

Methylation of Transposable Elements and Gene Expression in *Arabidopsis Thaliana*

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27 November 2025



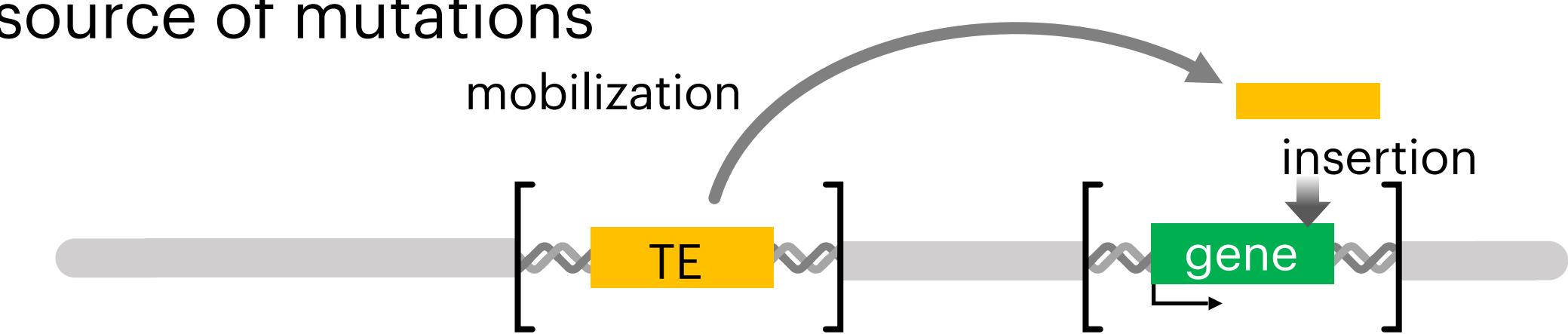
Contents

- Background: transposable elements and methylation
- Motivation: to explain GWAS findings
- Model: to understand methylation spreading
- Conclusions

Transposable Elements

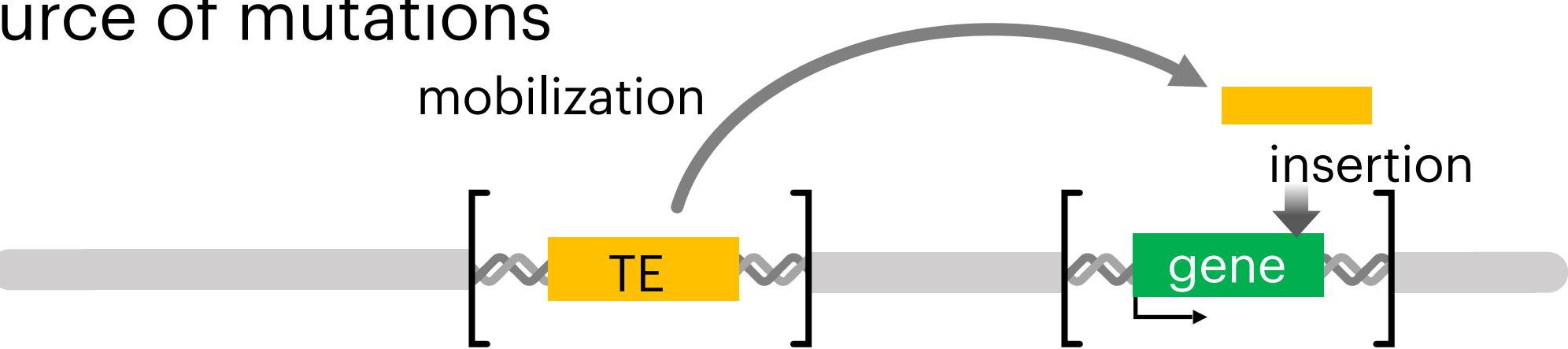
Transposable Elements

- Transposable Elements (TEs, “jumping genes”) are an important source of mutations
- TEs transpose by cut-and-paste or copy-and-paste mechanisms
- BUT: most TEs are degraded and do not transpose



Transposable Elements

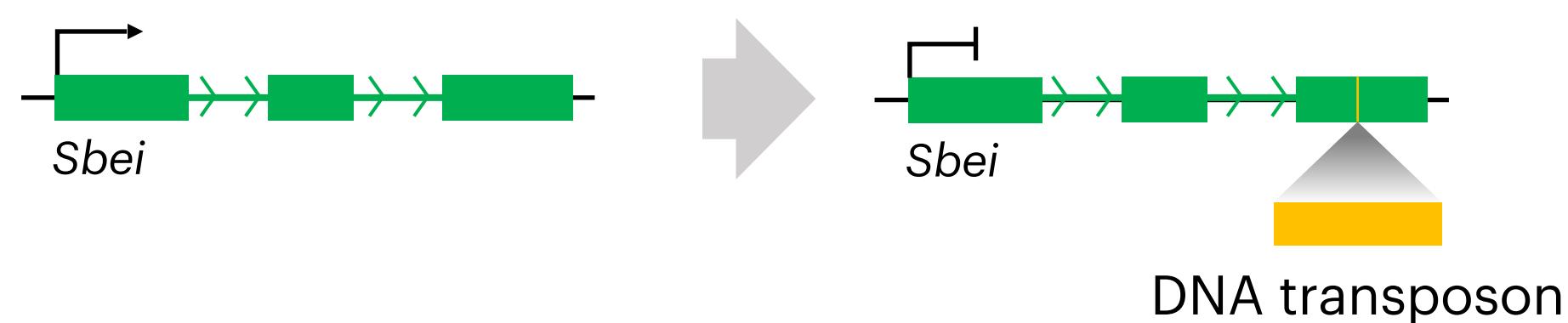
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Mutations may be **deleterious**...

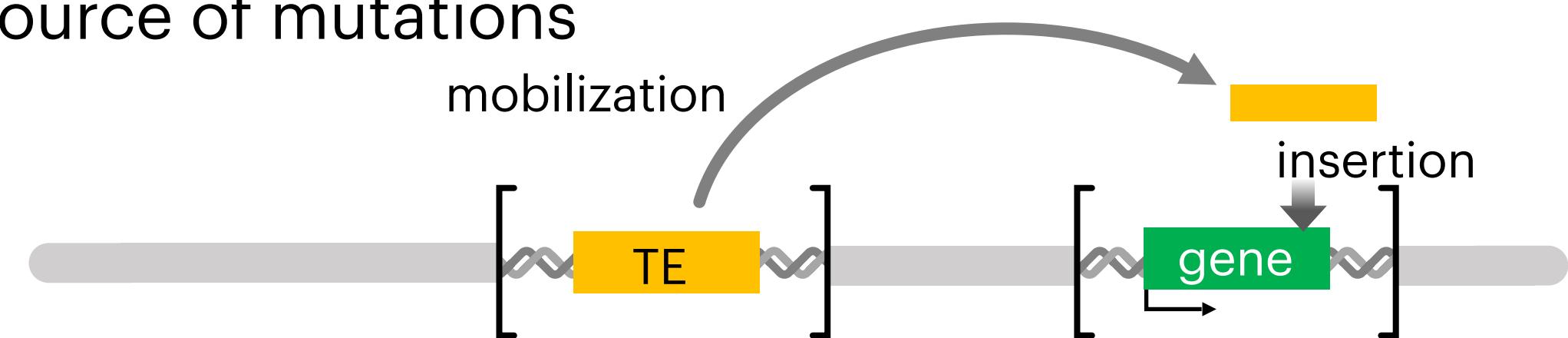


Bhattacharyya et al. Cell 1990



Transposable Elements

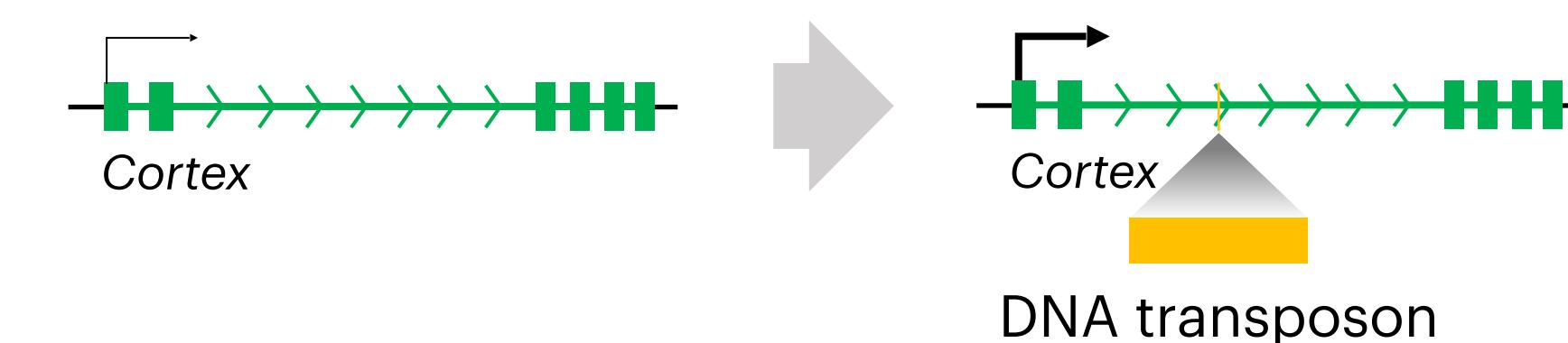
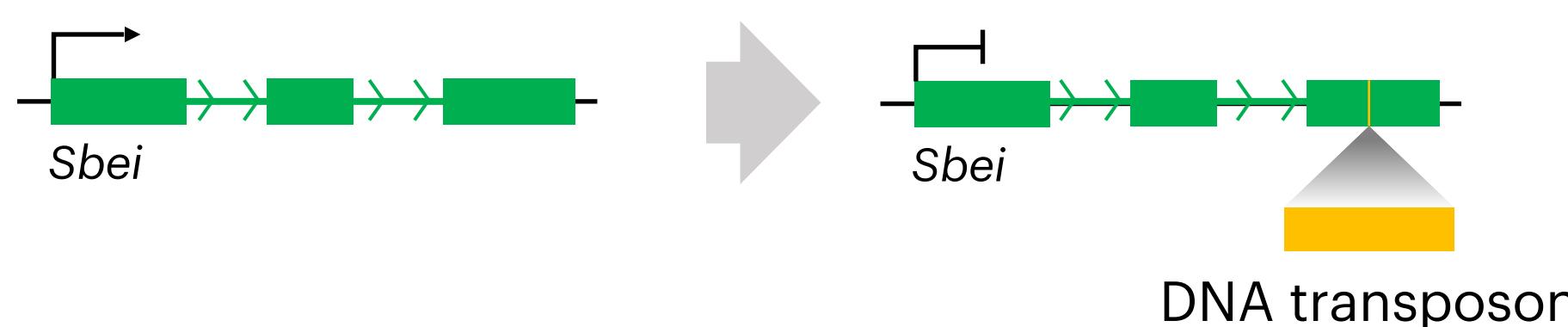
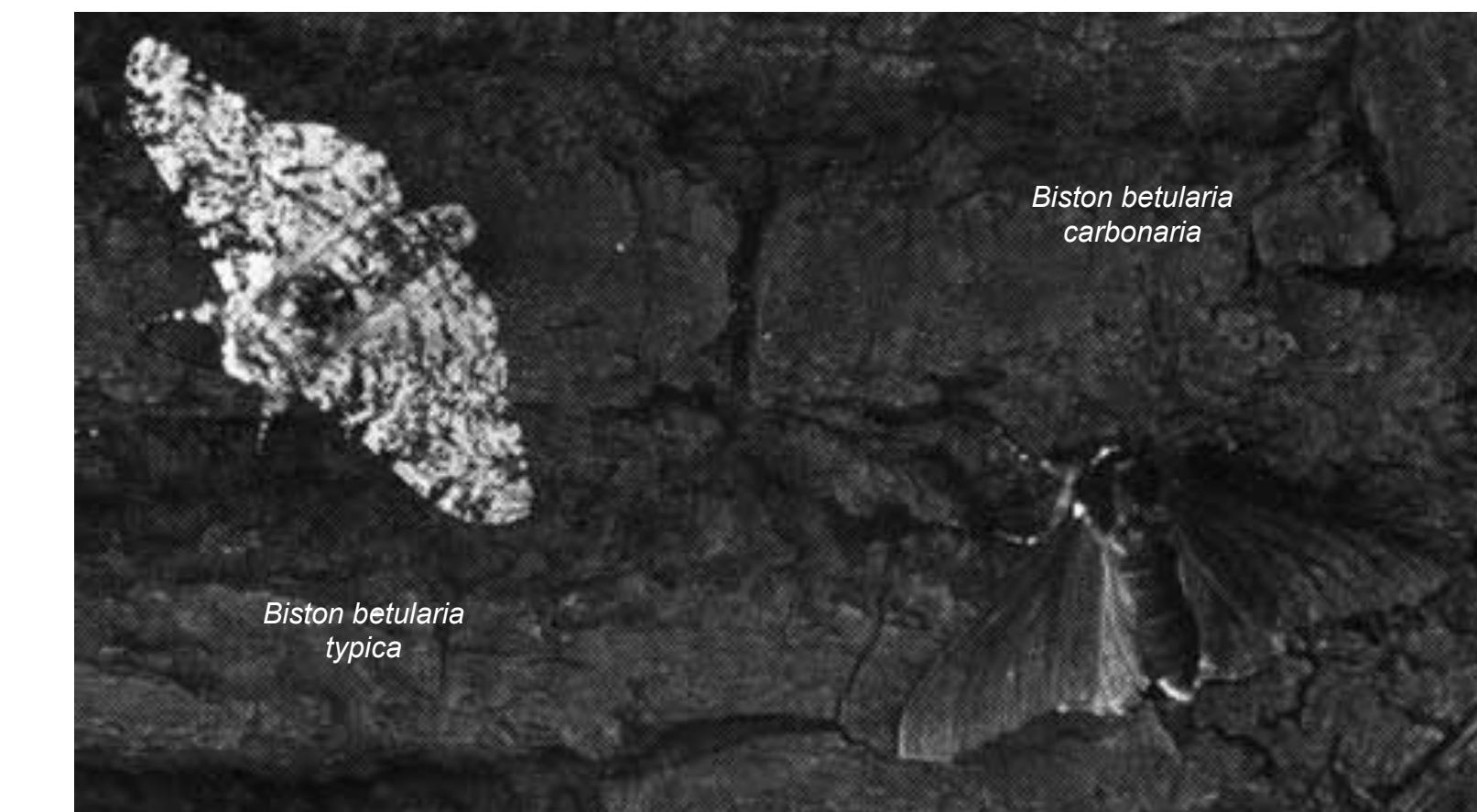
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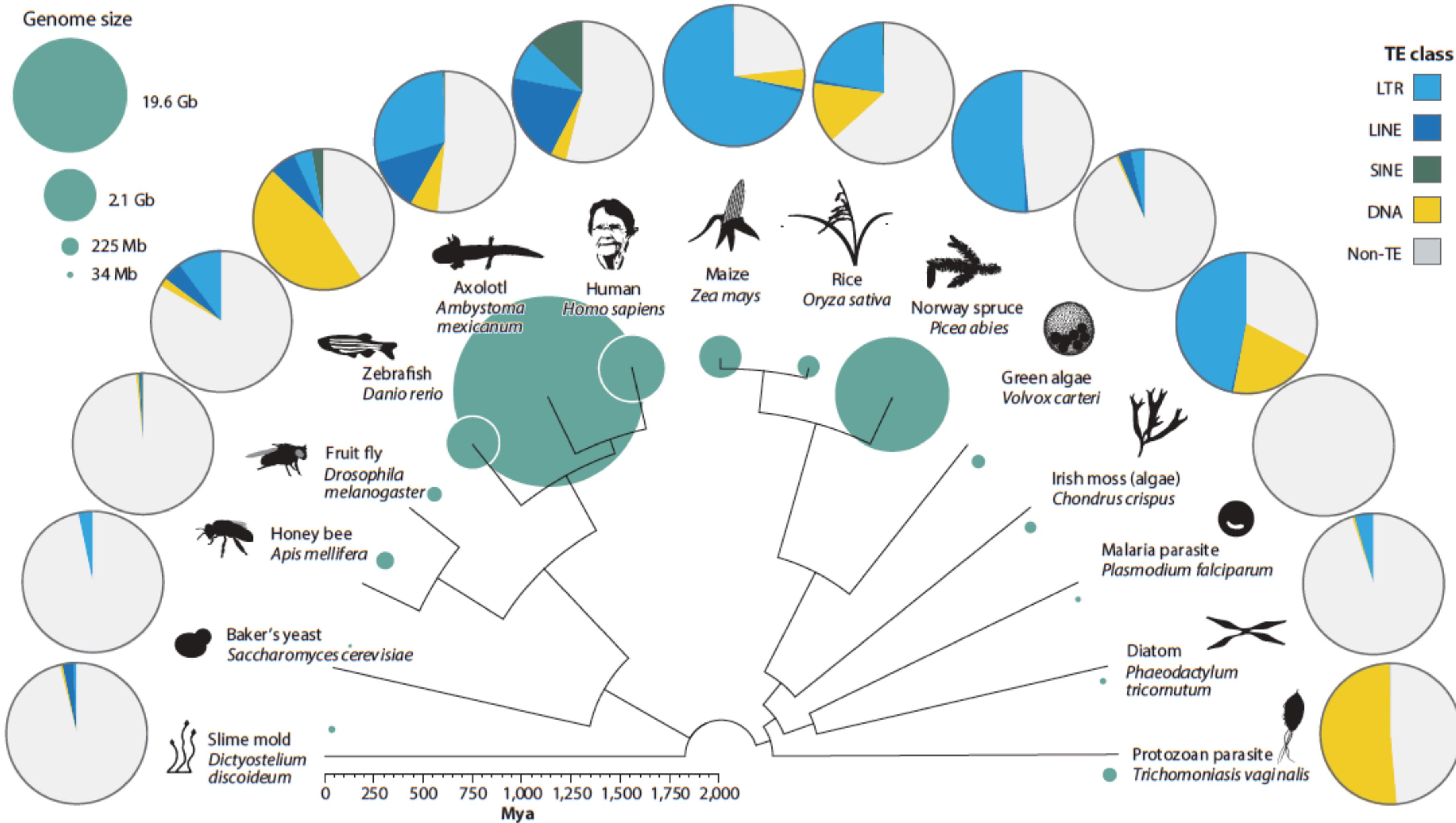
Mutations may be **deleterious**...



...yet sometimes **adaptive**



Transposable Elements

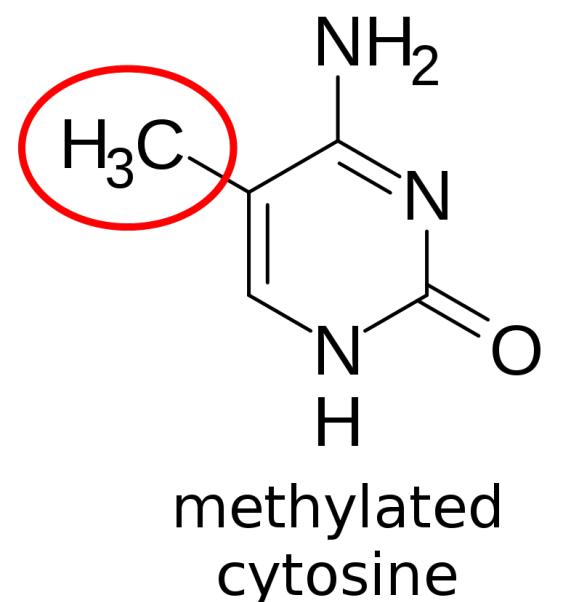
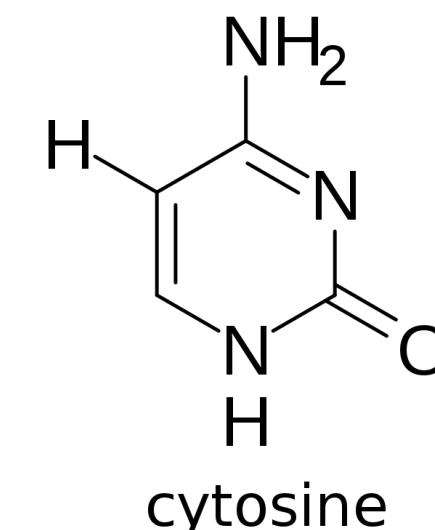
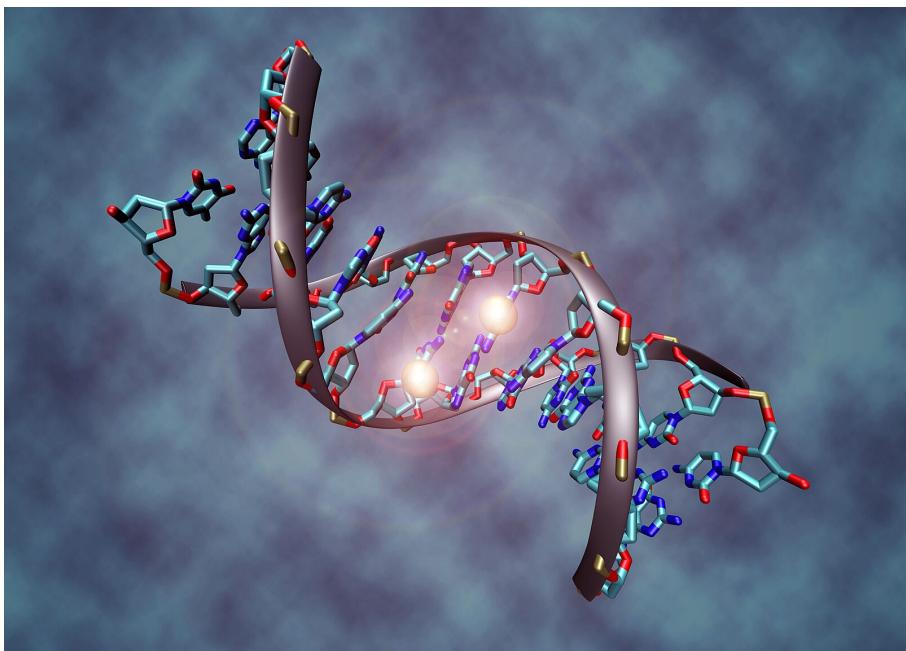


Epigenetic Regulation of Transposable Elements

DNA methylation:

- is an essential regulatory mechanism of TEs activity
- targets CG / CHG / CHH in plants
[H = anything besides G]
- affects TE / gene expression (silencing)
- may spread to flanking regions
- example:

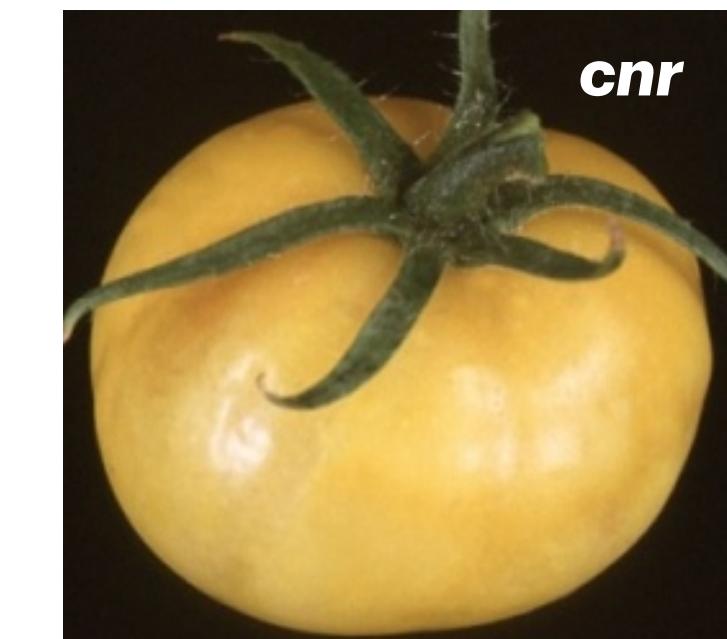
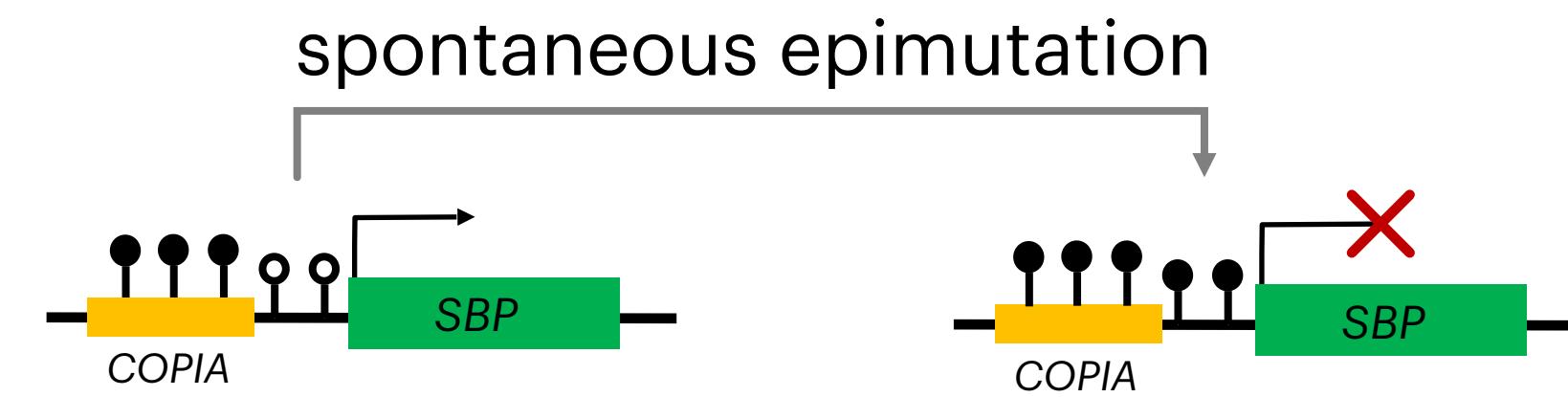
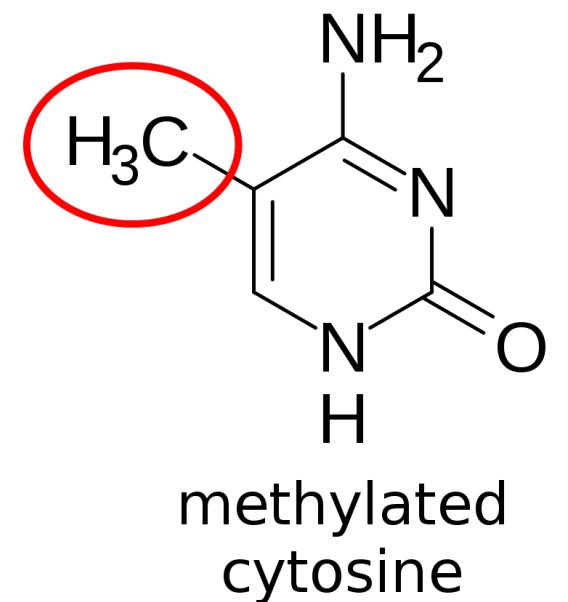
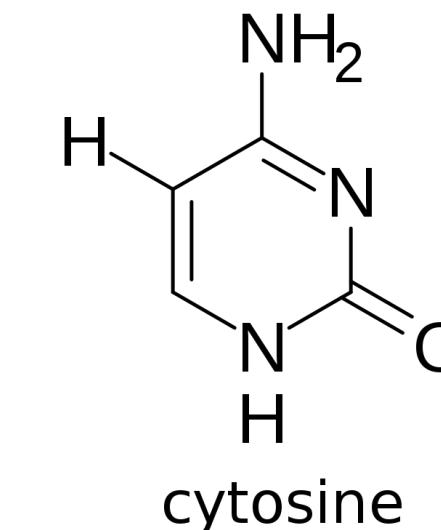
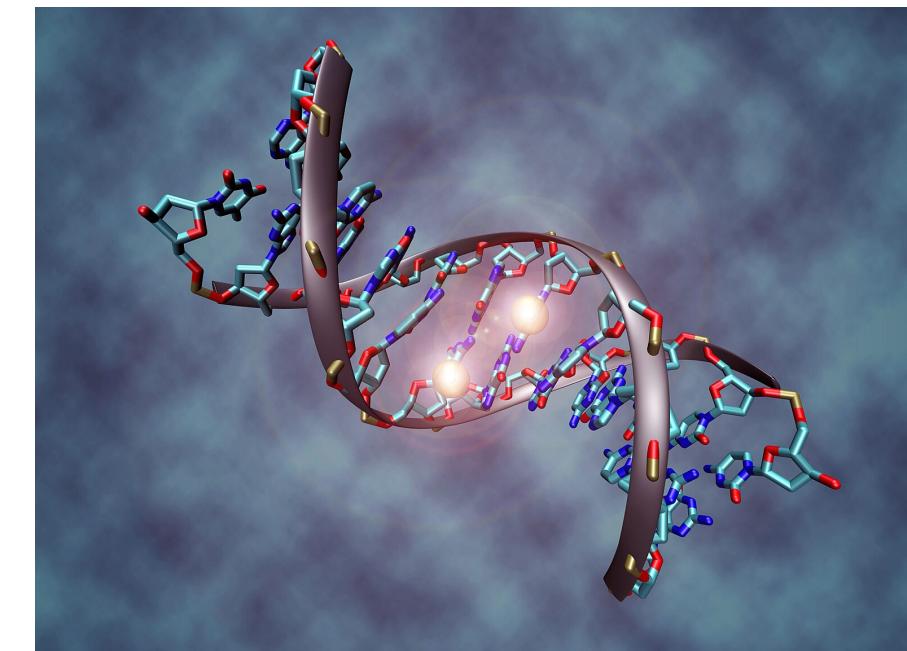
methylated promoter \Rightarrow no RNA \Rightarrow
 \Rightarrow no protein \Rightarrow no function



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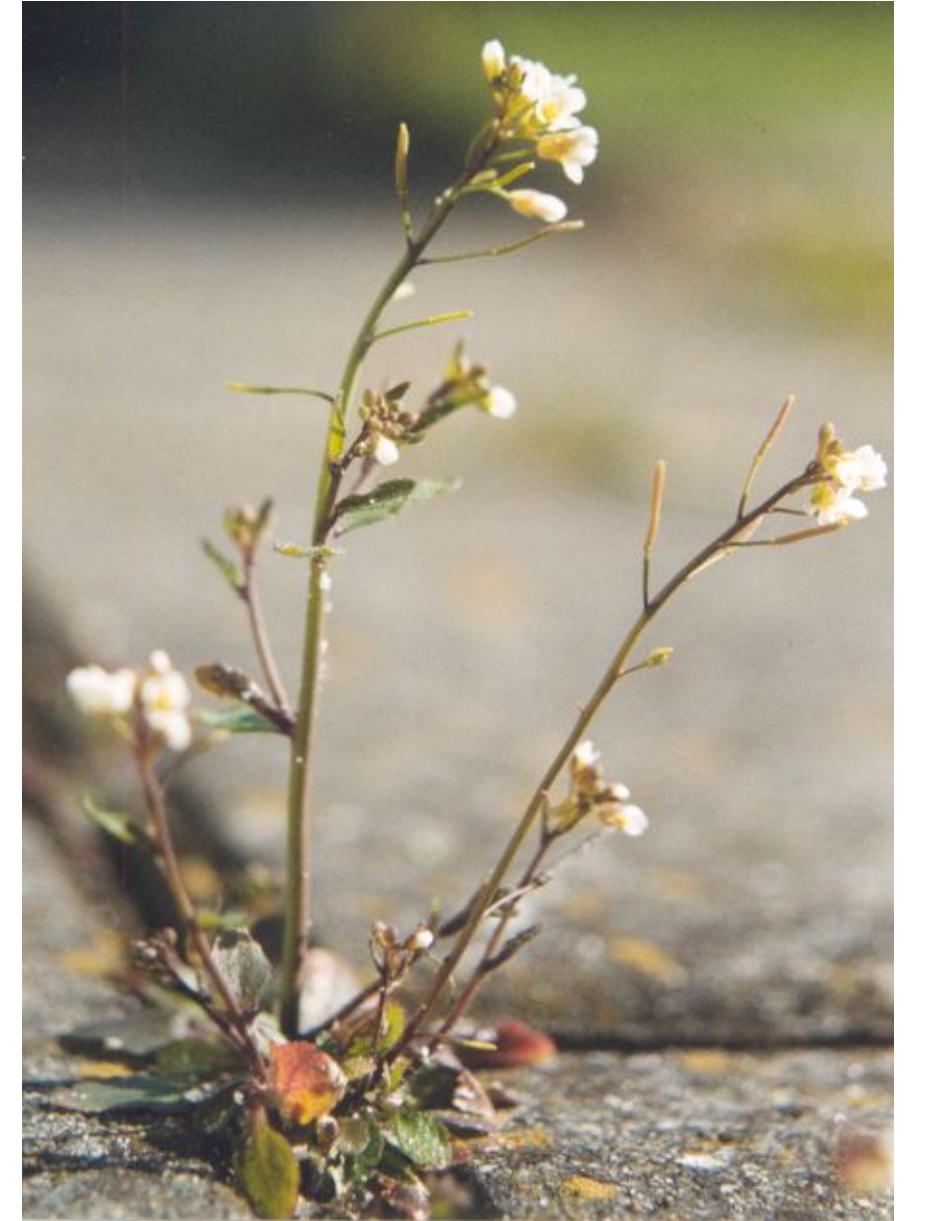


Manning et al., Nat Genet 2006

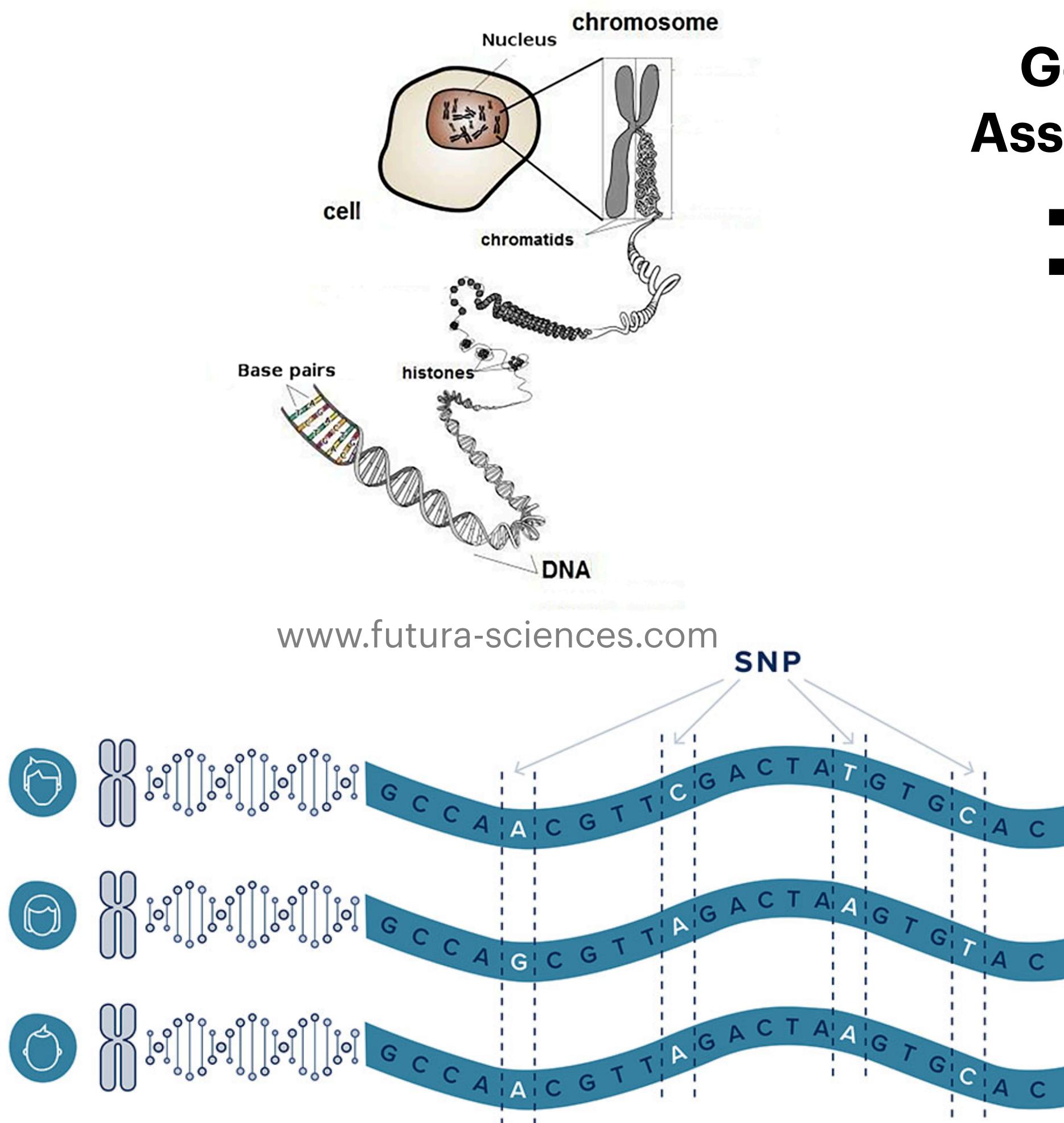
\Rightarrow perfect Mendelian segregation though no DNA changes observed

Our data: *Arabidopsis Thaliana*

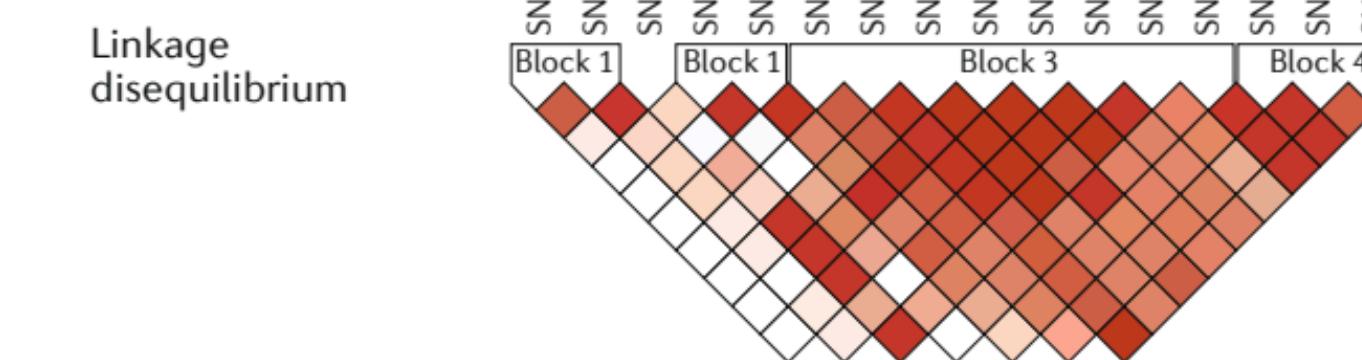
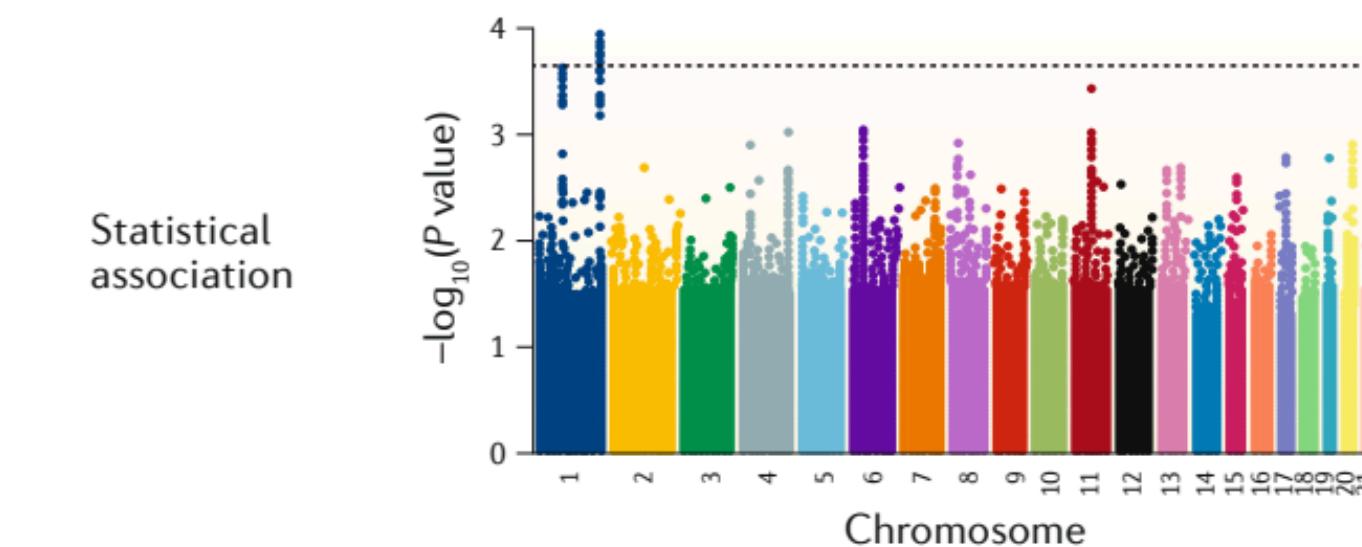
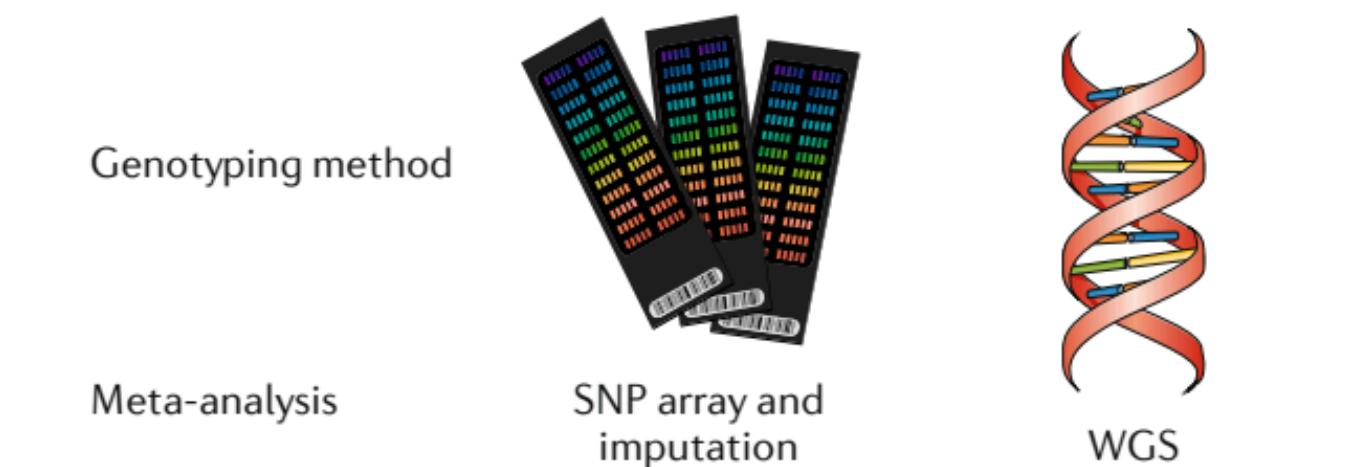
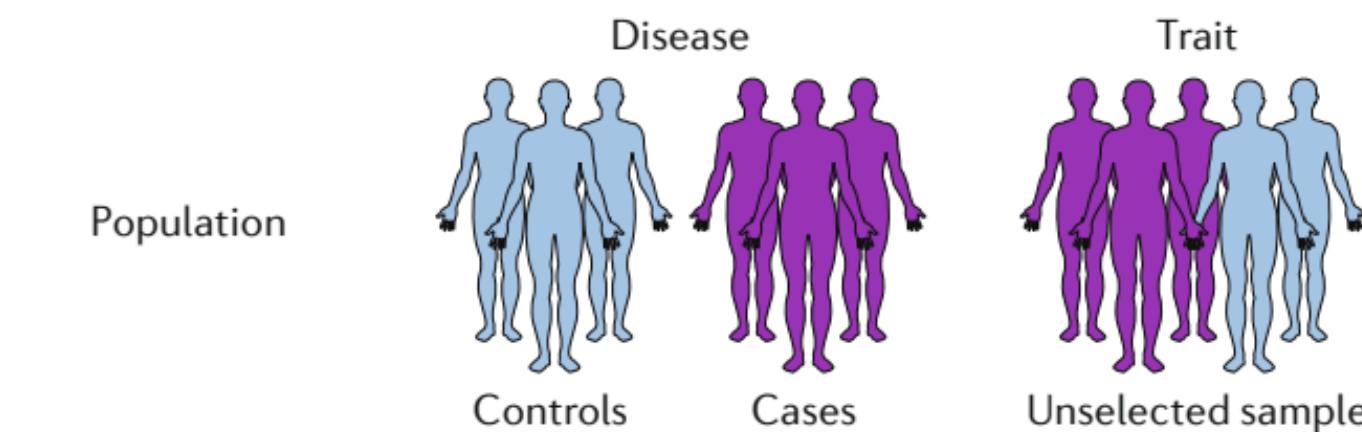
- 89 strains from throughout the world, **sequenced with ultra-long reads (Nanopore)**
- **TE annotation + Full methylation profiles** (for all contexts CG, CHG, CHH)
- Gene expression data



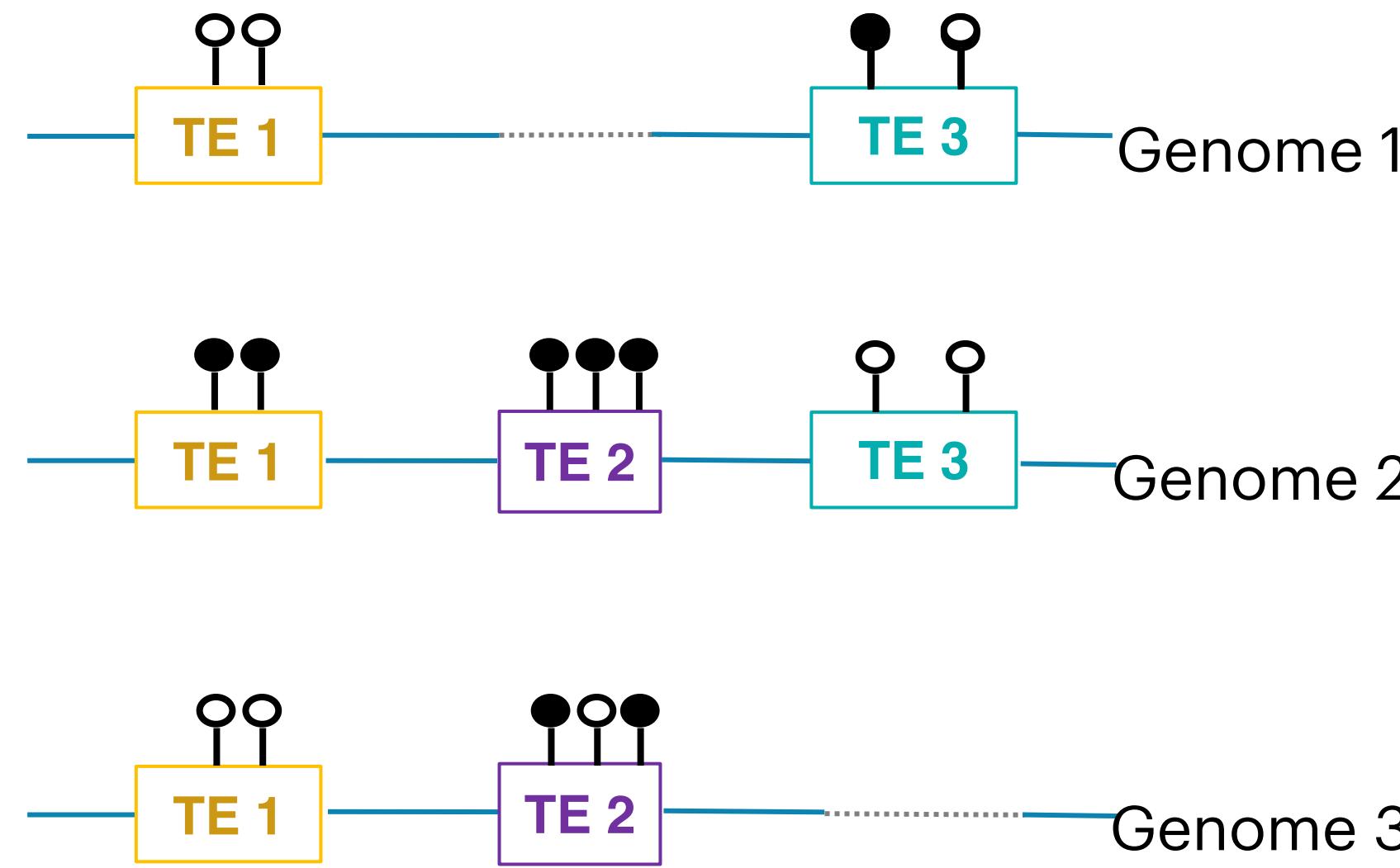
From genotype to phenotype



Genome-Wide Association Study



From epi-genotype to phenotype

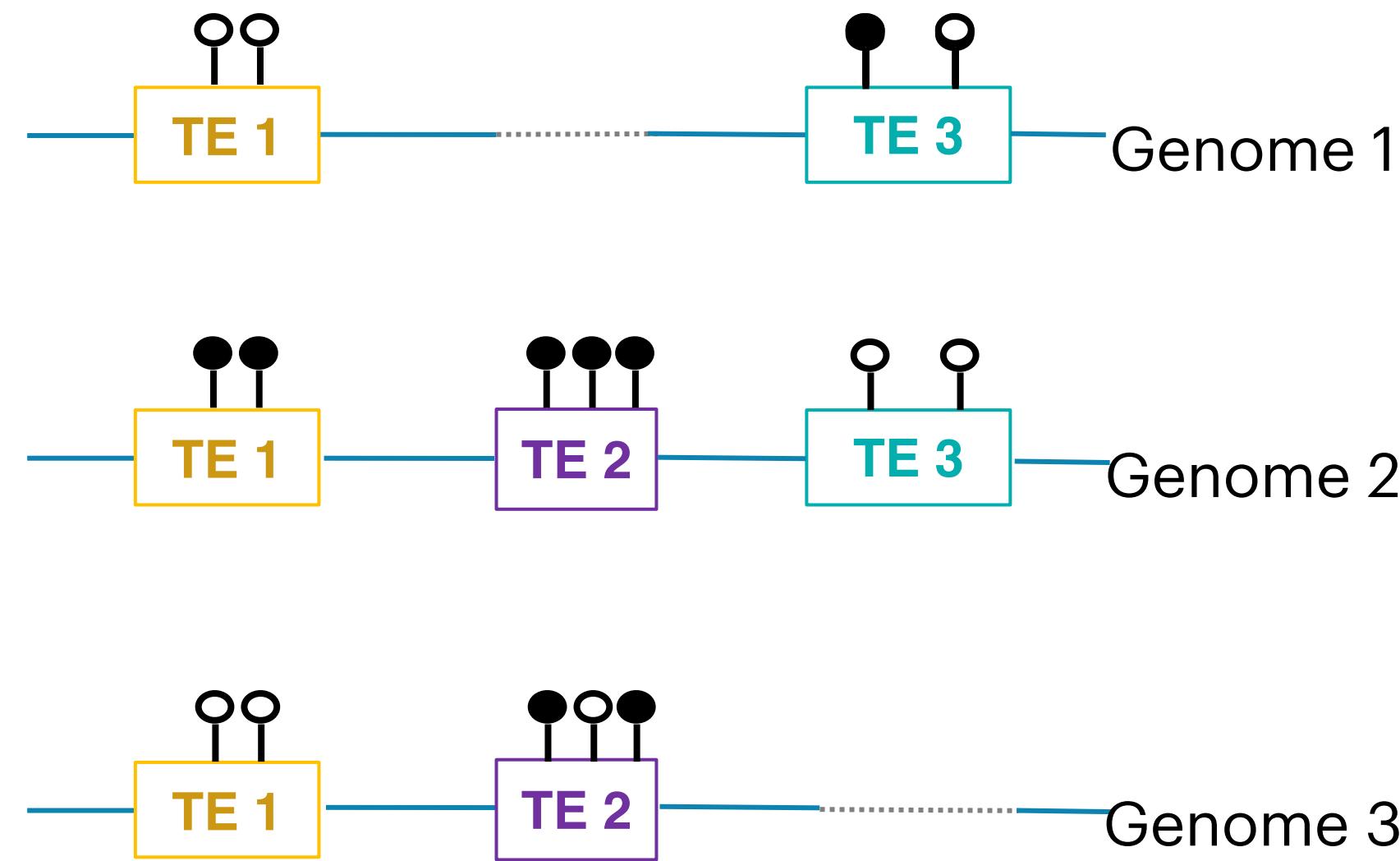


**Genome-Wide
Association Study**



	Gene A	Gene B	Gene C
Genome 1			
Genome 2			
Genome 3			

From epi-genotype to phenotype



Genome-Wide
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3 groups:

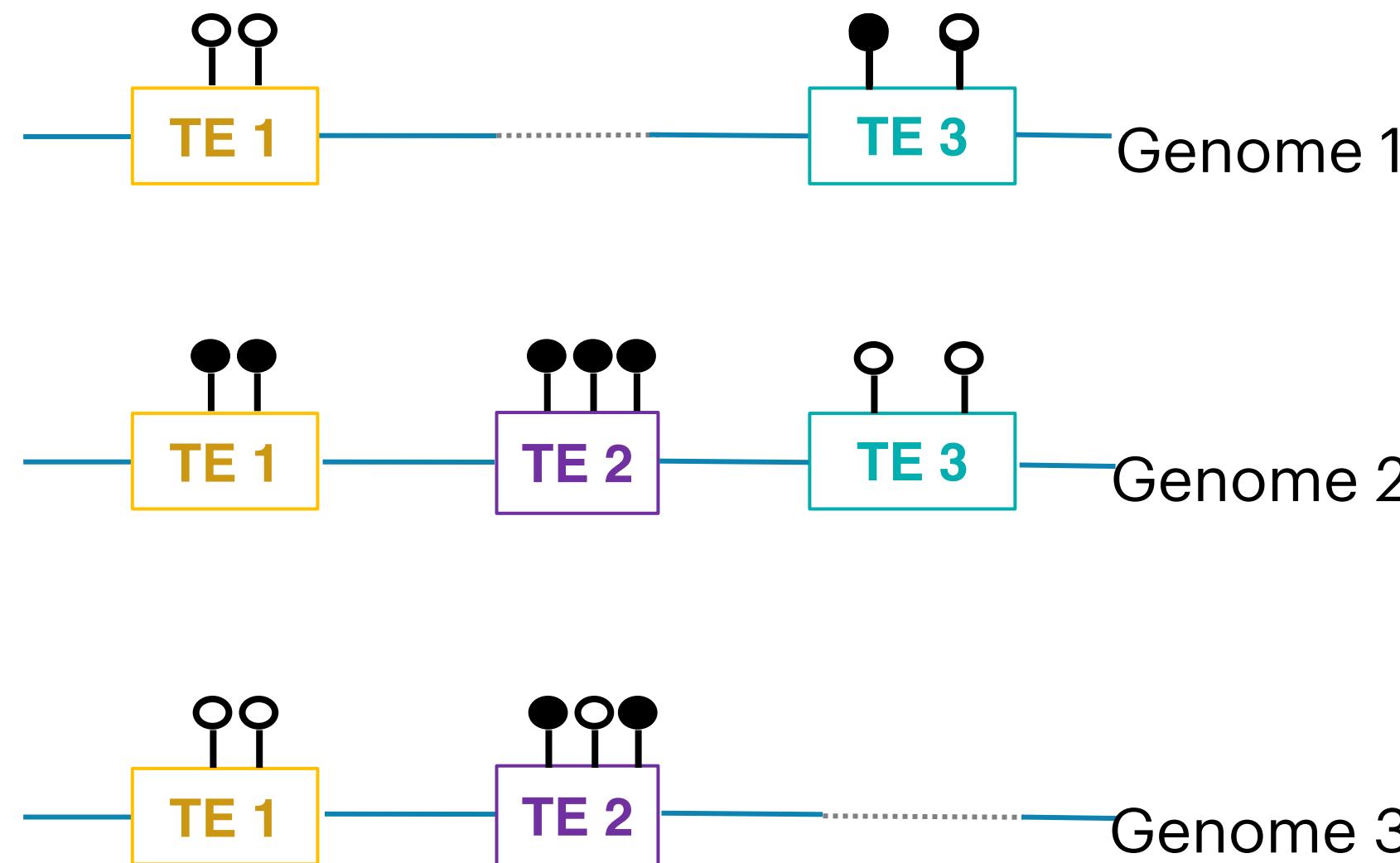
0 = absent

1 = present and not methylated (< 5%)

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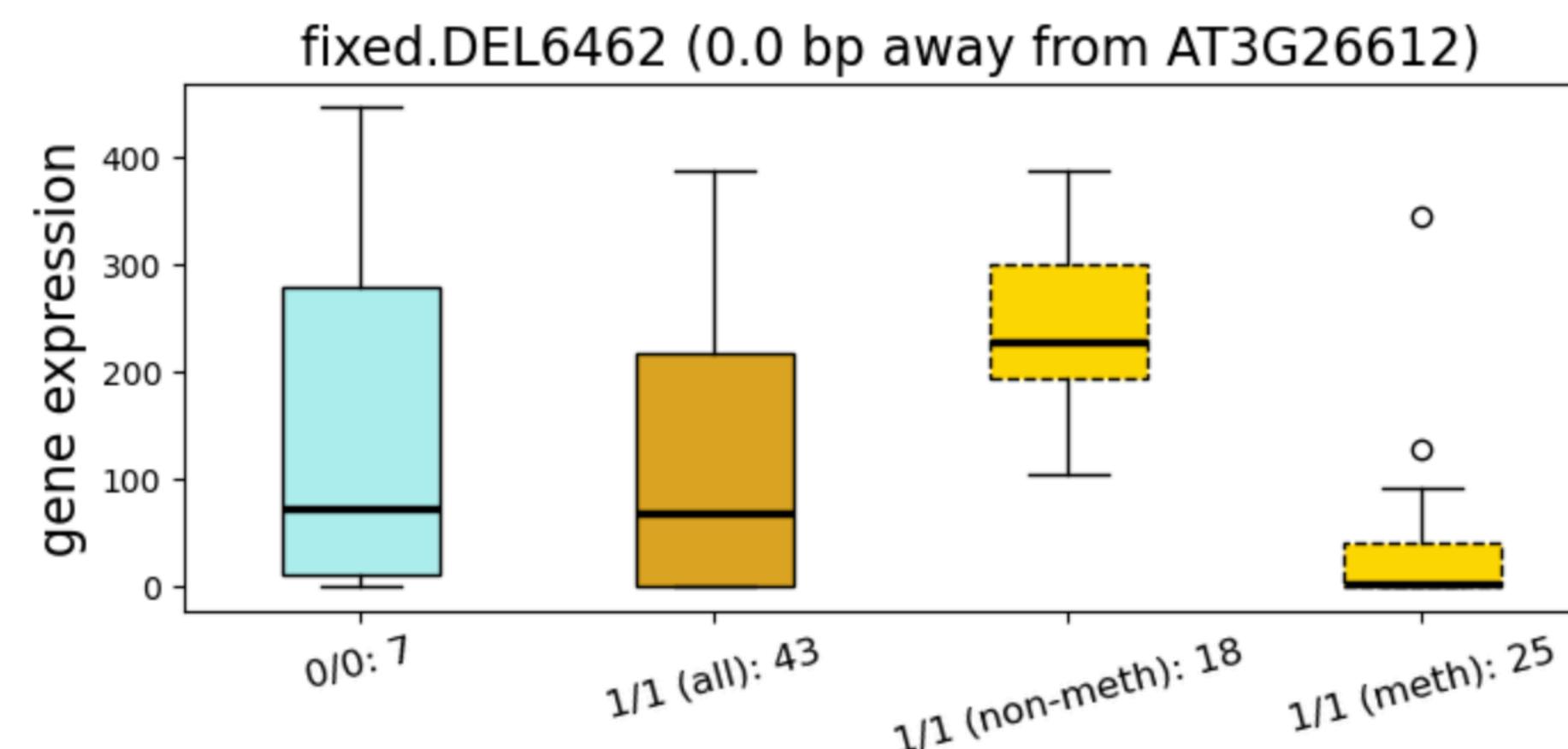
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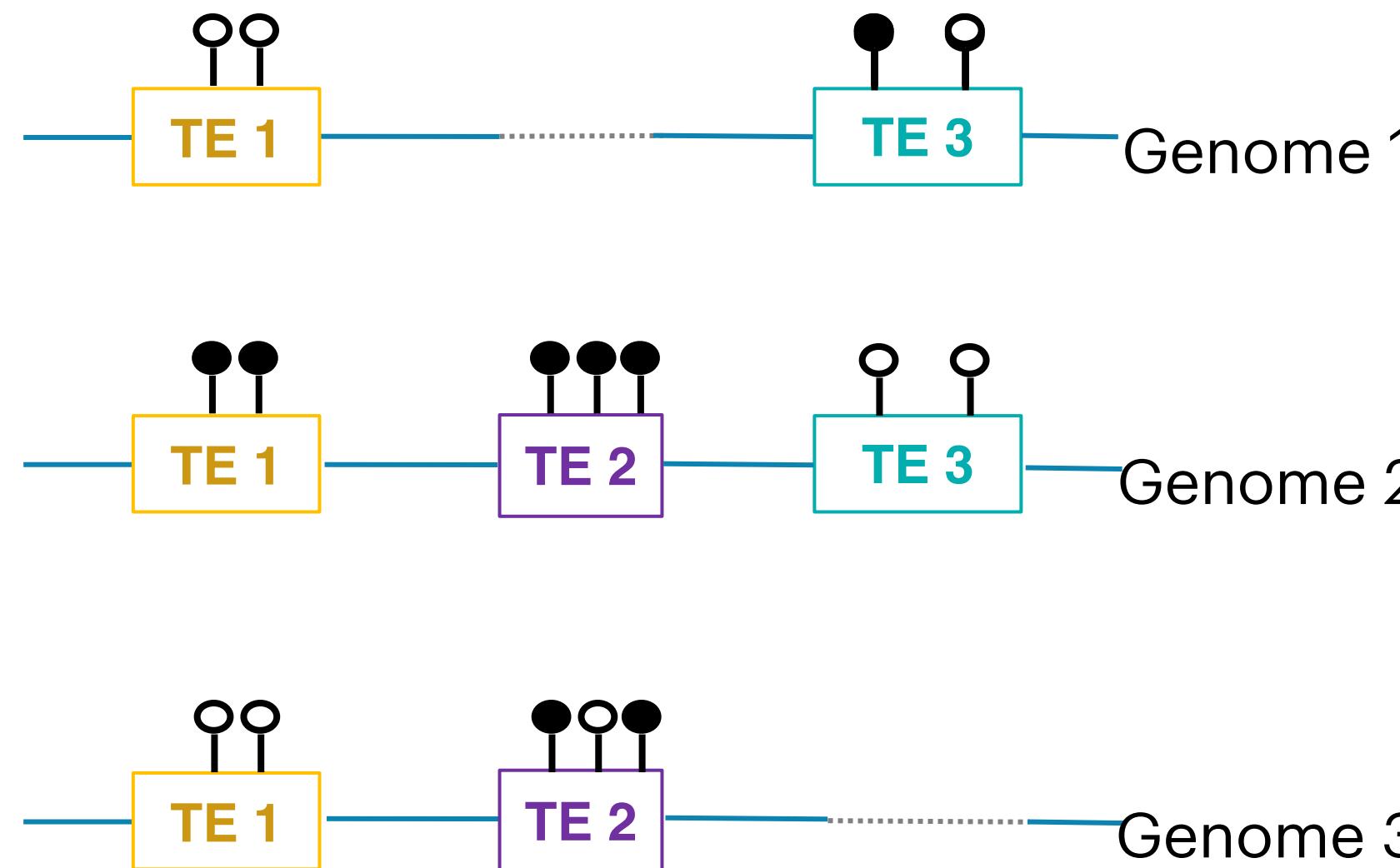
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For the moment, cis- effects only:

P_tip	P_meth	TIP	Chr	start	end	Distance from gene
2780	0.516462	0.000002	fixed.DEL6462	Chr3	9783357	NaN
						0.0



From epi-genotype to phenotype



Genome-Wide Association Study



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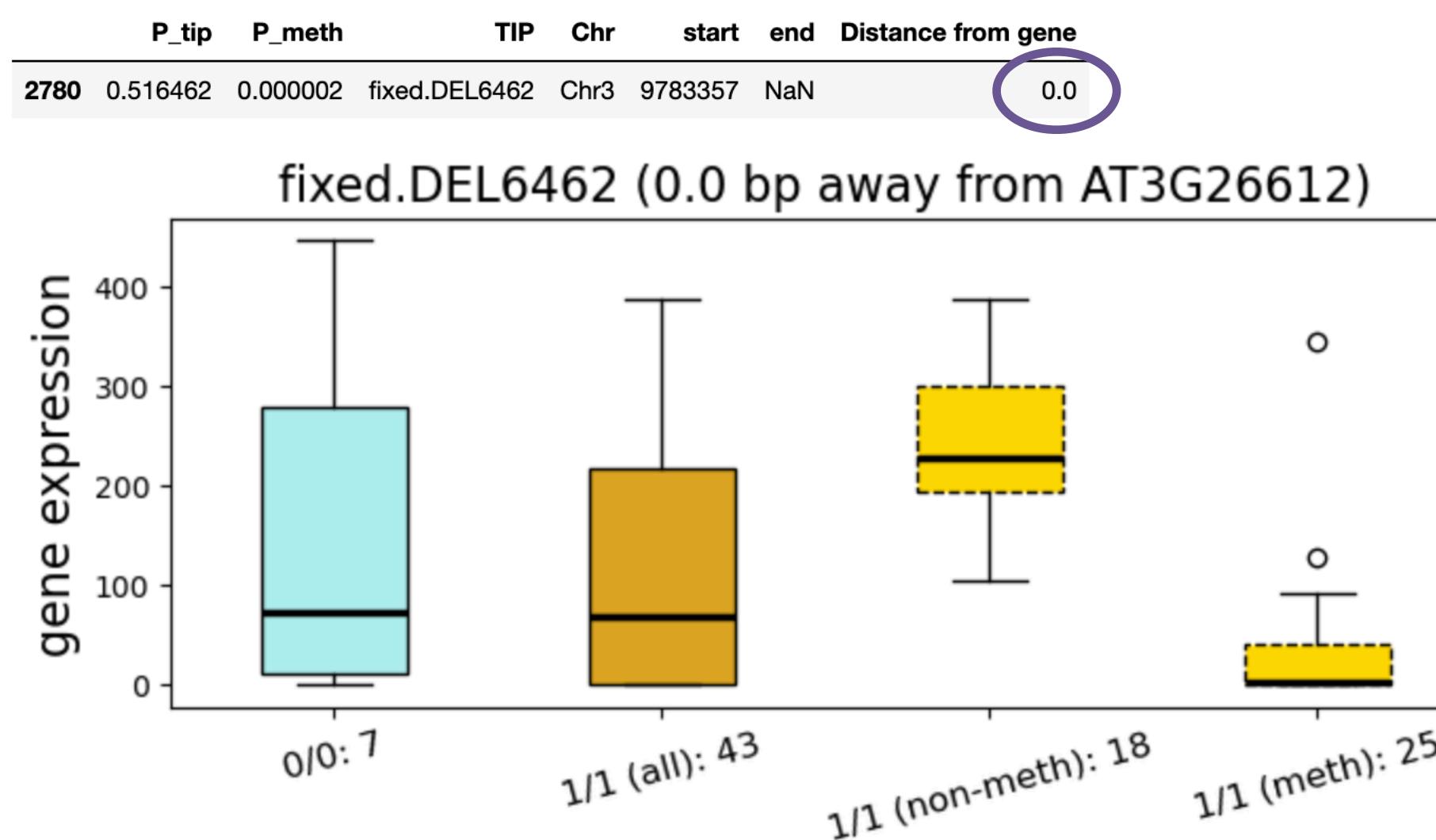
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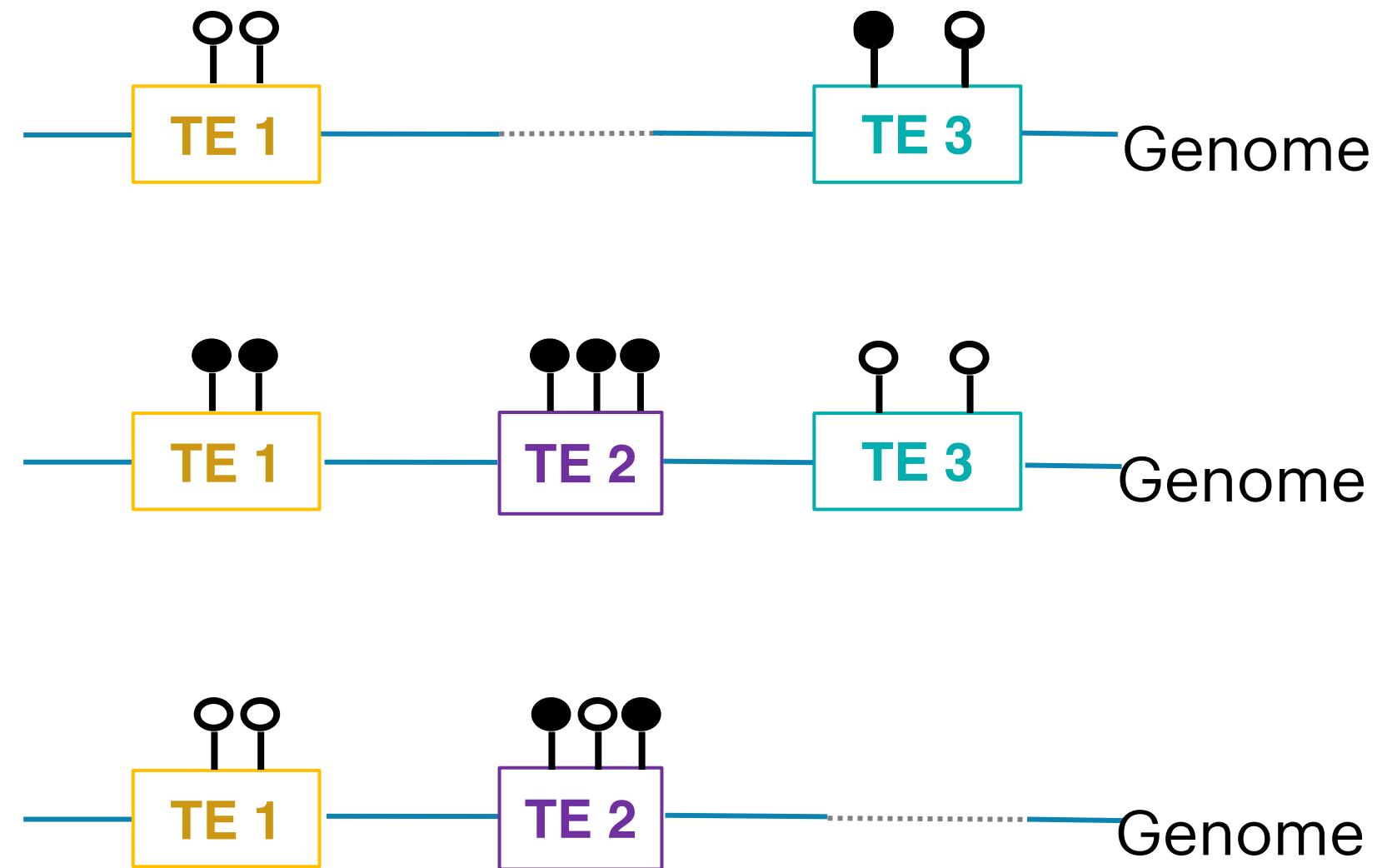
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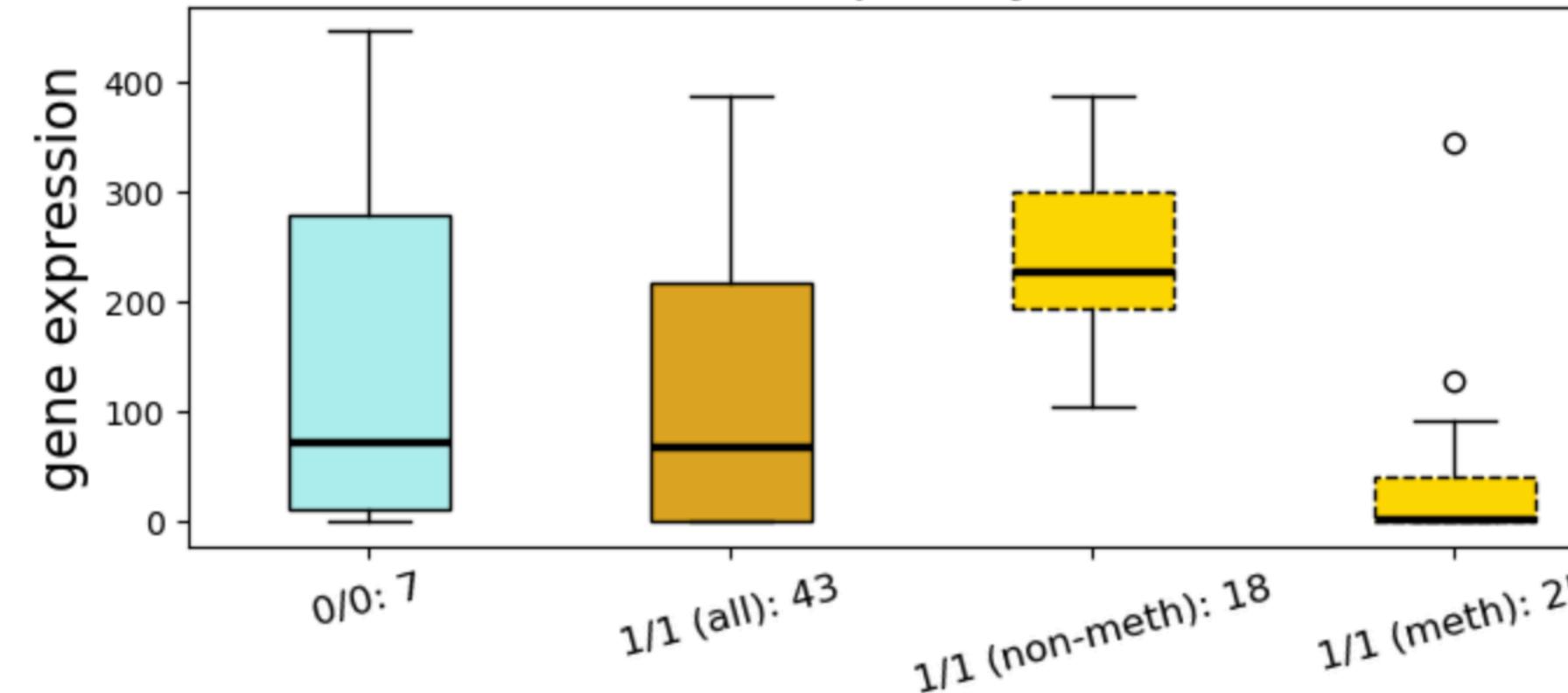
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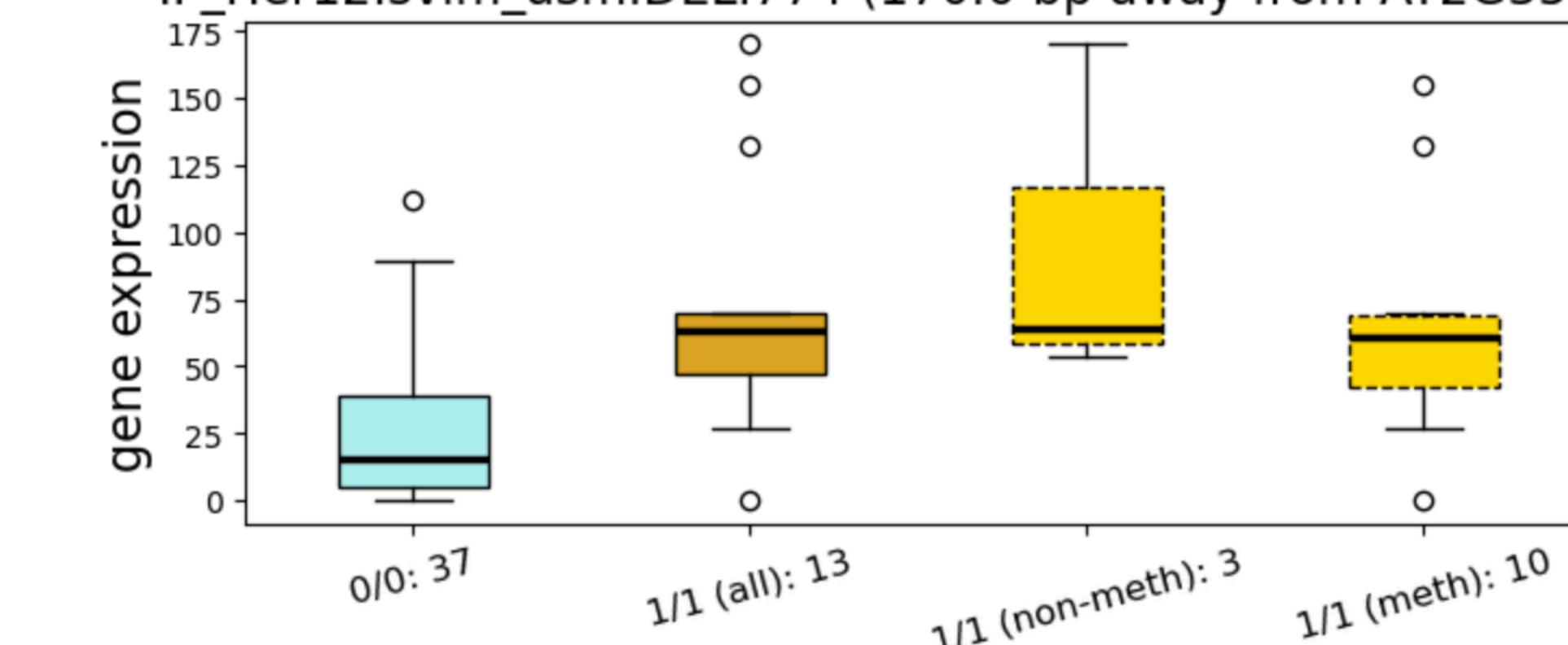
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fixed.DEL6462 (0.0 bp away from AT3G26612)

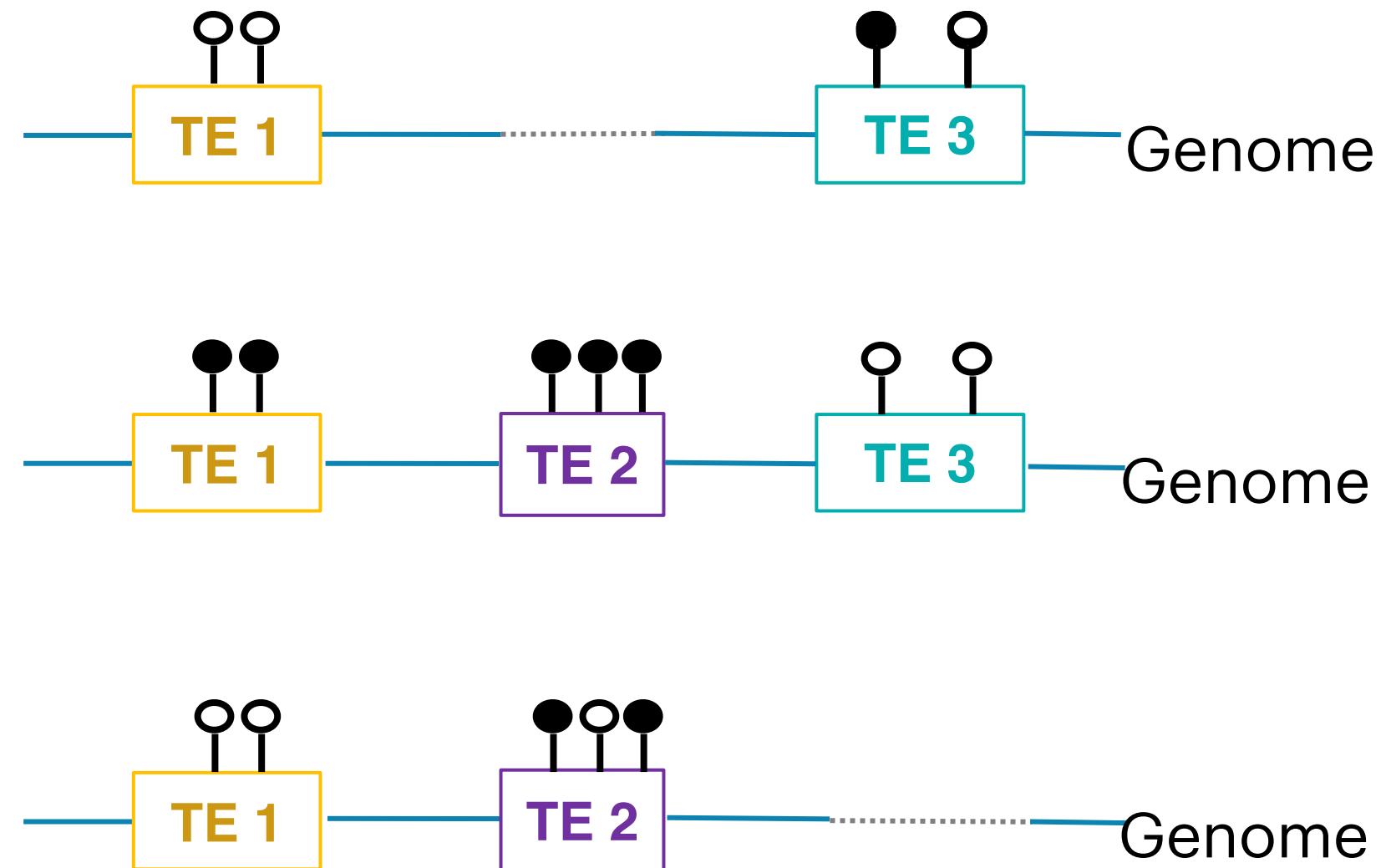


P_tip	P_meth	TIP	Chr	start	end	Distance from gene
535	0.000068	0.002327	IP_Her12.svim_asm.DEL.774	Chr2	15110051	15110322.0

IP_Her12.svim_asm.DEL.774 (170.0 bp away from AT2G35980)



From epi-genotype to phenotype



Genome-Wide Association Study



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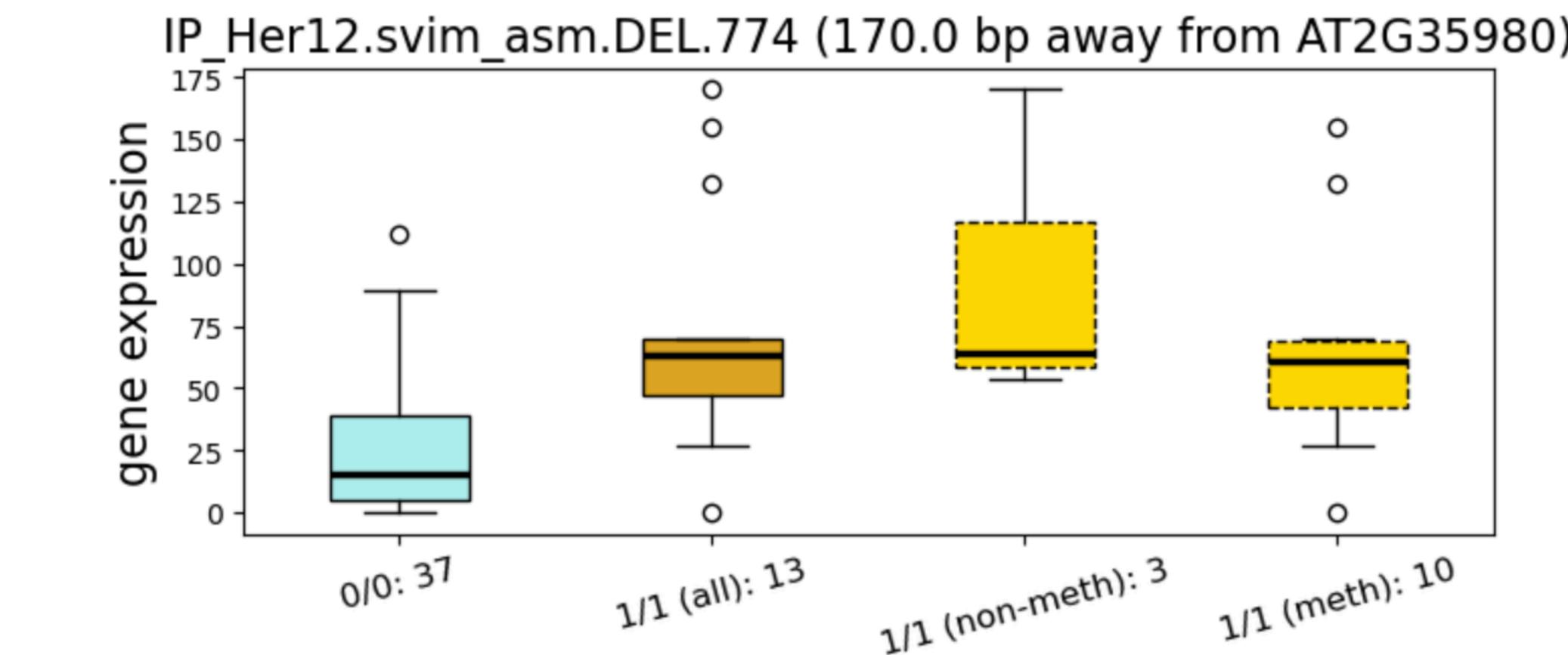
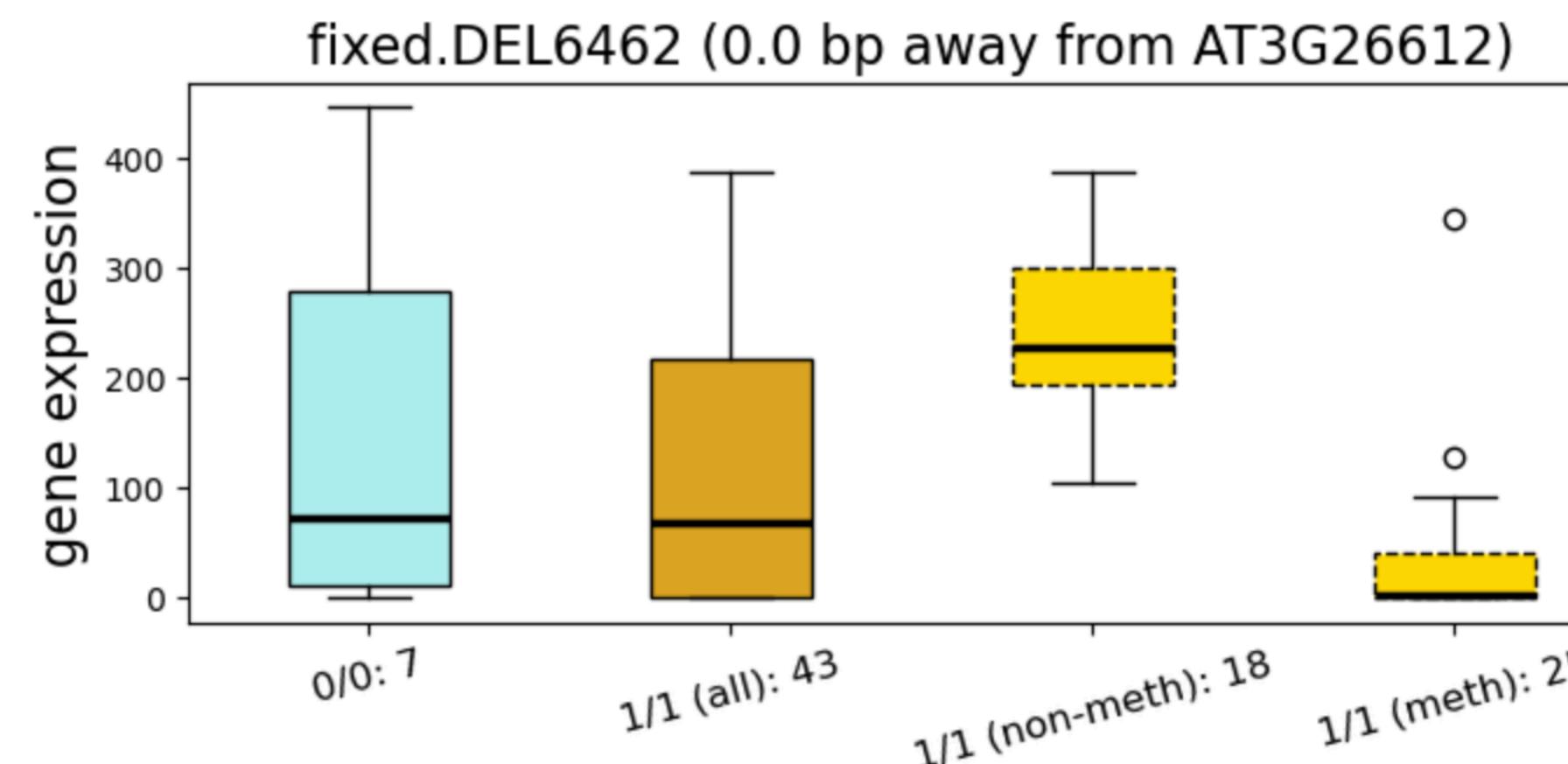
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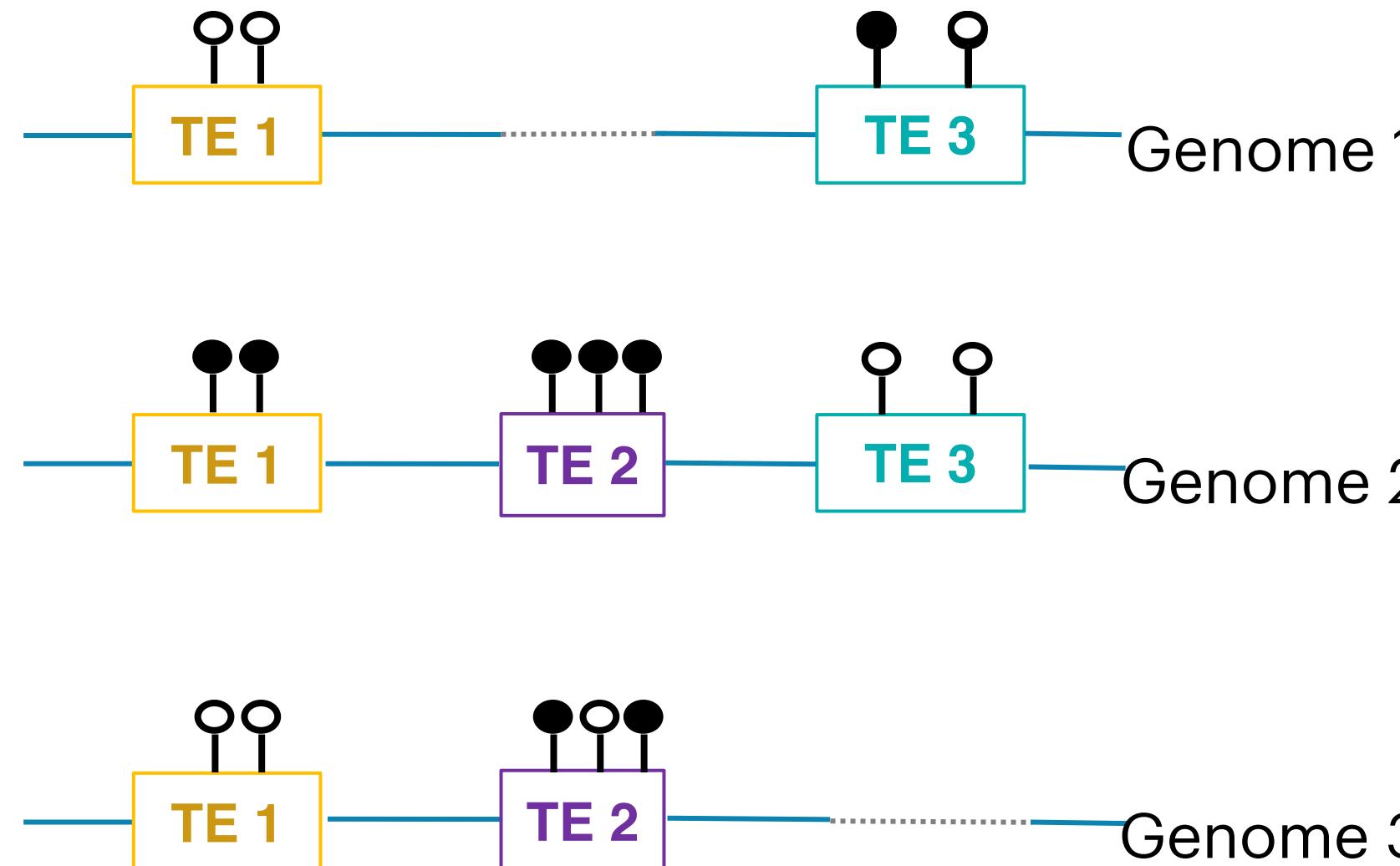
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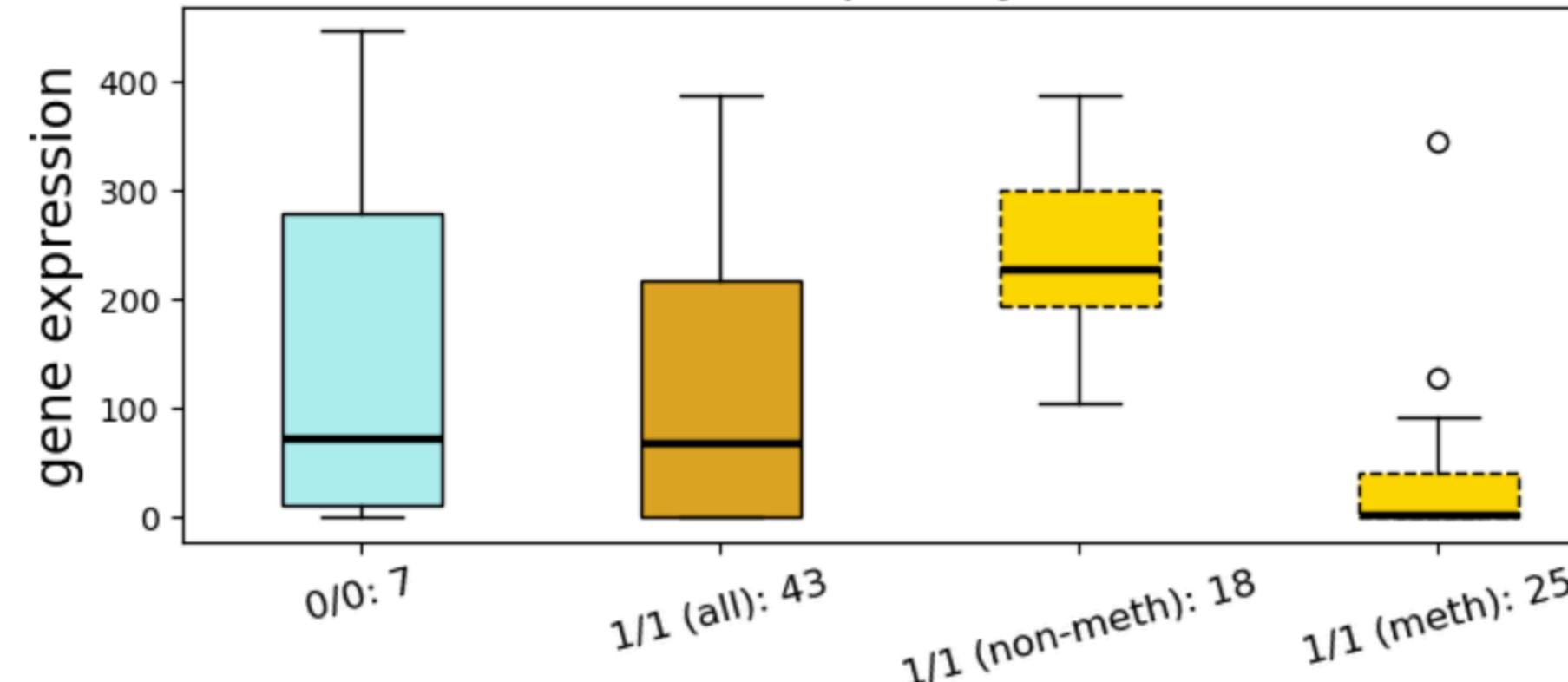
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Spreading ?

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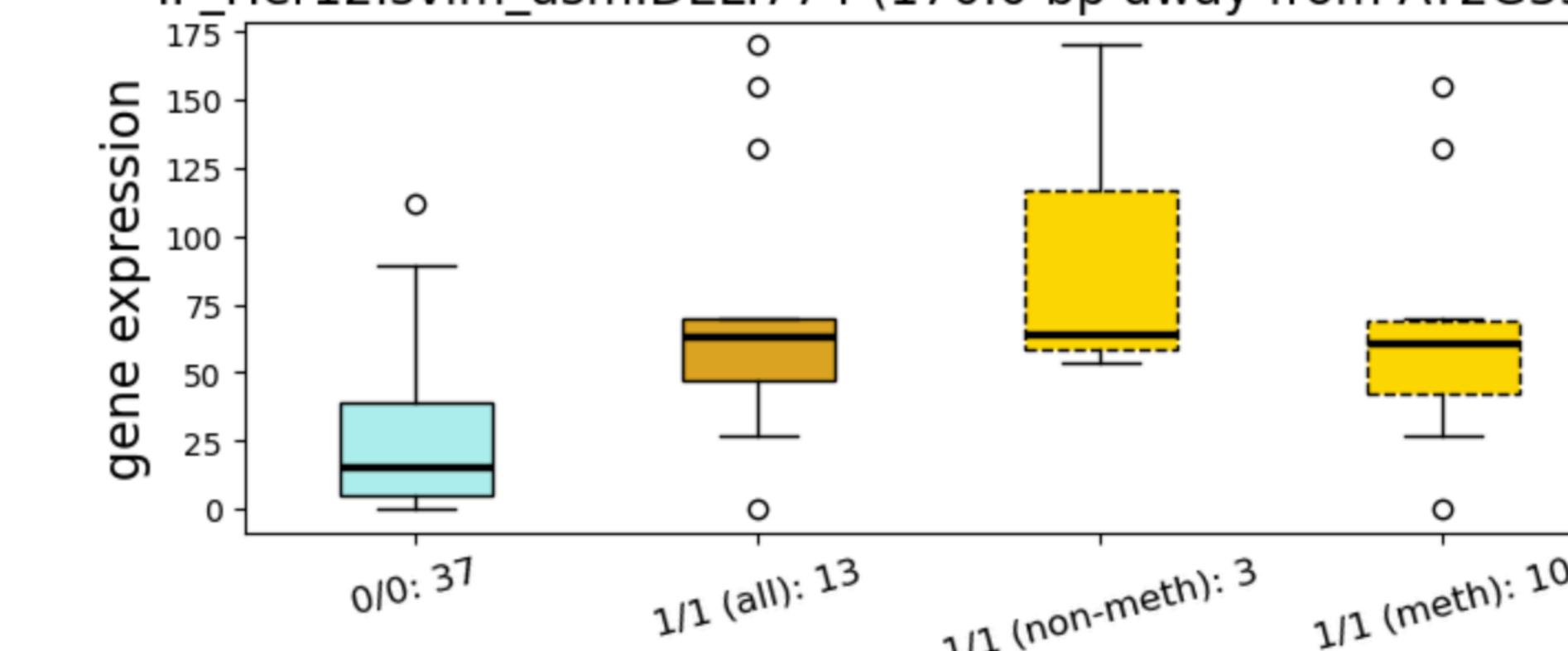
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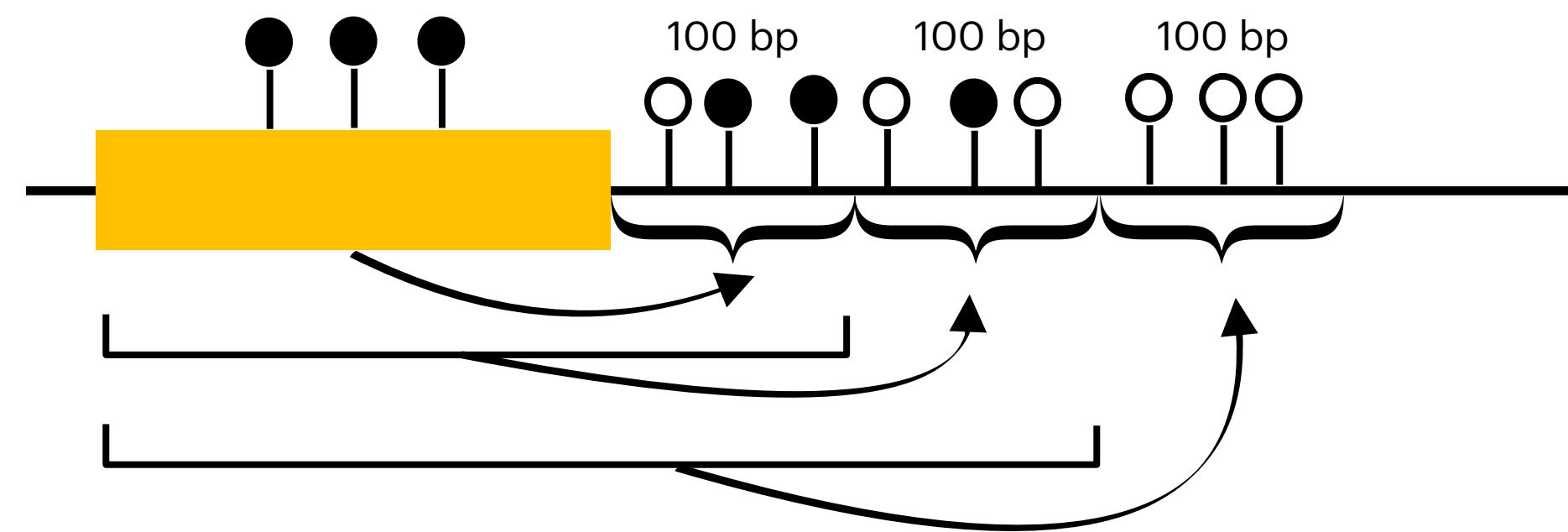


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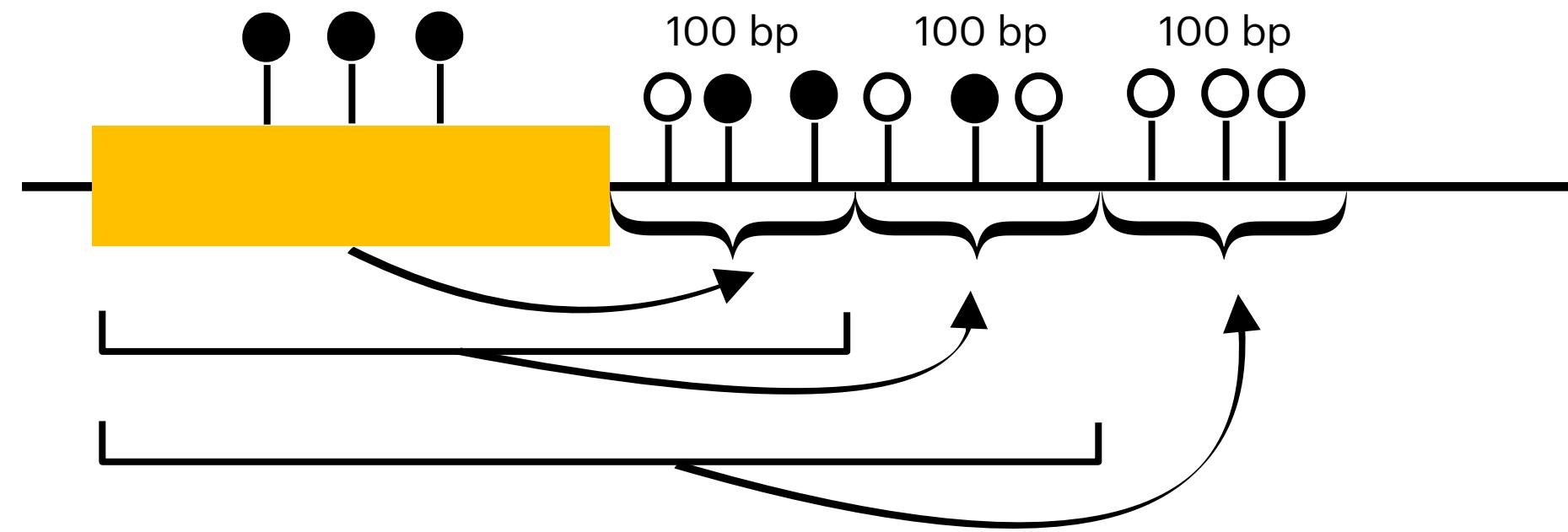
Prediction of methylation spreading



Model:

Random Forest

Prediction of methylation spreading



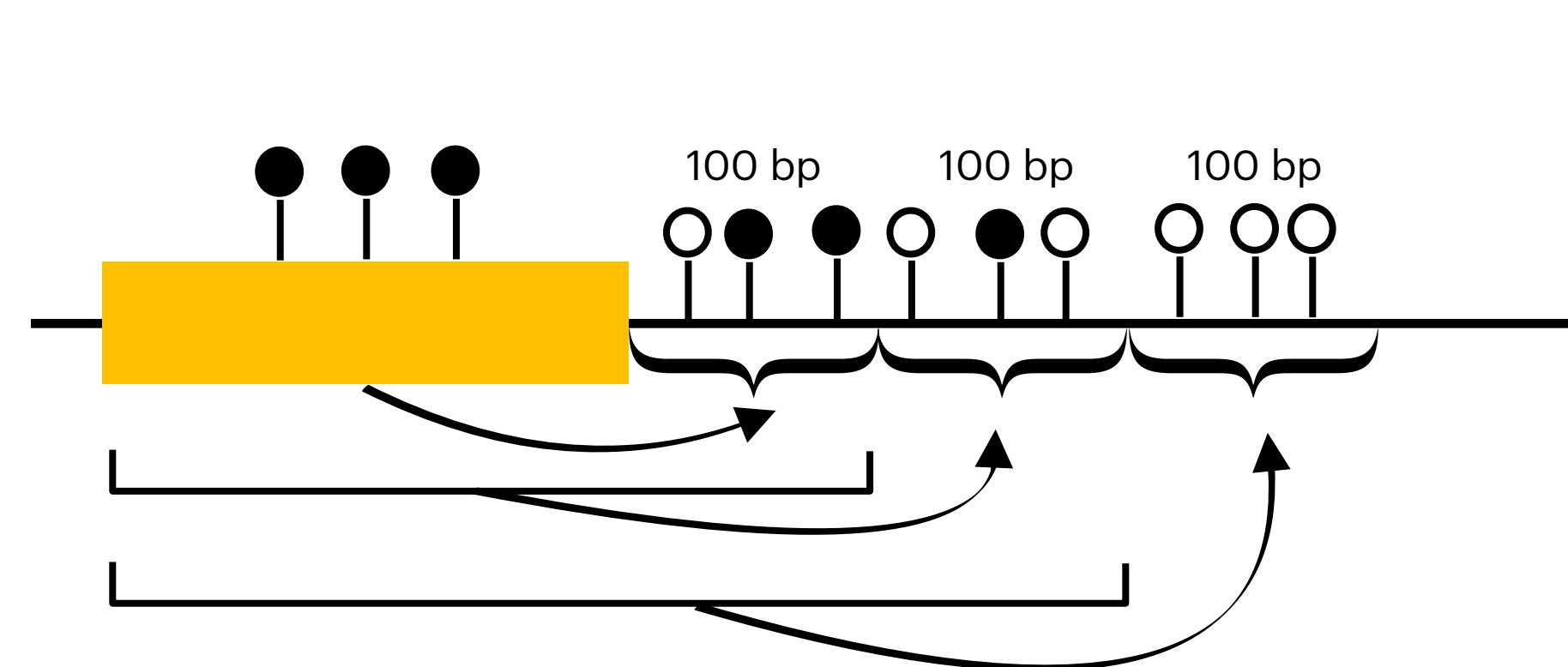
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Prediction of methylation spreading

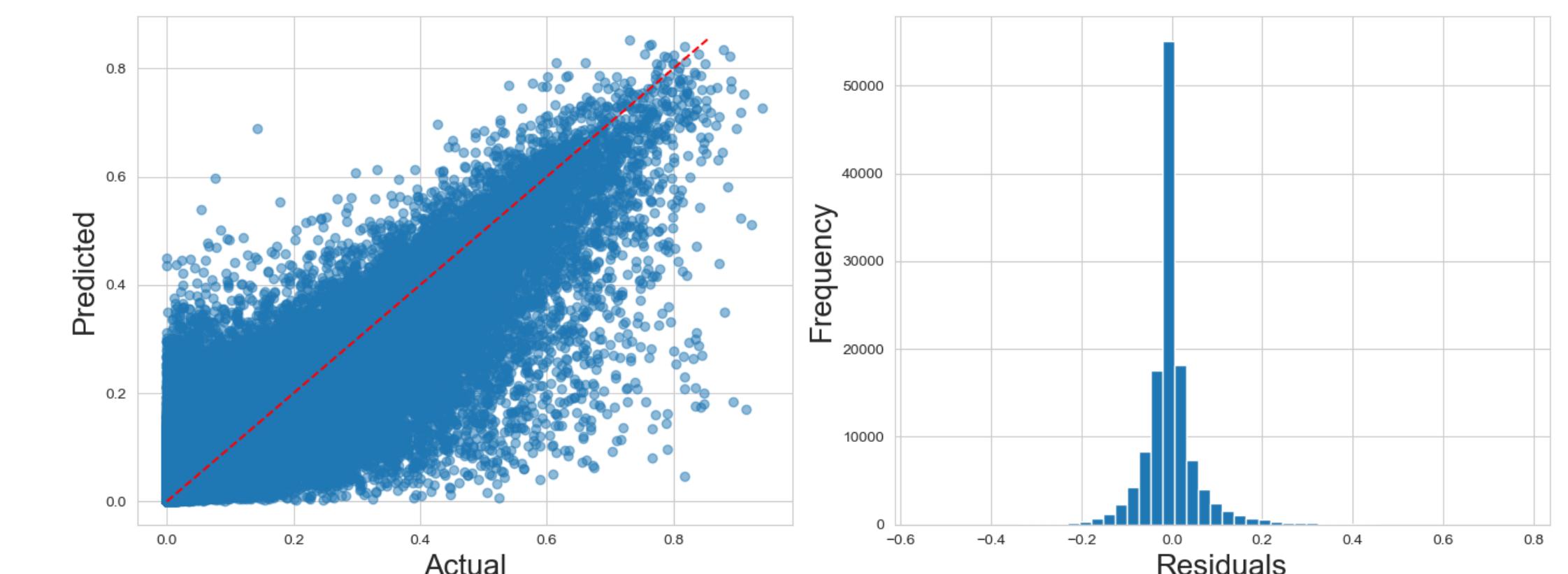
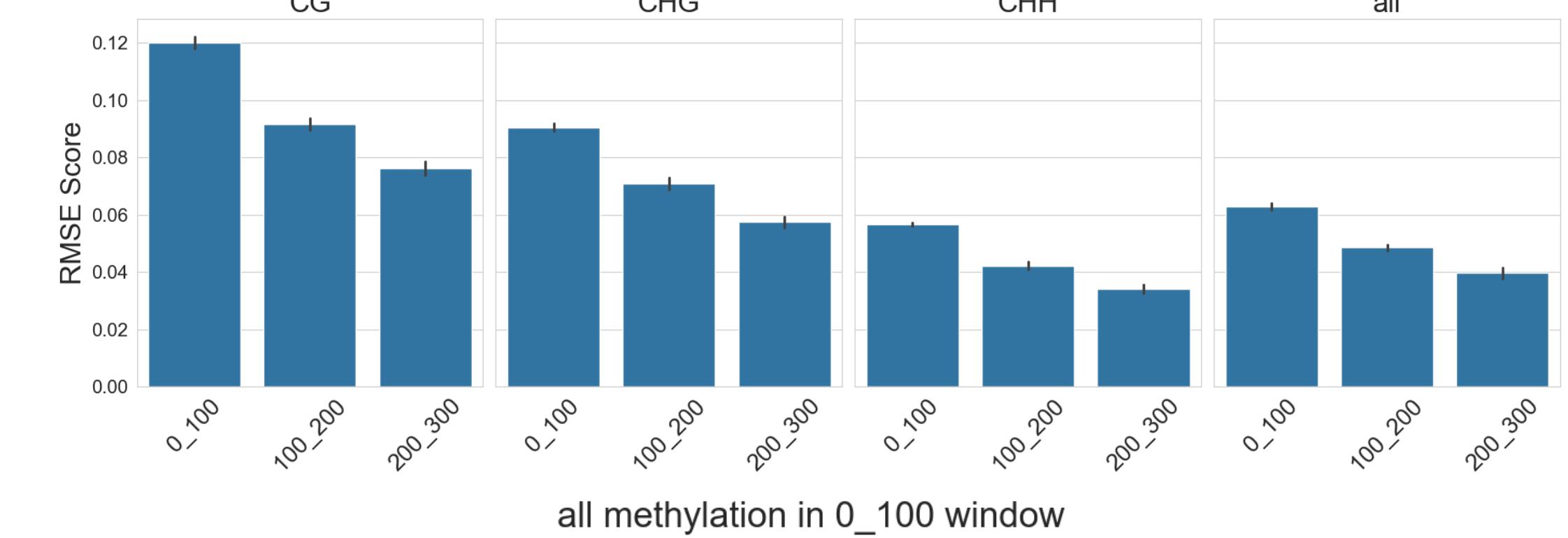
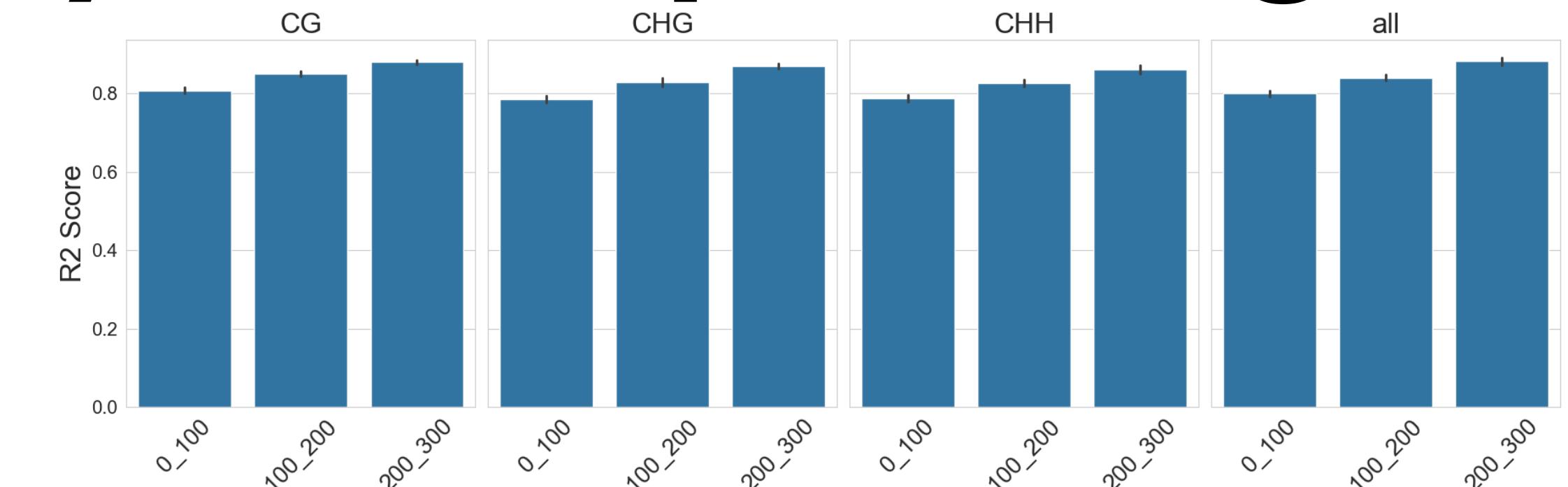


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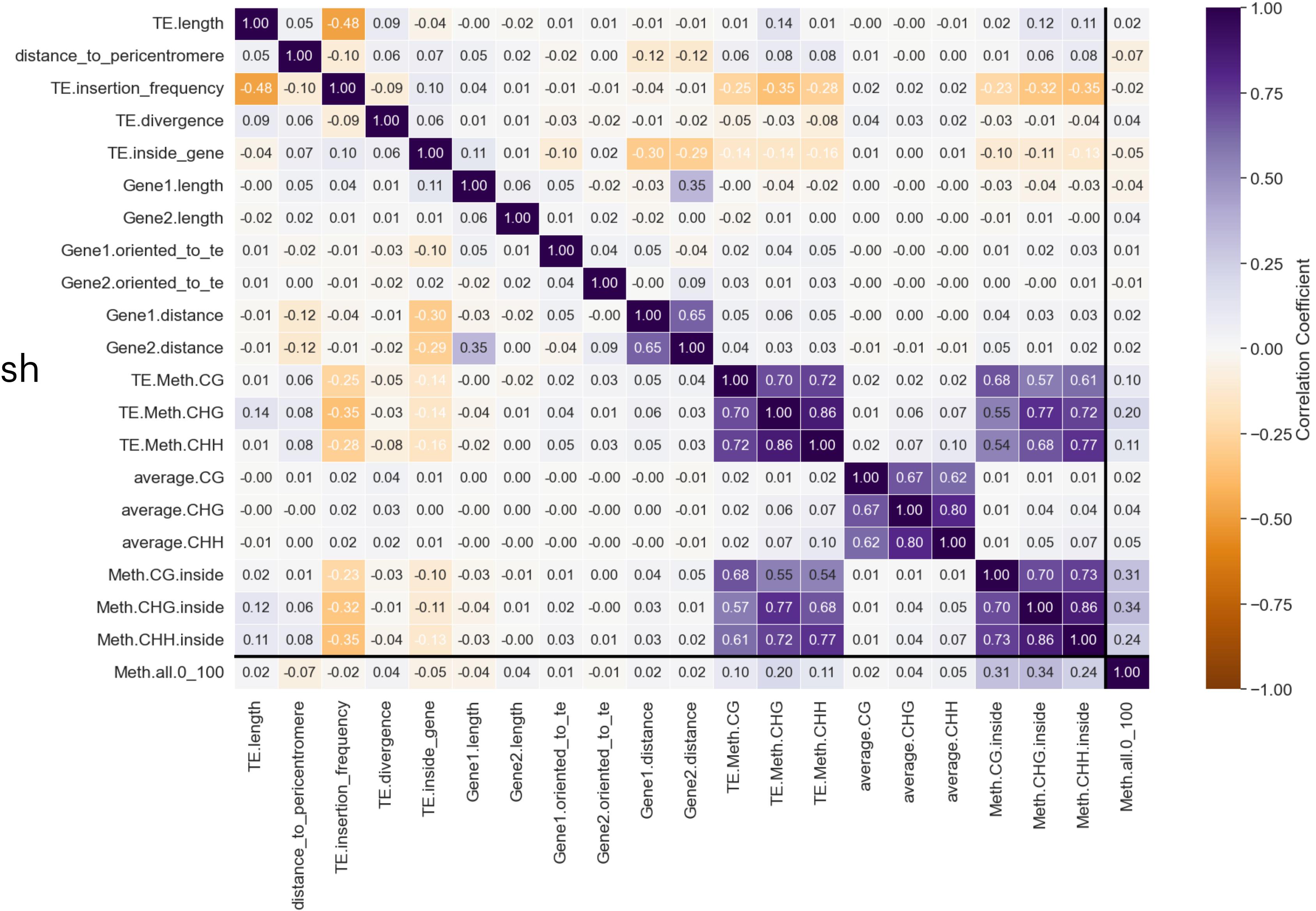


Prediction of methylation spreading

- The model predicts well, but which features define methylation level?

Prediction of methylation spreading

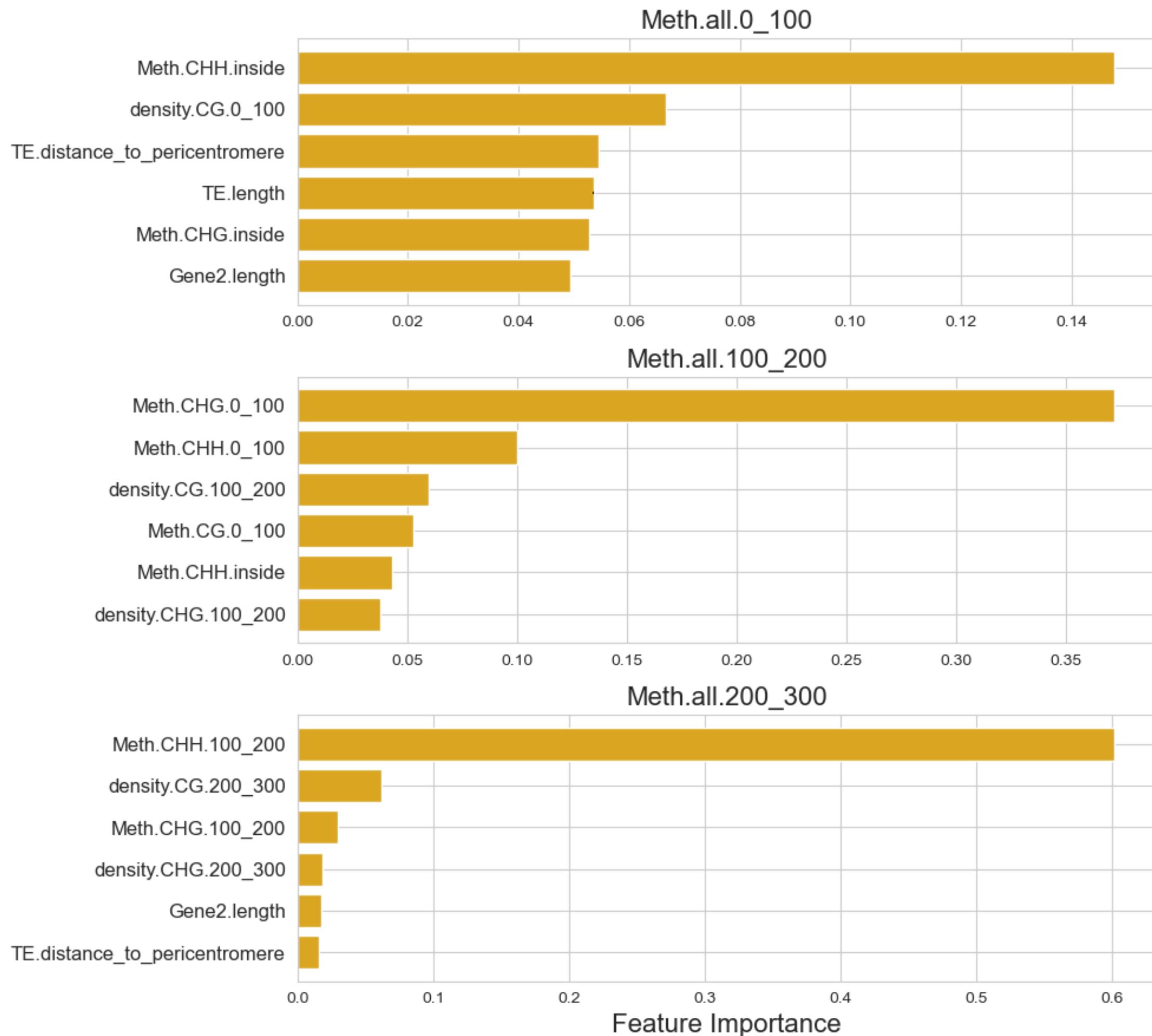
- The model predicts well, but which features define methylation level?
- Some features are highly correlated \implies hard to distinguish between them



Prediction of methylation spreading

Impurity-based feature importances

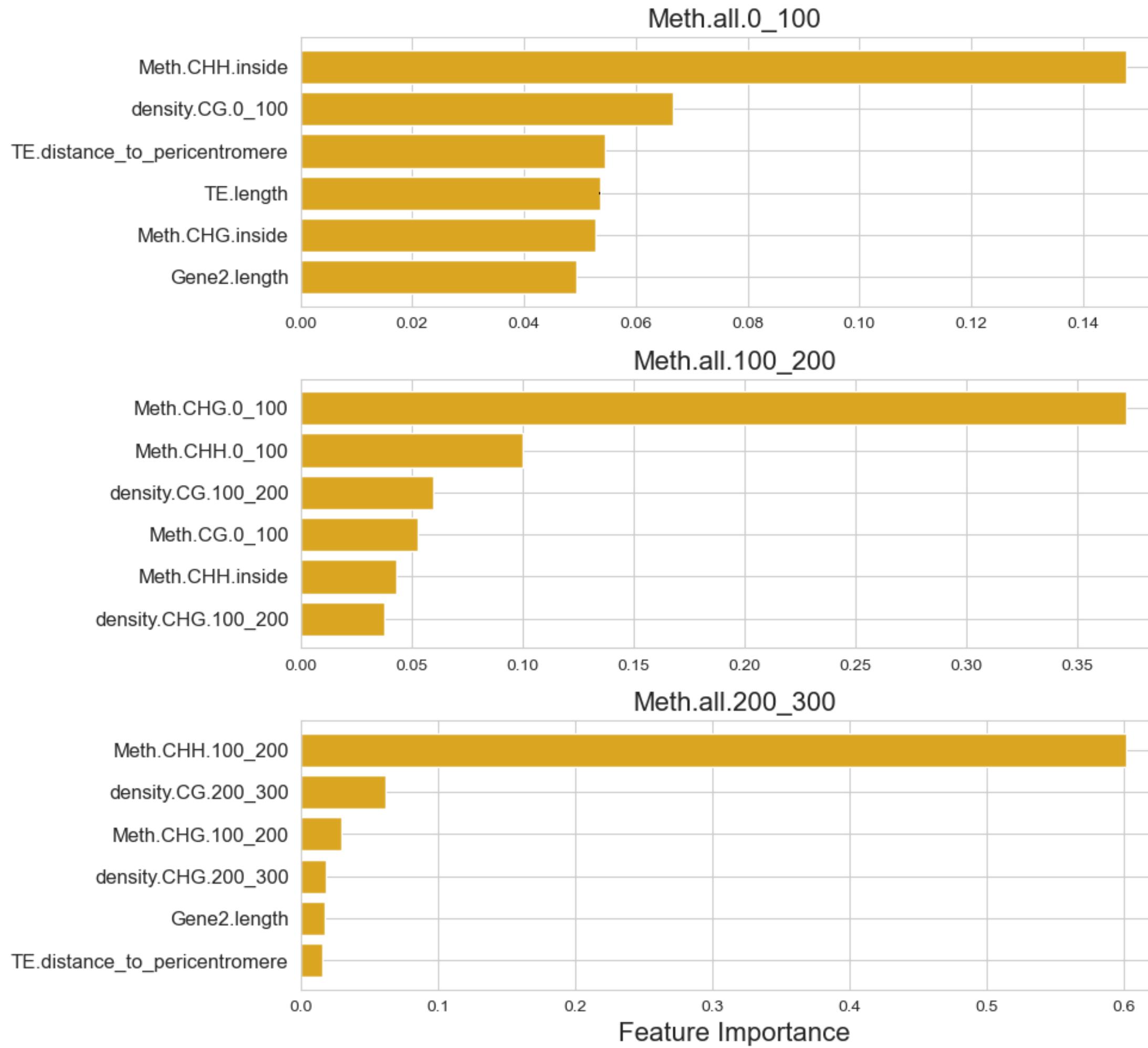
* 10 independent runs with different random seeds



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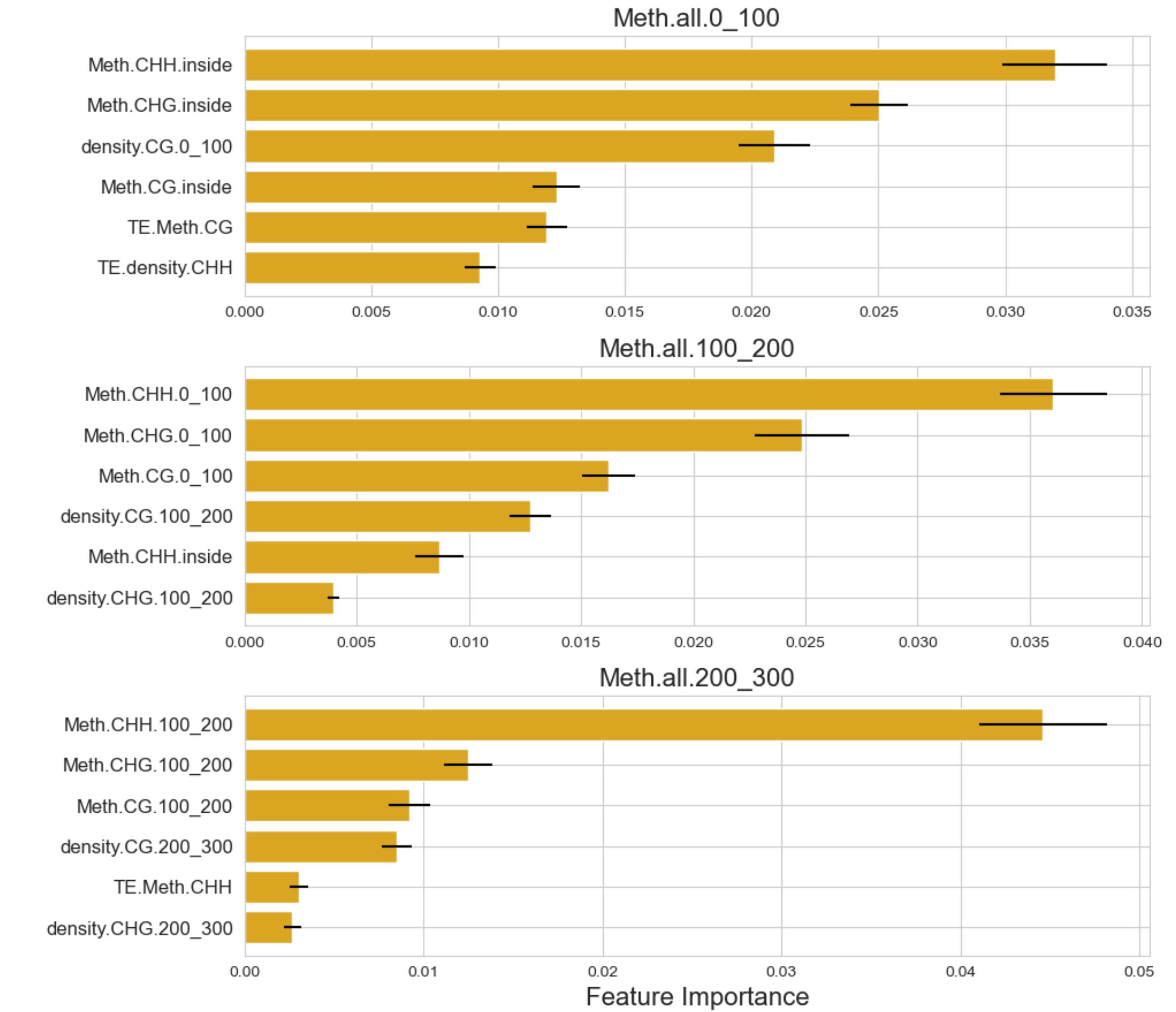
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SHAP values

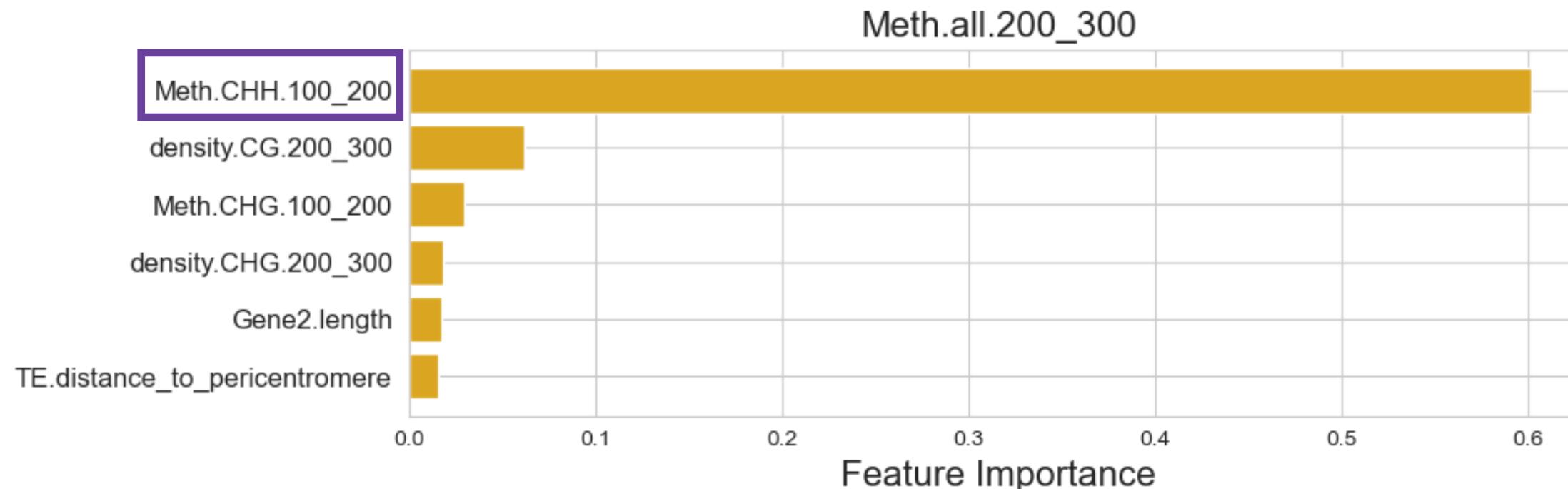
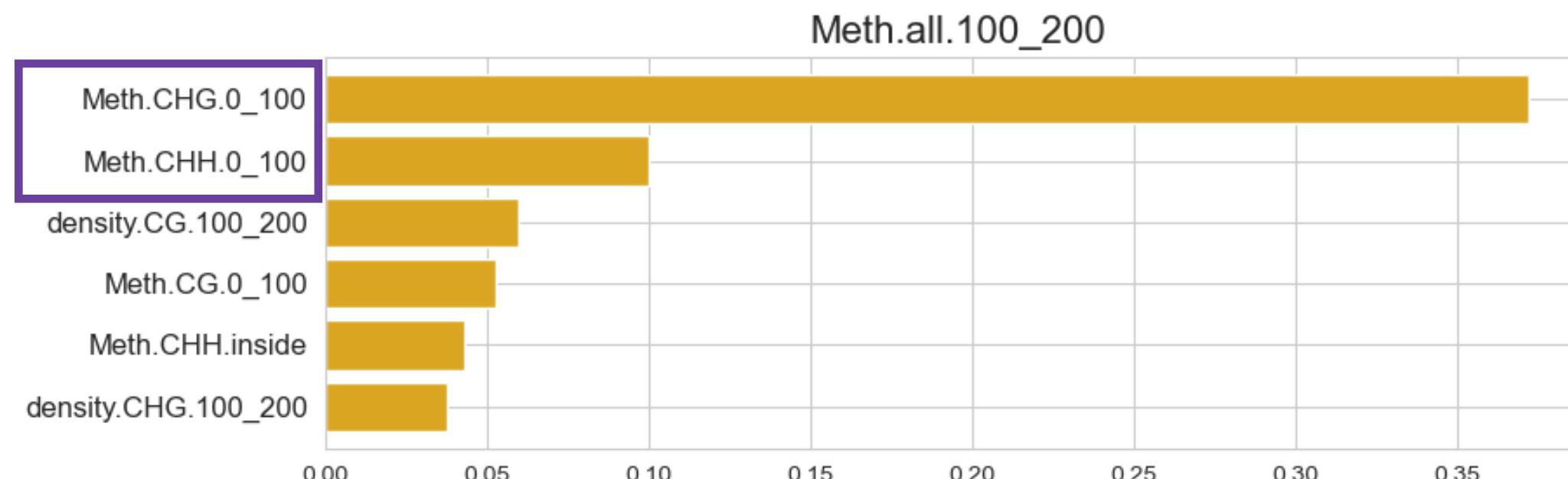
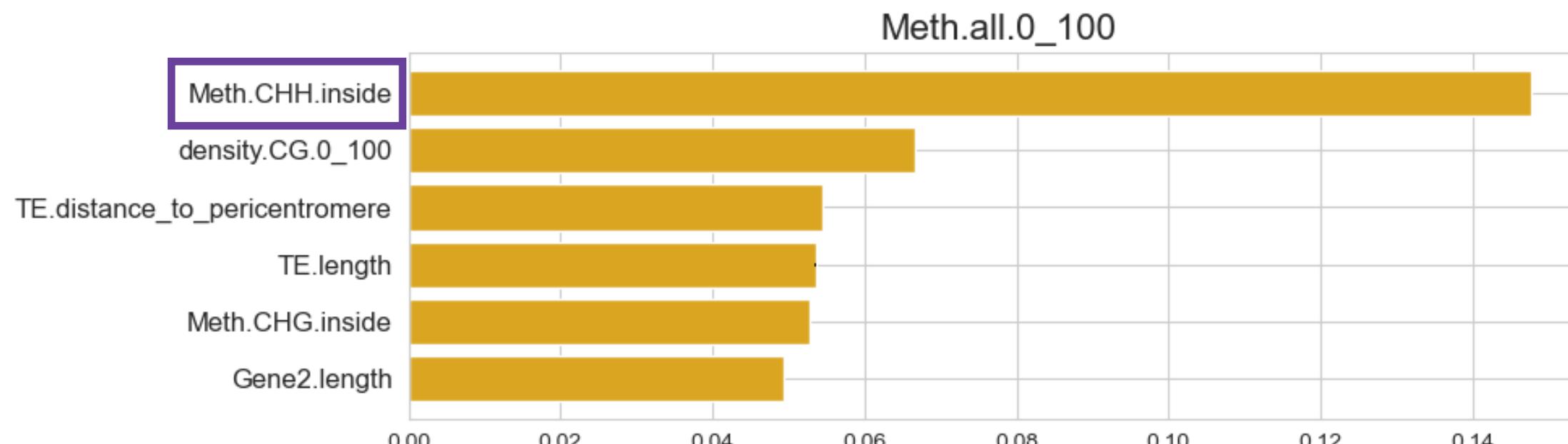
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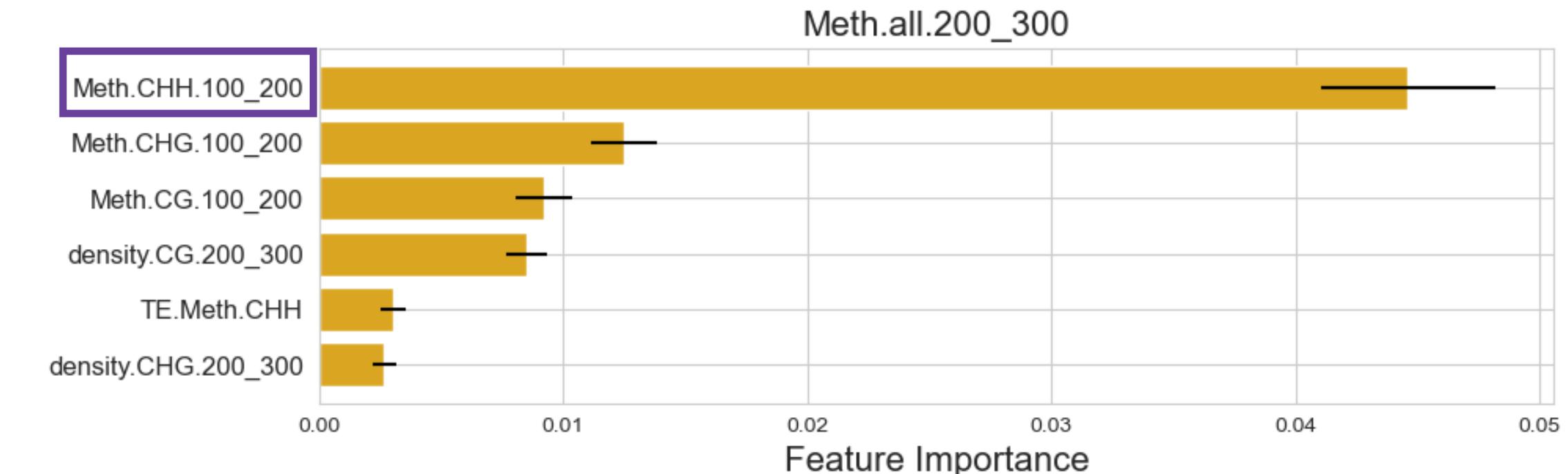
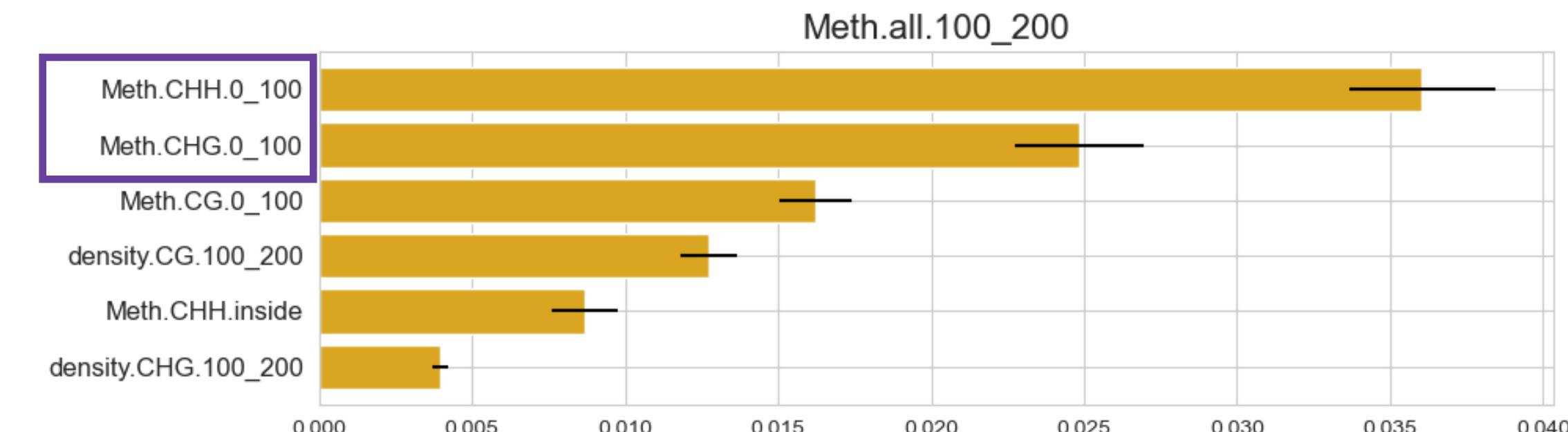
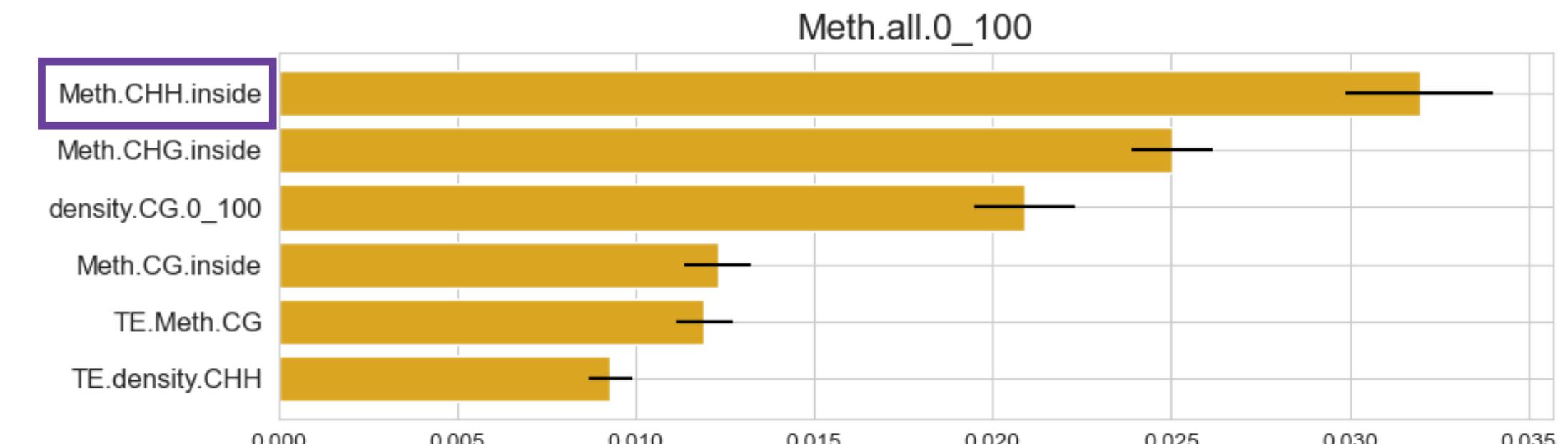
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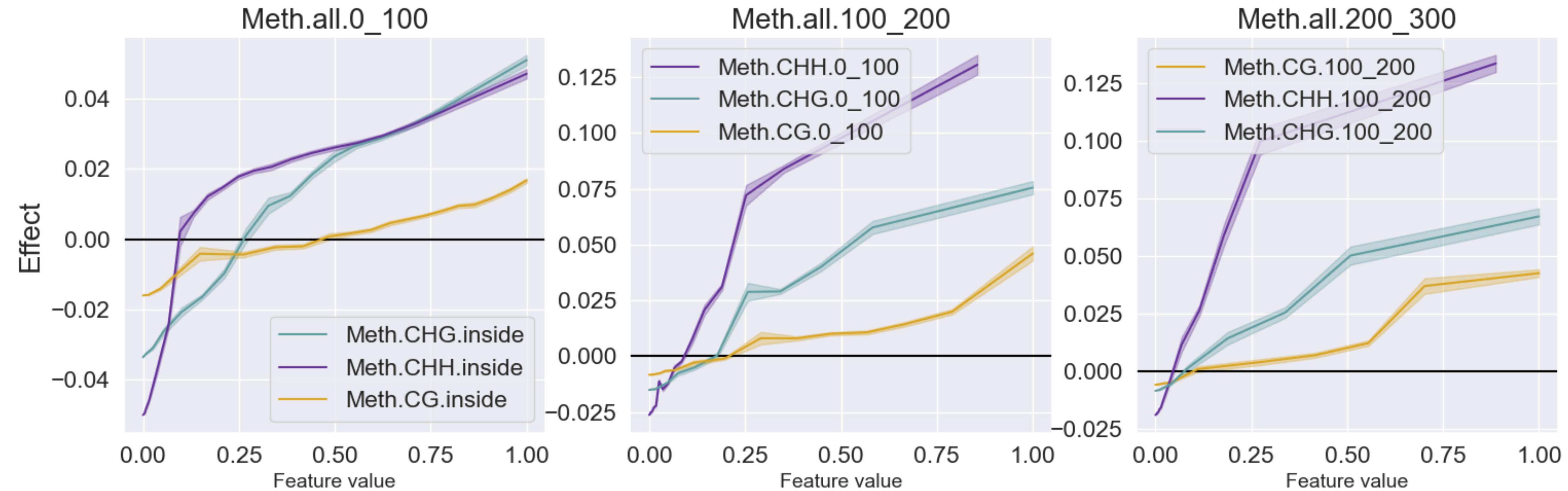
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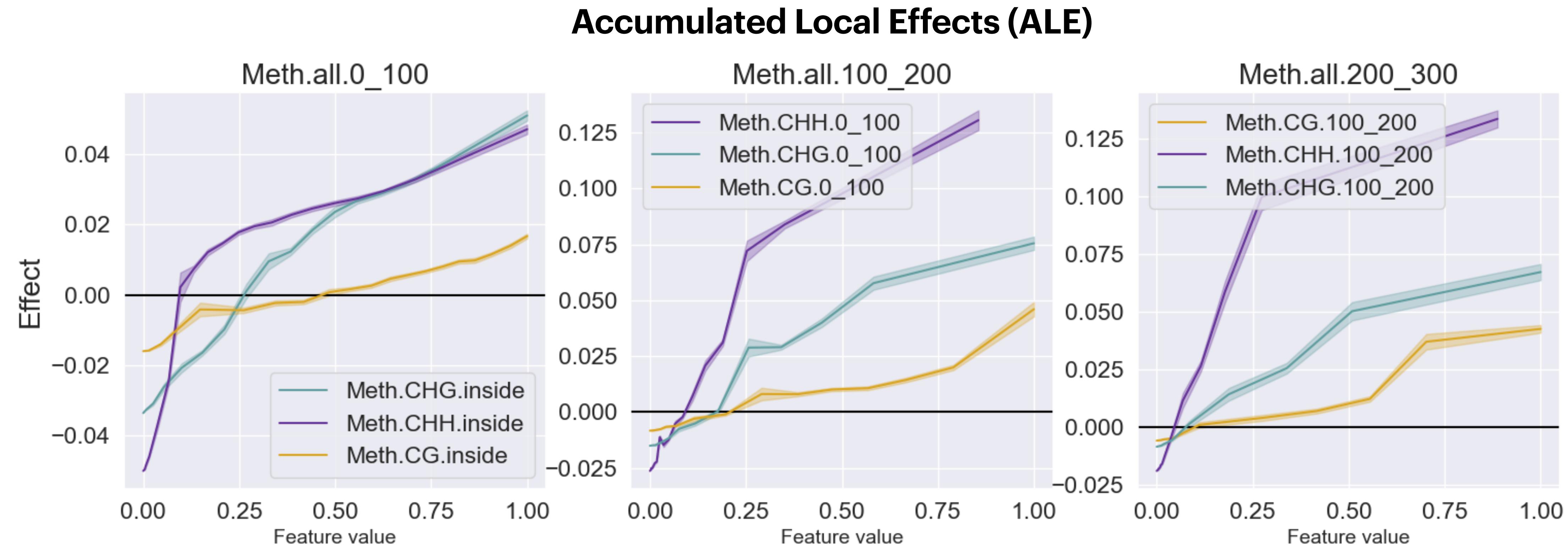


Prediction of methylation spreading

Accumulated Local Effects (ALE)



Prediction of methylation spreading

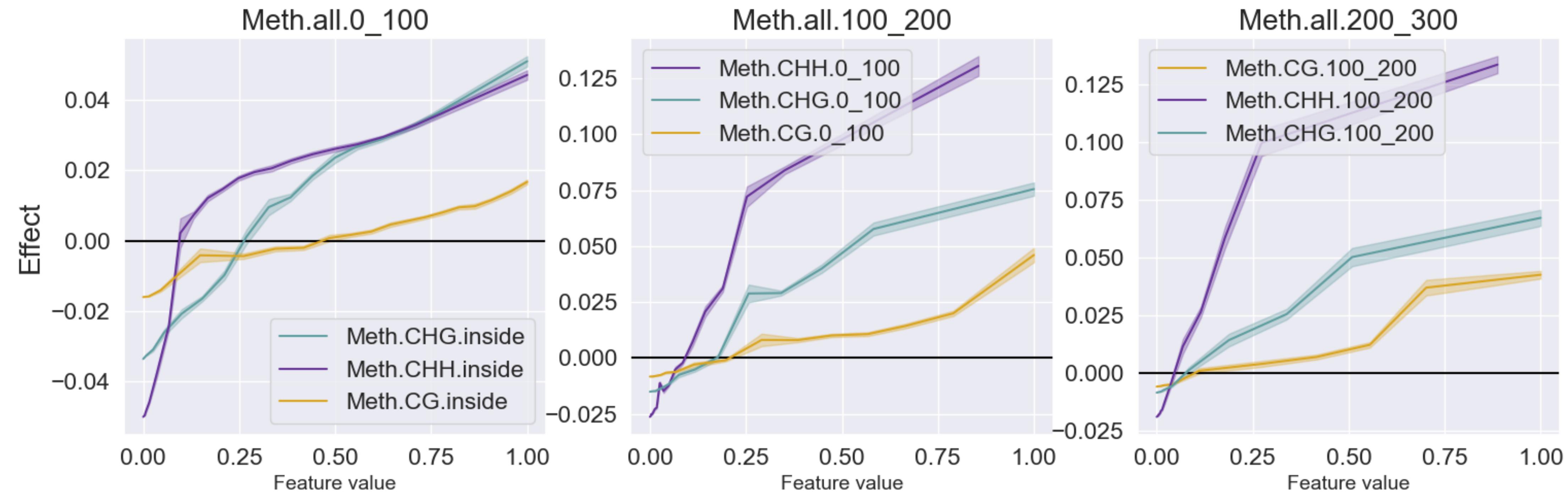


Conclusion:

- **Methylation of the TE inside ends** consistently comes as the most important feature with monotonous effect increase
- TE is **methylated on the ends** \rightarrow more likely to spread

Prediction of methylation spreading

Accumulated Local Effects (ALE)



Conclusion:

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Question:

- What defines the **methylation of the TE inside ends**?

Prediction of inside methylation

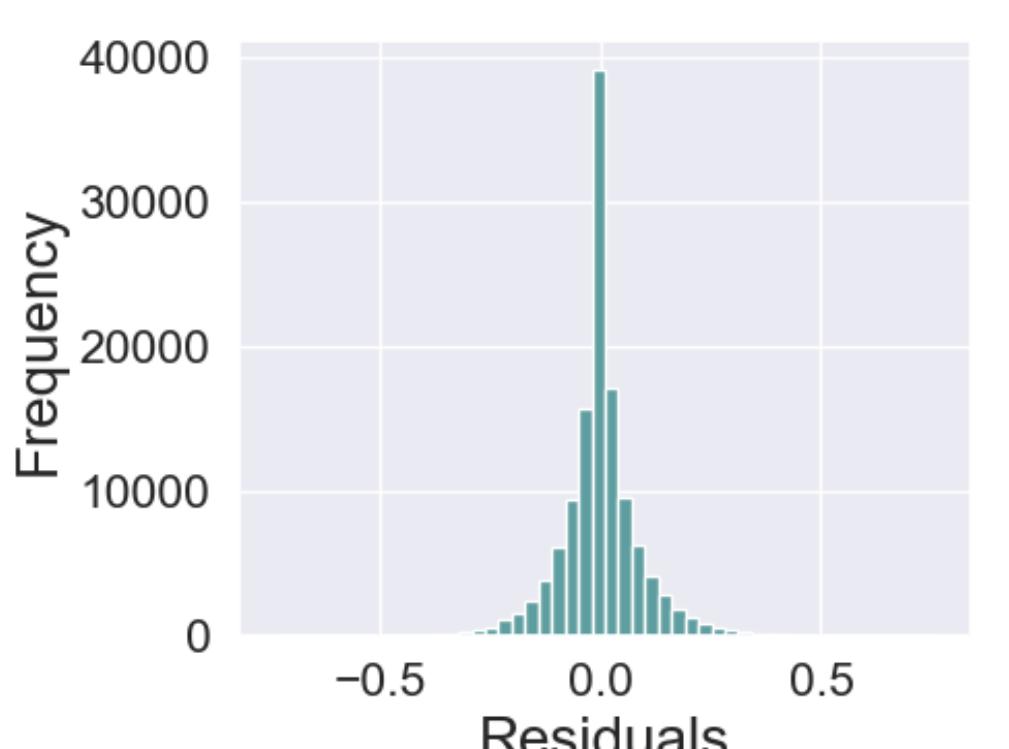
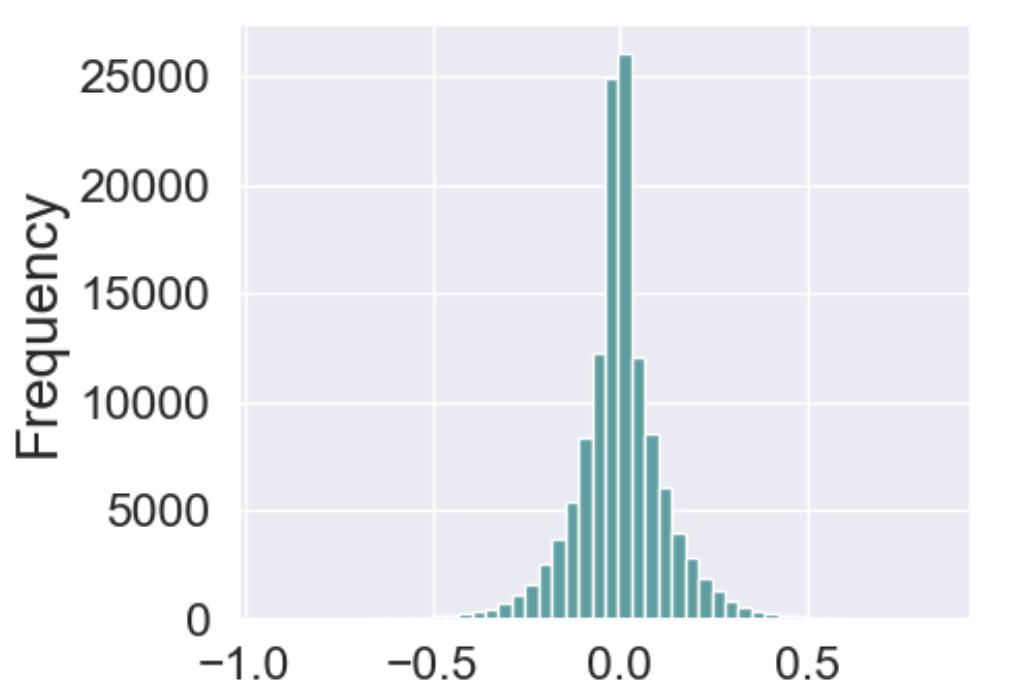
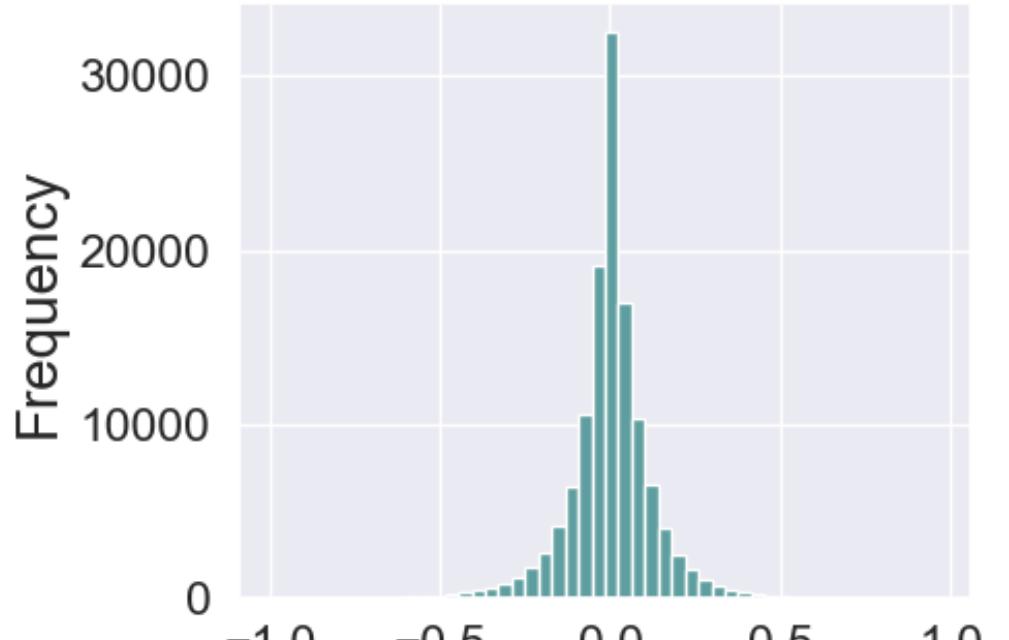
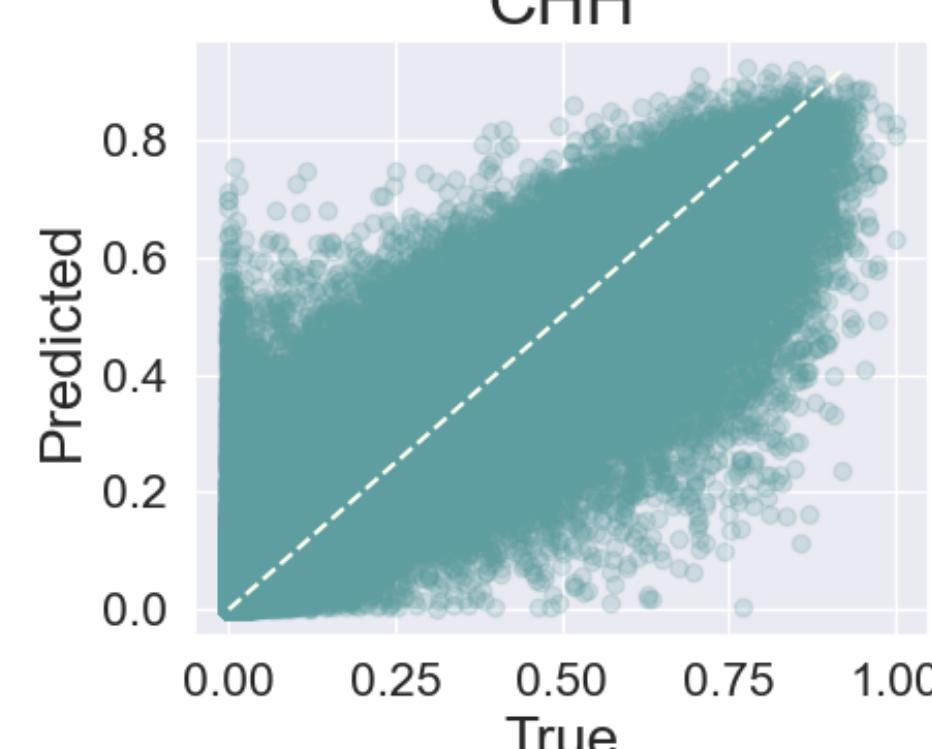
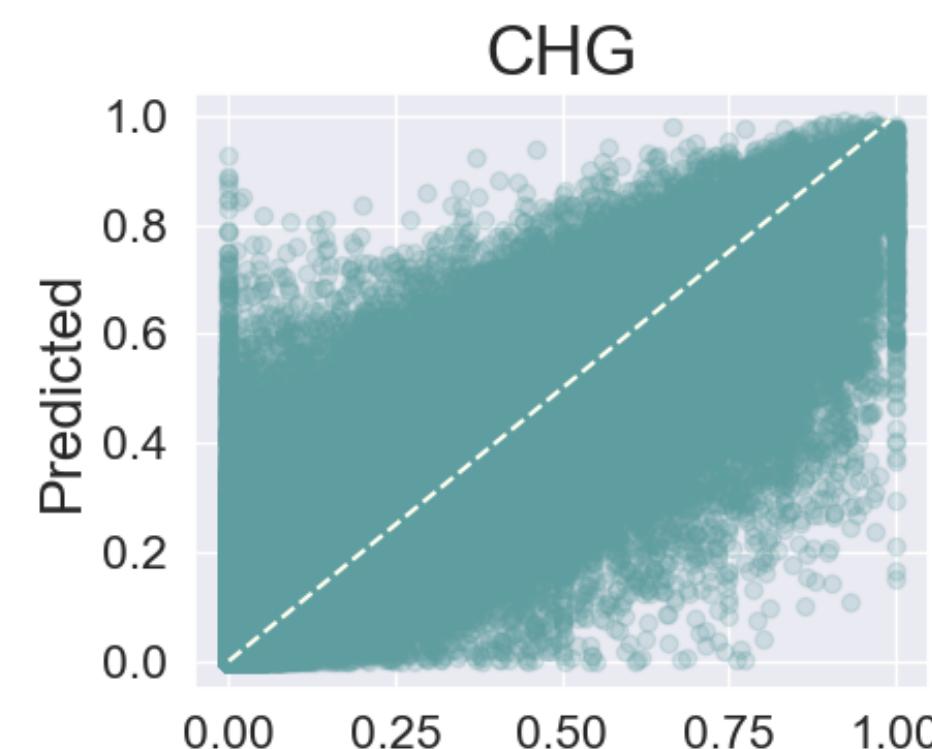
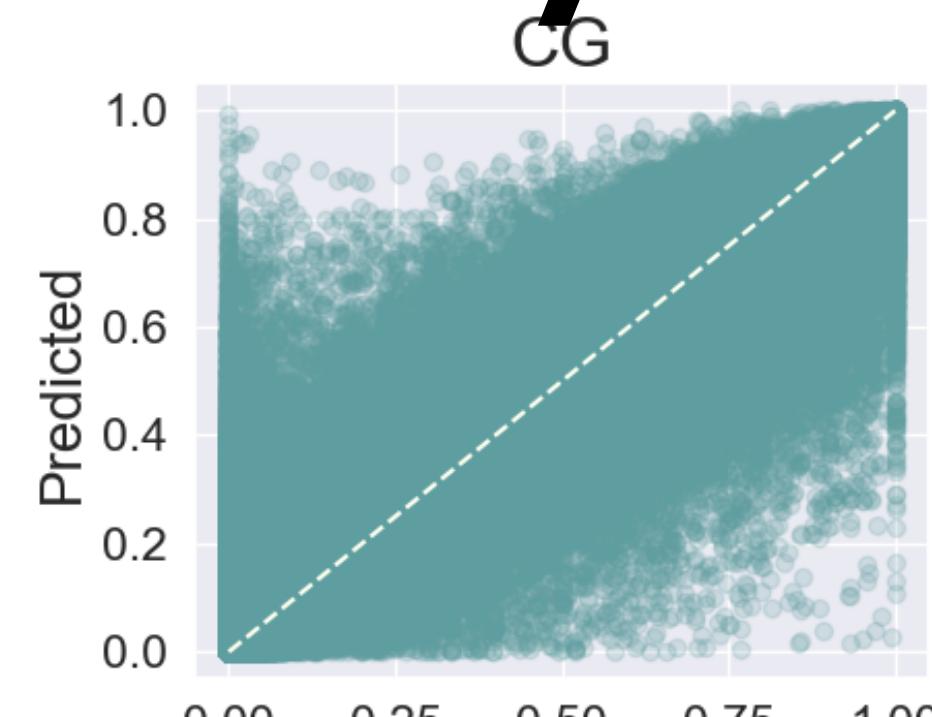
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Prediction of inside methylation

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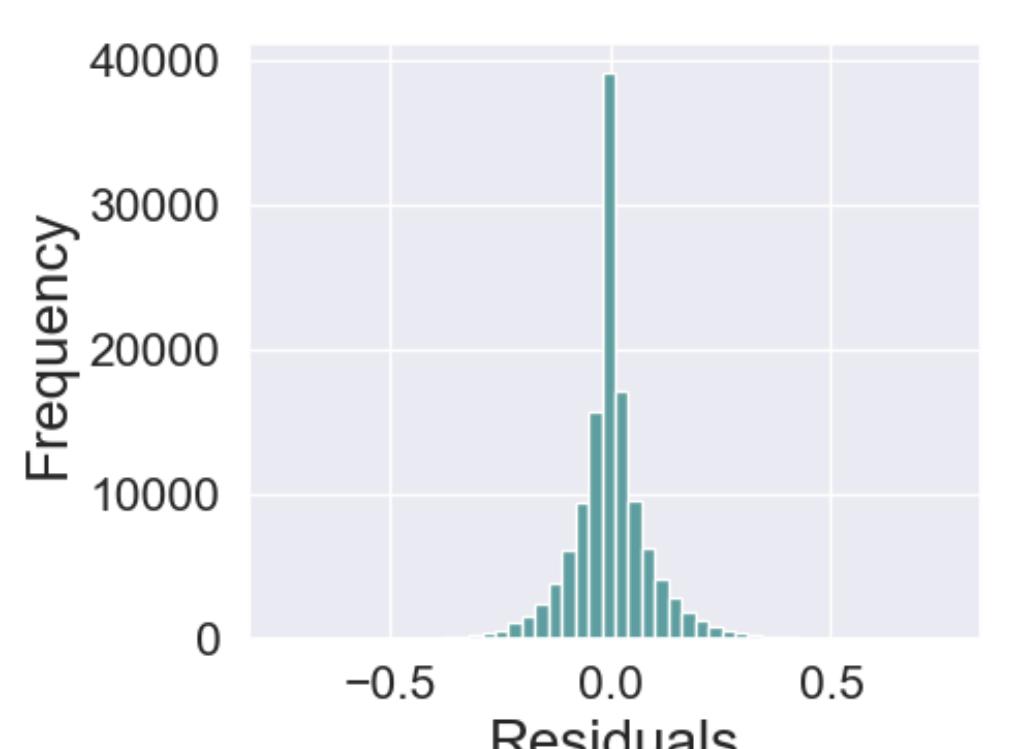
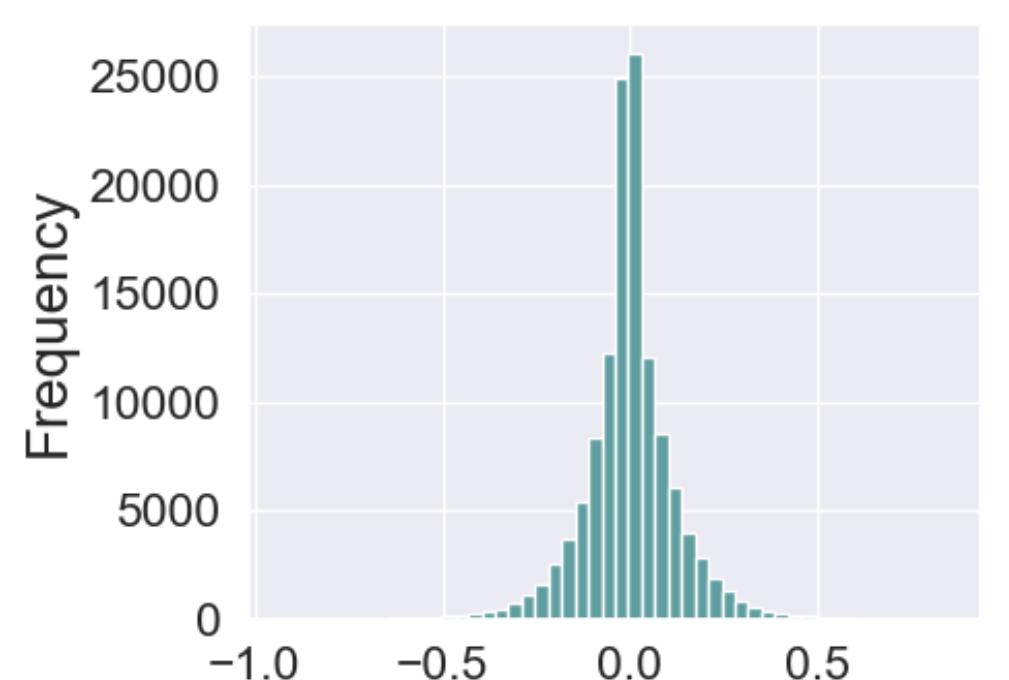
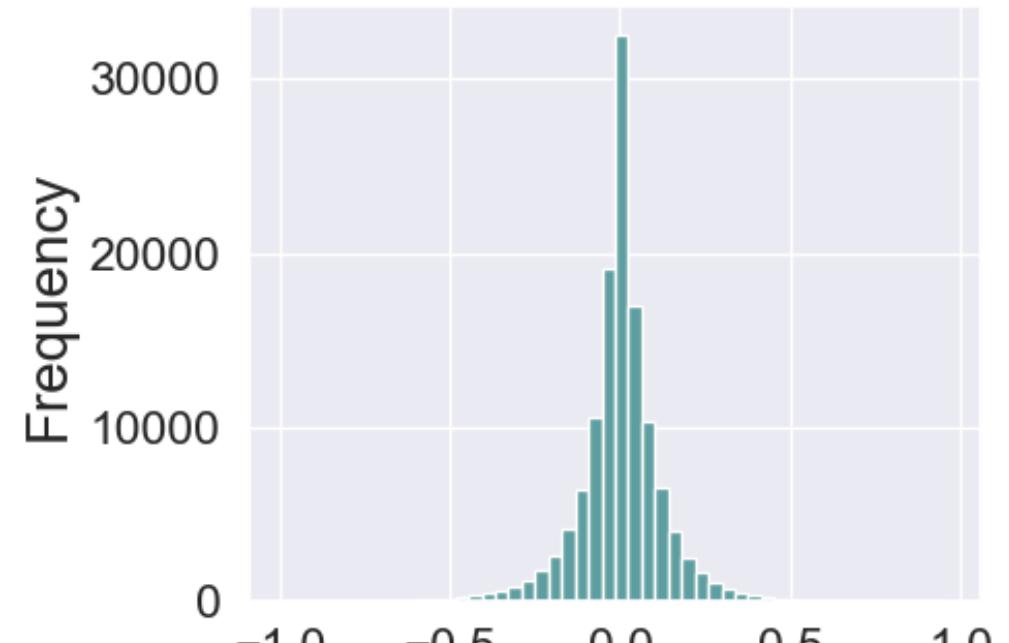
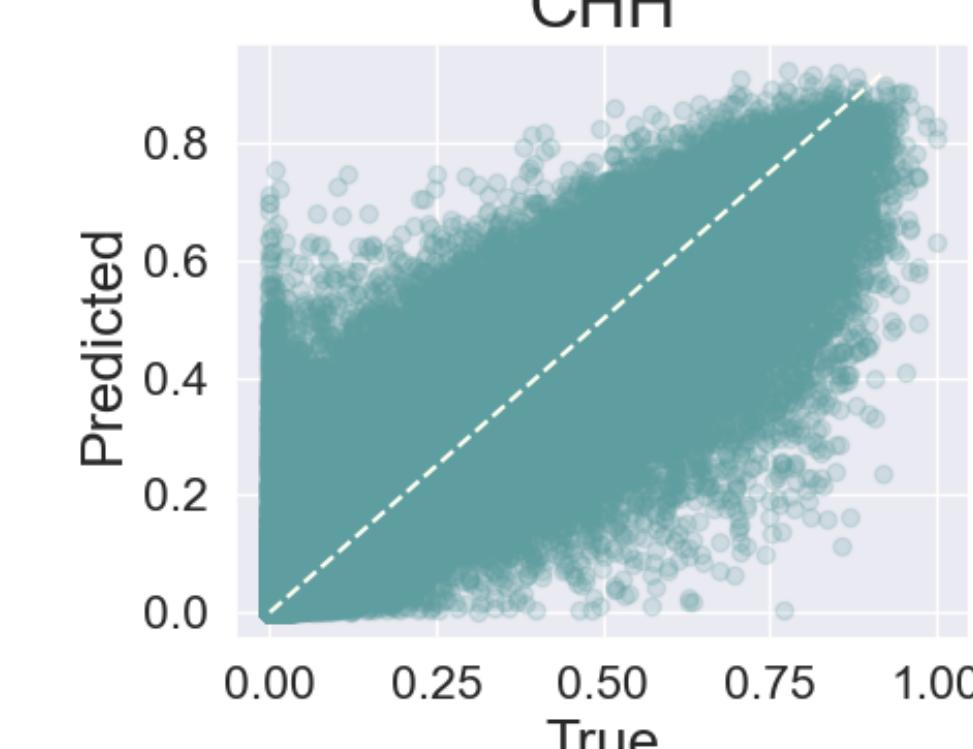
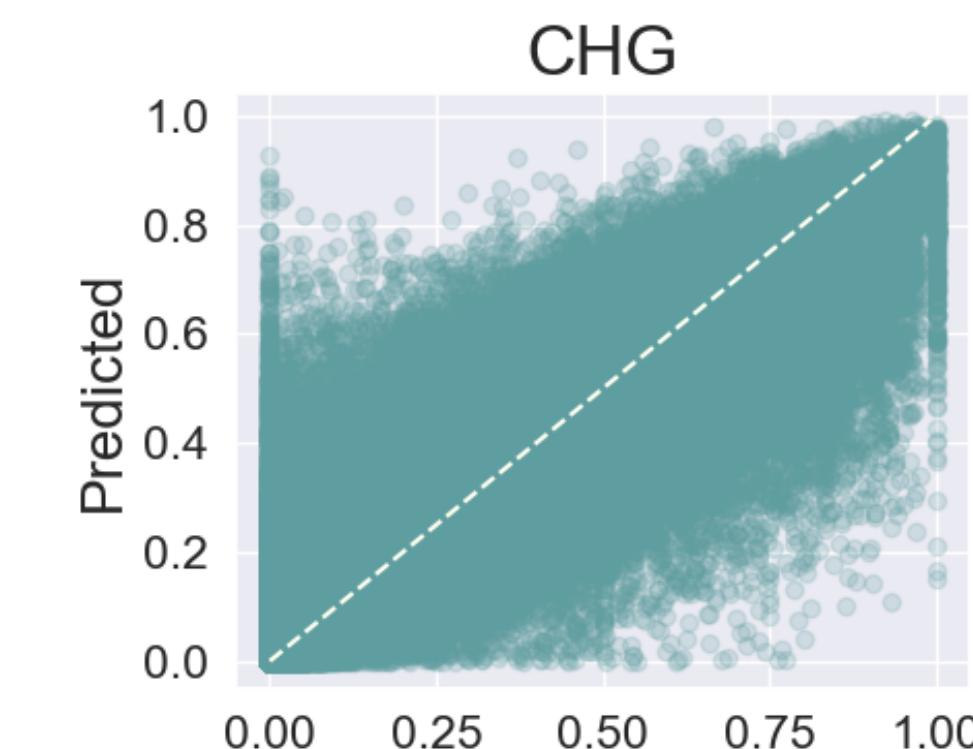
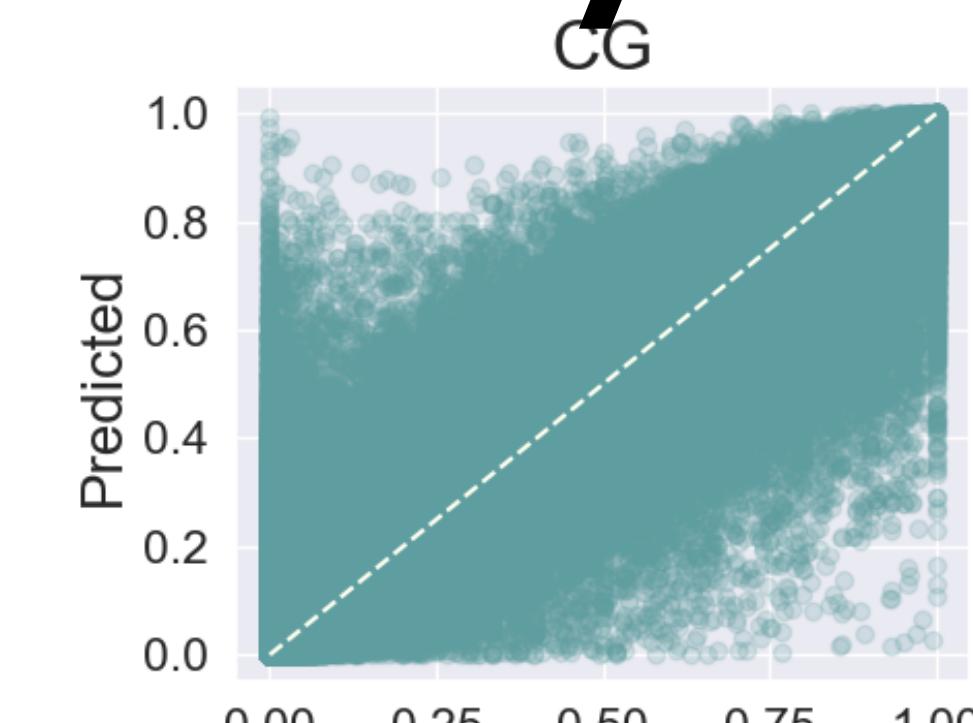
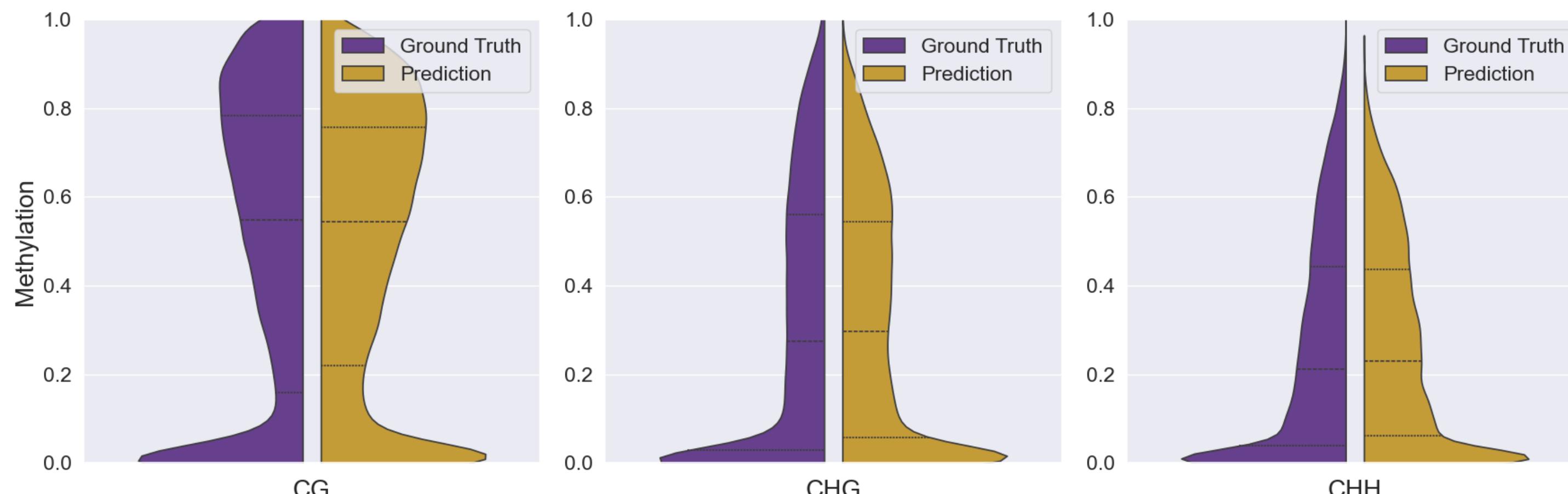
- **TE** (length, distance to pericentromere, superfamily, insertion frequency, divergence, if inside a gene)
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- **Densities** of CG, CHG, CHH contexts



Prediction of inside methylation

Features:

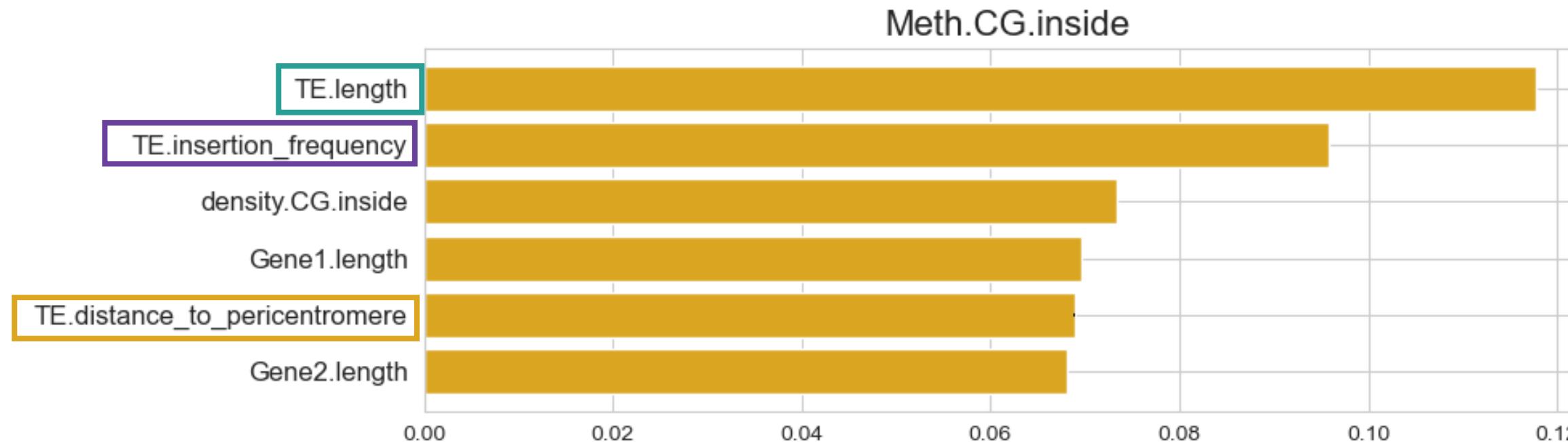
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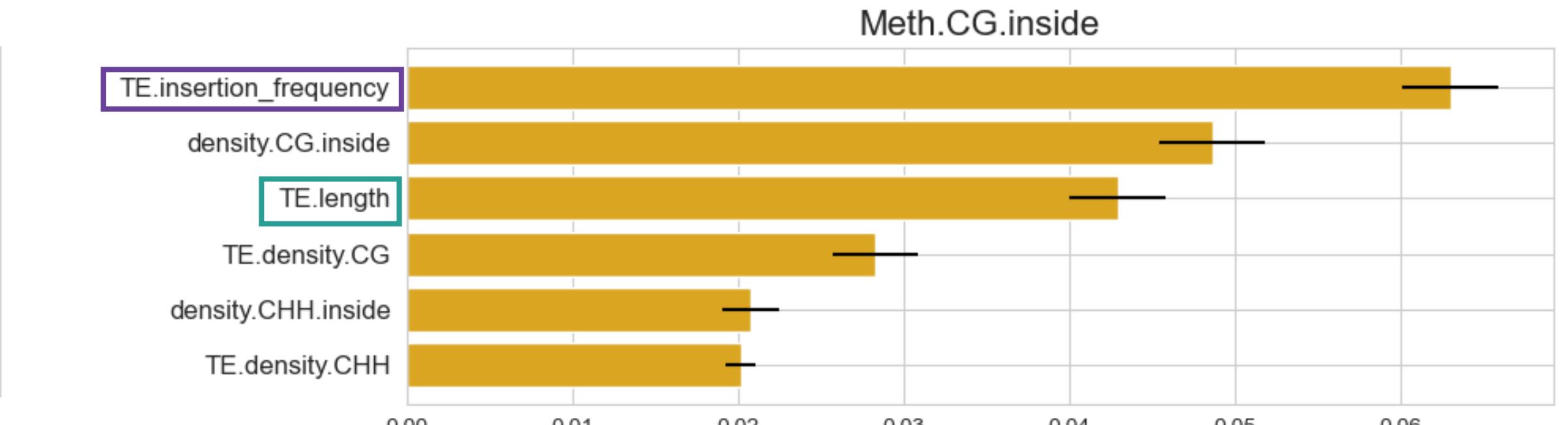
Impurity-based feature importances

* 10 independent runs with different random seeds

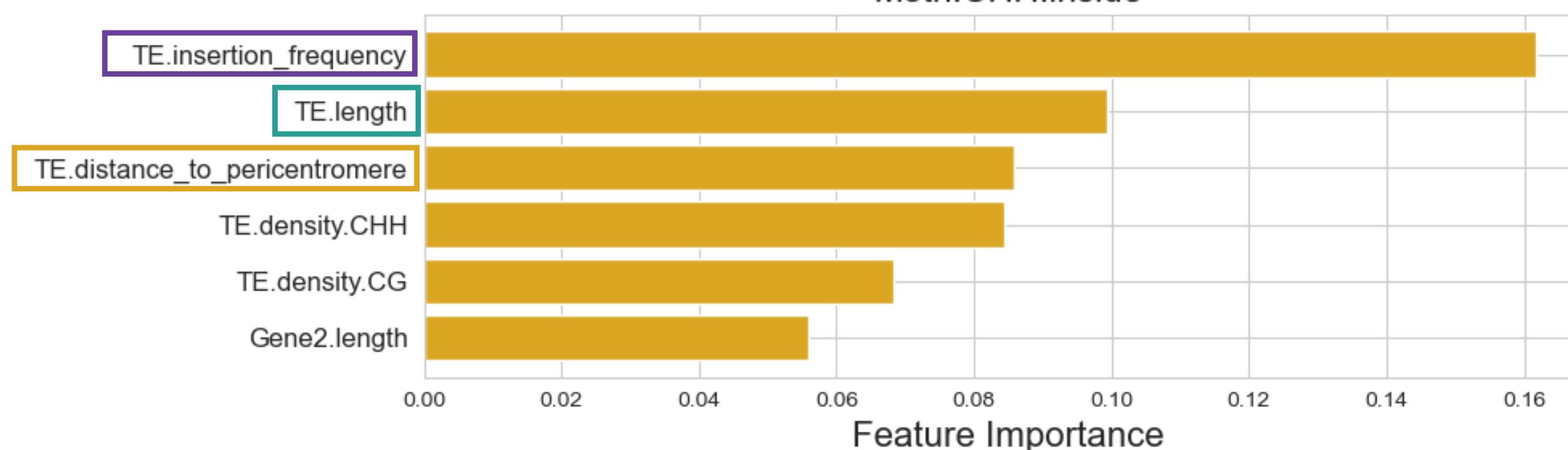


SHAP values

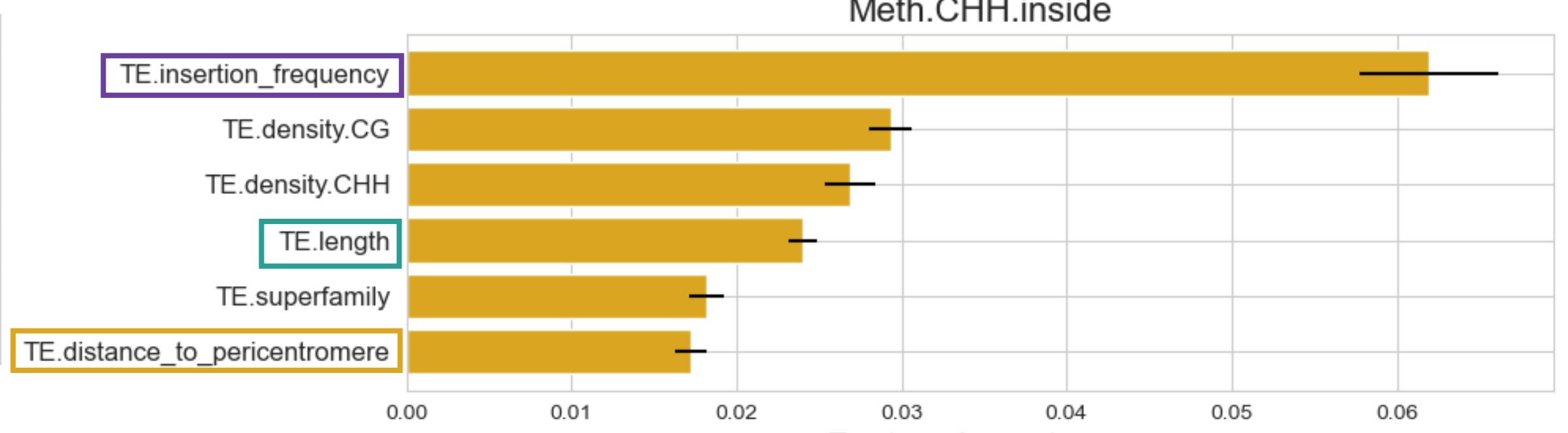
* 10 folds in a cross-validation manner



Meth.CHH.inside



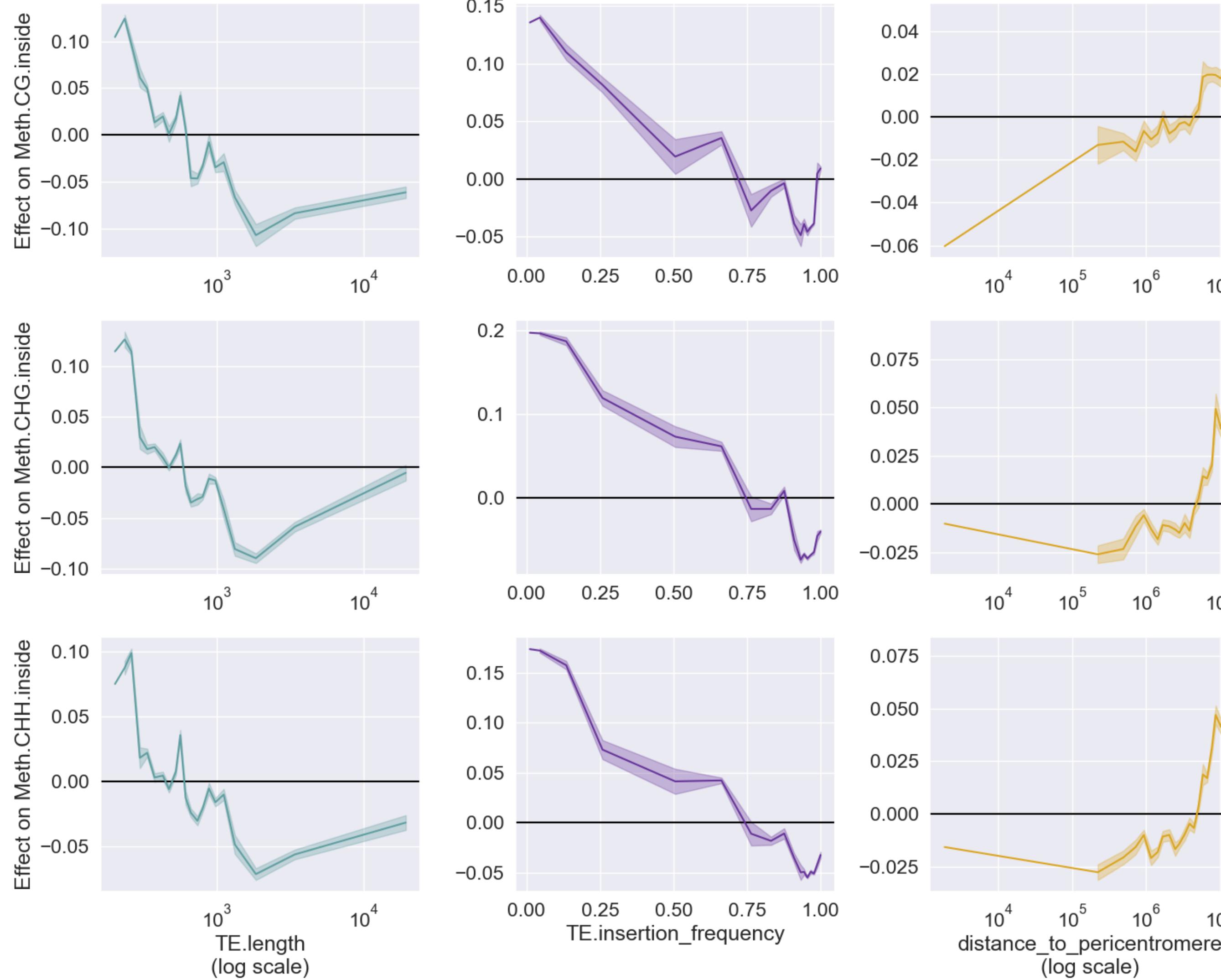
Meth.CHH.inside



Feature Importance

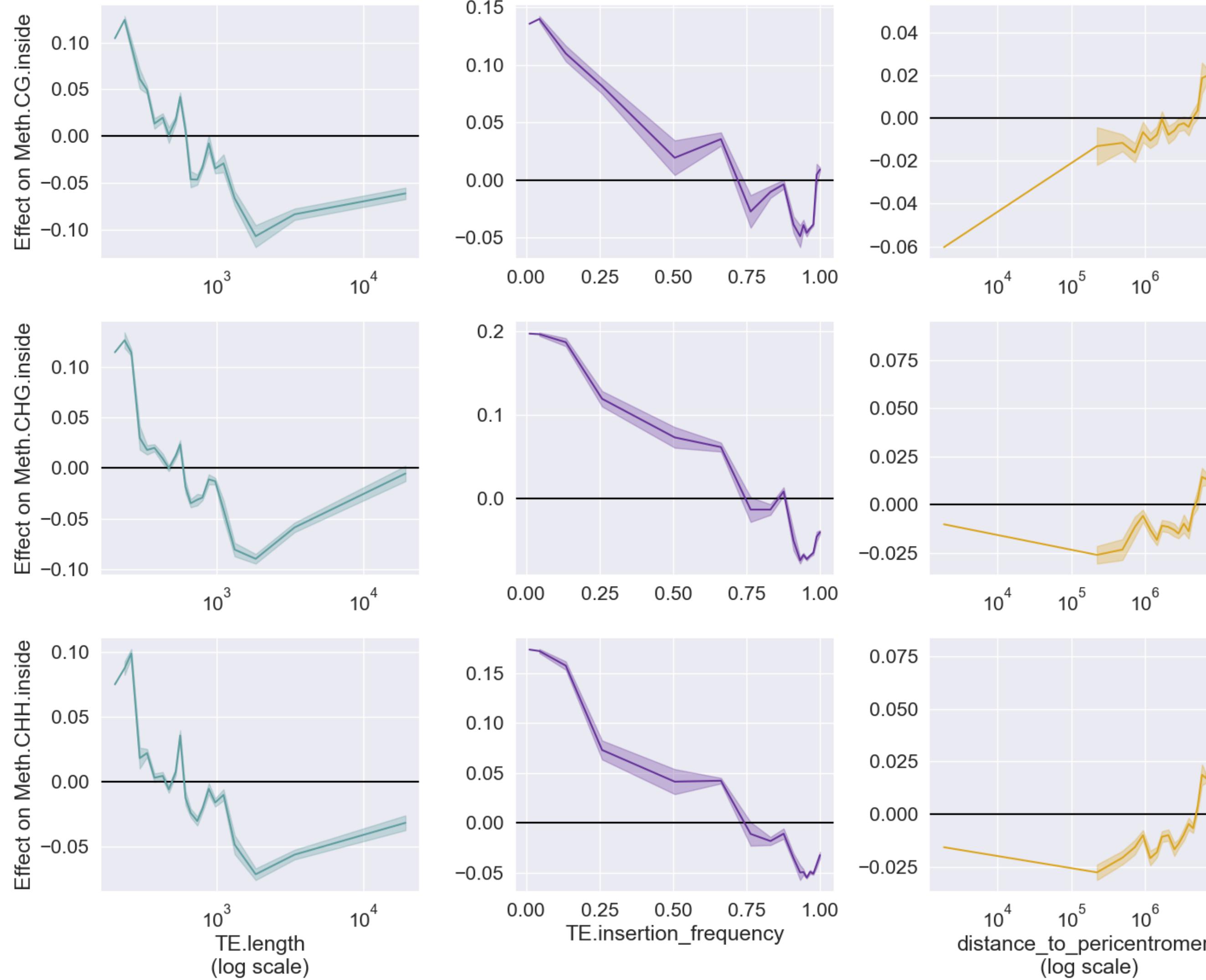
Prediction of inside methylation

Accumulated Local Effects



Prediction of inside methylation

Accumulated Local Effects



Conclusions

- Length of TE is unlikely to be a driving factor (rather a confounder)
- **Insertion frequency:** Rare (= new) TEs are targeted by methylation machinery

- **Distance to pericentromere:** More distant TEs are more likely to be targeted

Biological conclusions

- **The most important factors:**

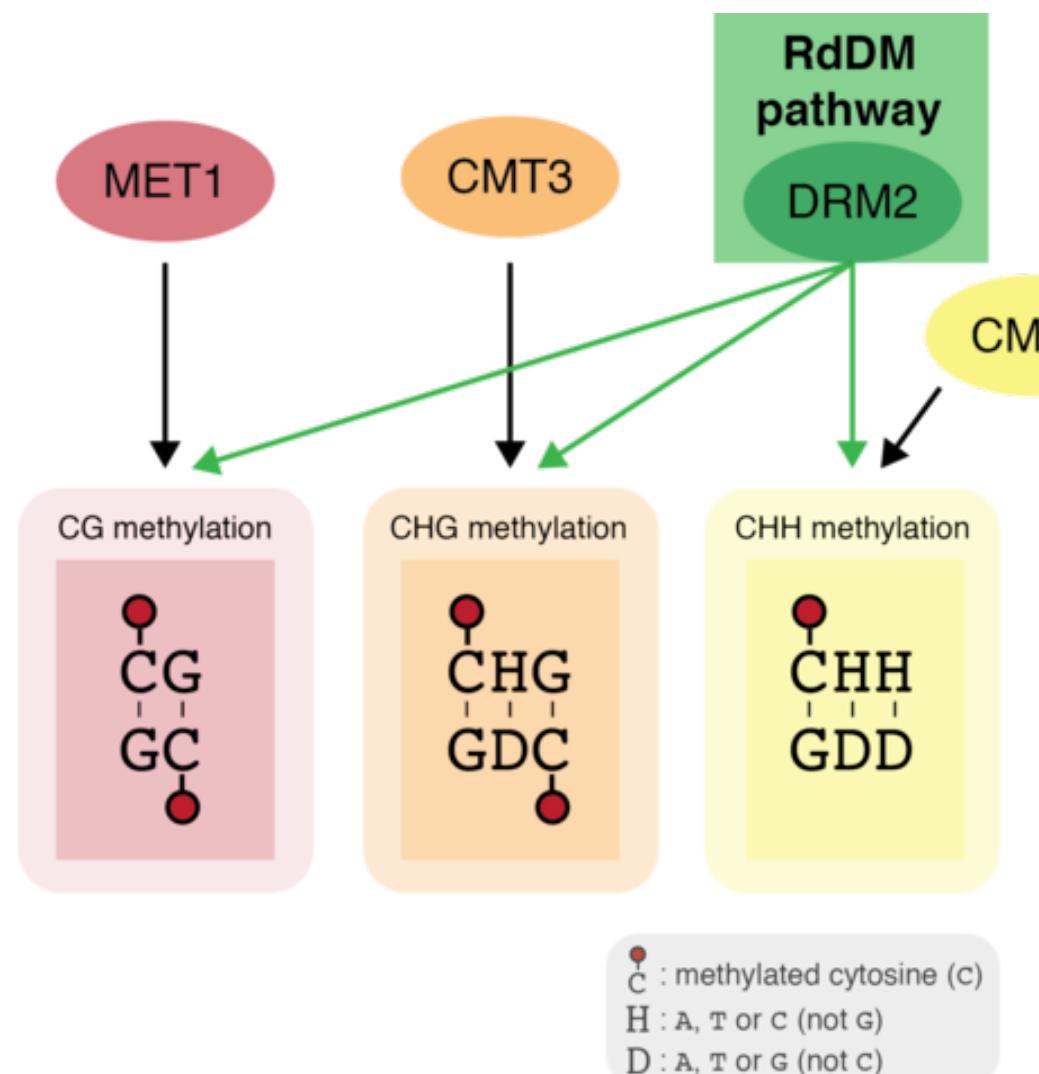
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Biological conclusions

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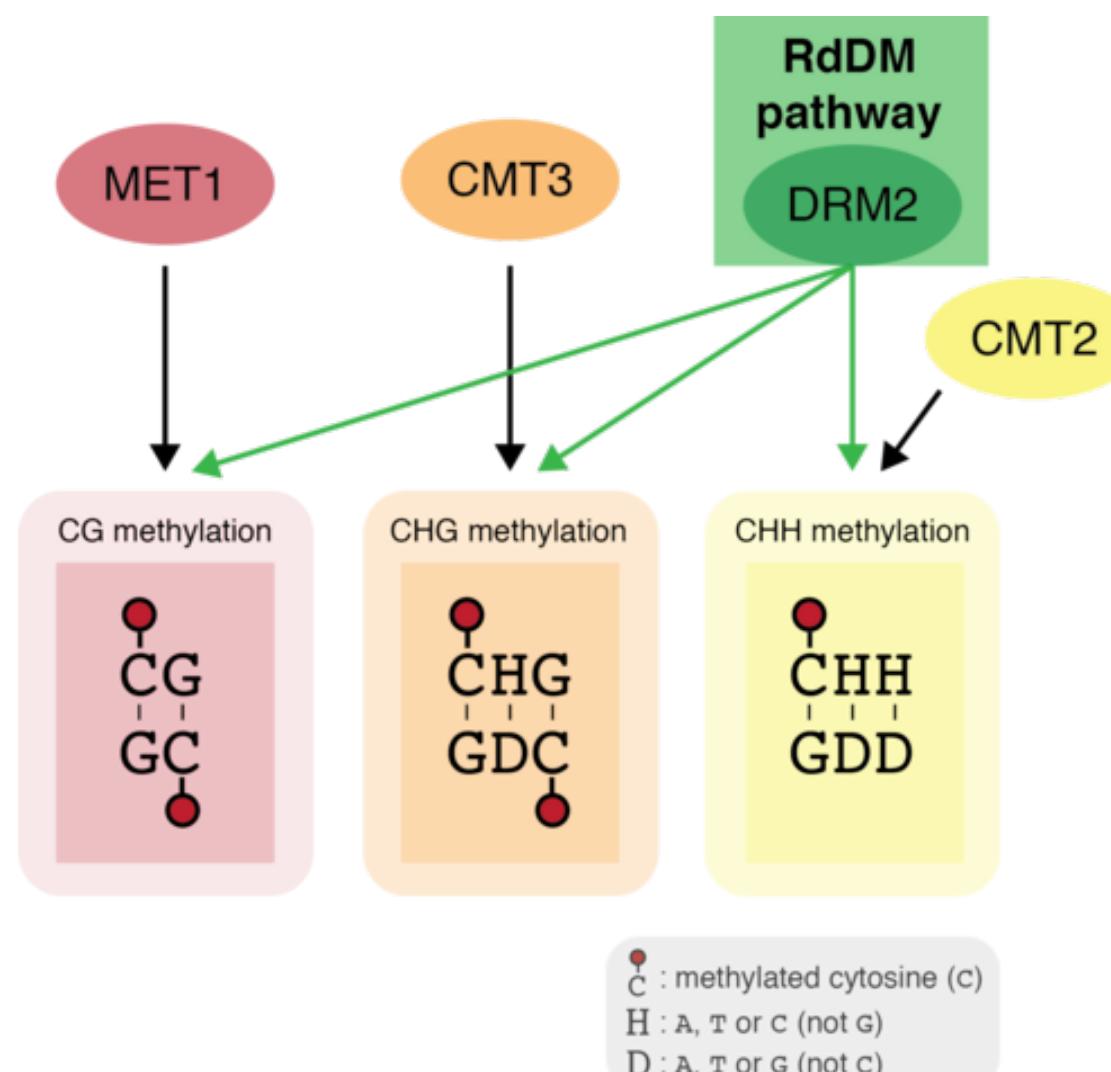
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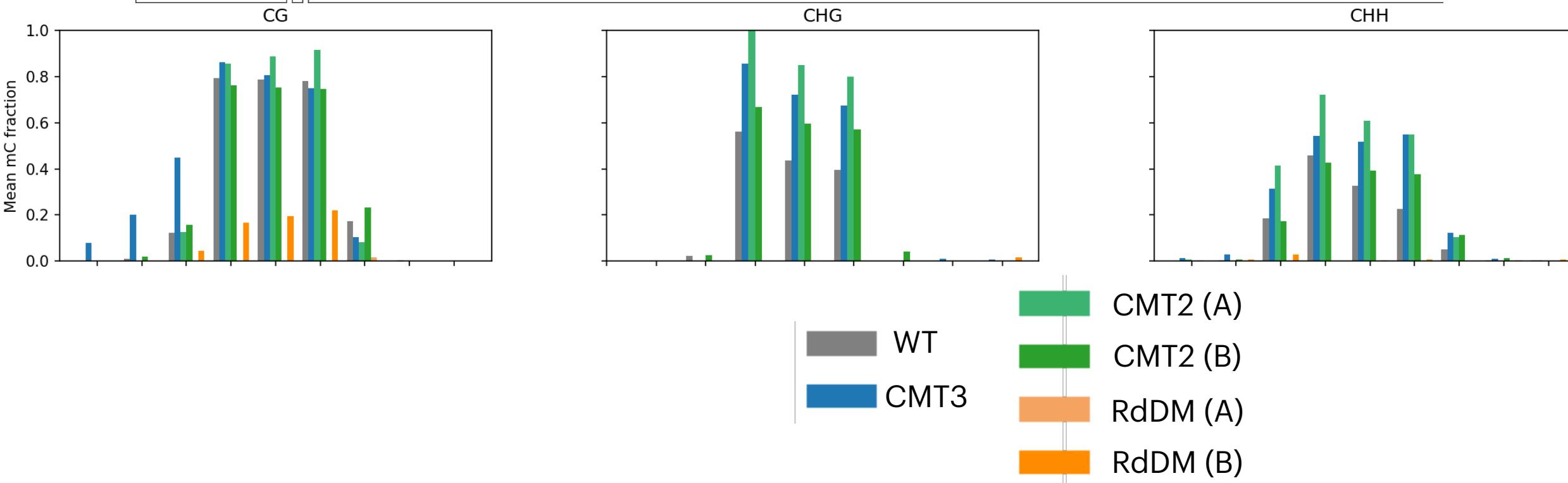
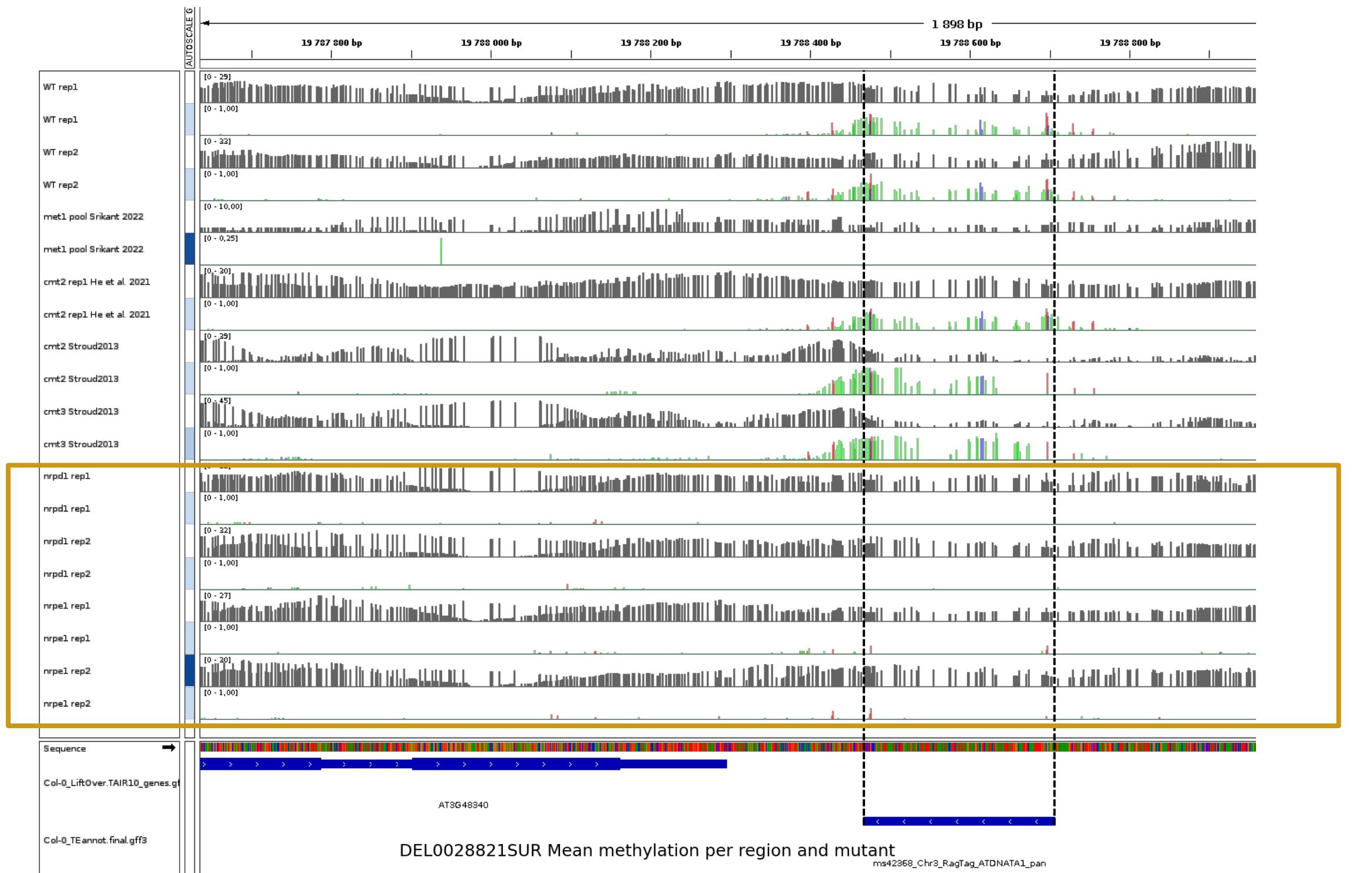
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- Test: mutants of Col-0 strain of *A. Thaliana* where different methylation pathways are knocked out

Biological confirmation

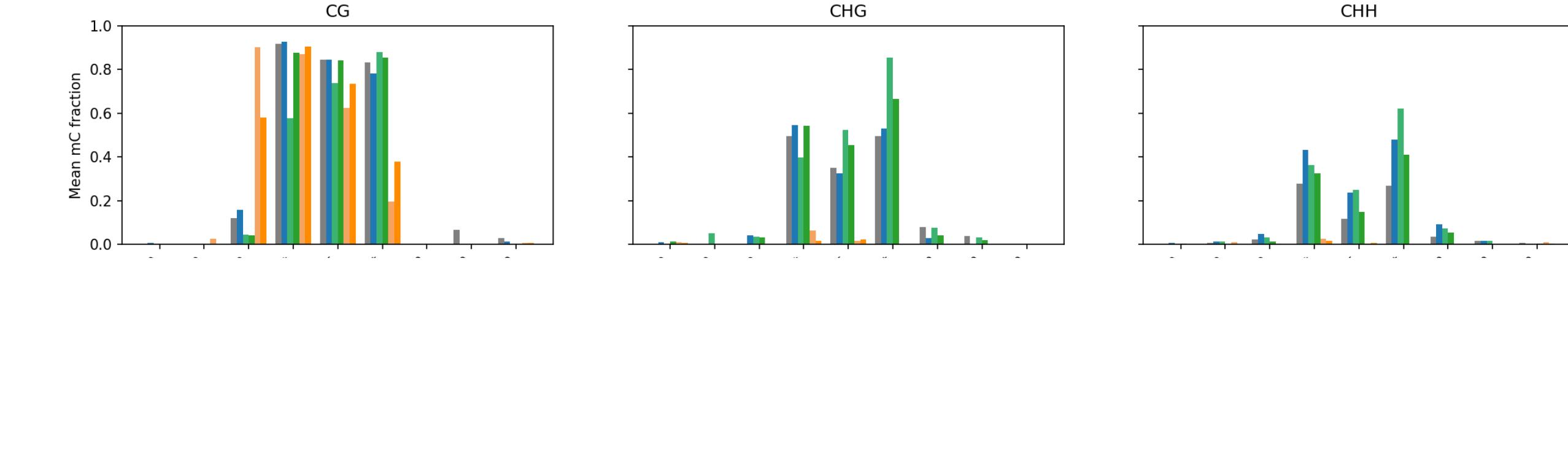
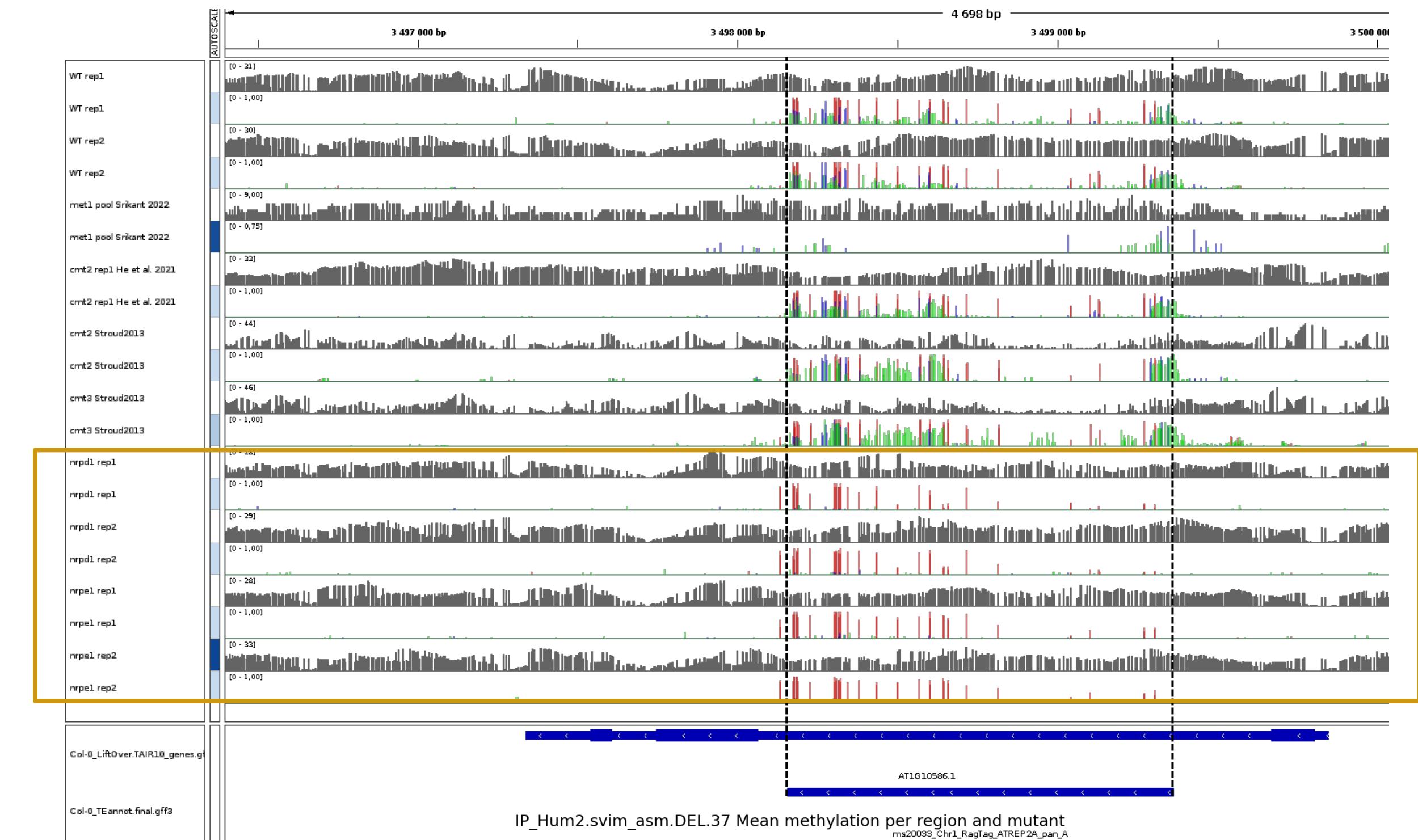
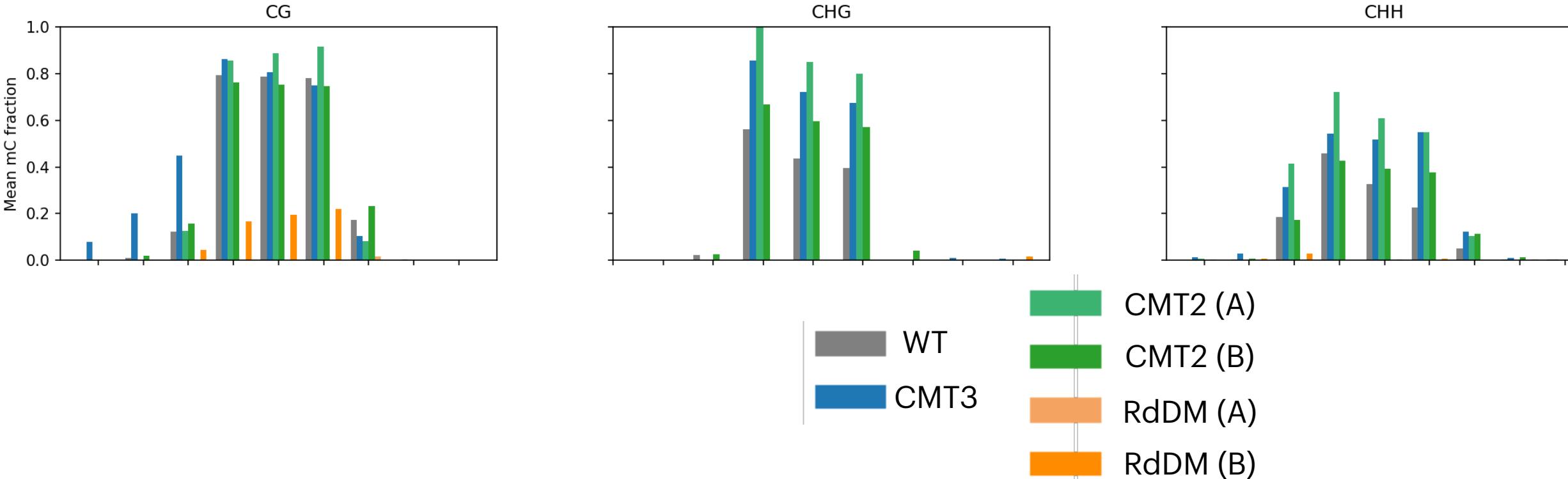
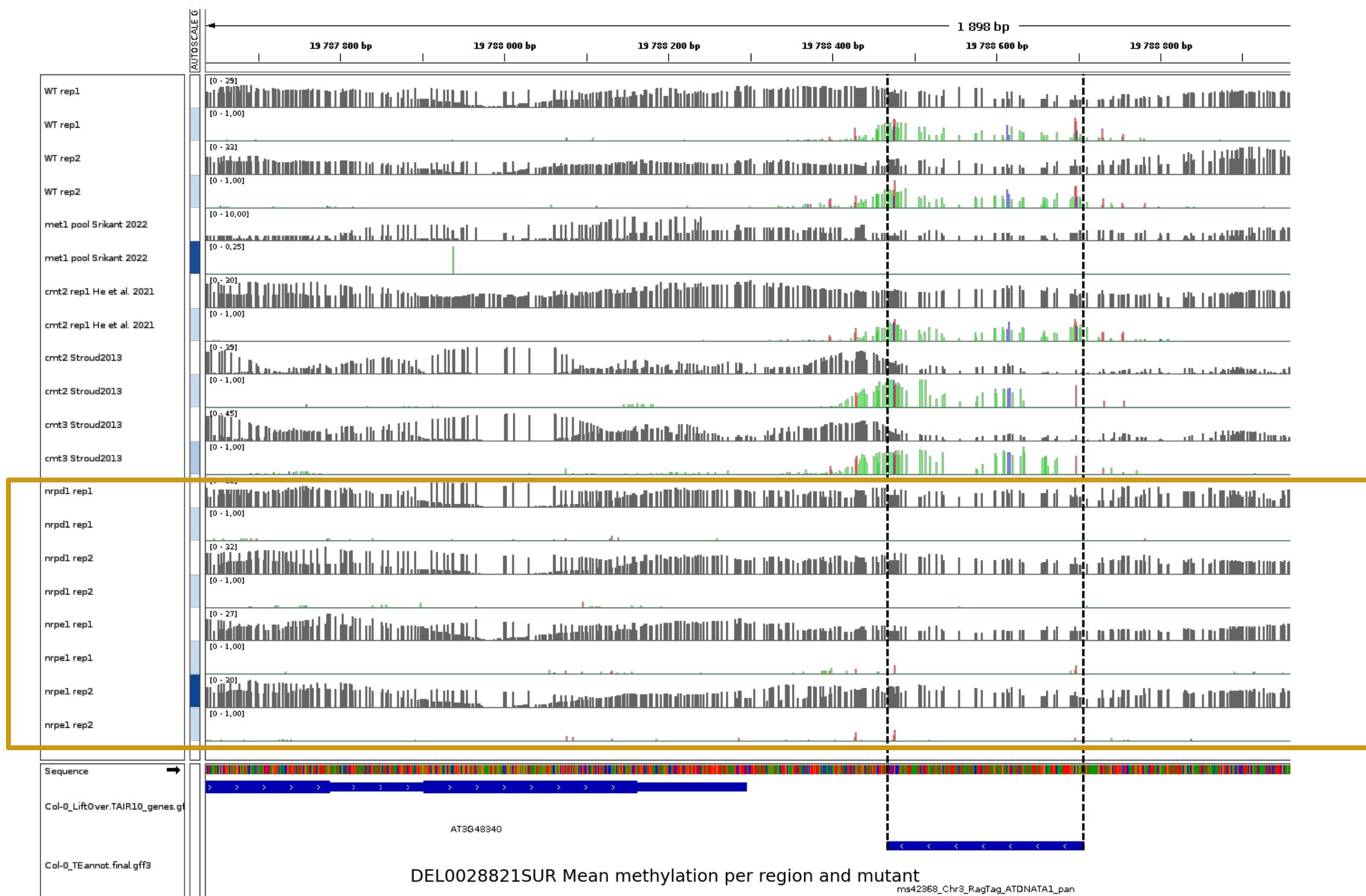
DEL0028821SUR



Biological confirmation

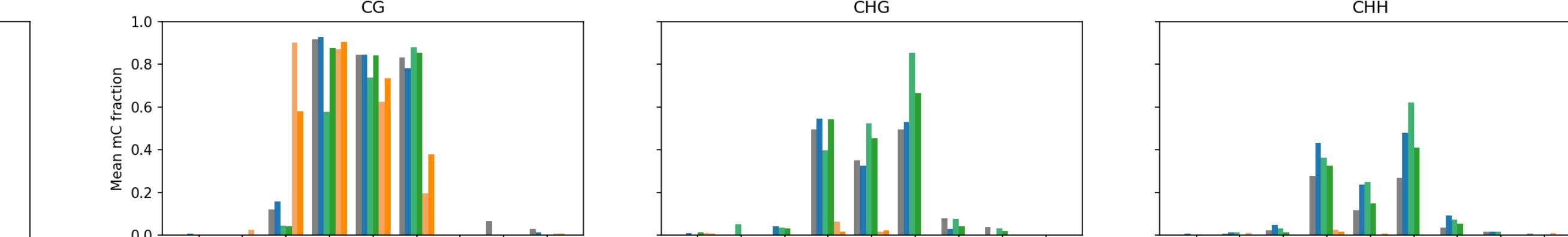
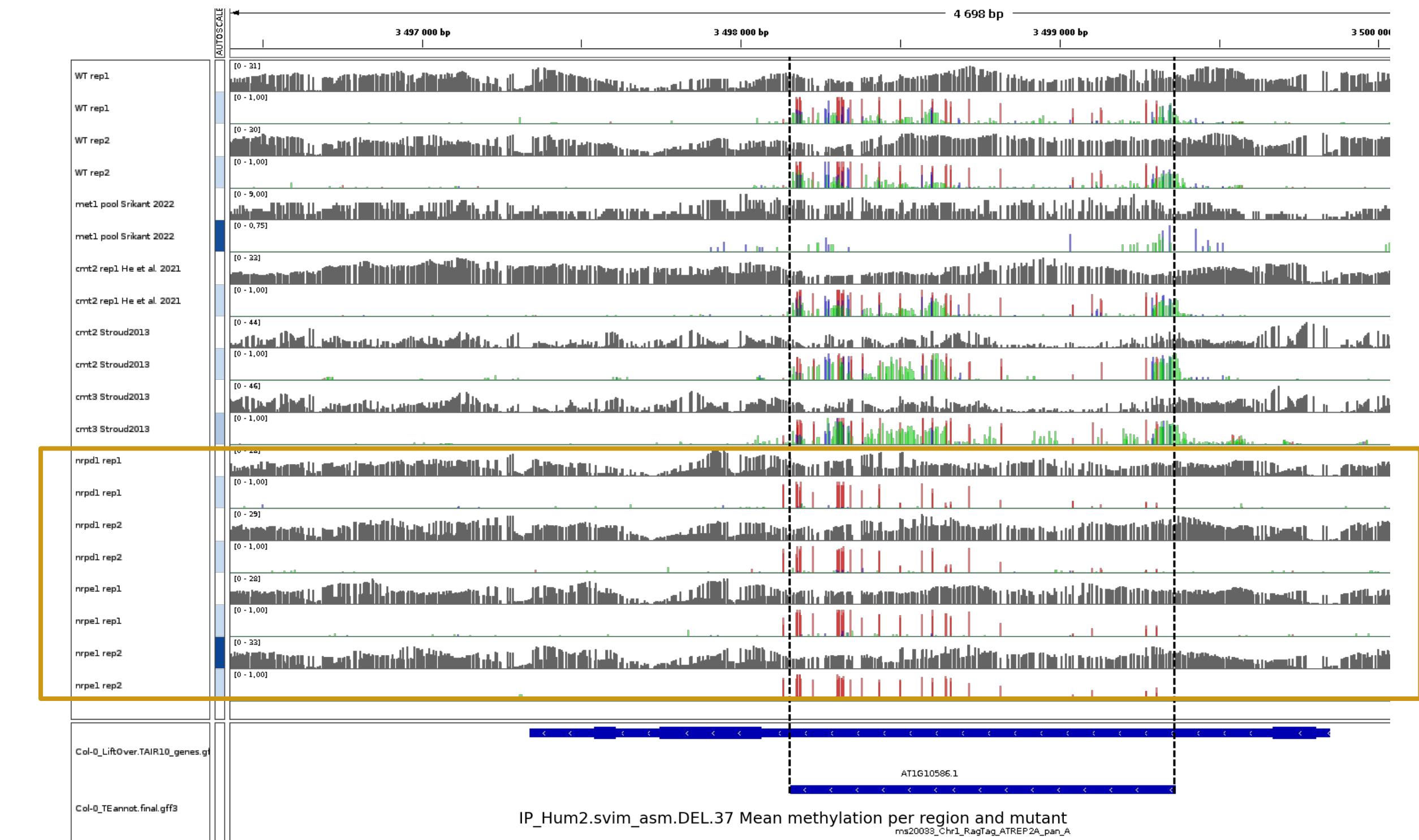
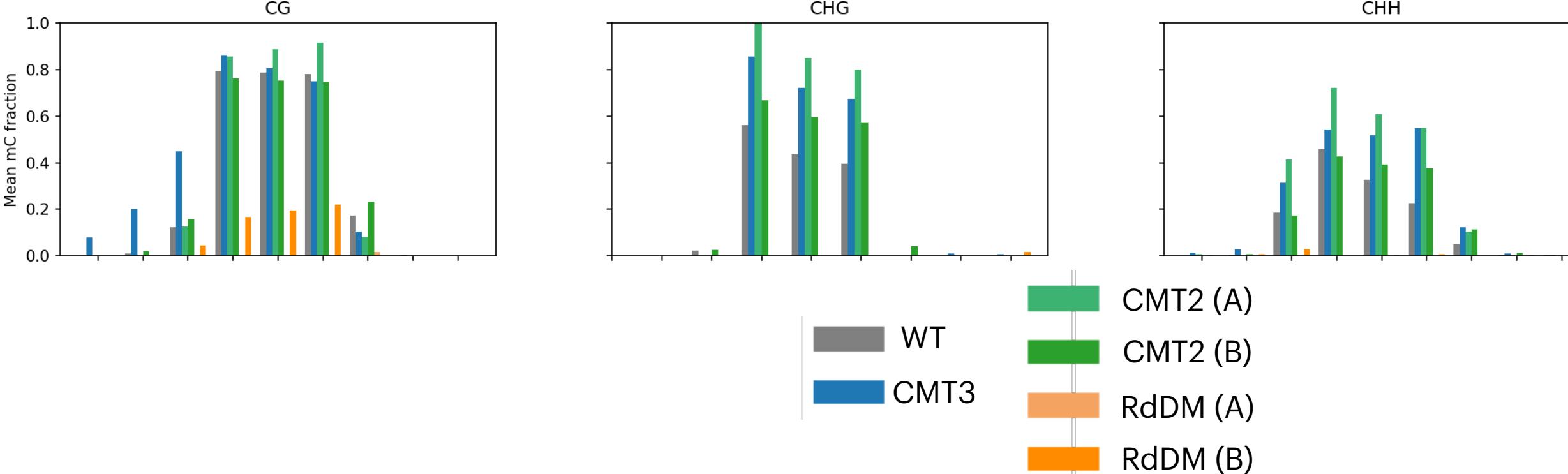
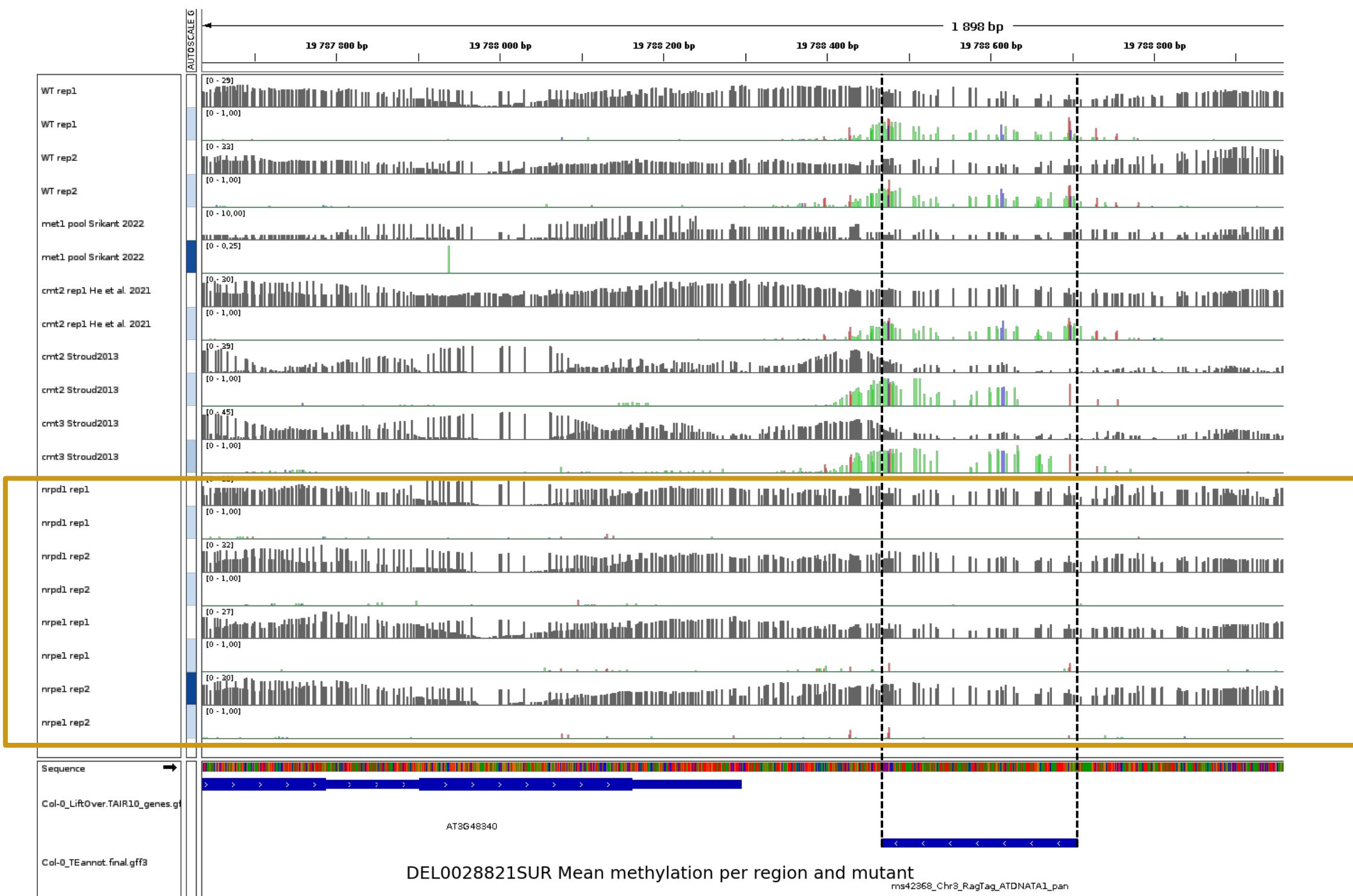
DE0028821SUR

IP_Hum2.svim_asm.DEL.37



Biological confirmation

DE0028821SUR



CHG and CHH methylation (and spreading!) disappear in RdDM mutants

Conclusions

- An example of the workflow:

biological phenomenon \Rightarrow machine learning model \Rightarrow explanations \Rightarrow real biological mechanisms

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- Different **explainability tools** have been explored, and they provide **consistent conclusions**
- Potential actor (**RdDM**) is identified
- We understand better one of the factors to **explain GWAS signals**

Acknowledgements

CBIO

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(and everyone else)



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