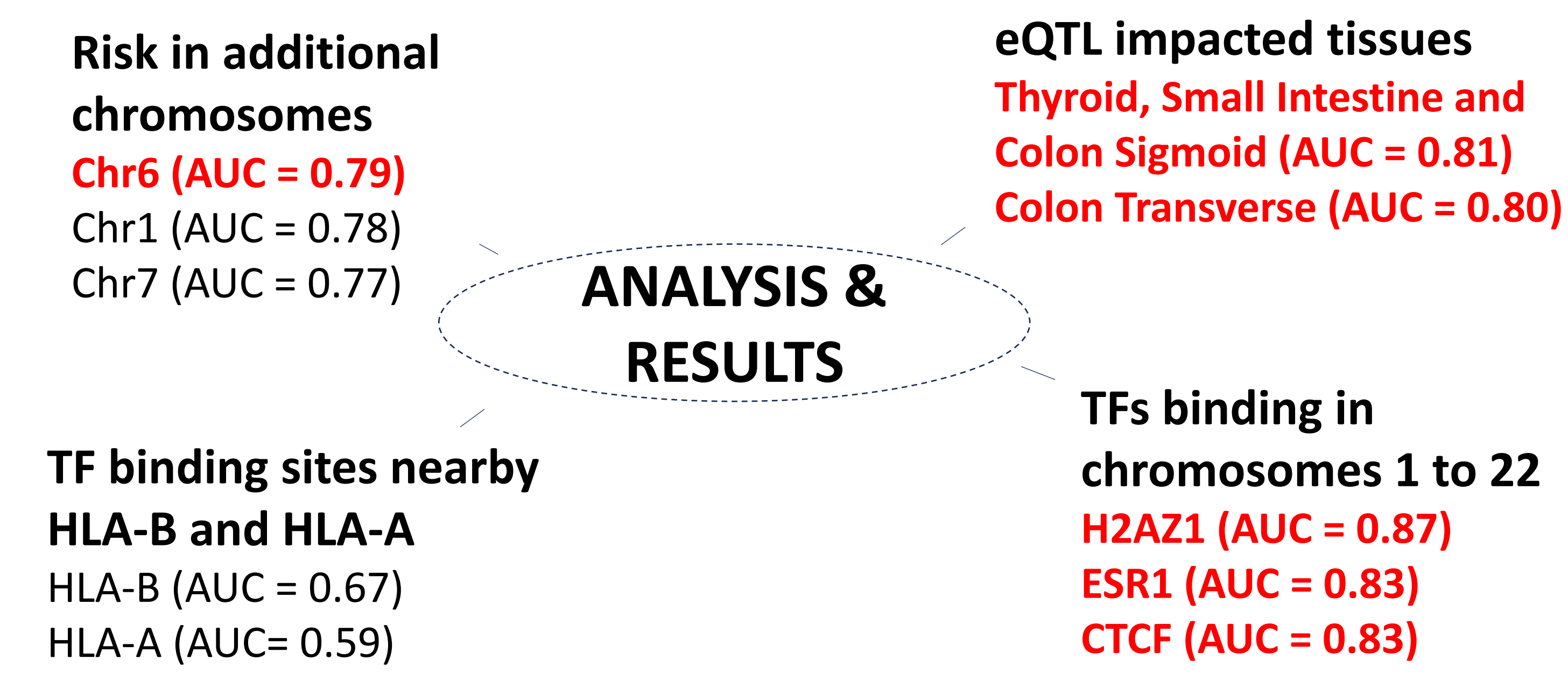
- Expression quantitative trait loci (eQTL) data was used to find a signal specific to a tissue type
- TF binding sites were obtained from the Gene Transcription Regulation Database (GTRD)

Results step 1: Best model



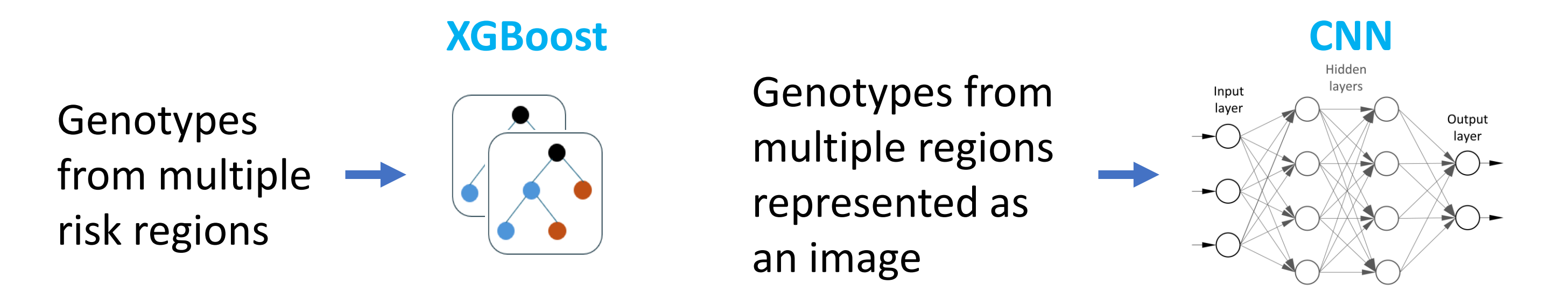
Result: XGBoost performed with an Area Under the Curve (AUC) 0.75 the best

Results step 2: Identifying risk modifying regions



Next steps: Use deep learning model to predict risk

Hypothesis: An image-based model (e.g., Convolutional Neural Network [CNN]) will capture relationships between risk modifying regions and outperform the XGBoost model



Conclusion

This study has identified regions in the DNA where genetic variations potentially modify risk of CBZ-induced SJS. This work will facilitate future targeted deep-learning approaches to identify robust risk biomarkers.

Acknowledgement

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References

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2. SJS image from <https://dermnetnz.org/topics/sjs-ten-images> licensed under Creative Commons Attribution-NonCommercial-NoDerivs 3.0 (New Zealand)

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