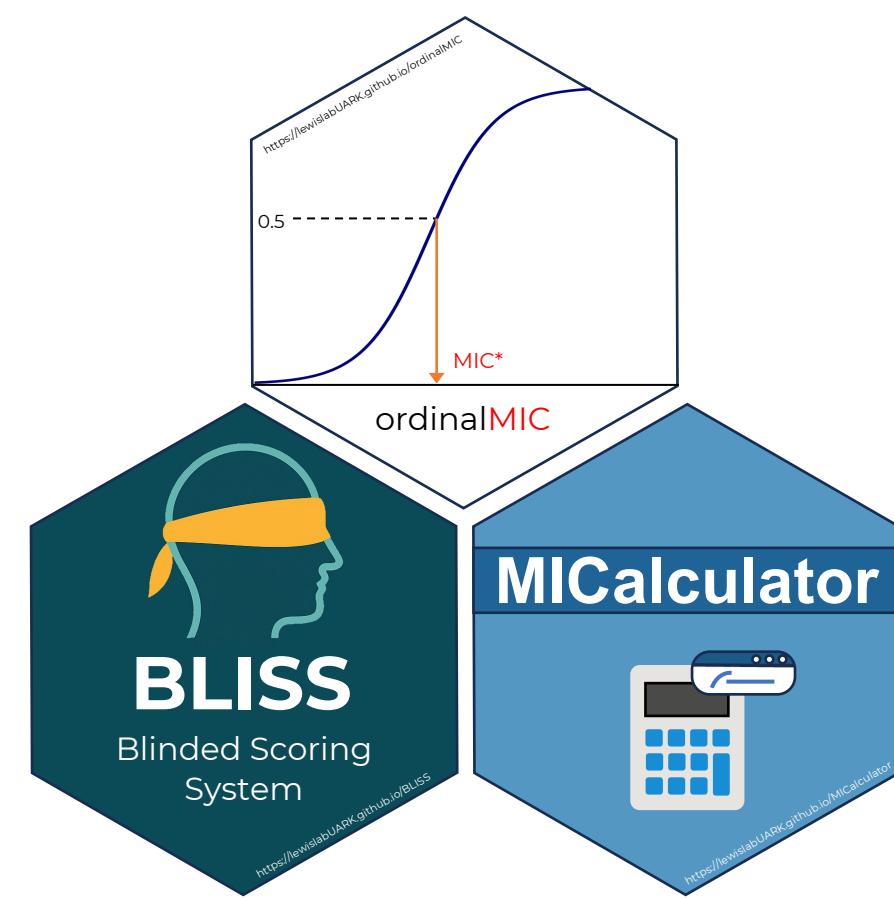


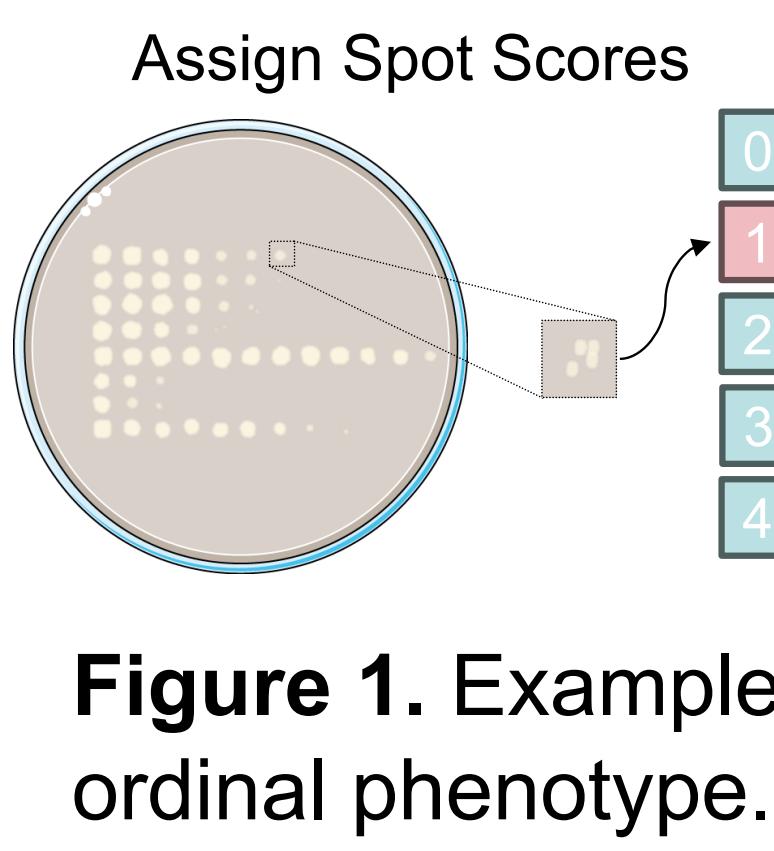
Interpretable Ordinal Analysis for Complex Designs in Cell and Molecular Biology

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The Challenge

- Complex phenotypes are commonly ordinally measured by scoring.
- Rapid to Collect
- Difficult to Analyze
- Testing Assumptions
- Ordinal Regression¹
- Powerful
- Uncommon²



The Solution

Convert ordinal scores into a quantitative, interpretable metric we call MIC^* .

$$P(Y \leq 0 | X, X_C = MIC^*) = 0.5$$

- Derived from ordinal regression model
- Biologically interpretable
- $MIC = \text{Minimum Inhibitory Concentration}$

Statistical Framework

Proportional Odds (PO) Model:

$$\text{logit}(P(Y_i \leq j)) = \alpha_j - X_i \beta$$

The MIC^* Metric:

$$MIC^* = g^{-1} \left(\frac{\alpha_0 - \sum \beta_k X_k}{\beta_C + \sum \beta_{Ck} X_k} \right)$$

Group Differences:

$$\Delta MIC^* = MIC^*_{B} - MIC^*_{A}$$

$$\Delta \log_2 MIC^* = \log_2 MIC^*_{B} - \log_2 MIC^*_{A}$$

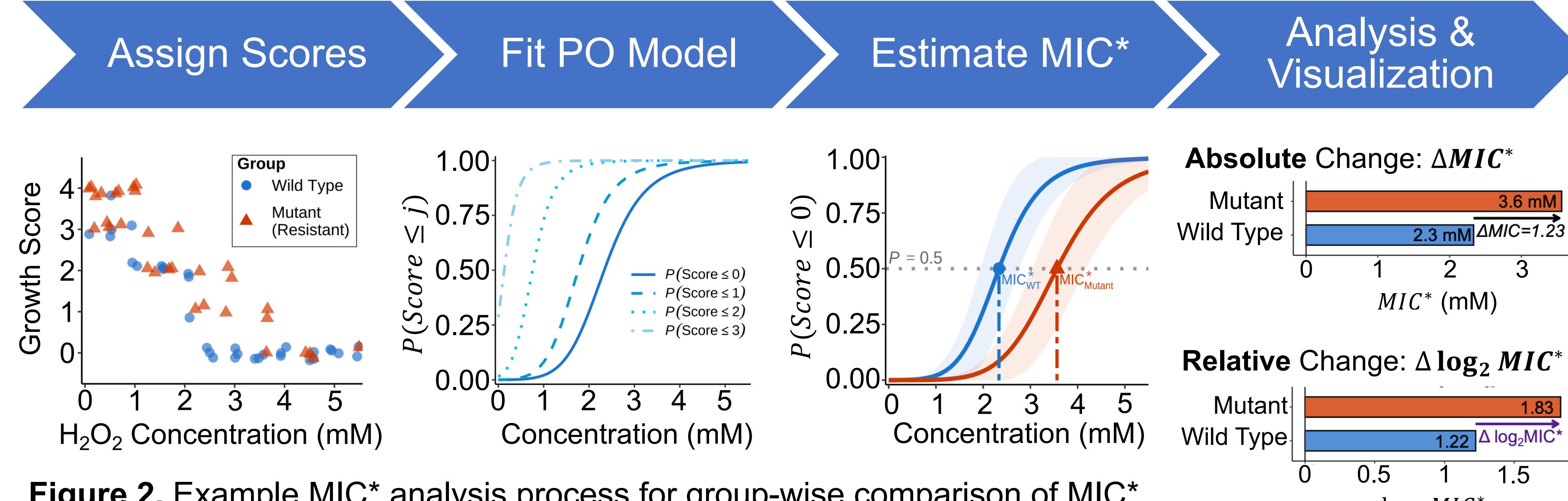
Variance Estimation:

$$V(\Delta MIC^*) = (\nabla g_B - \nabla g_A)^T \sum (\nabla g_B - \nabla g_A)$$

The *ordinalMIC* Software Suite

Tool	Purpose	Benefit	Access
BLISS Web App	Blinded Scoring Assess Consistency	Reduce Bias Browser Based	lewislabUARK.github.io/ BLISS
ordinalMIC R package	Perform Core Statistical Analysis	Powerful & Flexible	lewislabUARK.github.io/ ordinalMIC
MICCalculator Web App	Point-and-Click Tool for MIC^* Analysis	Increased Accessibility	lewislabUARK.github.io/ MICCalculator

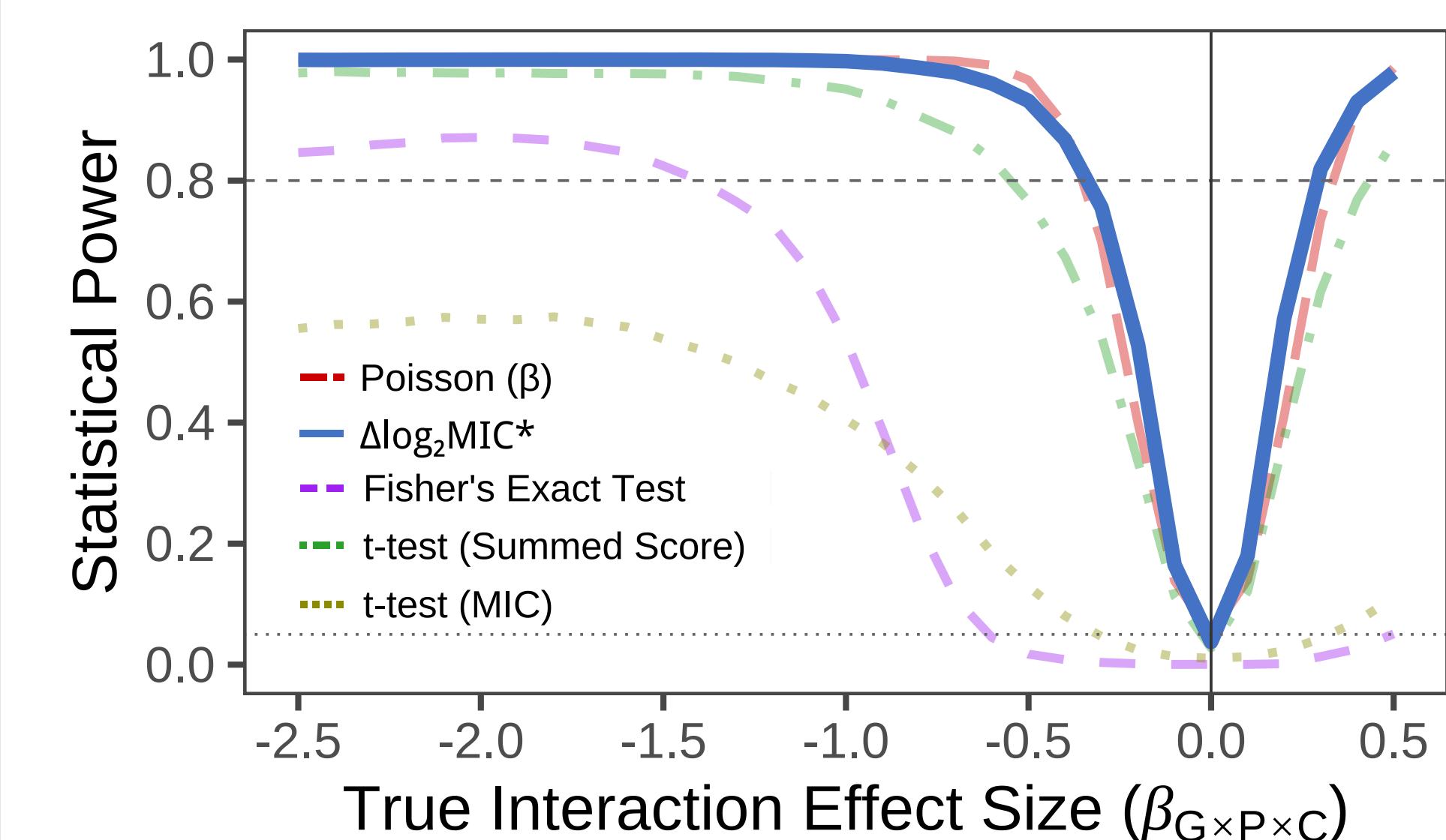
Table 1: Resources available for ordinal scoring and MIC^* analysis.



Estimator Performance

MIC^* is Powerful Estimator

Monte Carlo Simulations ($N=10^7$) of factorial GxE design confirm framework's utility.



- Sensitive:** Achieves >80% power to detect 41% increase in expected count.
- Specific:** Nominal Type I error rate (~5%)
- Accurate:** Low bias with mean bias 0.0026 (IQR: -0.011 to 0.018) for MIC^* estimates

Ex 1: Quantifying GxE

GxE = Genotype × Environment Interaction

Question: Does deletion of *CTT1* abolish salt-induced peroxide resistance in yeast?

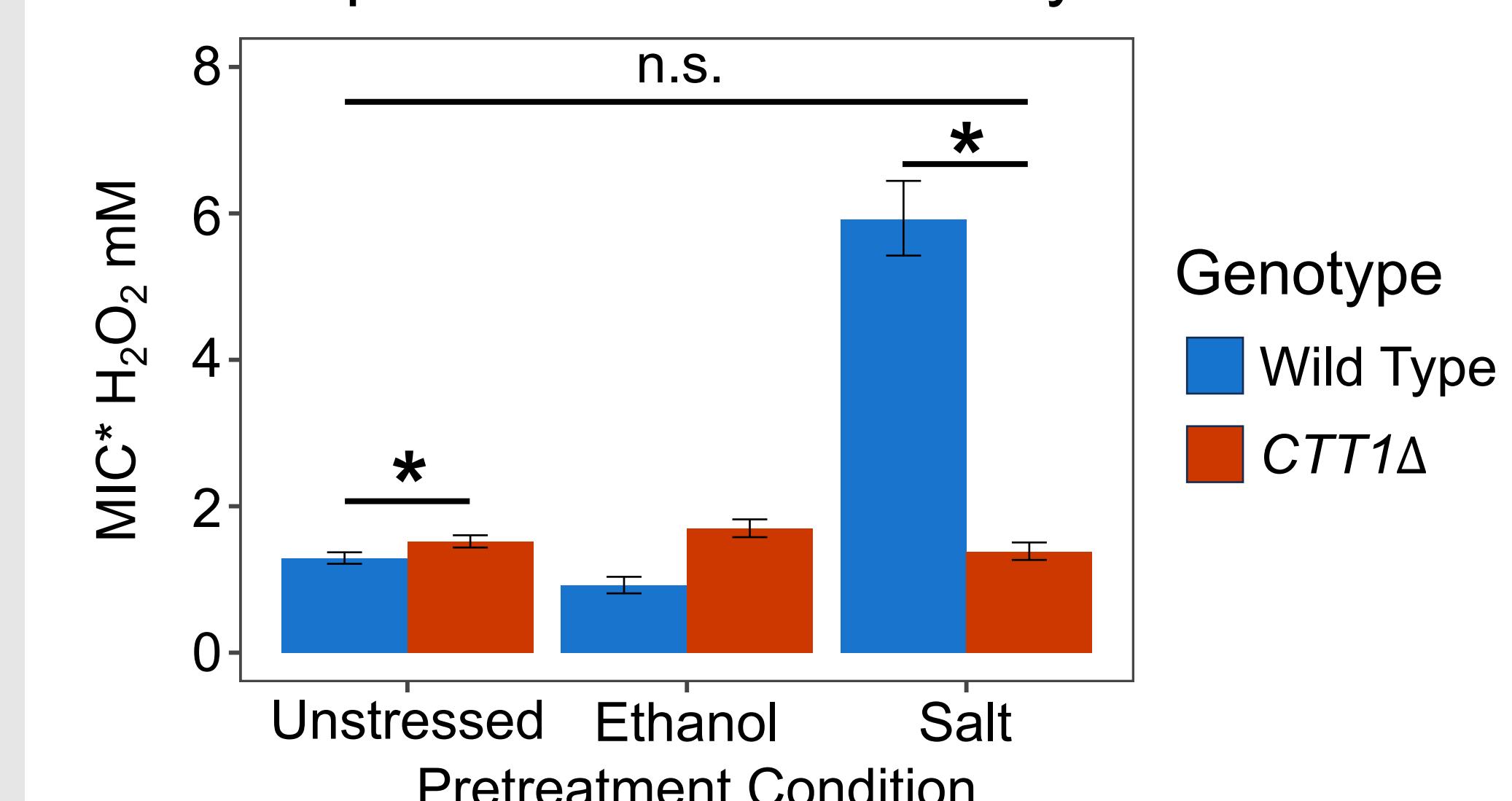


Figure 3. MIC^* for H_2O_2 resistance of Wild Type and $CTT1\Delta$ mutants following salt pretreatment.³ Error bars represent 95% CI, * denotes $p < 0.05$.

Finding: Yes. Resistance phenotype for salt is completely dependent on the gene *CTT1*.
 $\Delta MIC^* = 0.09$ (95% CI: -0.04 to 0.23 mM H_2O_2)

Ex 2: Trait Mapping

Question: Can spot assays quantify ethanol-induced peroxide resistance inheritance?

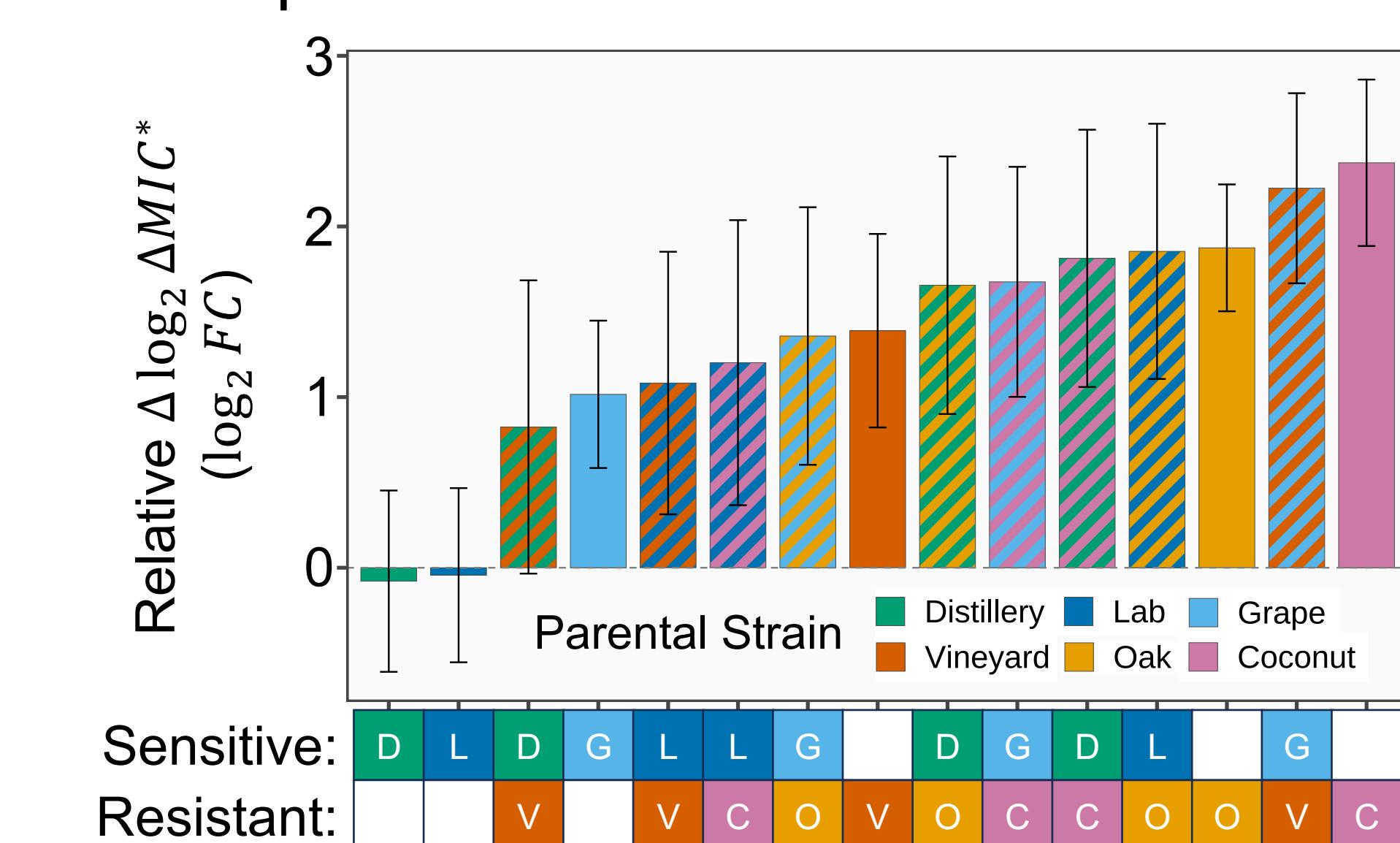


Figure 4. MIC^* for H_2O_2 resistance of wild yeast strains and crosses following ethanol pretreatment versus no pretreatment. Error bars represent 95% CI

Finding: MIC^* reveals majority exhibit an intermediate phenotype, while the Grape × Vineyard cross exhibited enhanced resistance.

Take-Home Messages

- Stop Summing Scores.** Conventional analyses of ordinal data are underpowered
- Translate Scores to Biology.** The MIC^* framework converts ordinal scores into intuitive quantitative metric.
- New Tools Available to Help.** We provide complete, open-source workflow for research

References

- [1] Agresti, A., 2010. *Analysis of ordinal categorical data*. John Wiley & Sons.
- [2] Agresti, A. and Tarantola, C., 2018. Simple ways to interpret effects in modeling ordinal categorical data. *Statistica Neerlandica*, 72(3), pp.210-223.
- [3] Scholes, A.N., Stueker, T.N., Hood, S.E., Locke, C.J., Stacy, C.L., Zhang, Q. and Lewis, J.A., 2024. Natural variation in yeast reveals multiple paths for acquiring higher stress resistance. *BMC biology*, 22(1), p.149.

Acknowledgements

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