

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

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27 November 2025



CBIO



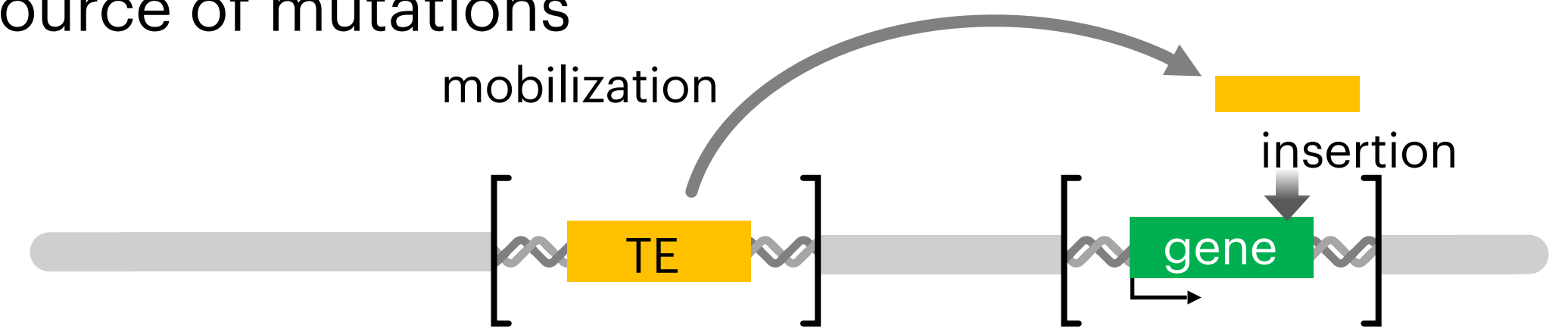
Contents

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

Transposable Elements

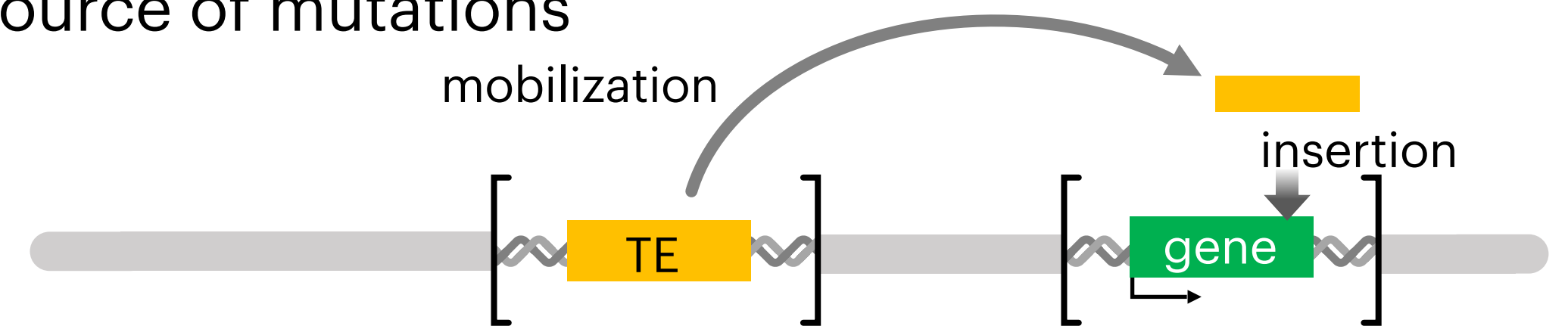
Transposable Elements

- 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

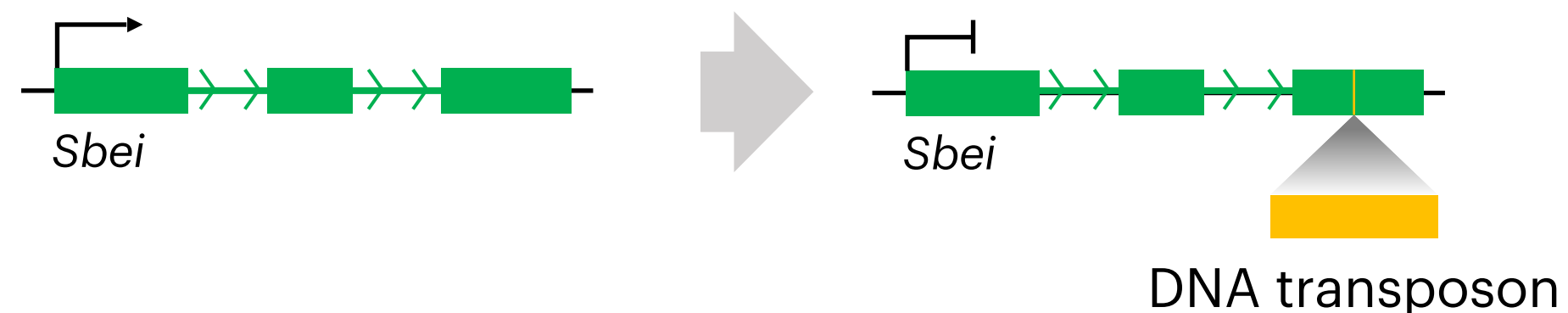
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Mutations may be **deleterious**...

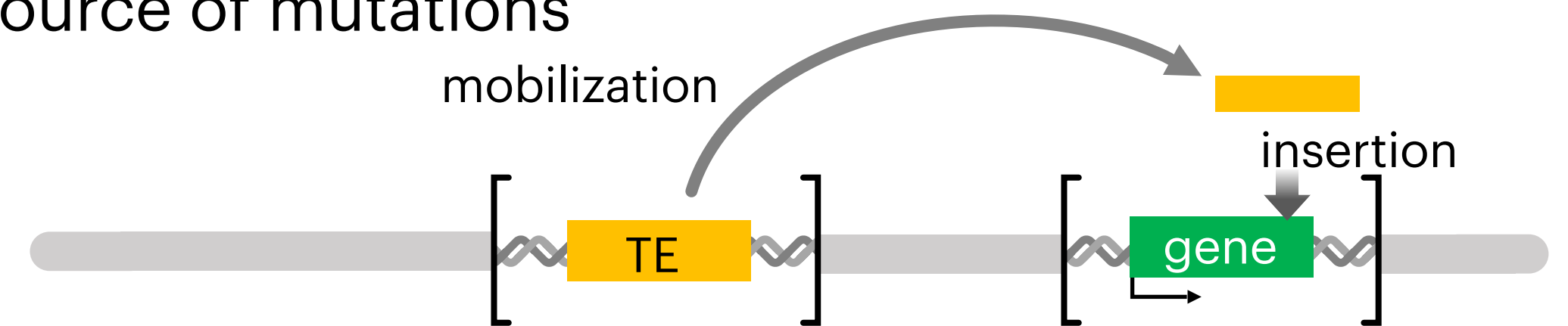


Bhattacharyya et al. Cell 1990



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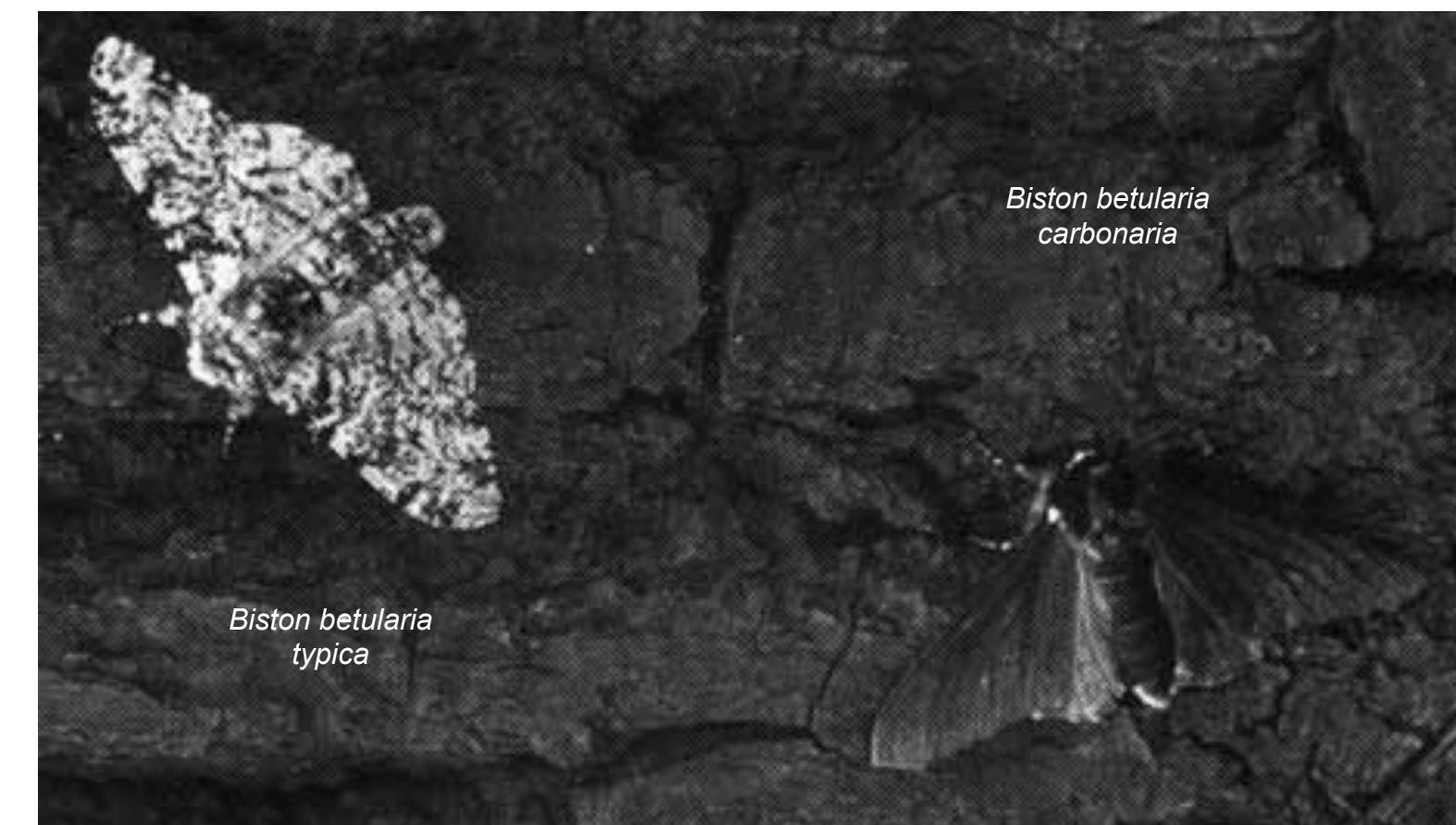
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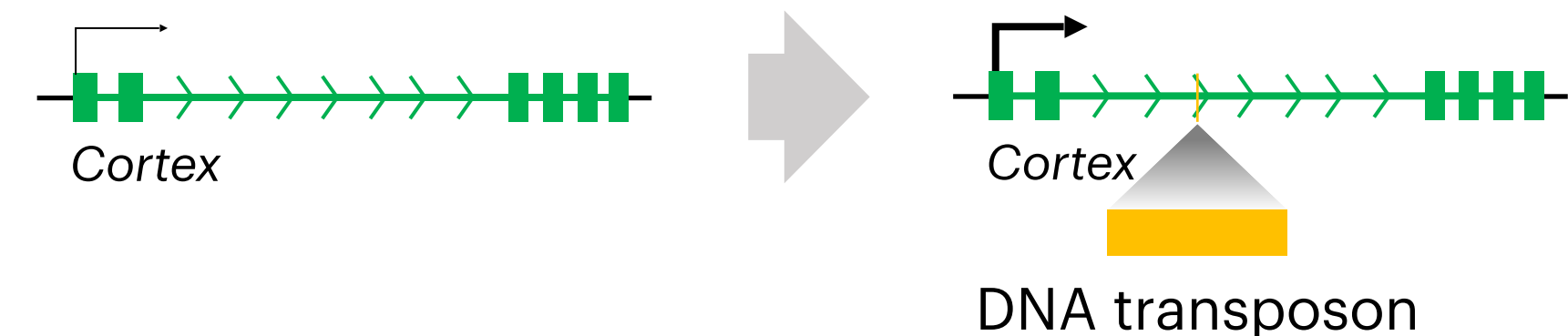
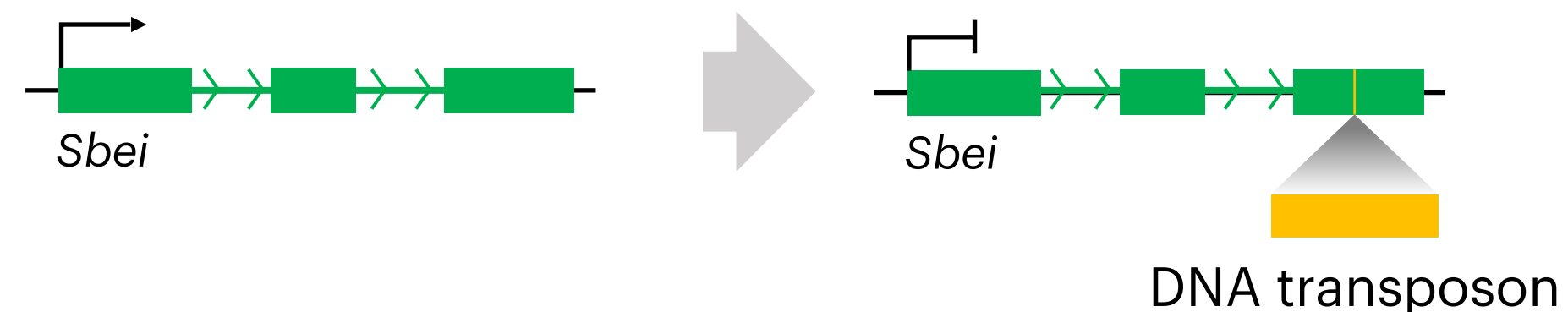
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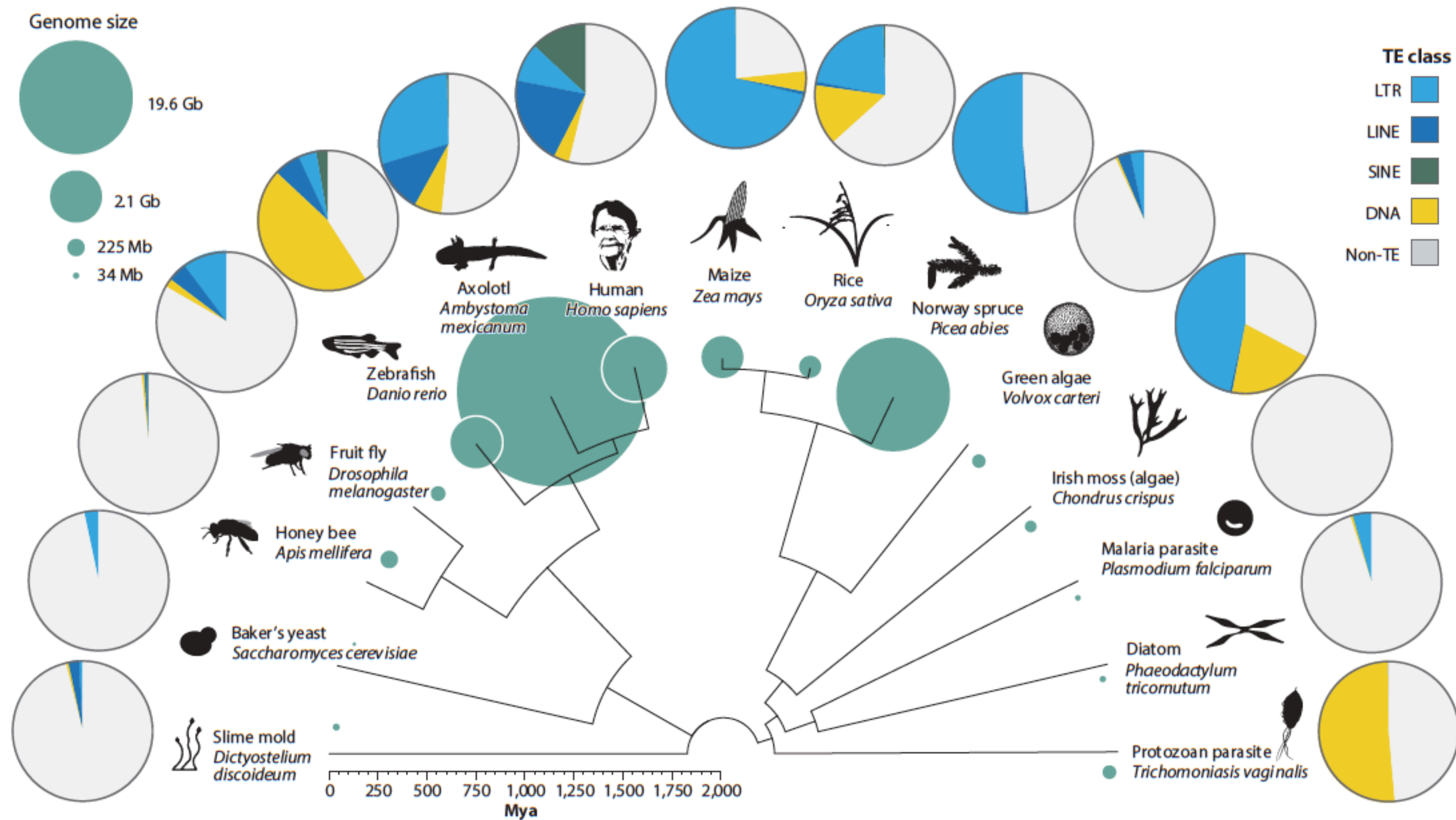
...yet sometimes **adaptive**



Kettlewell. *Heredity* 1956; van't Hof et al. *Nature* 2016



Transposable Elements

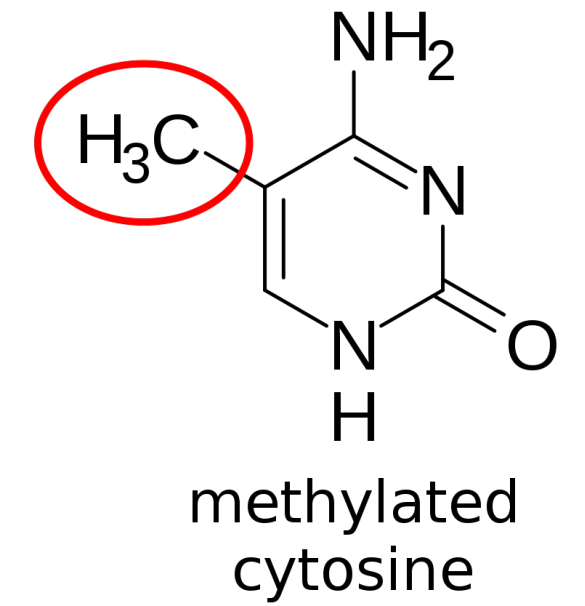
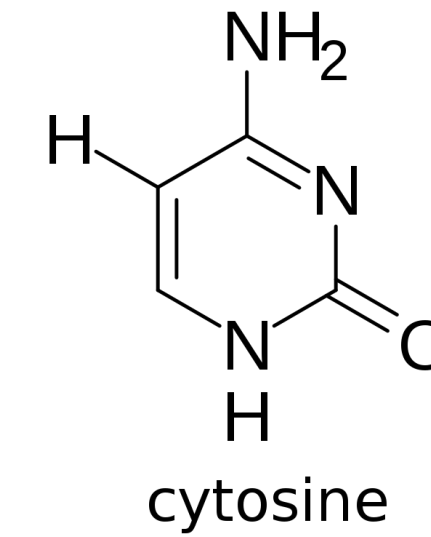
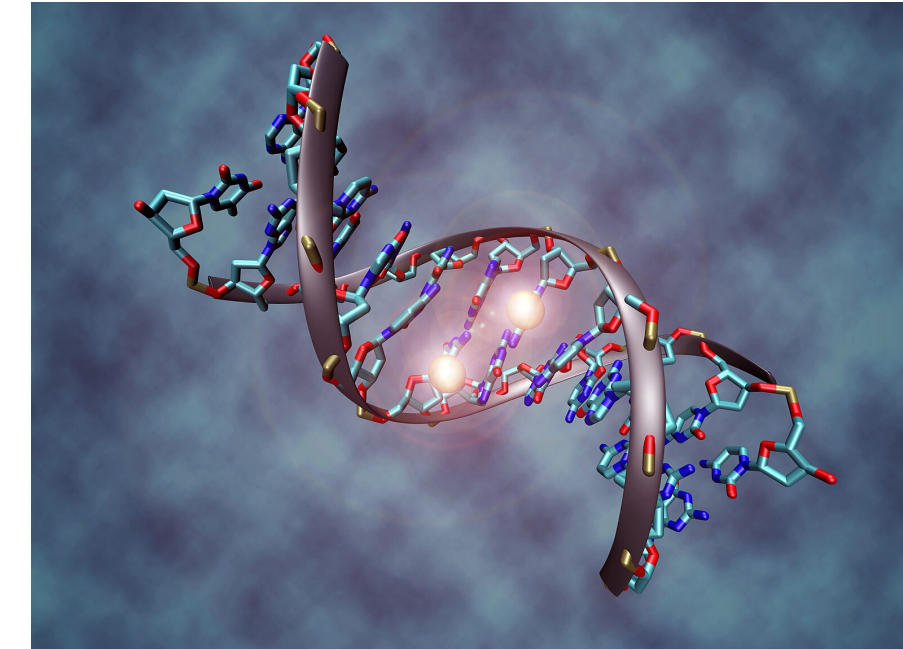


Epigenetic Regulation of Transposable Elements

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:

methyated promoter \implies no RNA \implies
 \implies no protein \implies no function

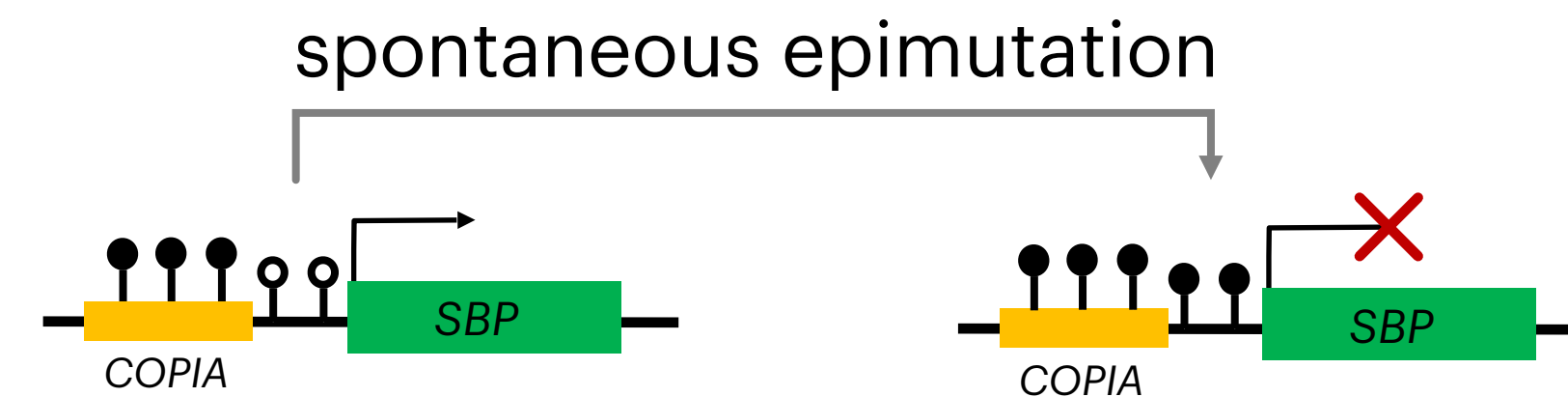
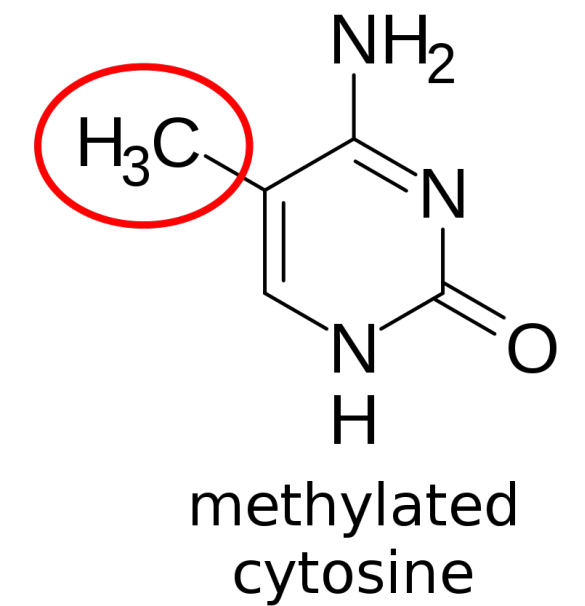
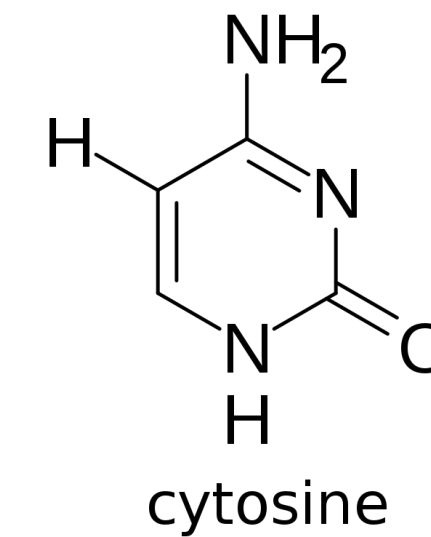
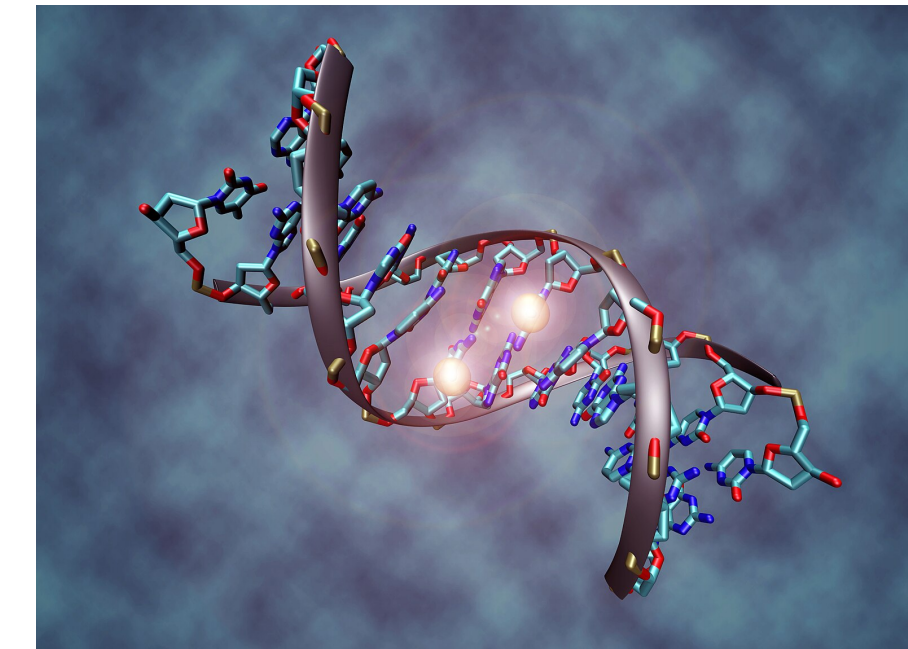


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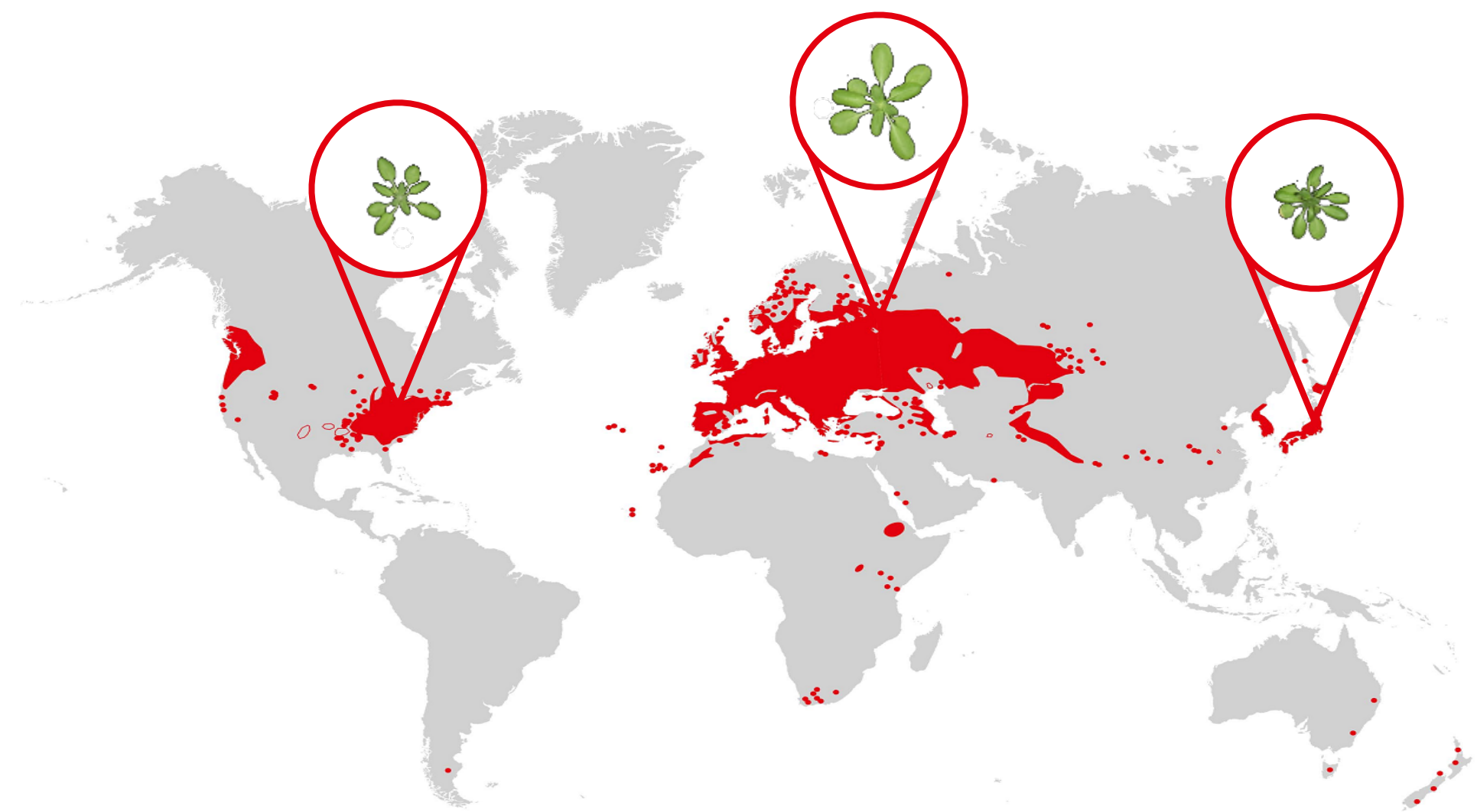
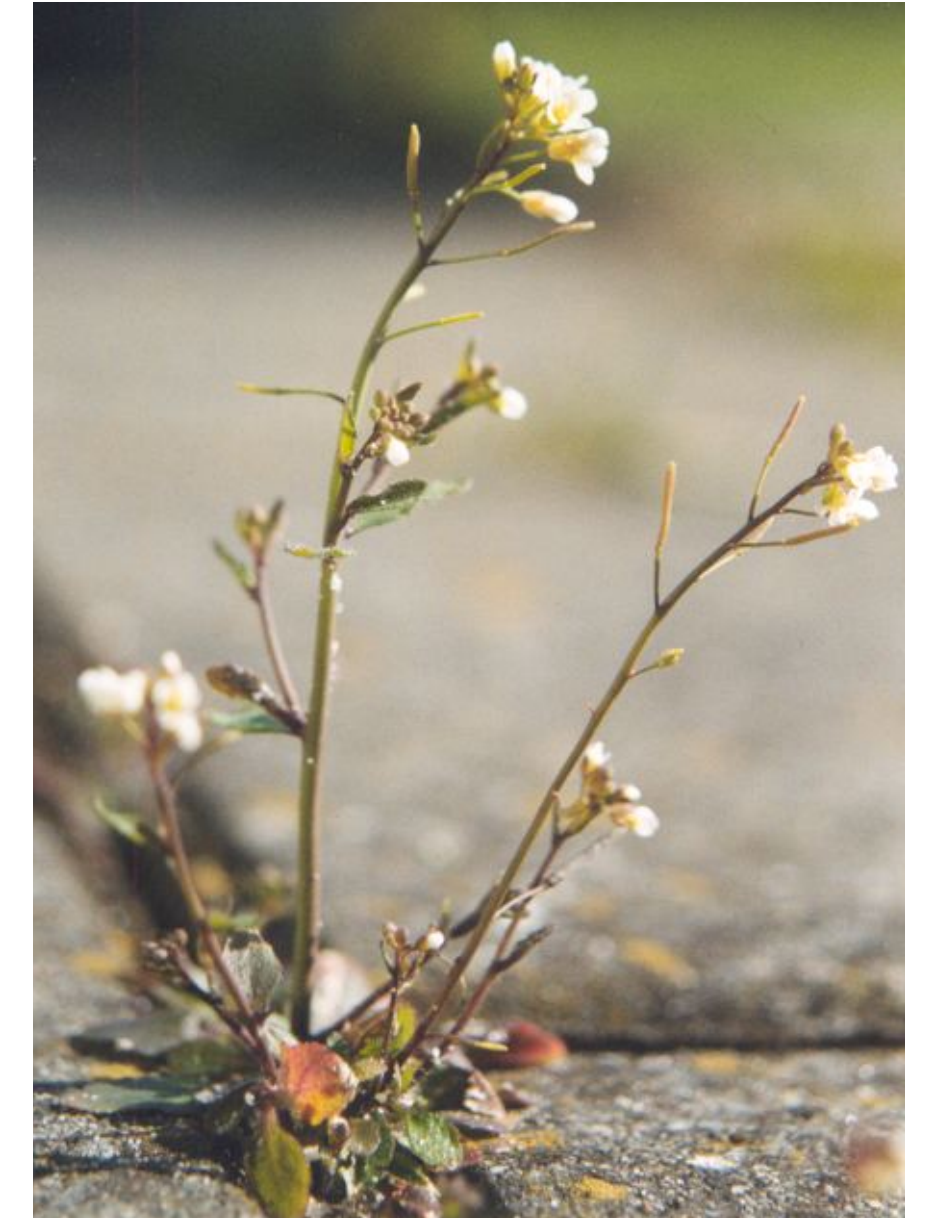


Manning et al., *Nat Genet* 2006

\implies perfect Mendelian segregation though no DNA changes observed

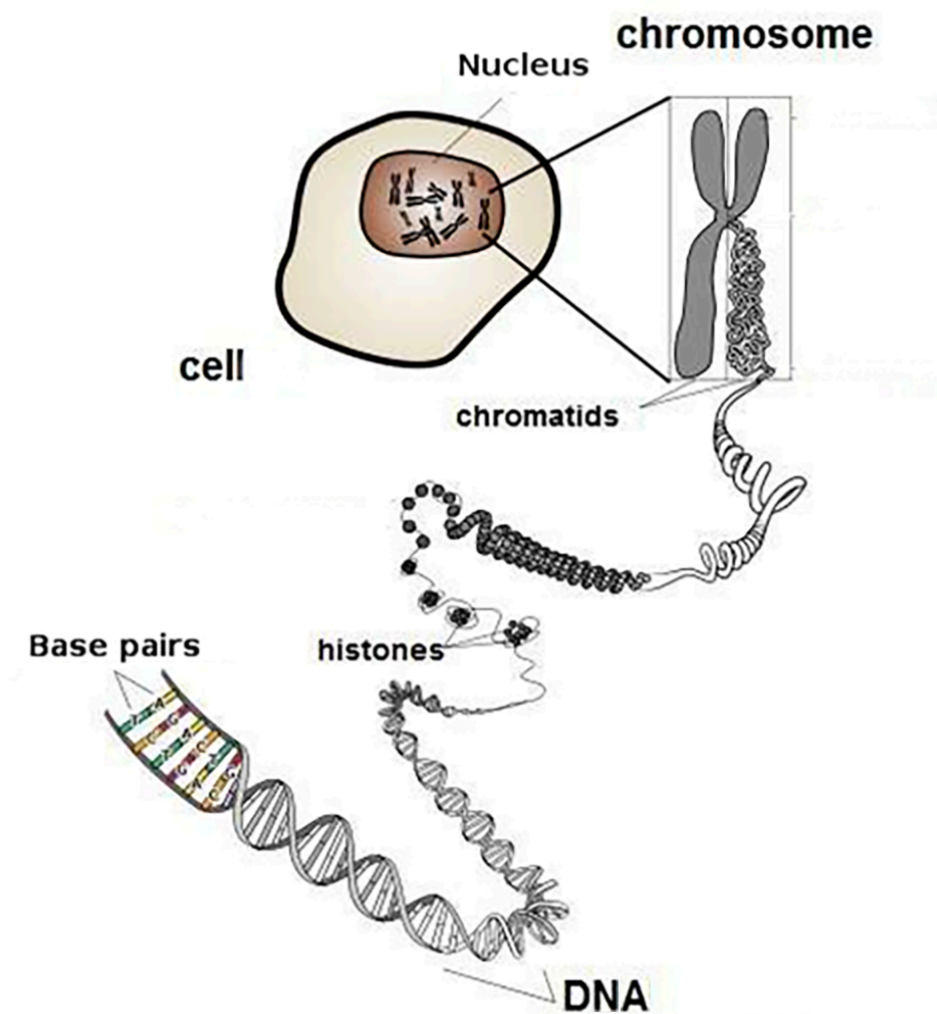
Our data: *Arabidopsis Thaliana*

- 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

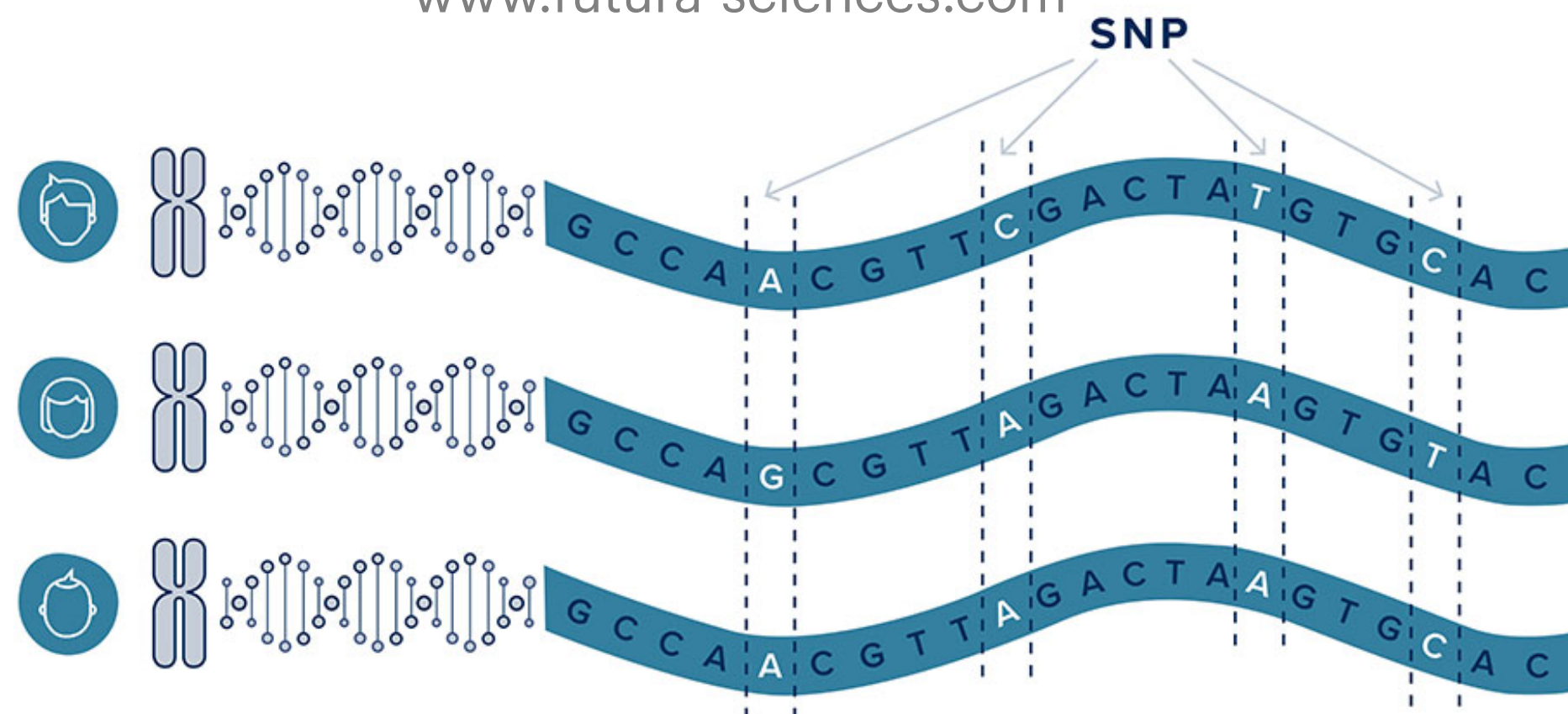


Kwakatsu et al. *Cell* 2016, Alonso-Blanco et al. *Cell* 2016, Quadrana et al. *eLife* 2016

From genotype to phenotype

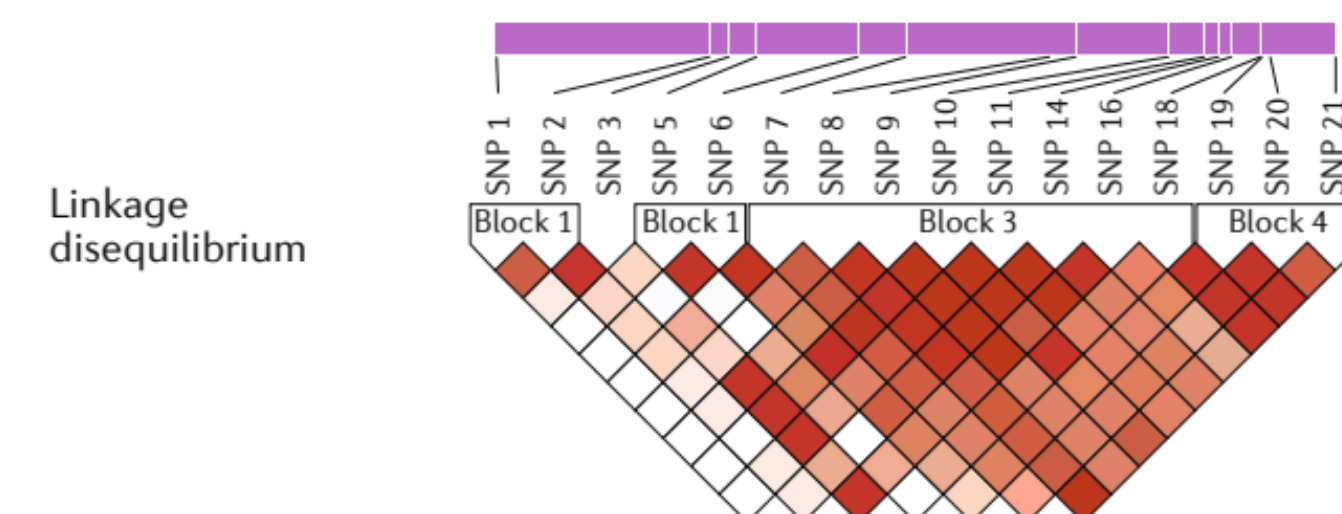
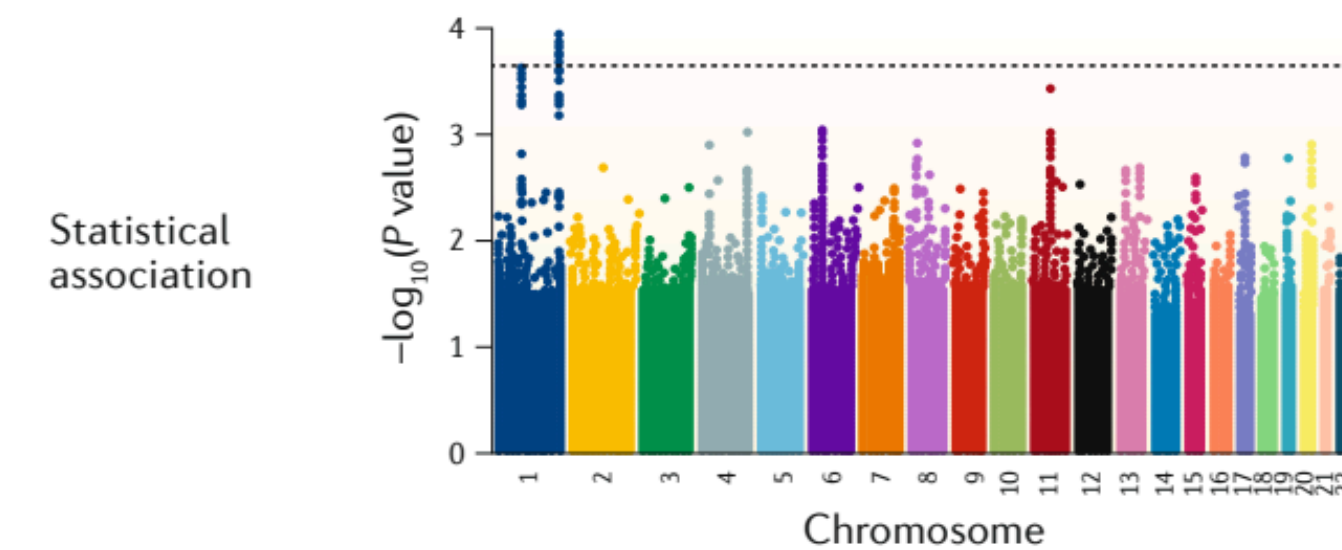
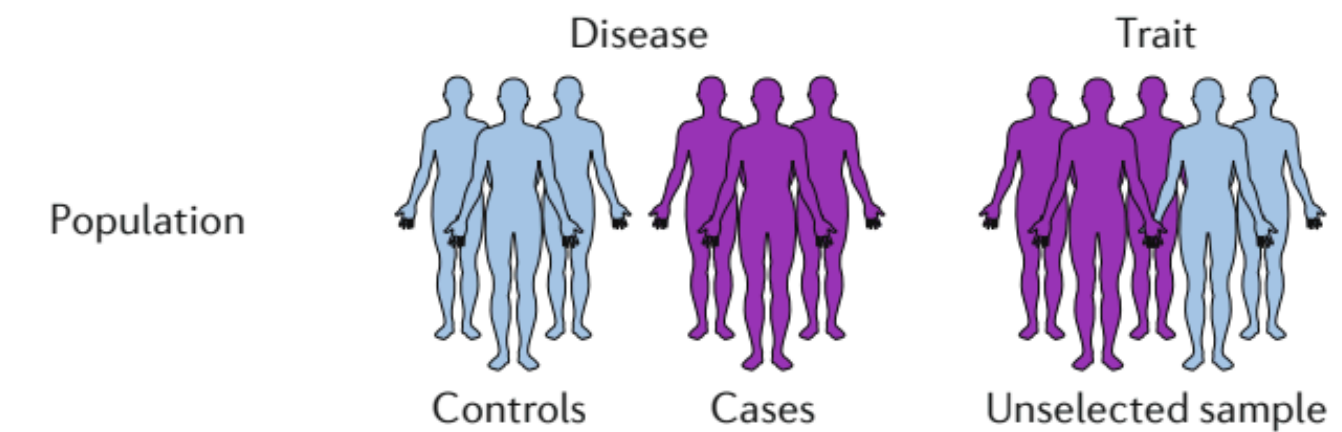


www.futura-sciences.com

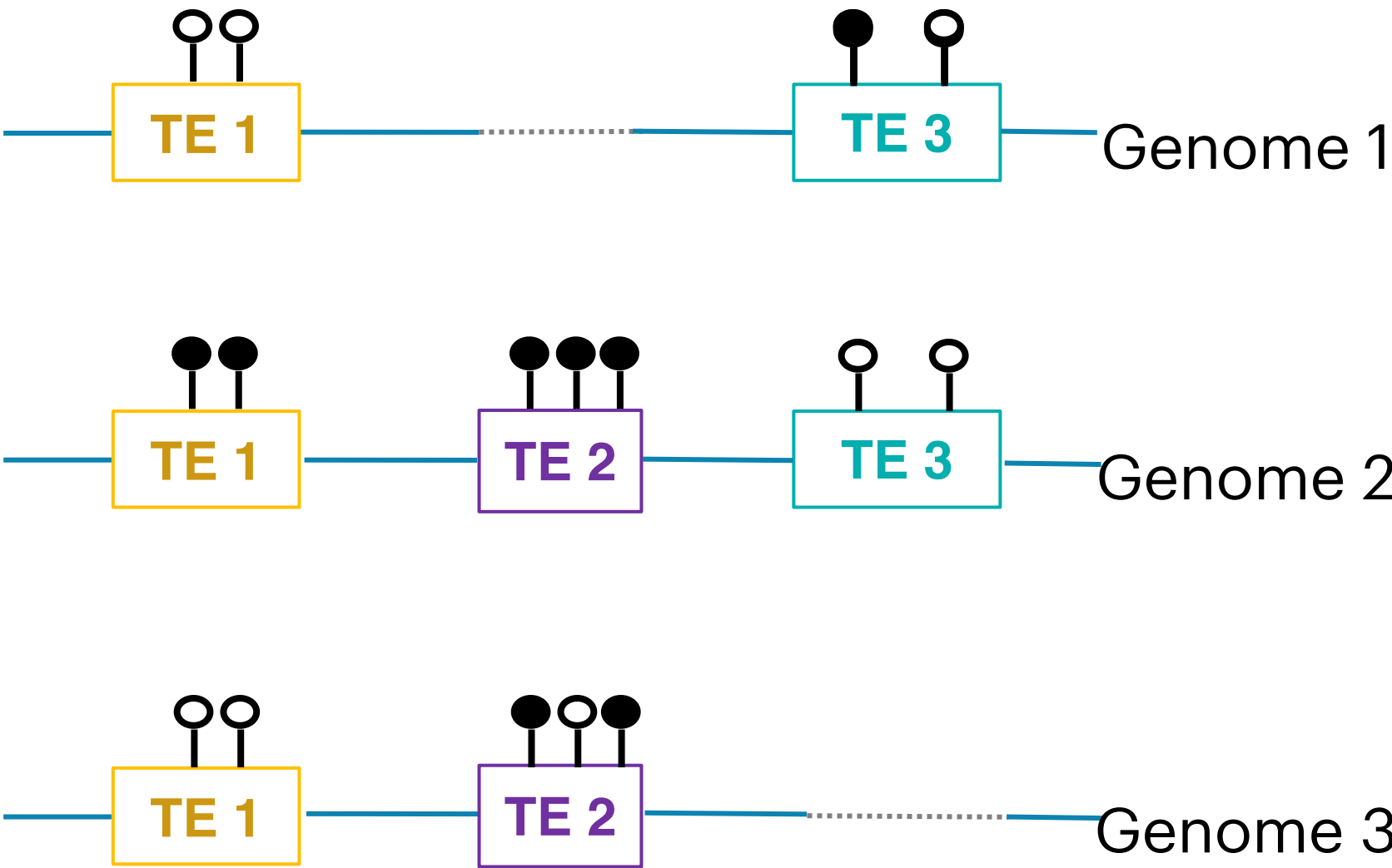


Scientific DX GmbH, 2020

Genome-Wide Association Study



From epi-genotype to phenotype

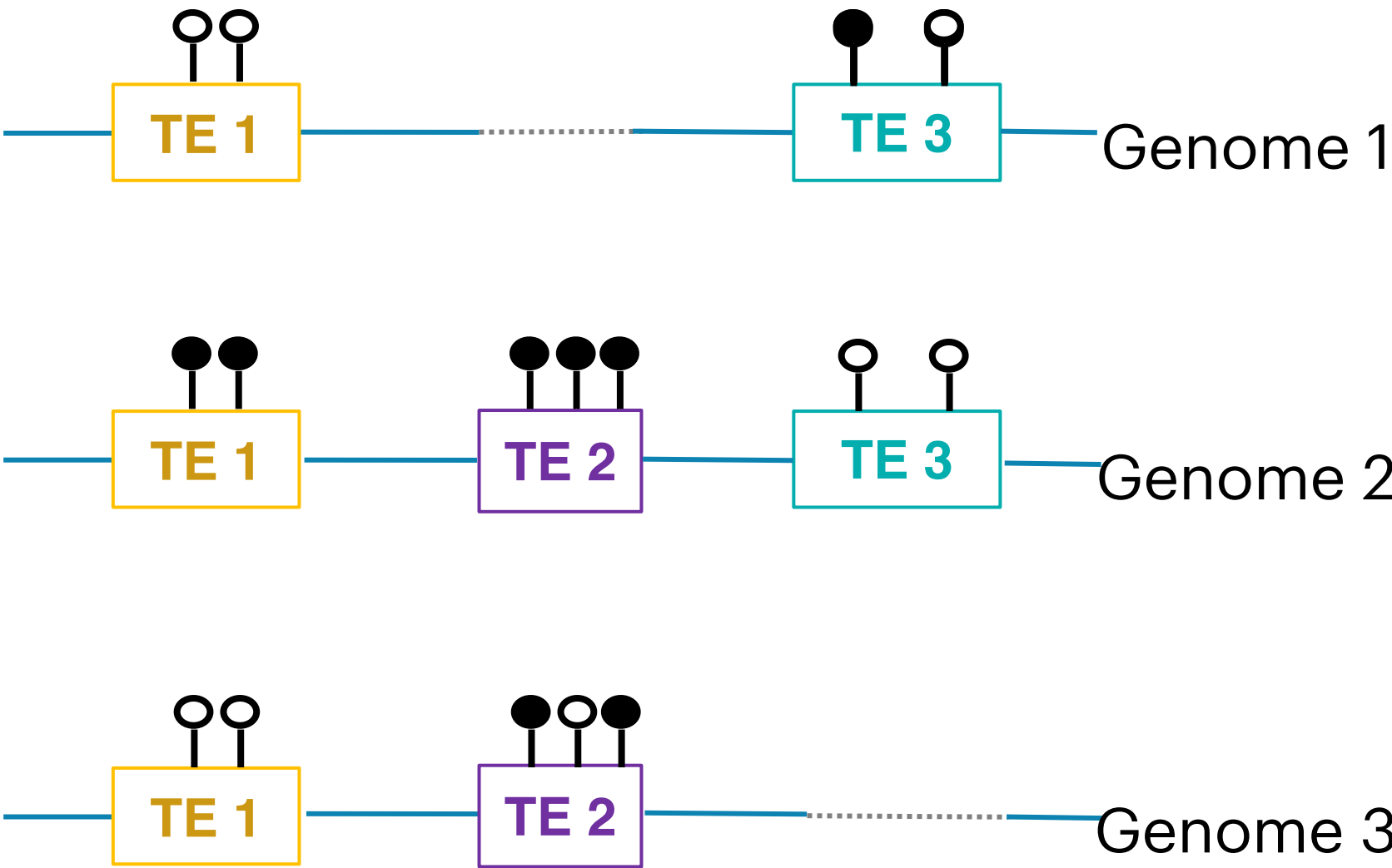


**Genome-Wide
Association Study**



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

From epi-genotype to phenotype



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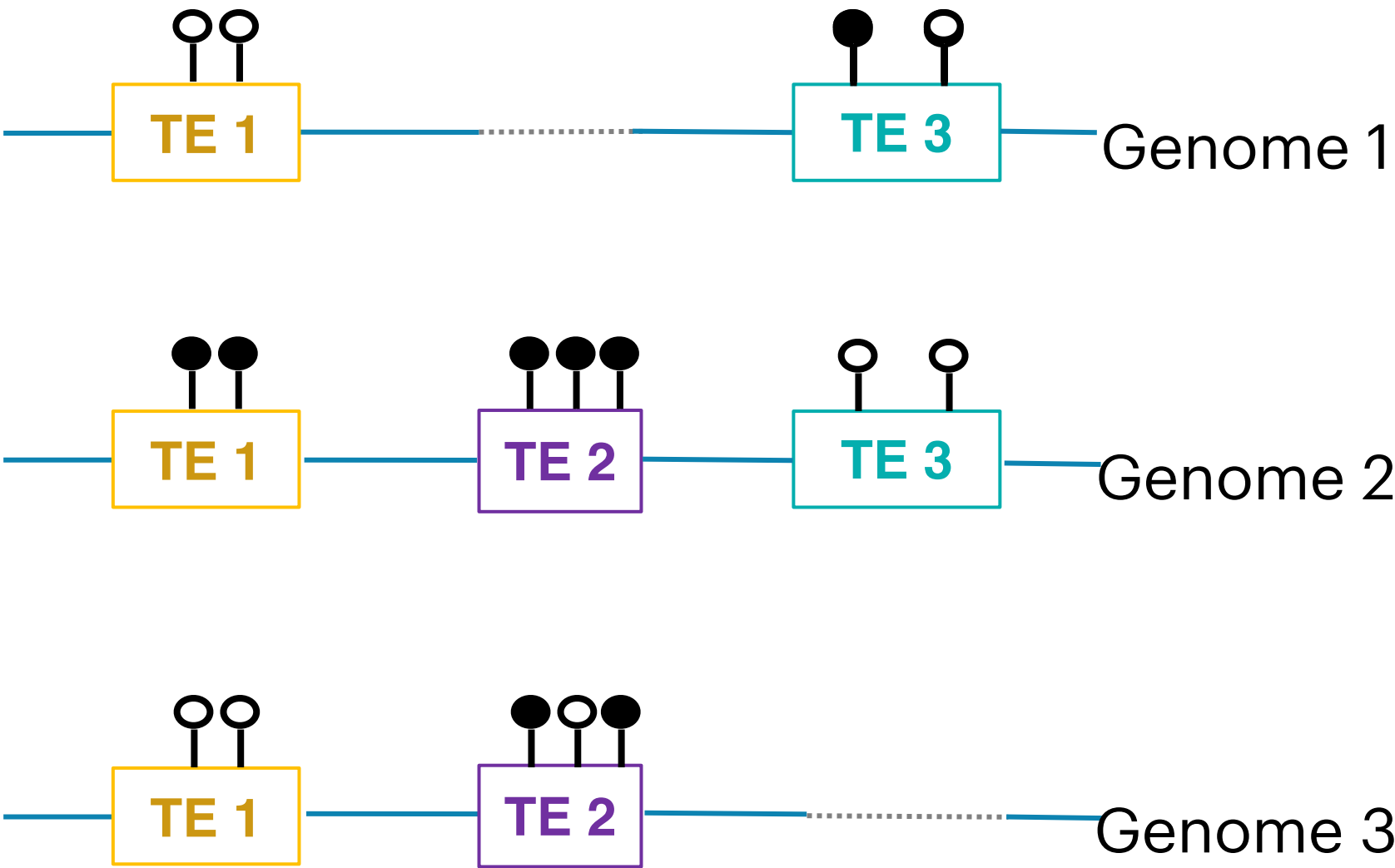


3 groups:

- 0 = absent
- 1 = present and not methylated (< 5%)
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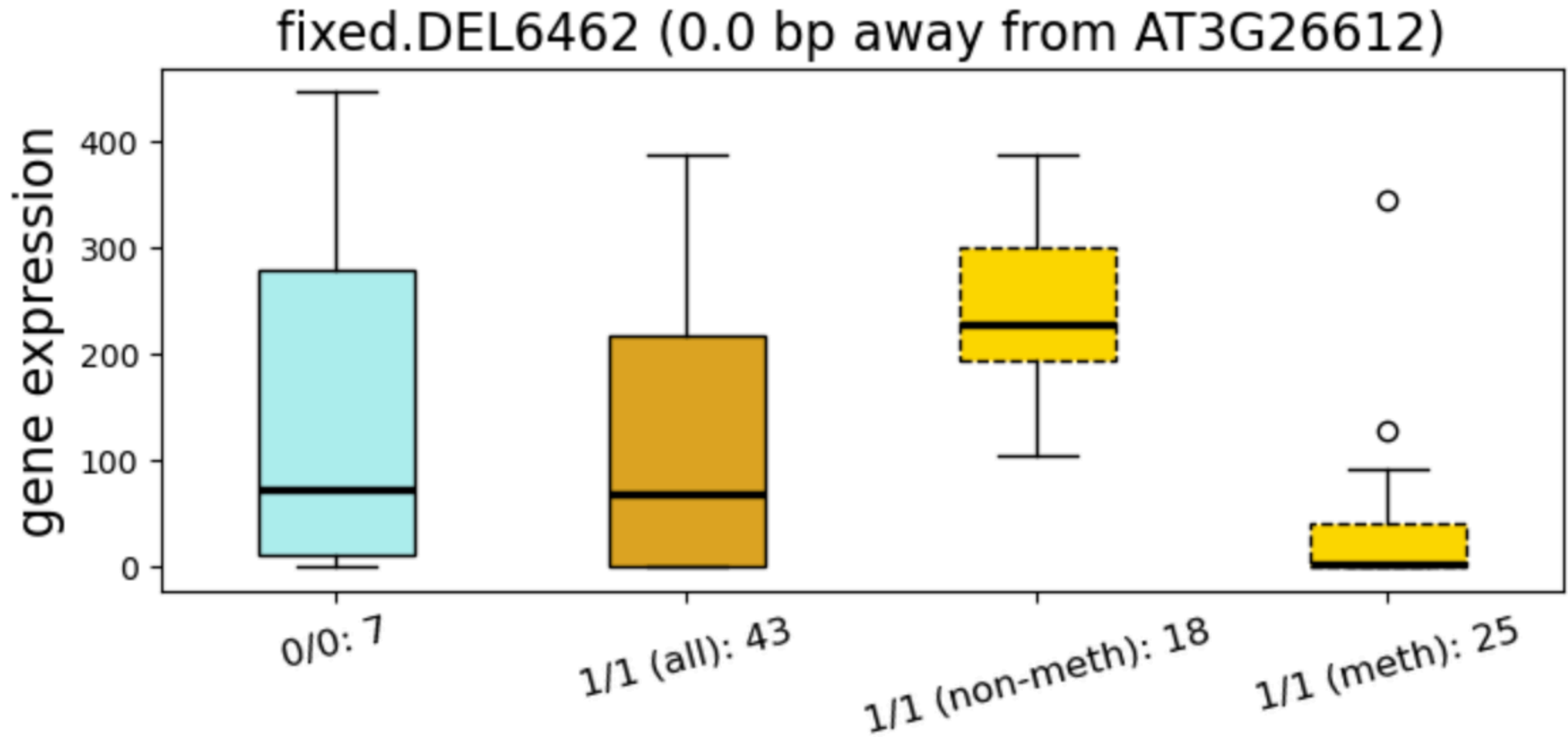
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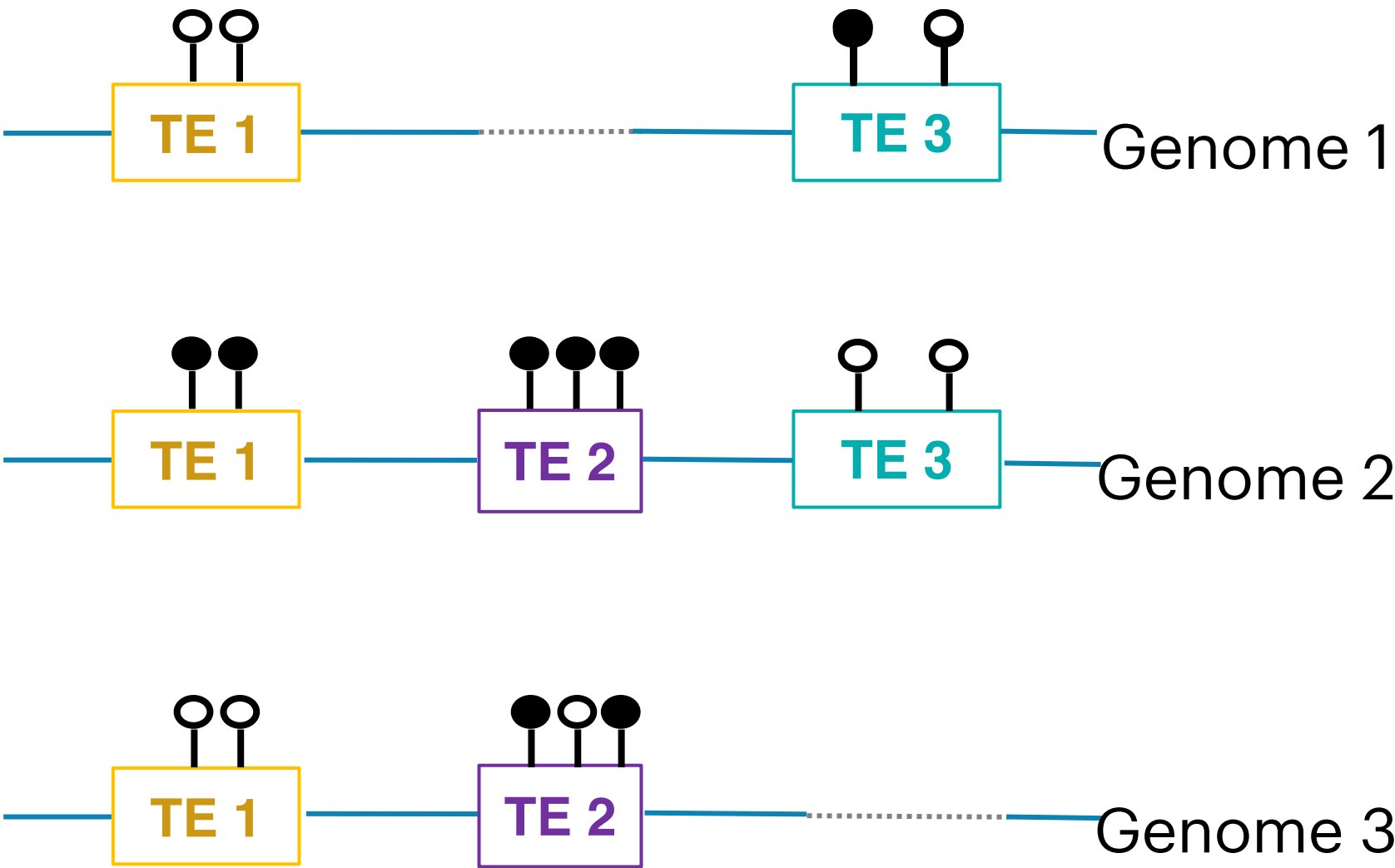
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For the moment, cis- effects only:

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



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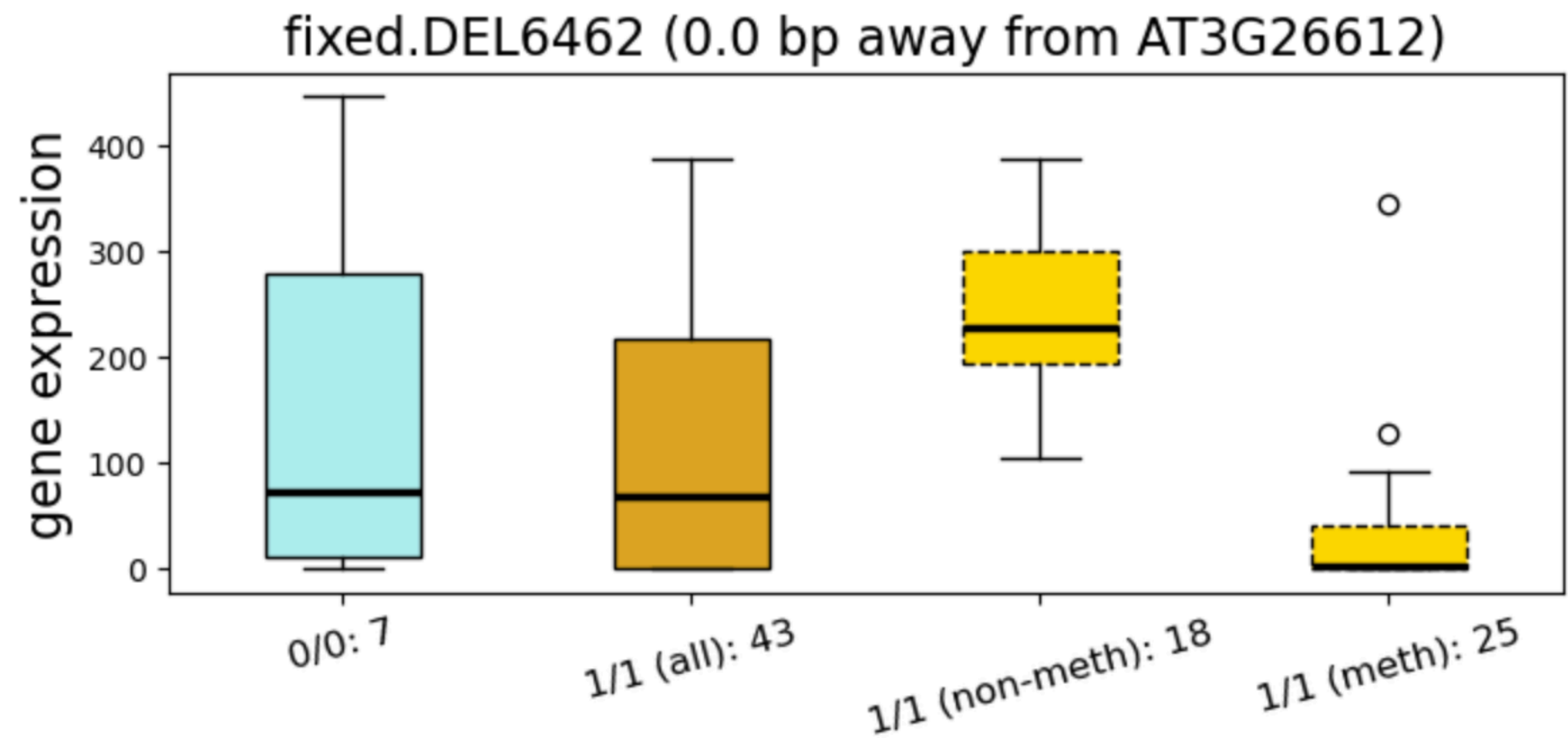
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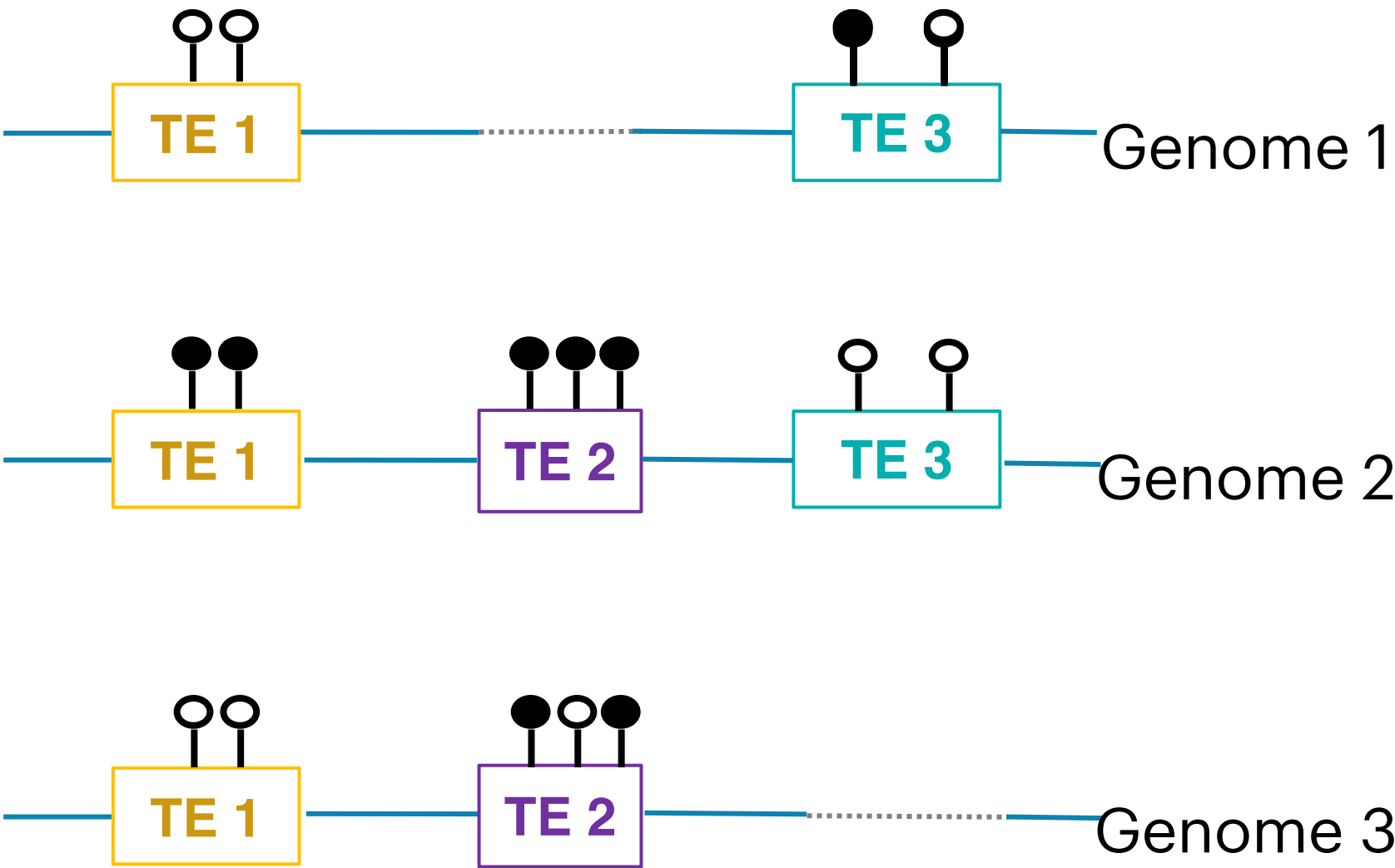
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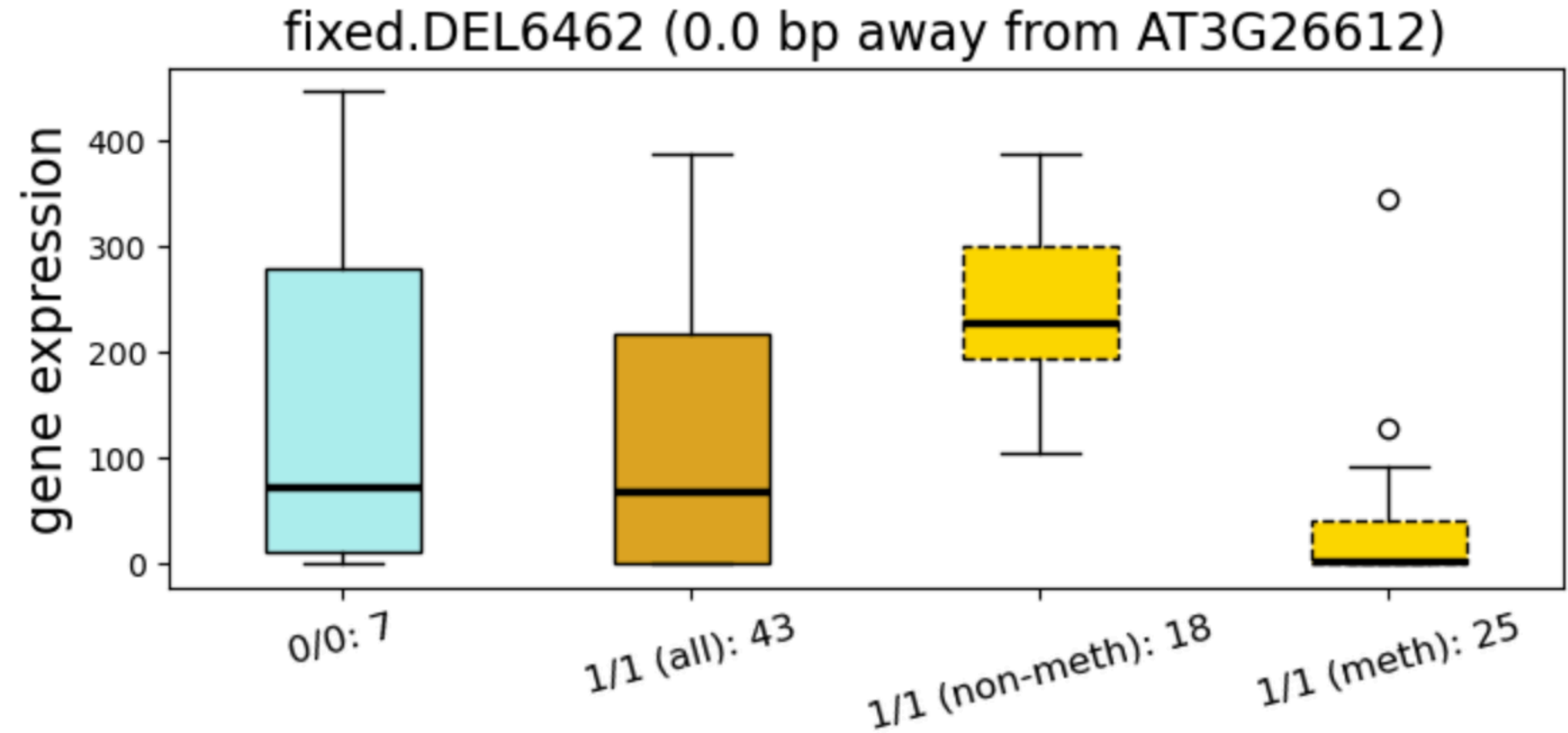
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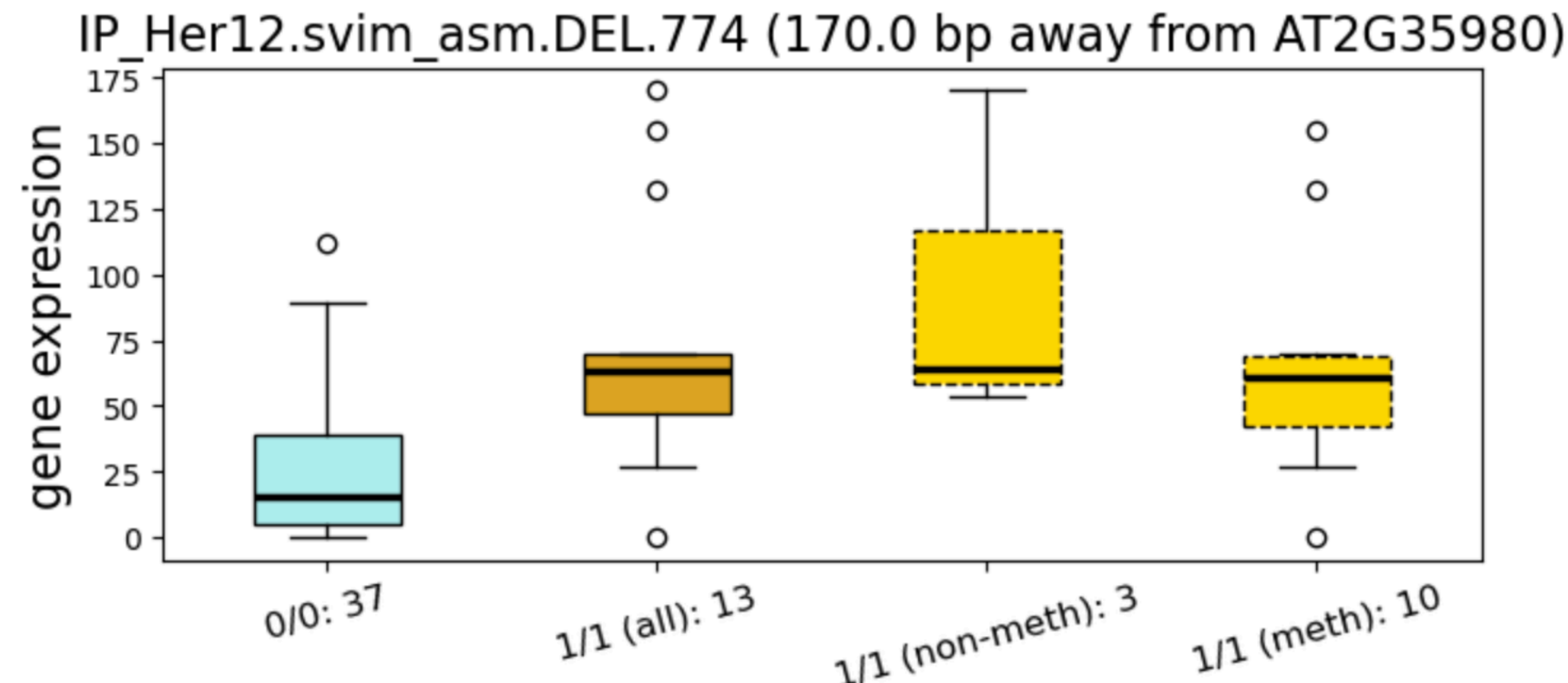
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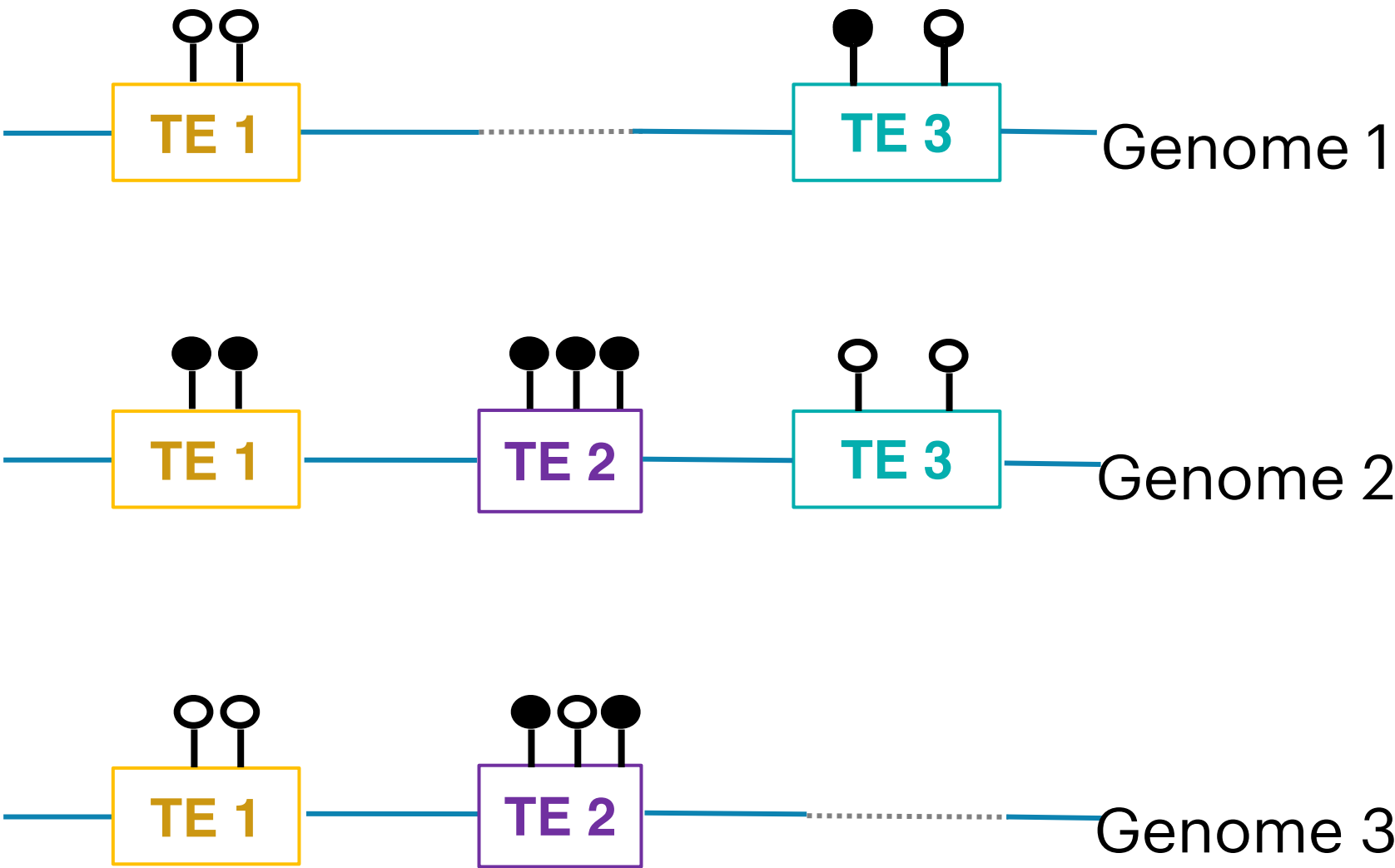
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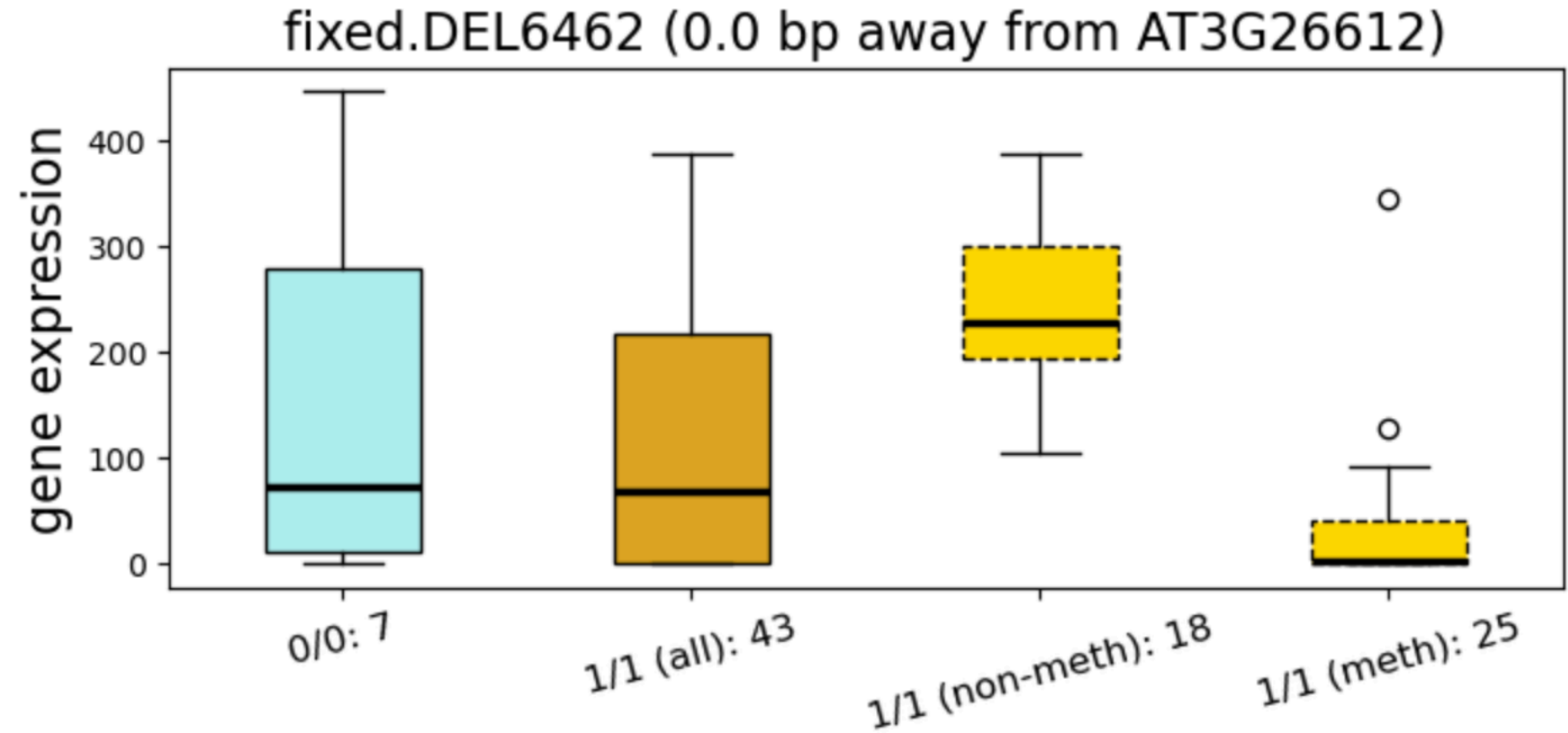
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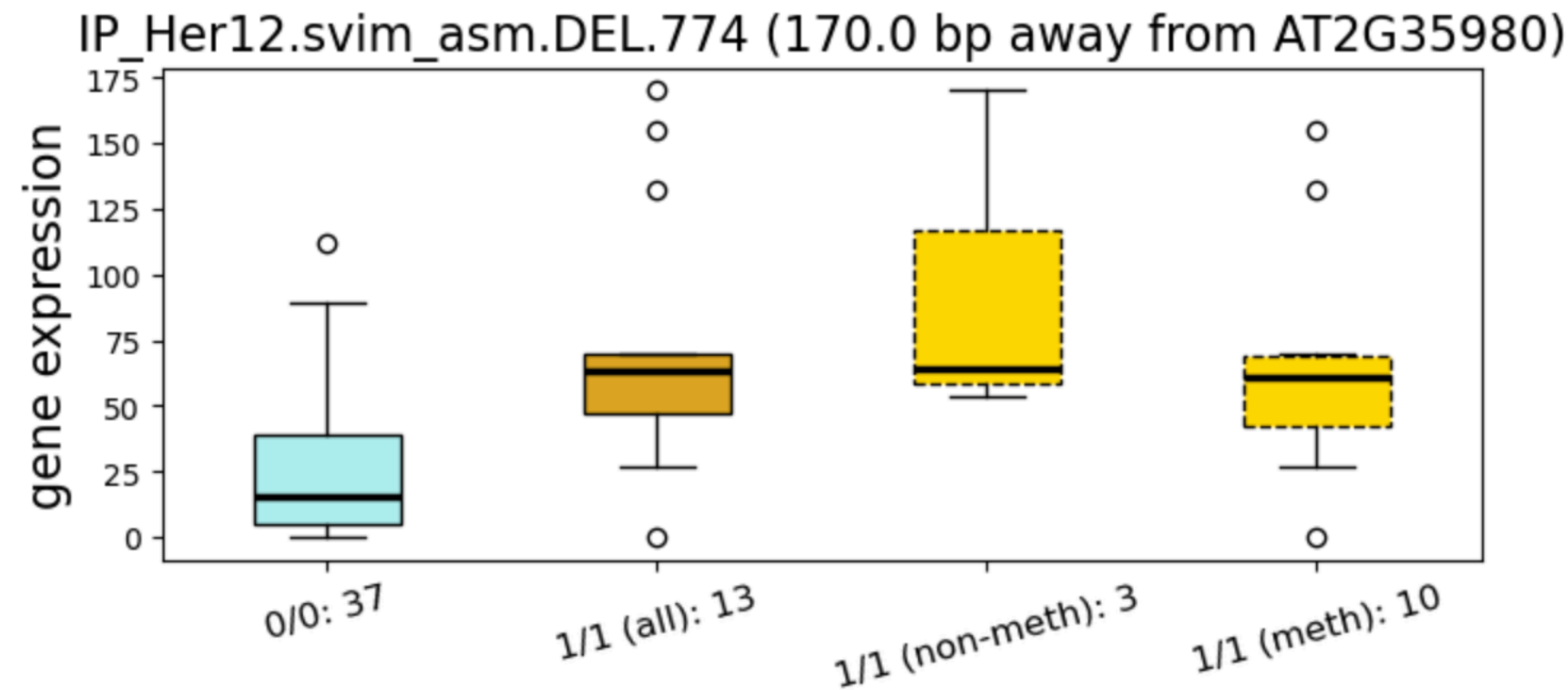
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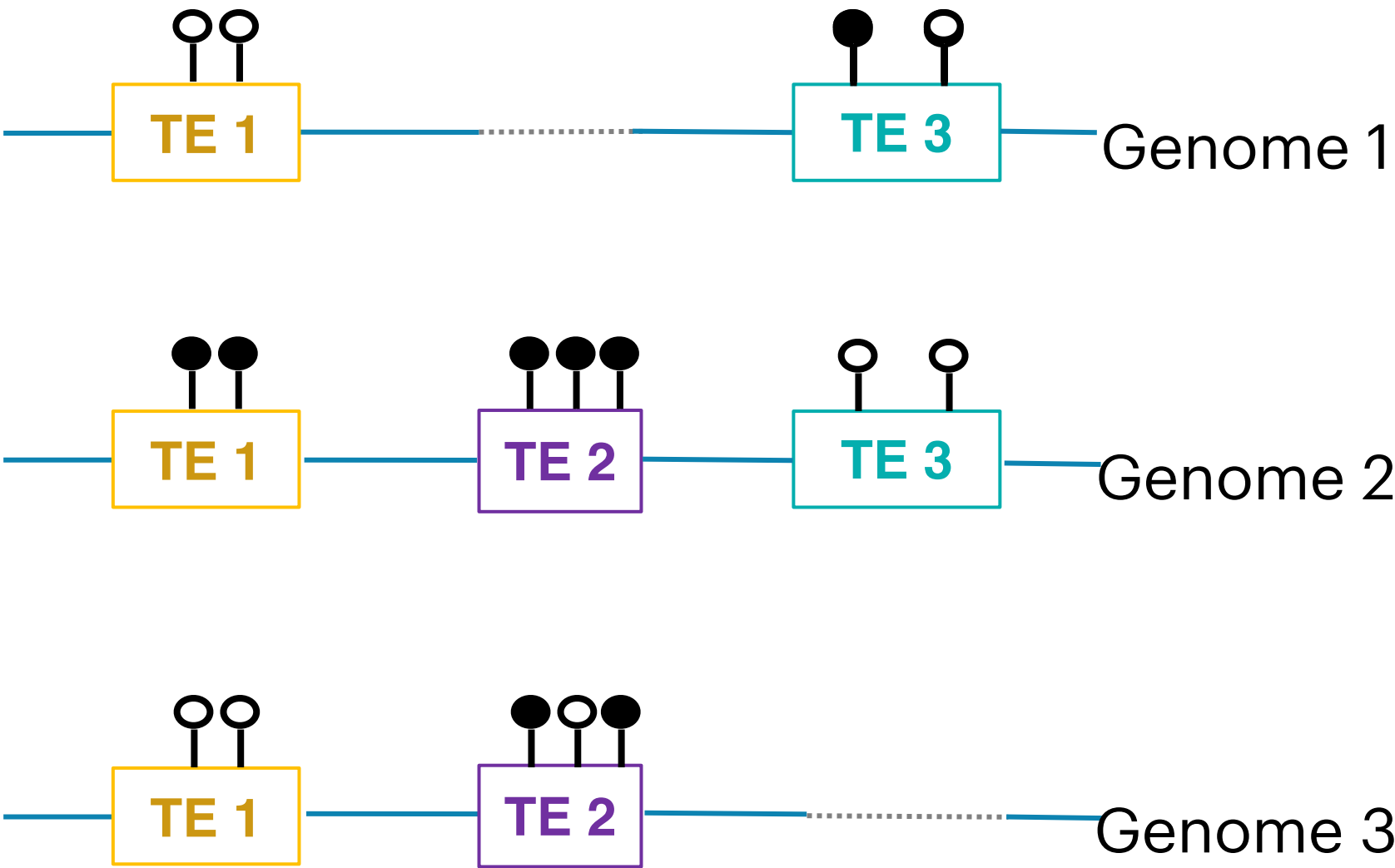
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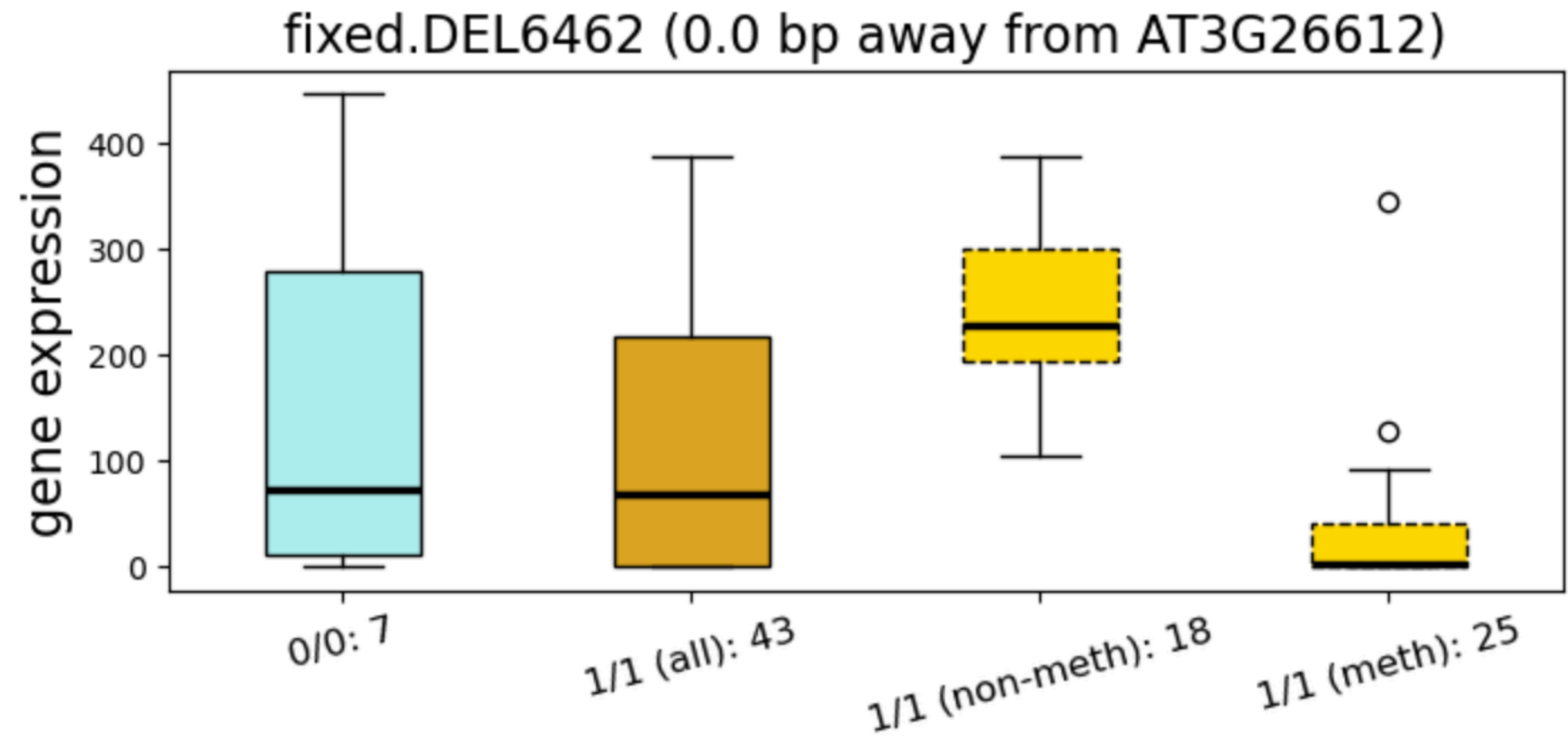
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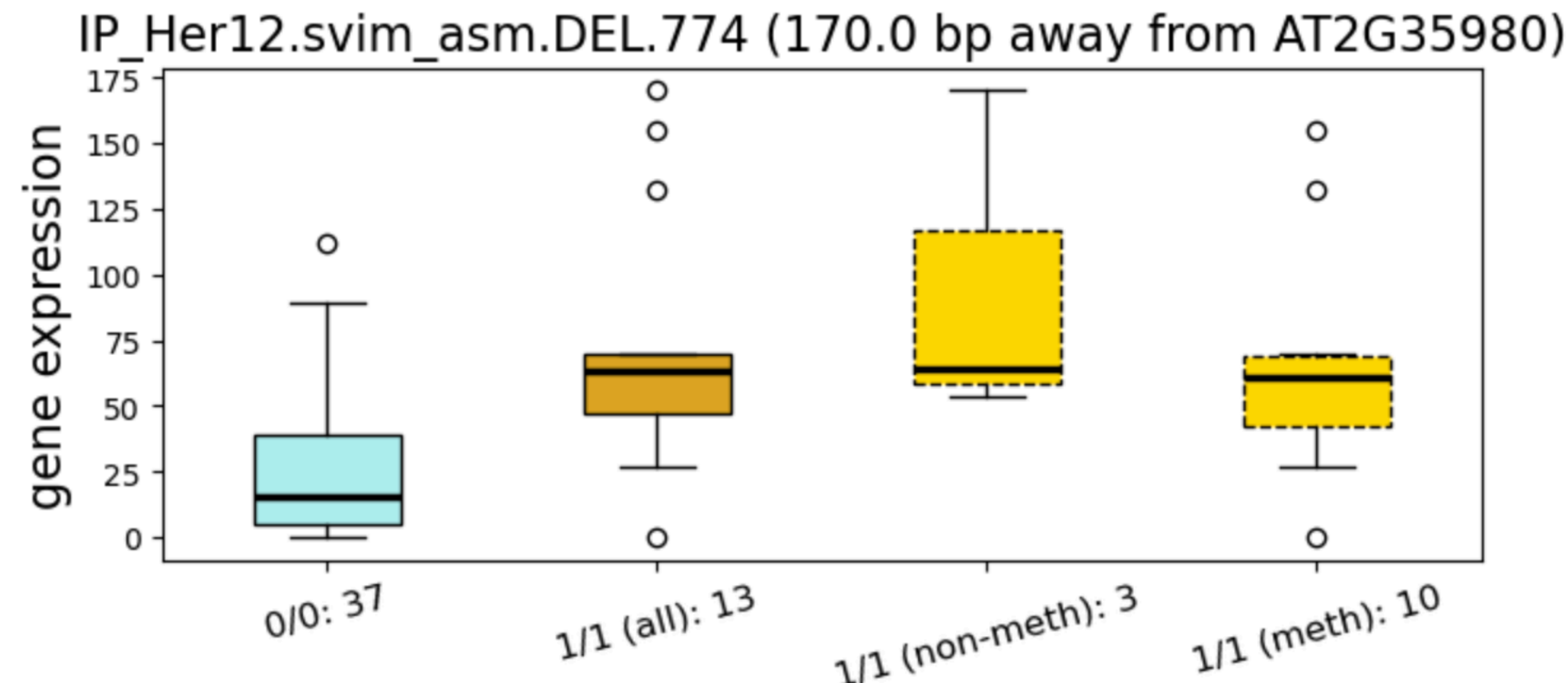
Spreading ?

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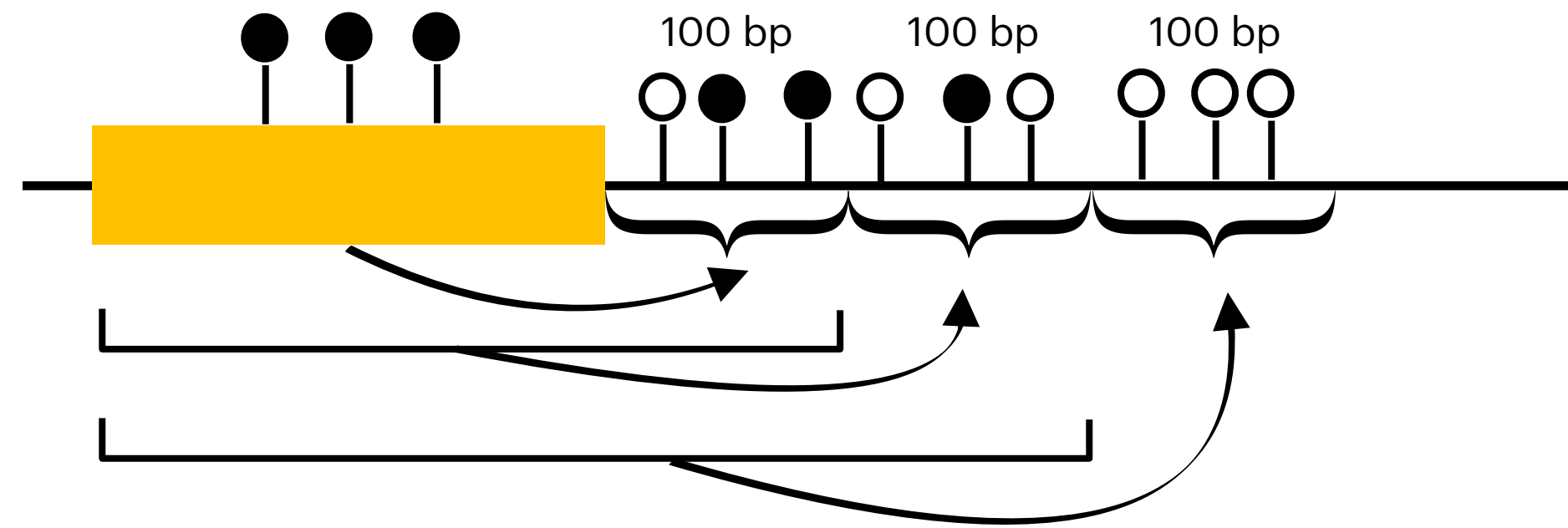
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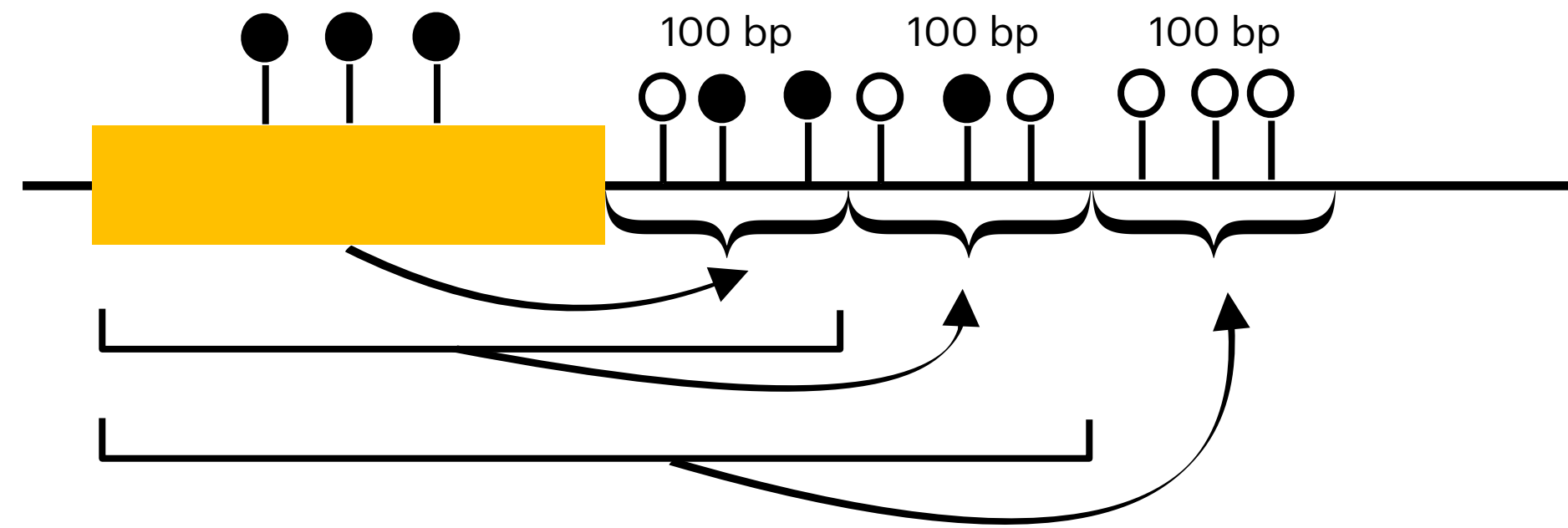


Prediction of methylation spreading



Model:
Random Forest

Prediction of methylation spreading



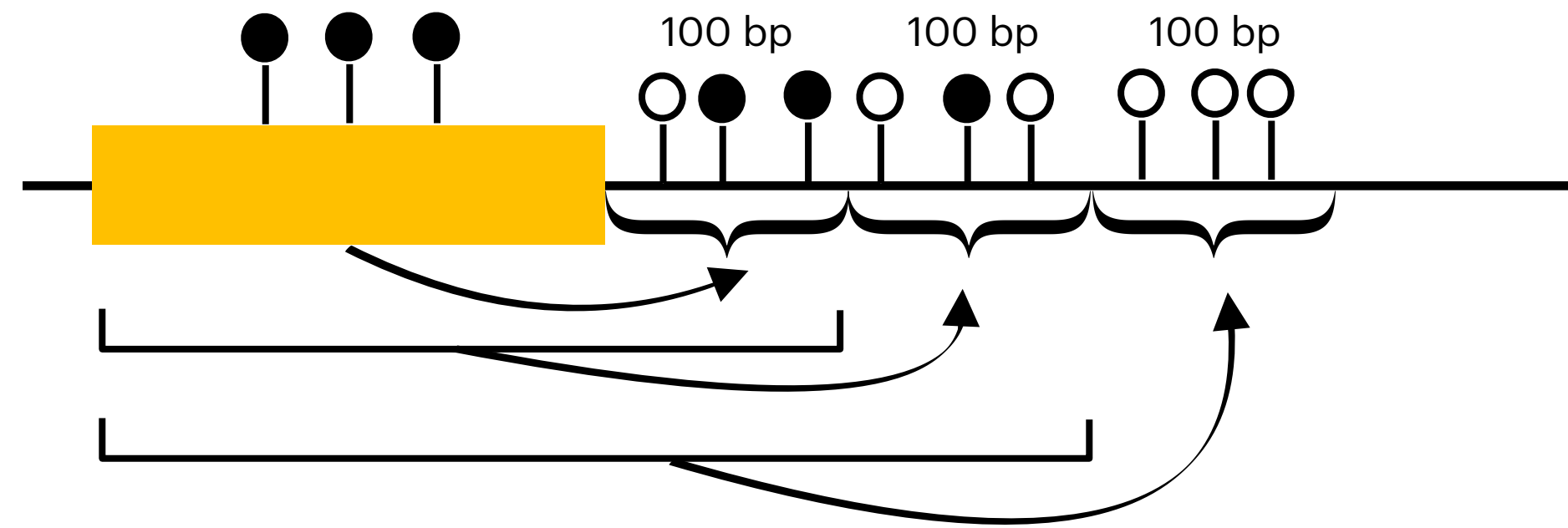
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Features:

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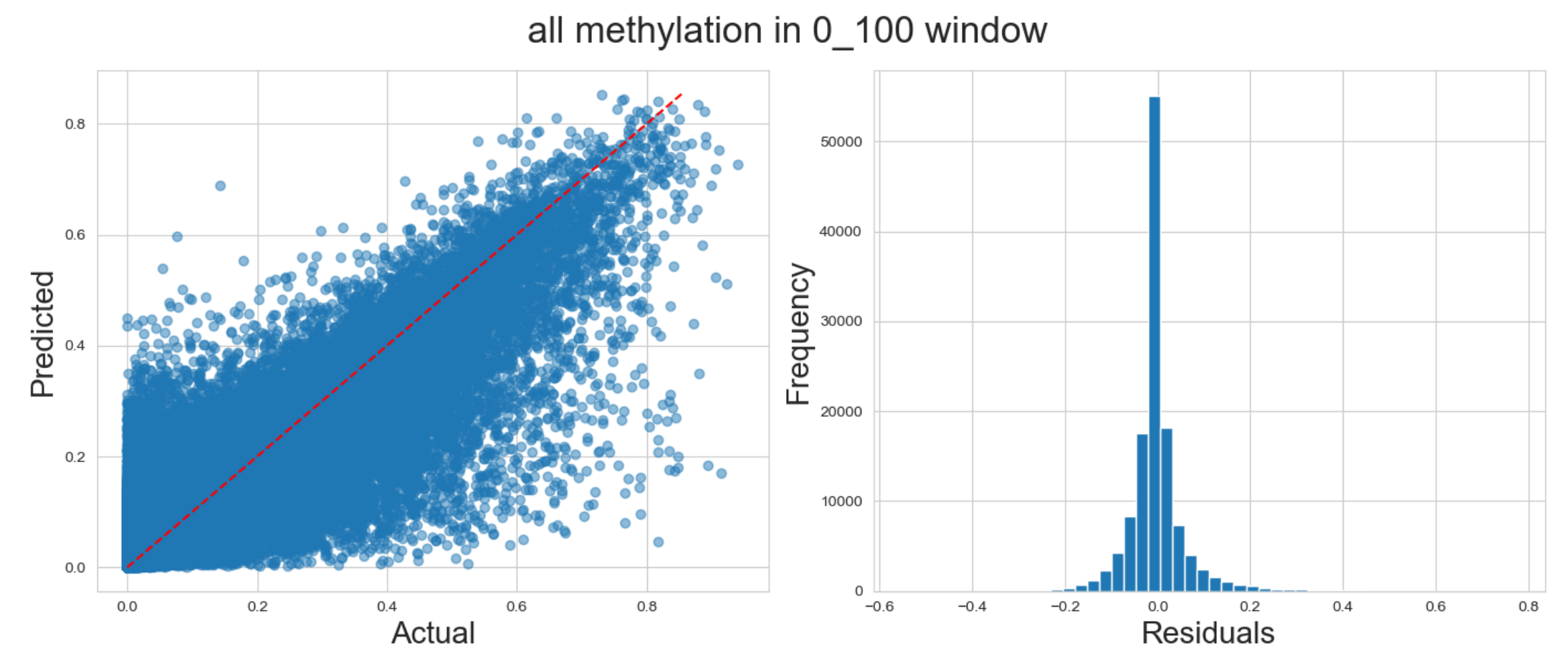
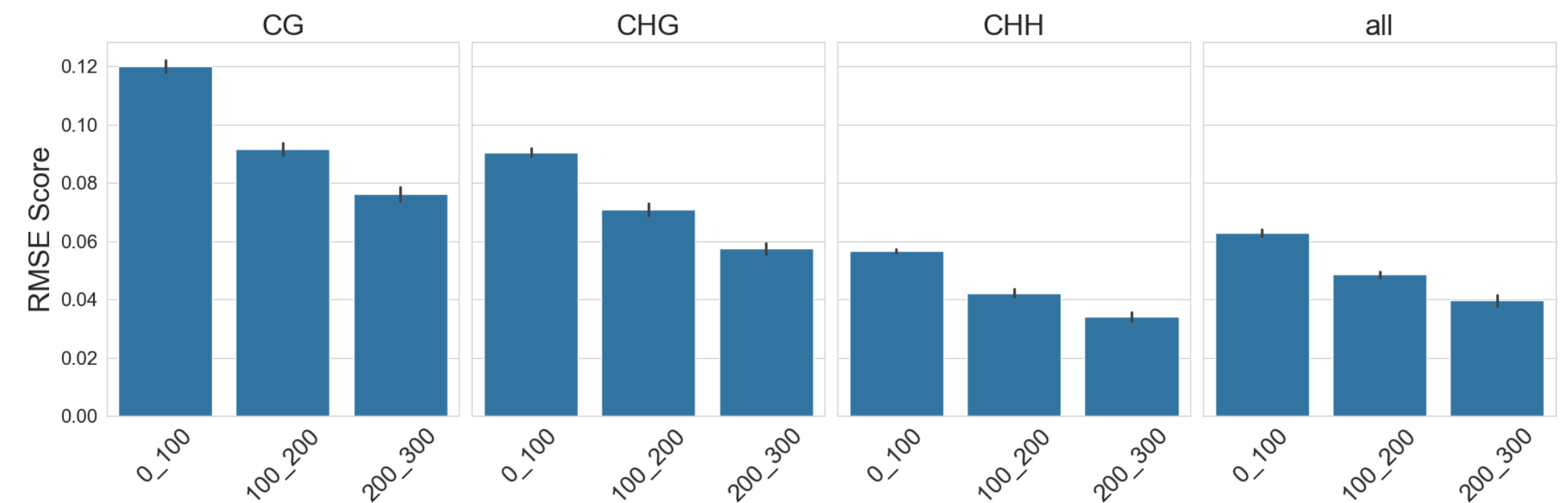
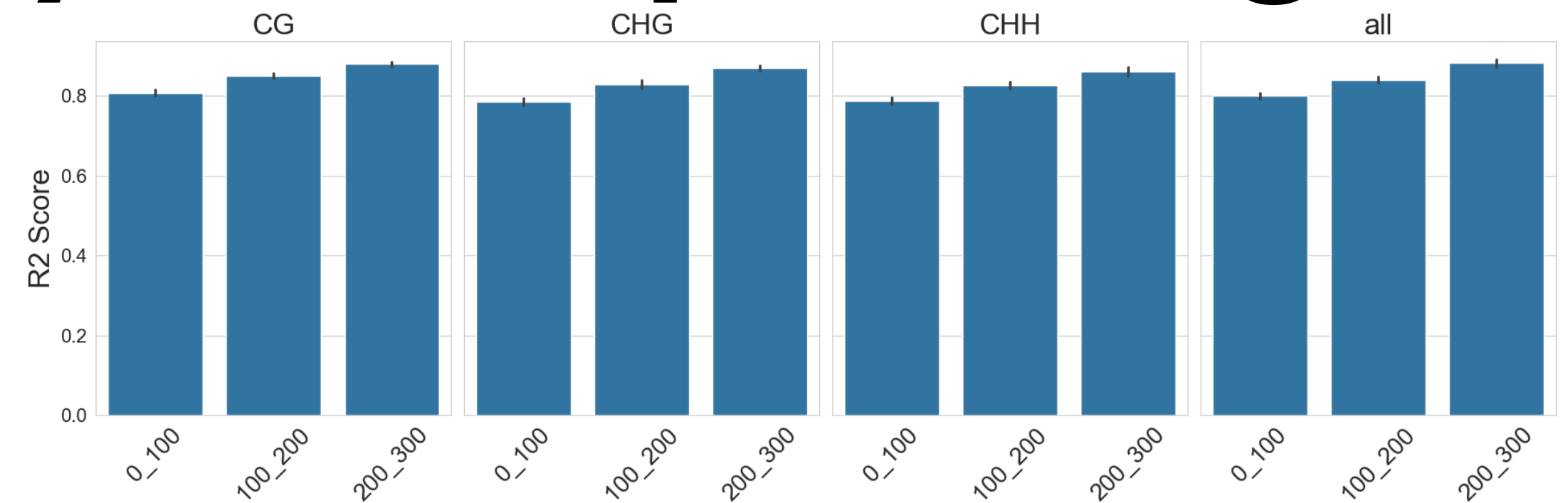
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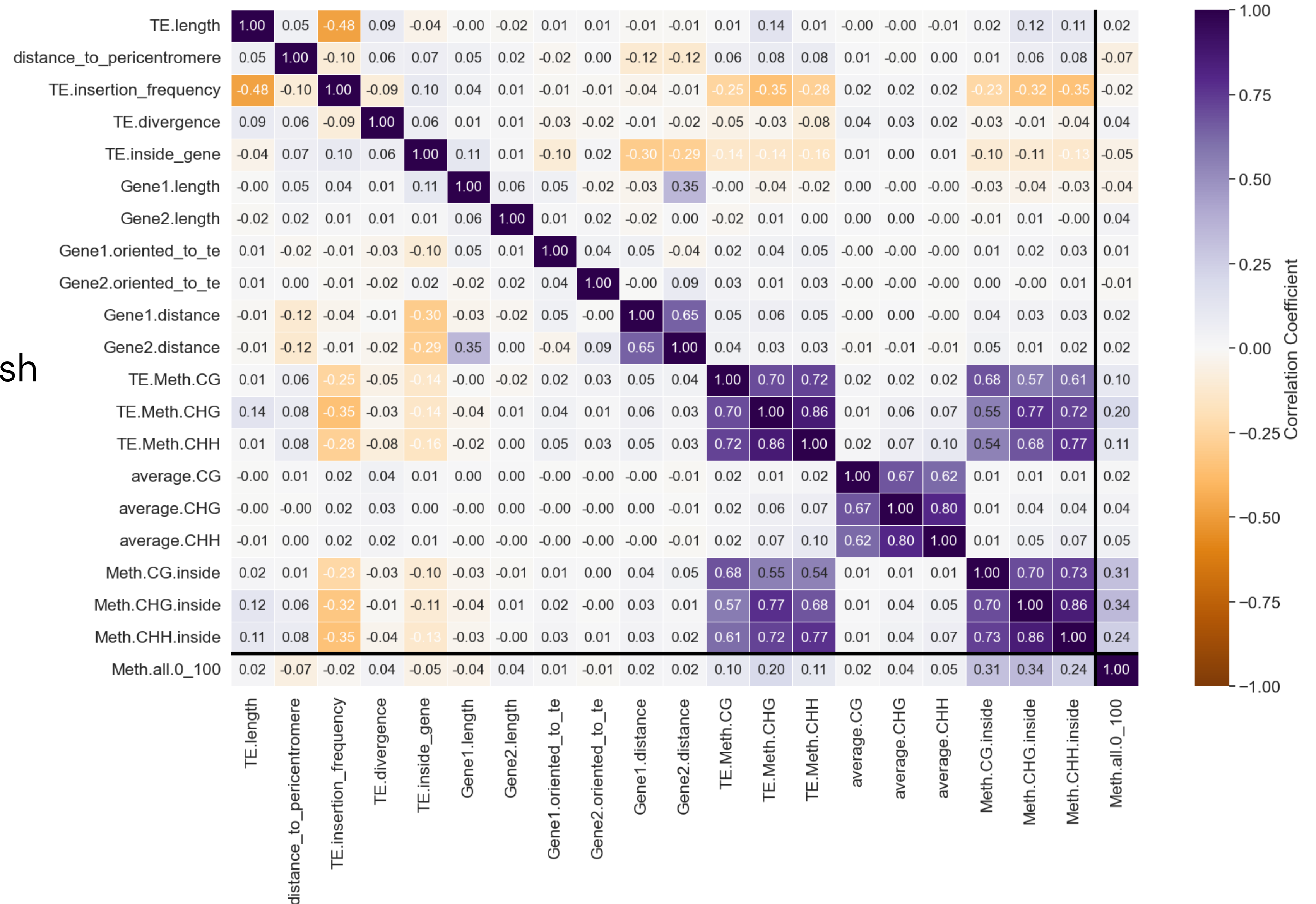


Prediction of methylation spreading

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

Prediction of methylation spreading

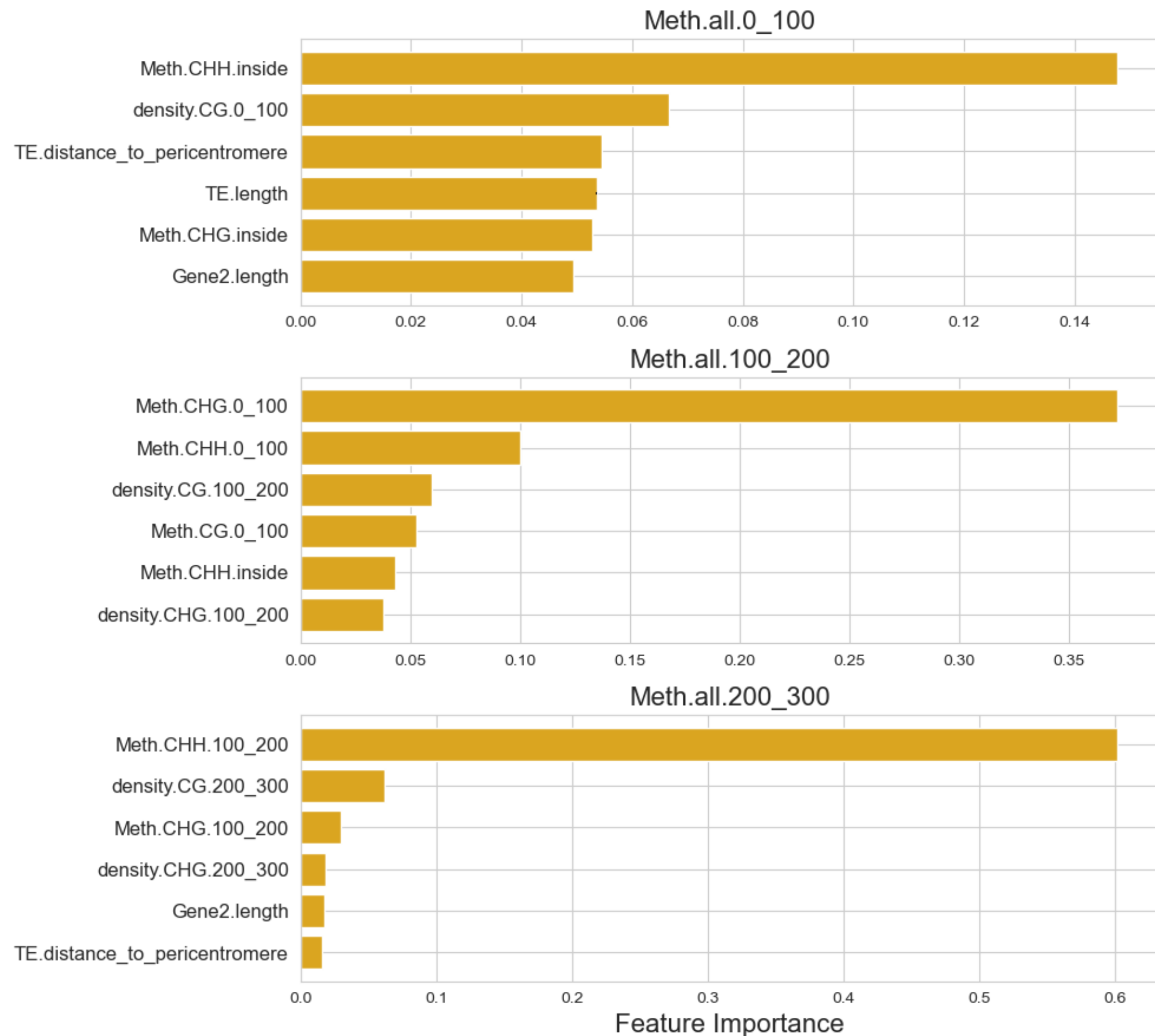
- The model predicts well, but which features define methylation level?
- Some features are highly correlated \implies hard to distinguish between them



Prediction of methylation spreading

Impurity-based feature importances

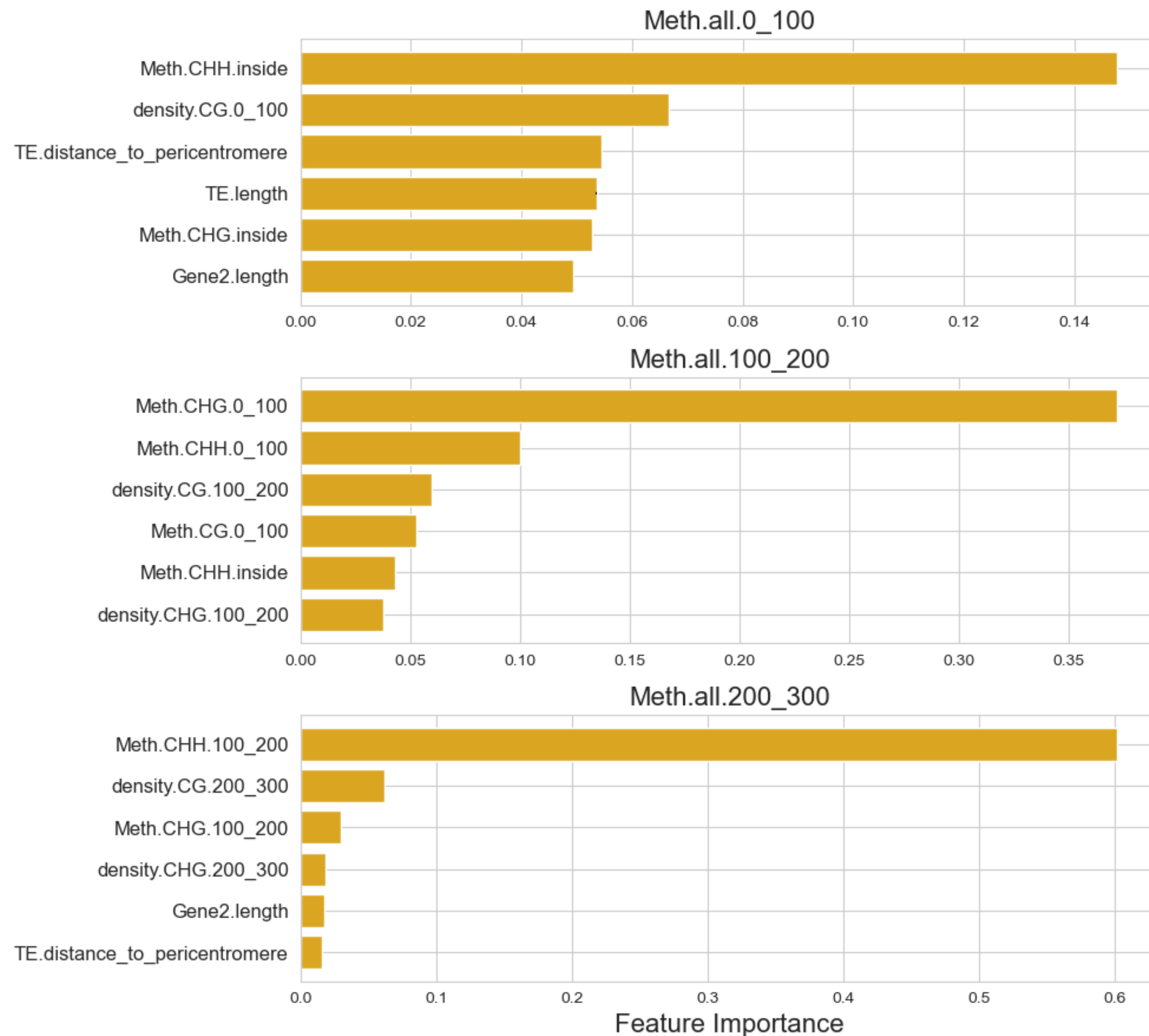
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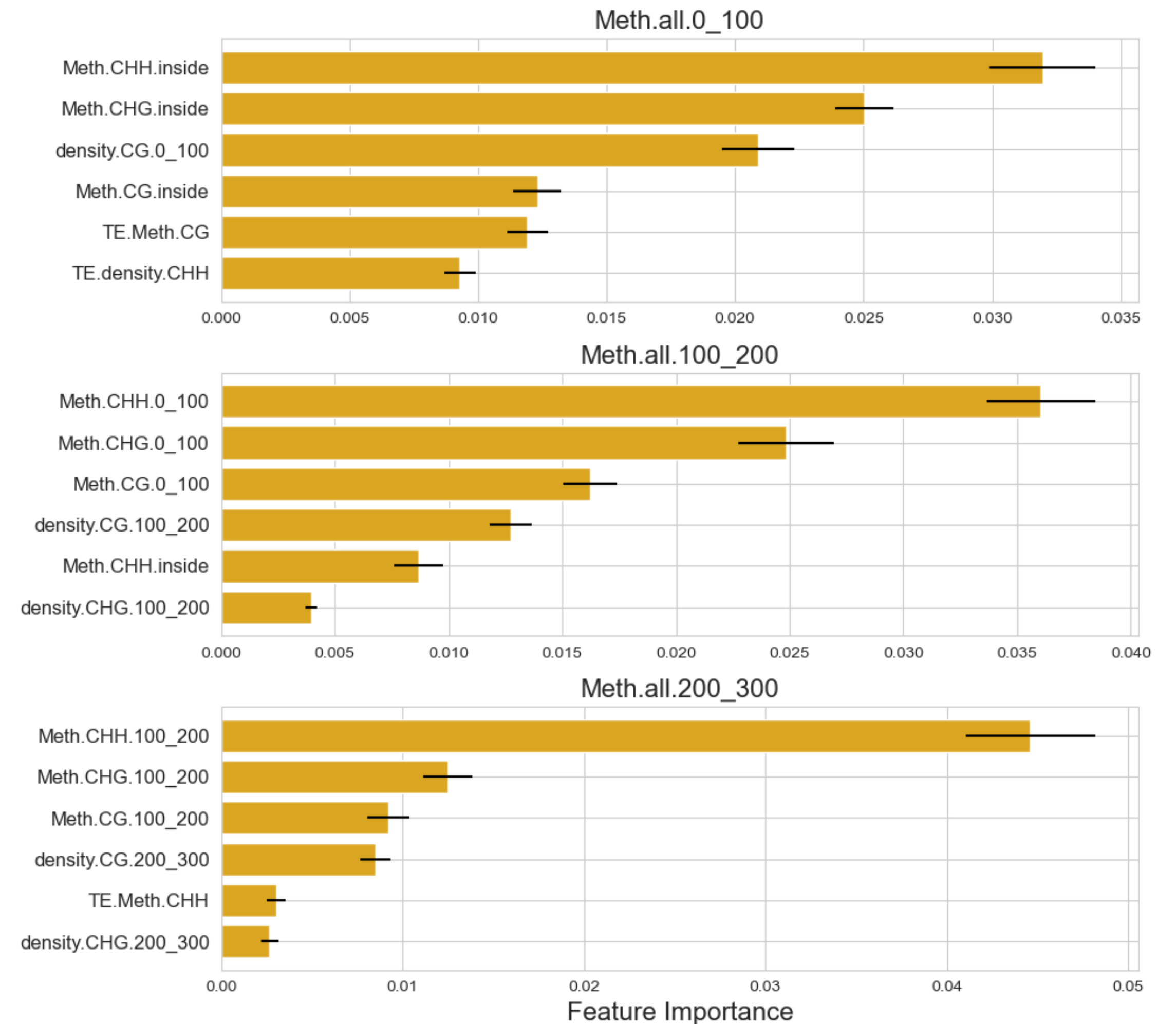
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SHAP values

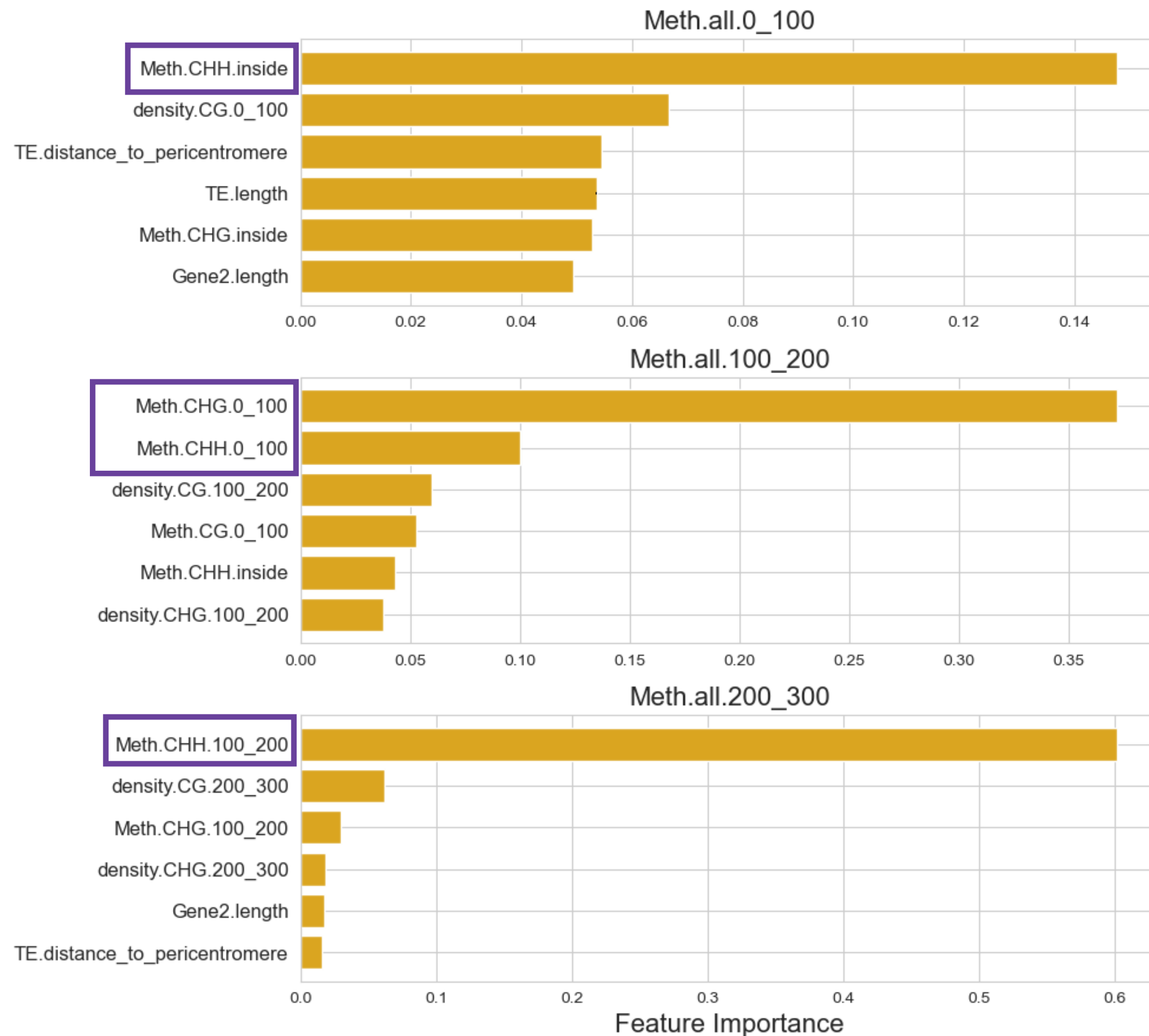
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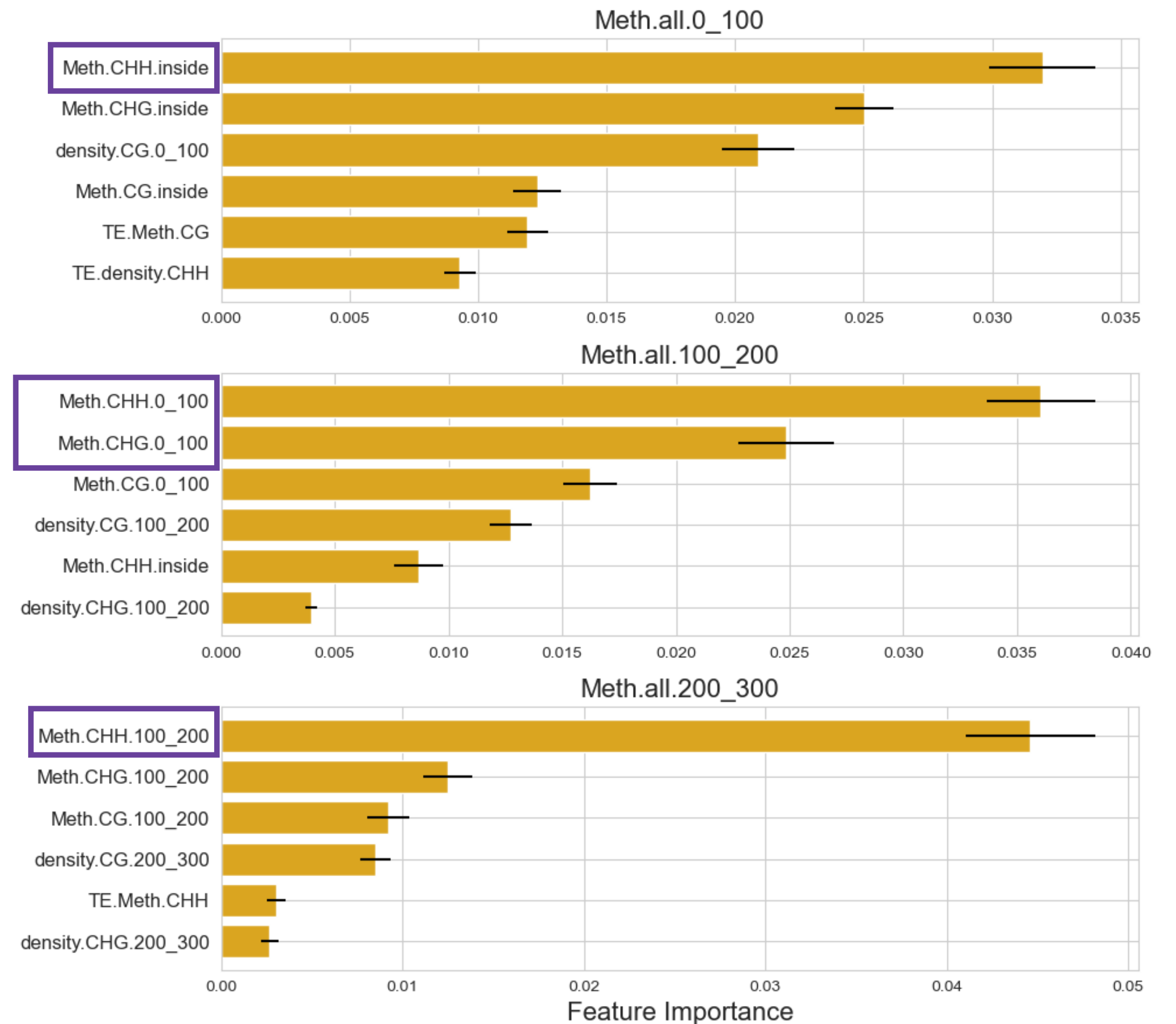
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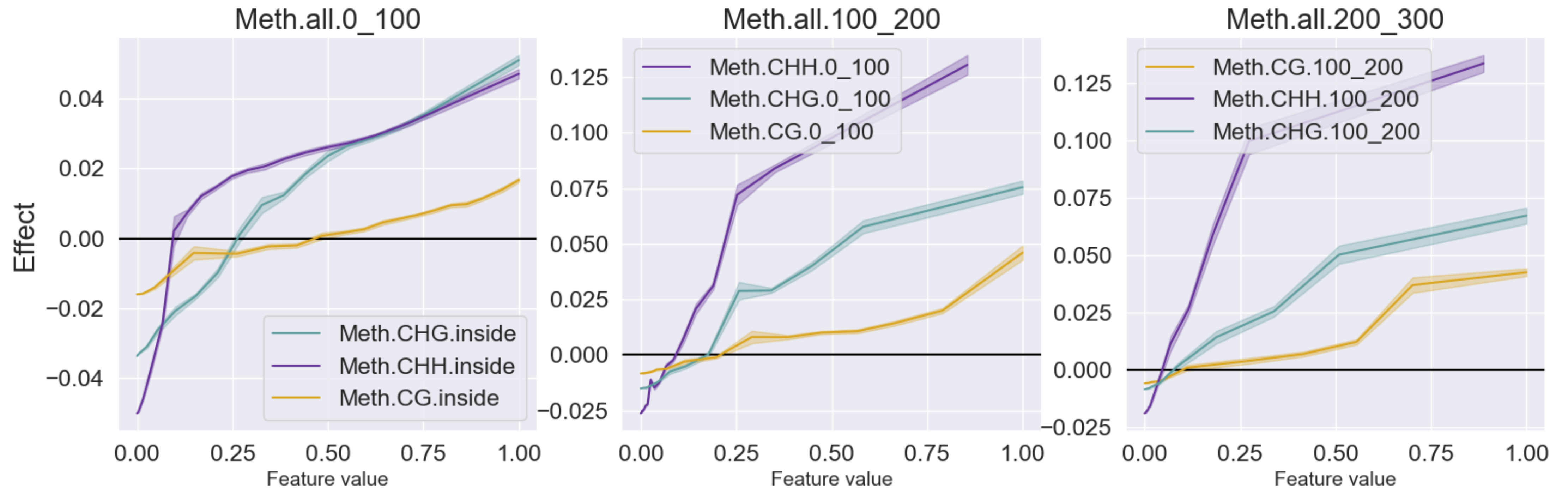
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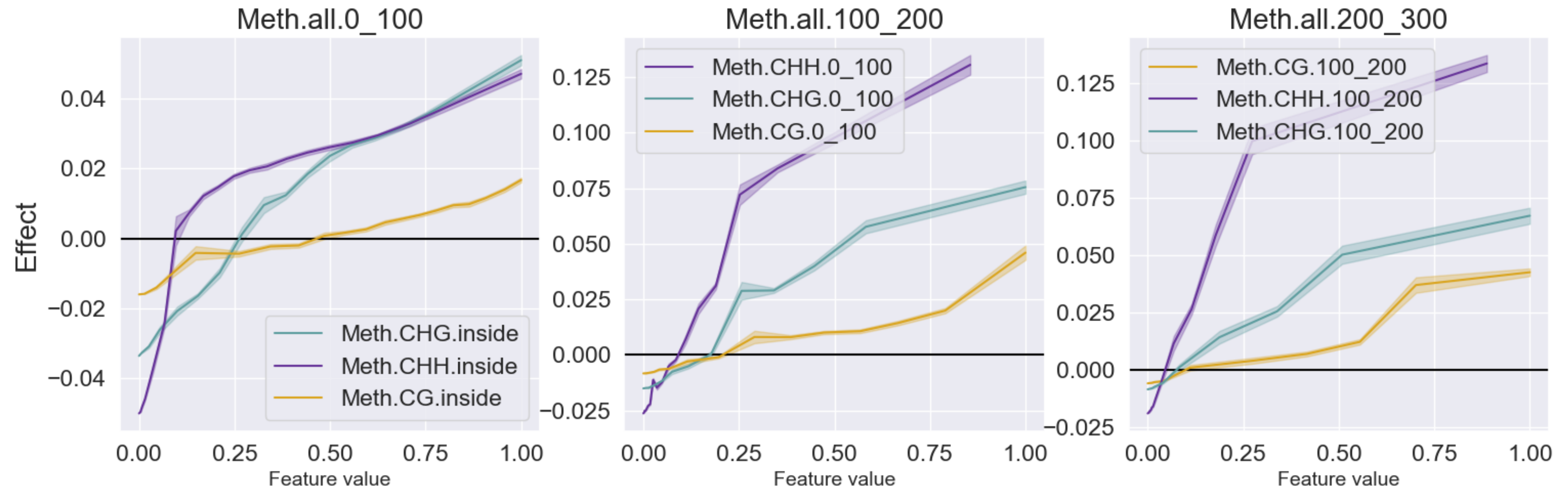
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Accumulated Local Effects (ALE)



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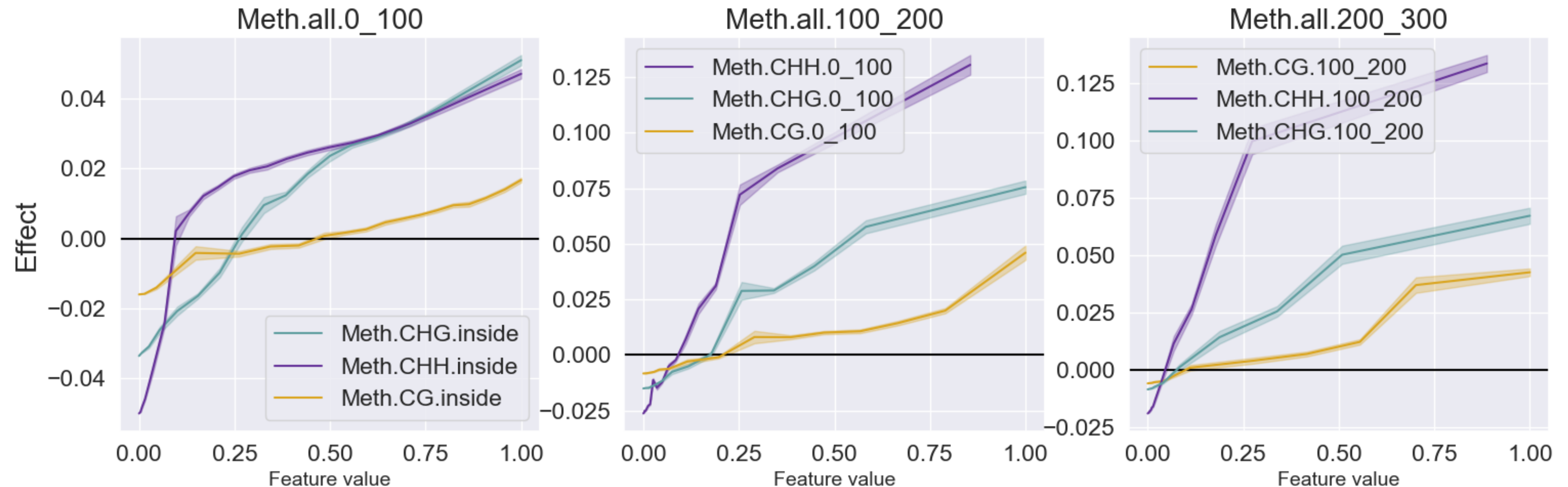


Conclusion:

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

Prediction of methylation spreading

Accumulated Local Effects (ALE)



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Question:

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

Prediction of inside methylation

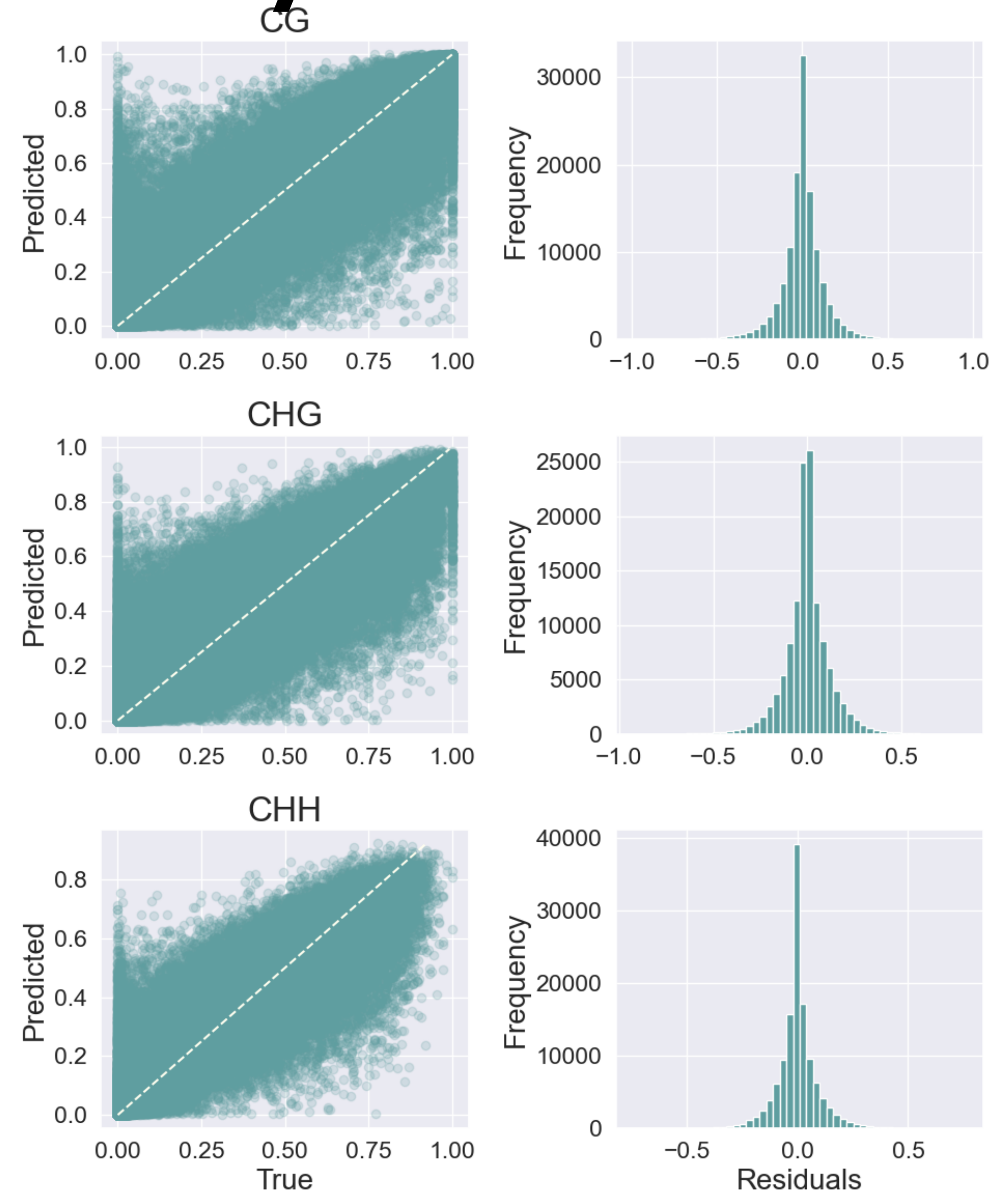
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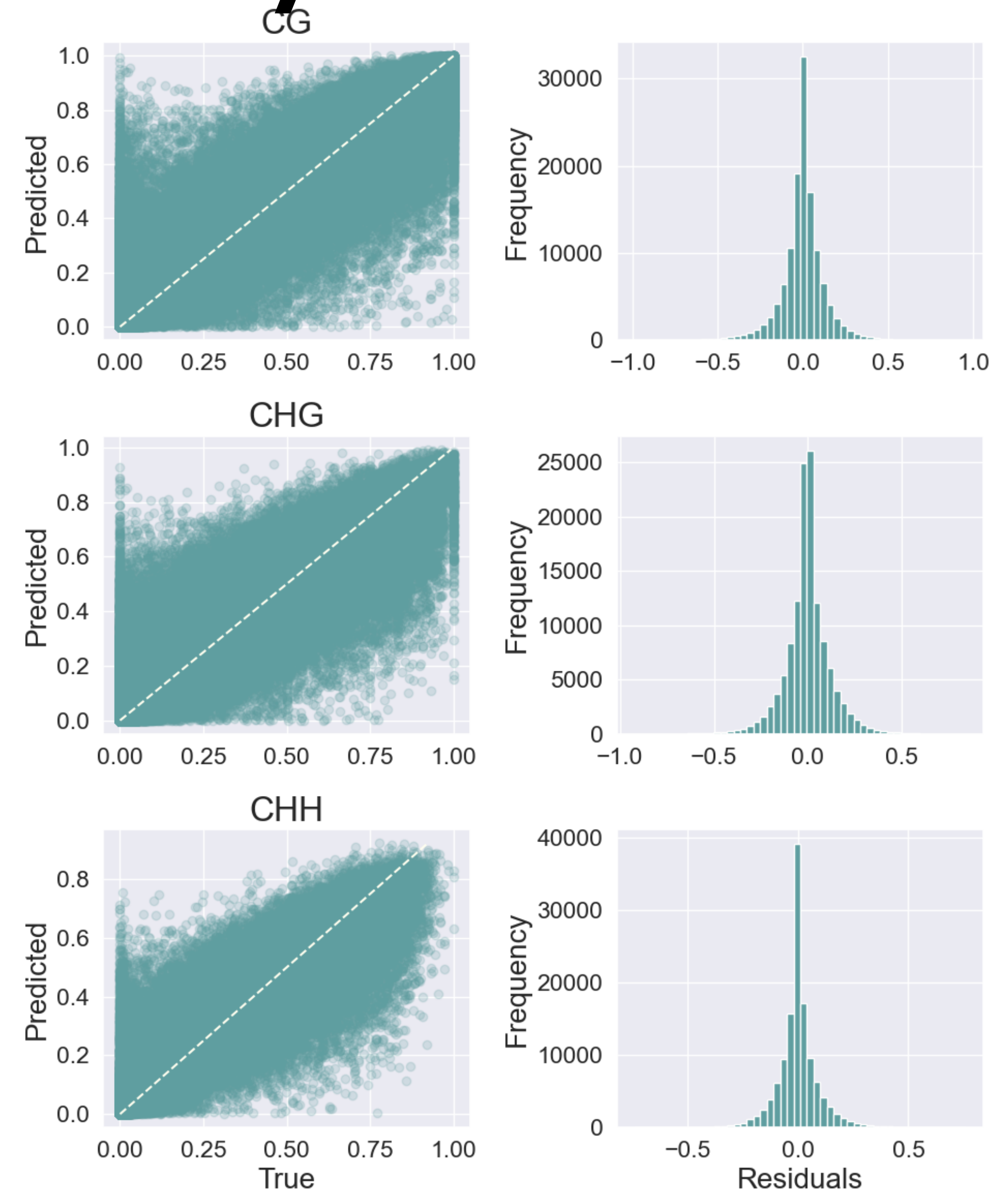
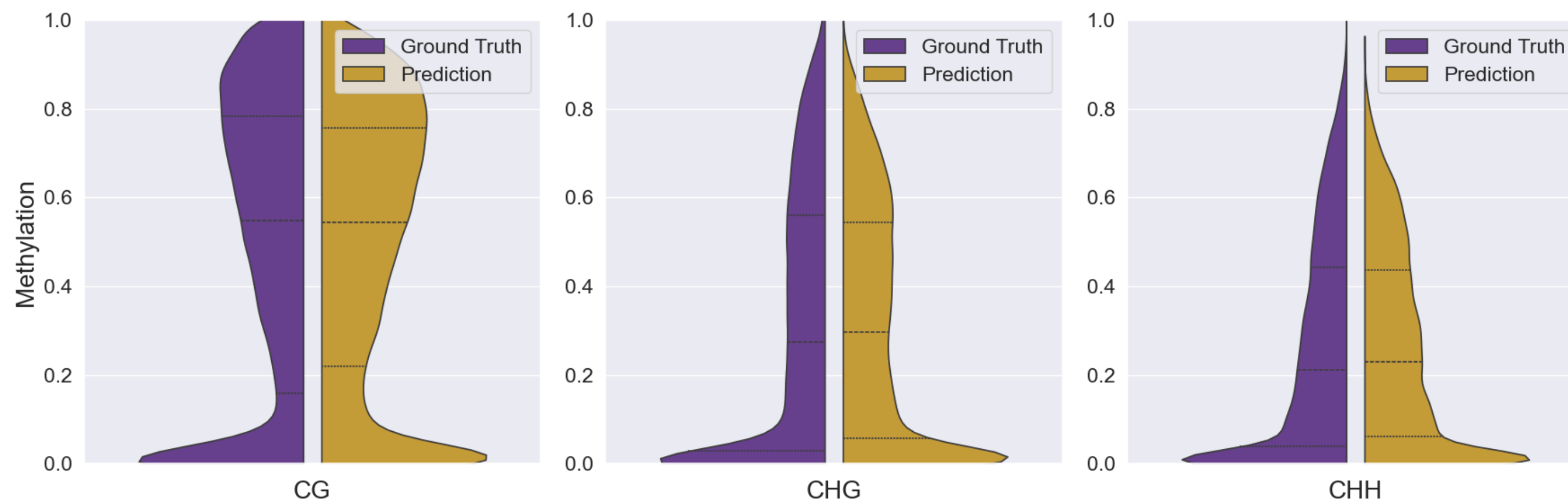
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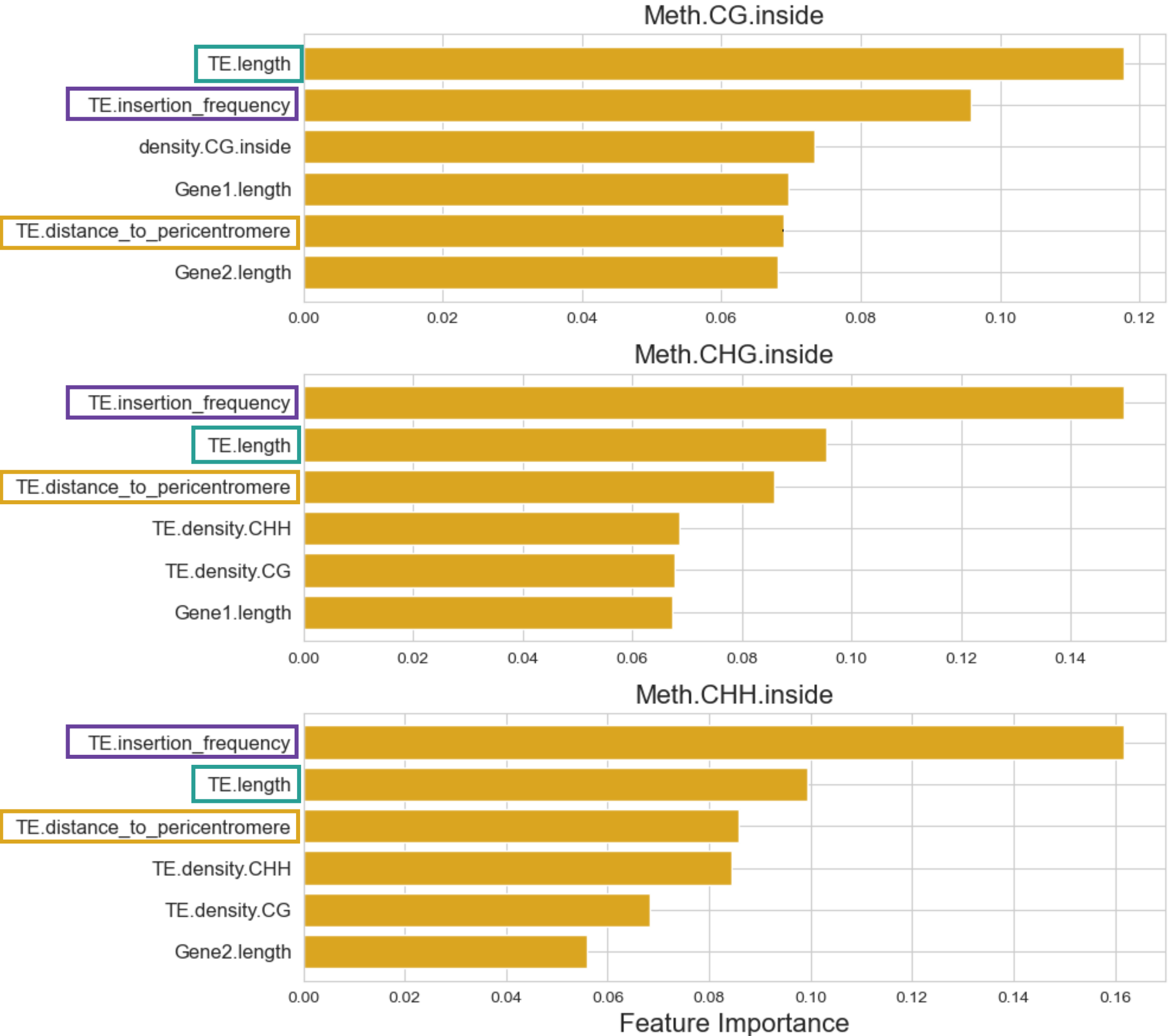
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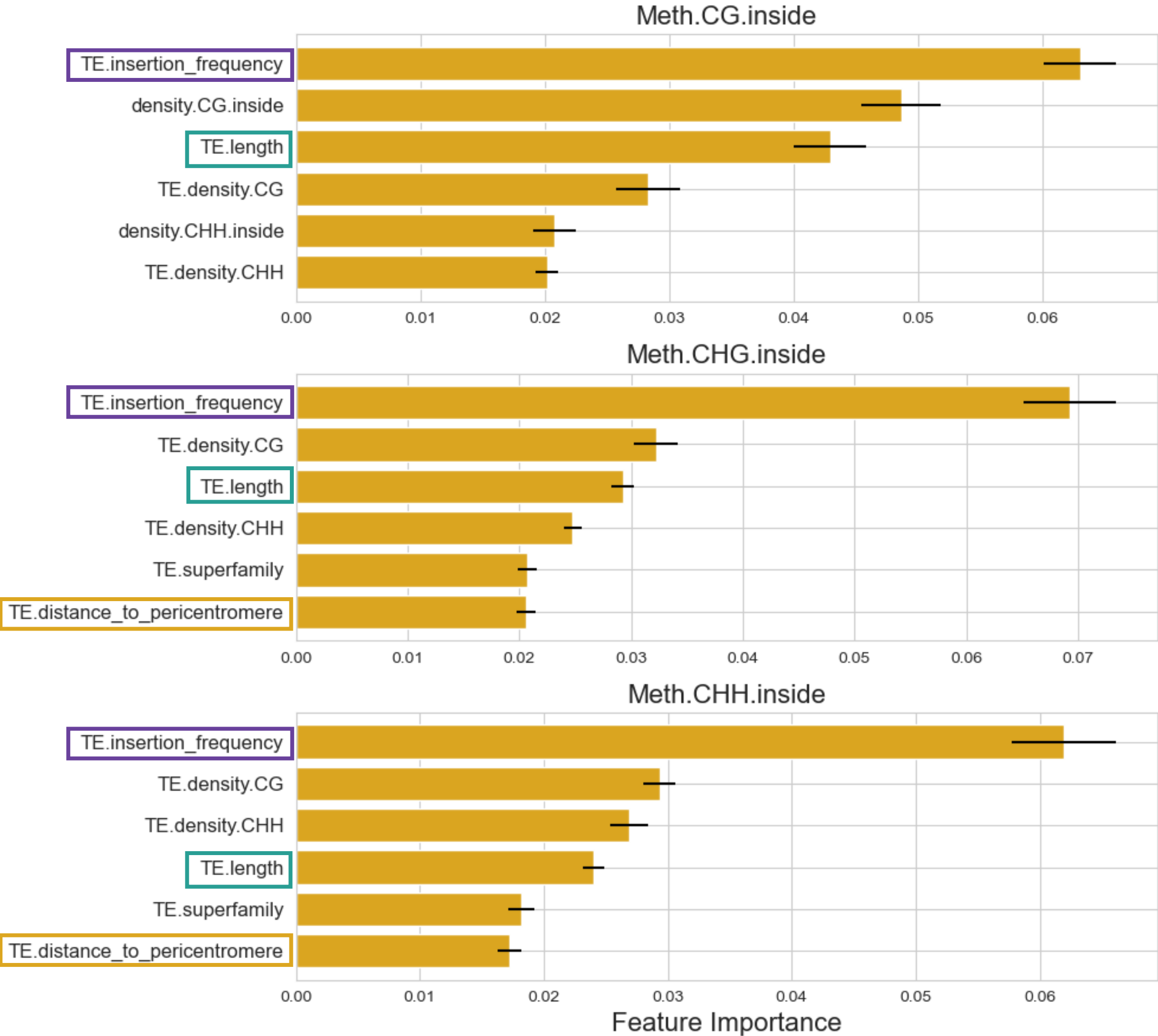
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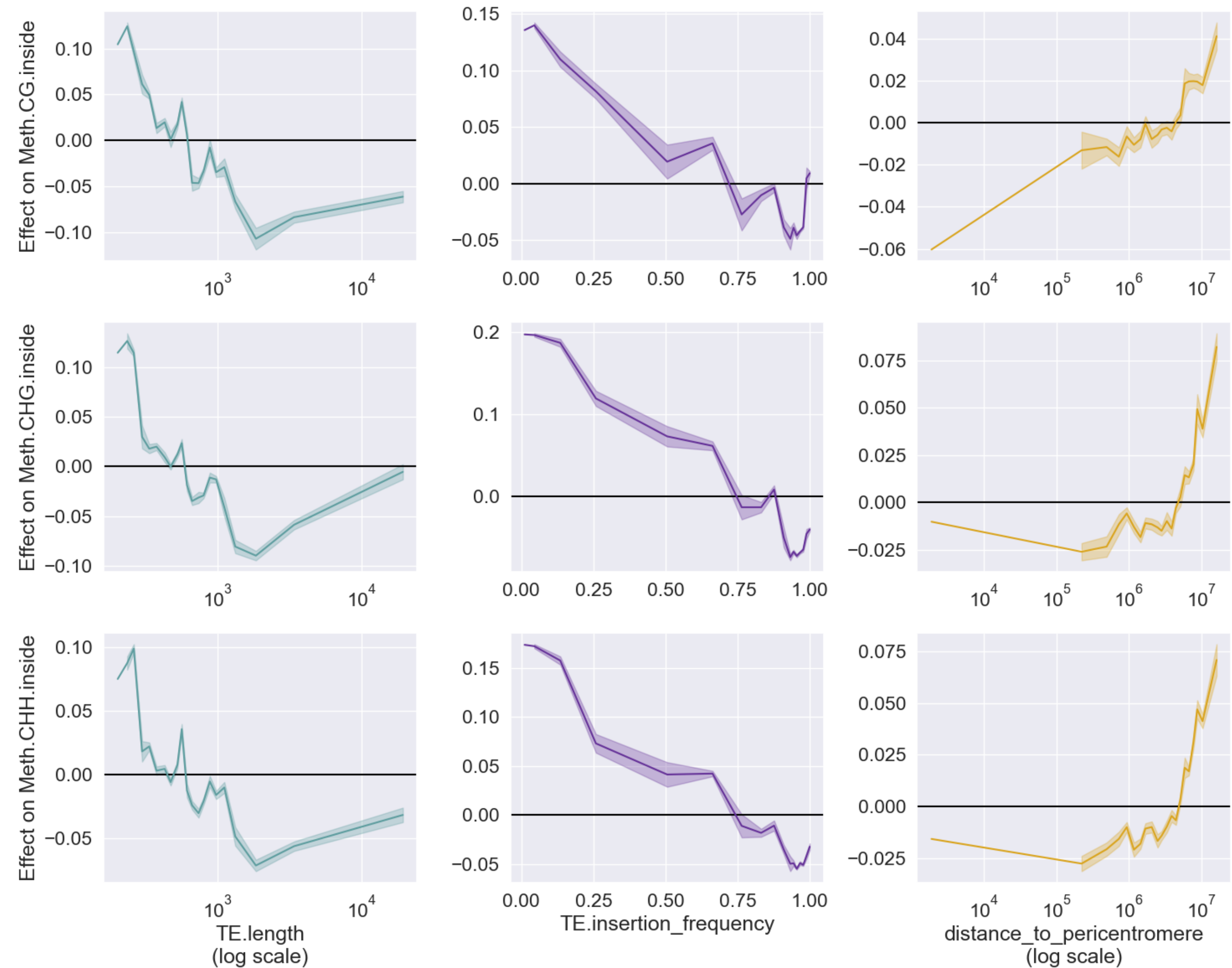
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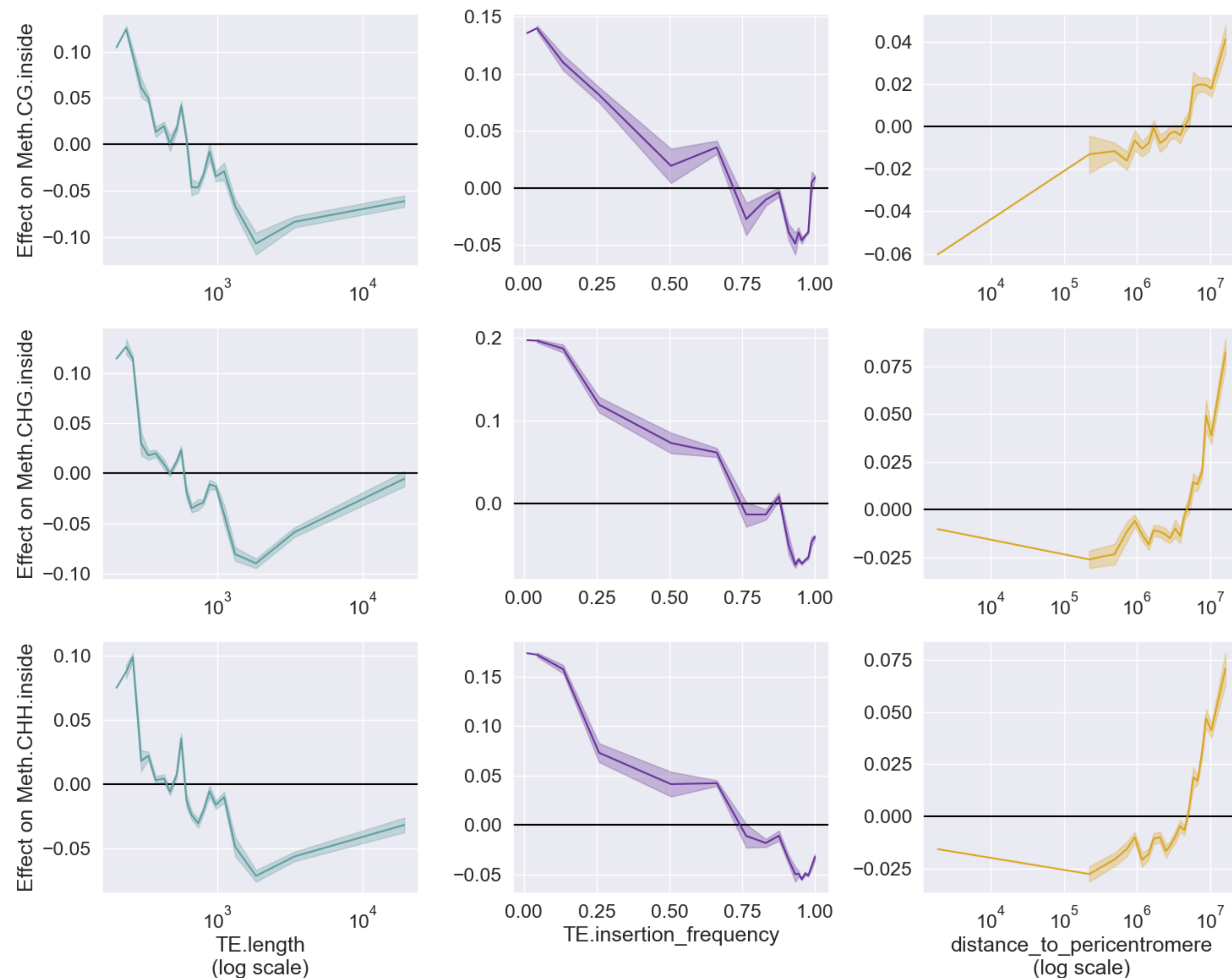
Prediction of inside methylation

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

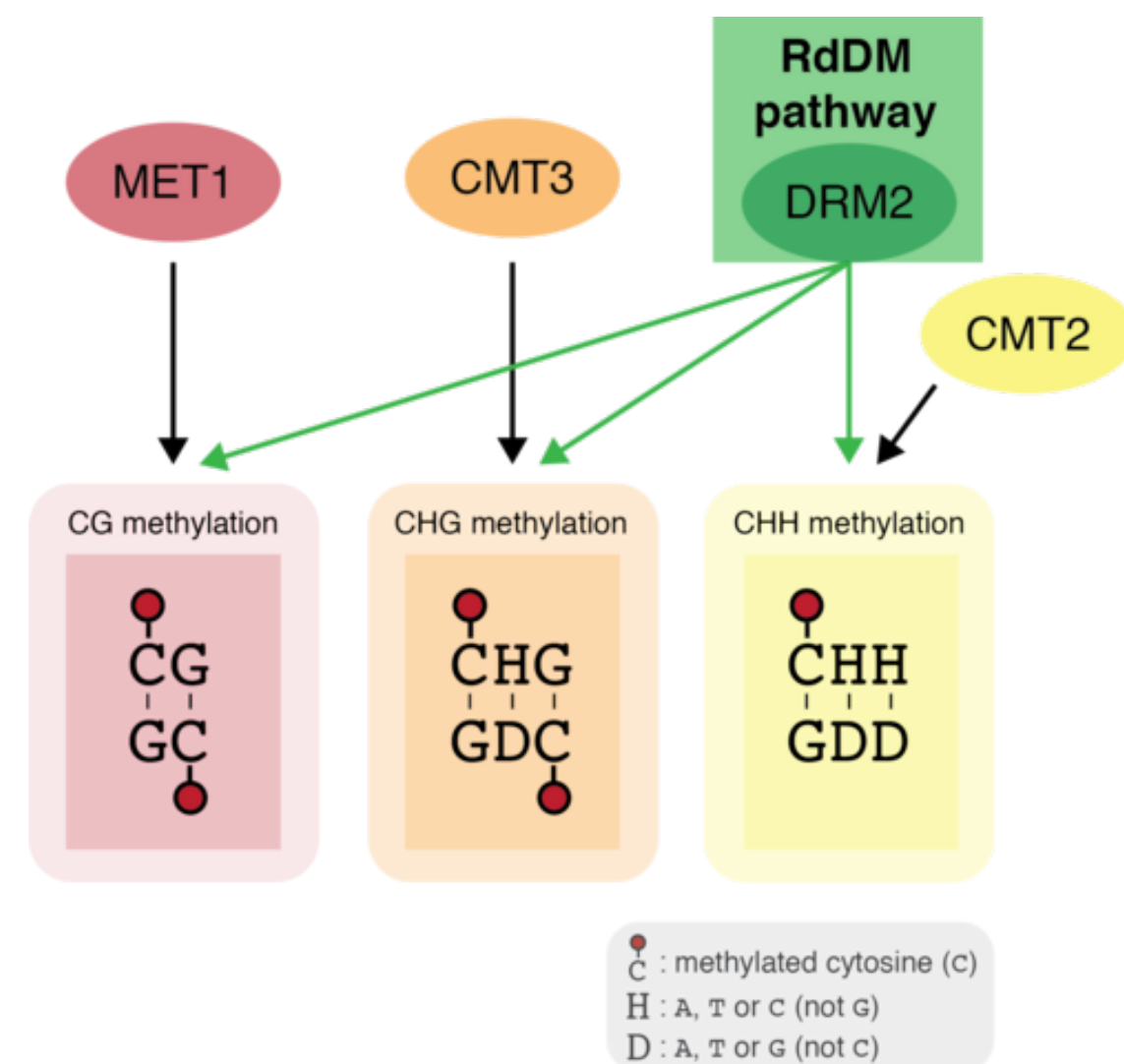
● The most important factors:

- methylation of insides in the CHG and CHH contexts
- insertion frequency
- distance to pericentromere

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Wikipedia

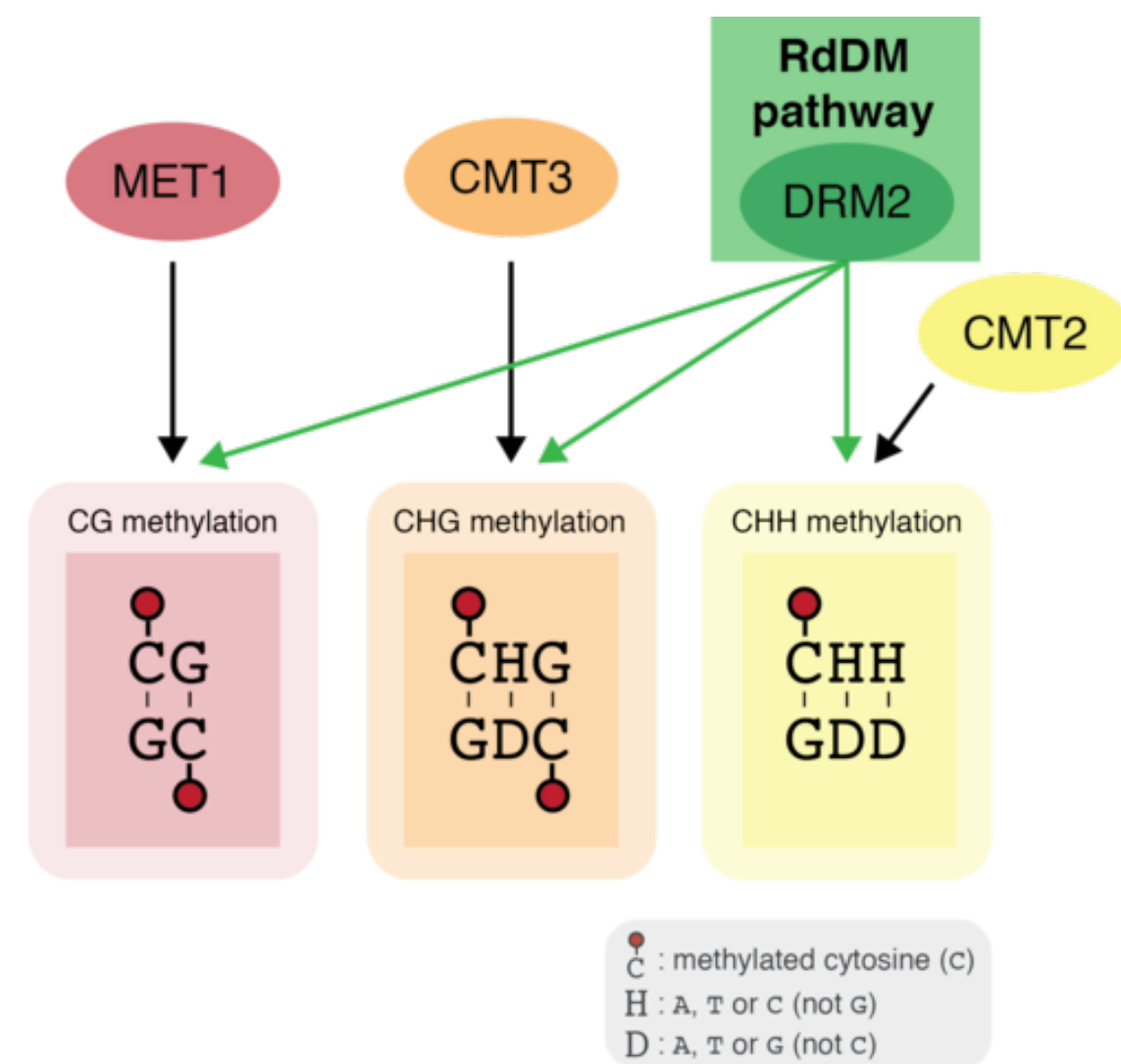
- **Hypothesis:** the **RdDM** machinery is responsible for spreading

- targets all contexts
- the only pathway capable of adding DNA methylation *de novo*
- targets rather chromosome arms

Biological conclusions

- **The most important factors:**

- methylation of insides in the CHG and CHH contexts
- insertion frequency
- distance to pericentromere



Wikipedia

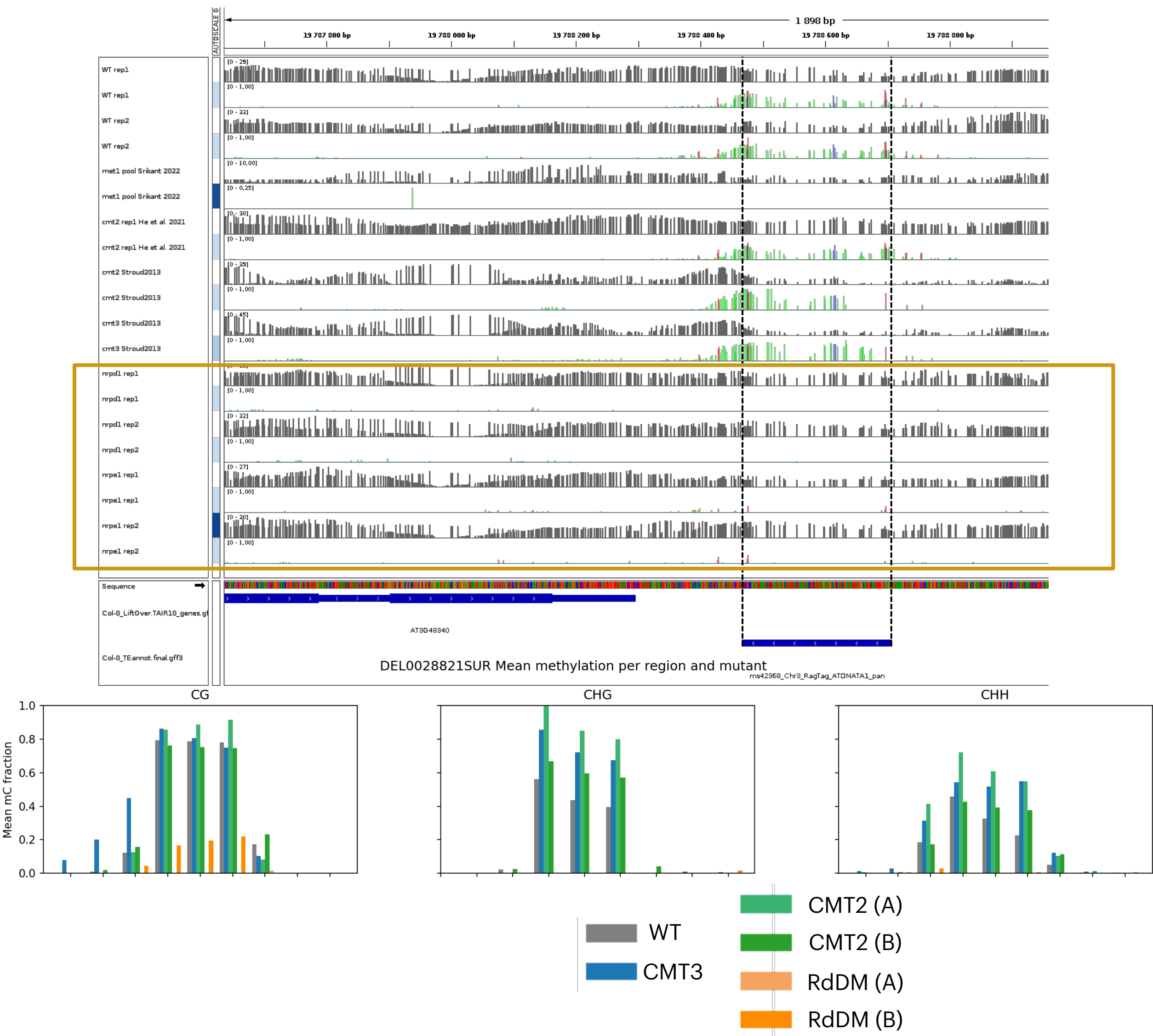
- **Hypothesis:** the **RdDM** machinery is responsible for spreading

- targets all contexts
- the only pathway capable of adding DNA methylation *de novo*
- targets rather chromosome arms

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

Biological confirmation

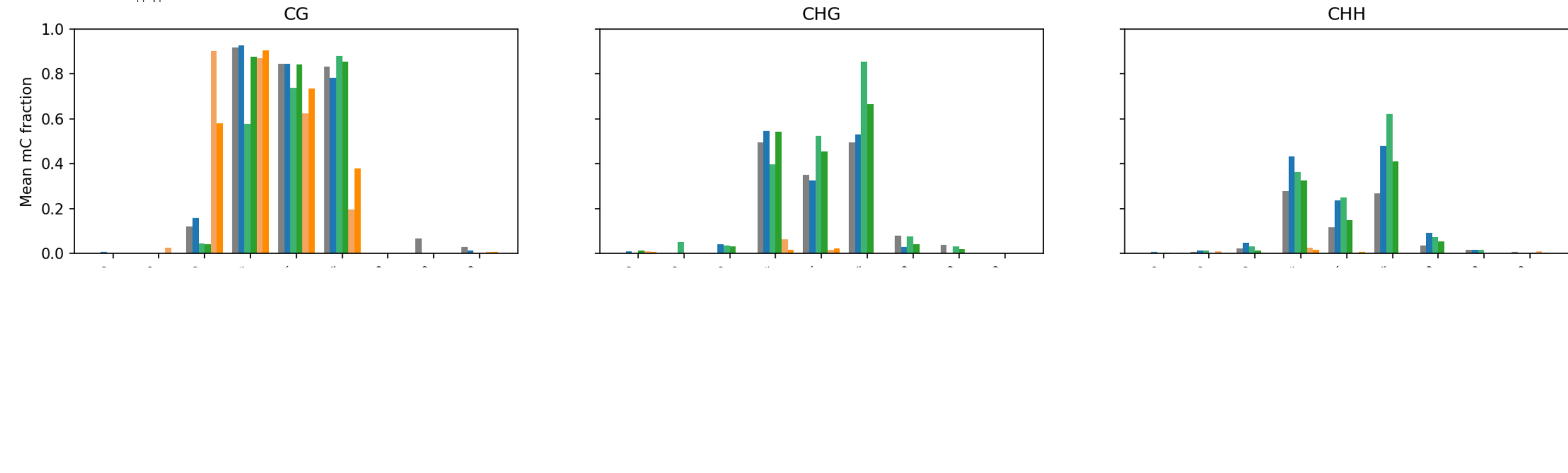
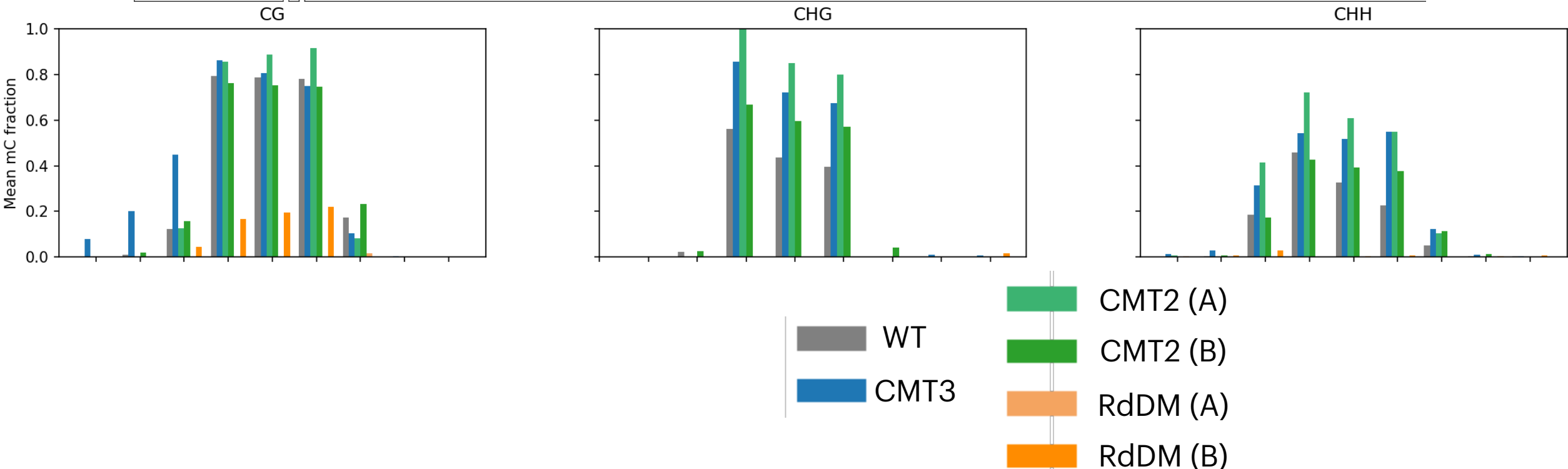
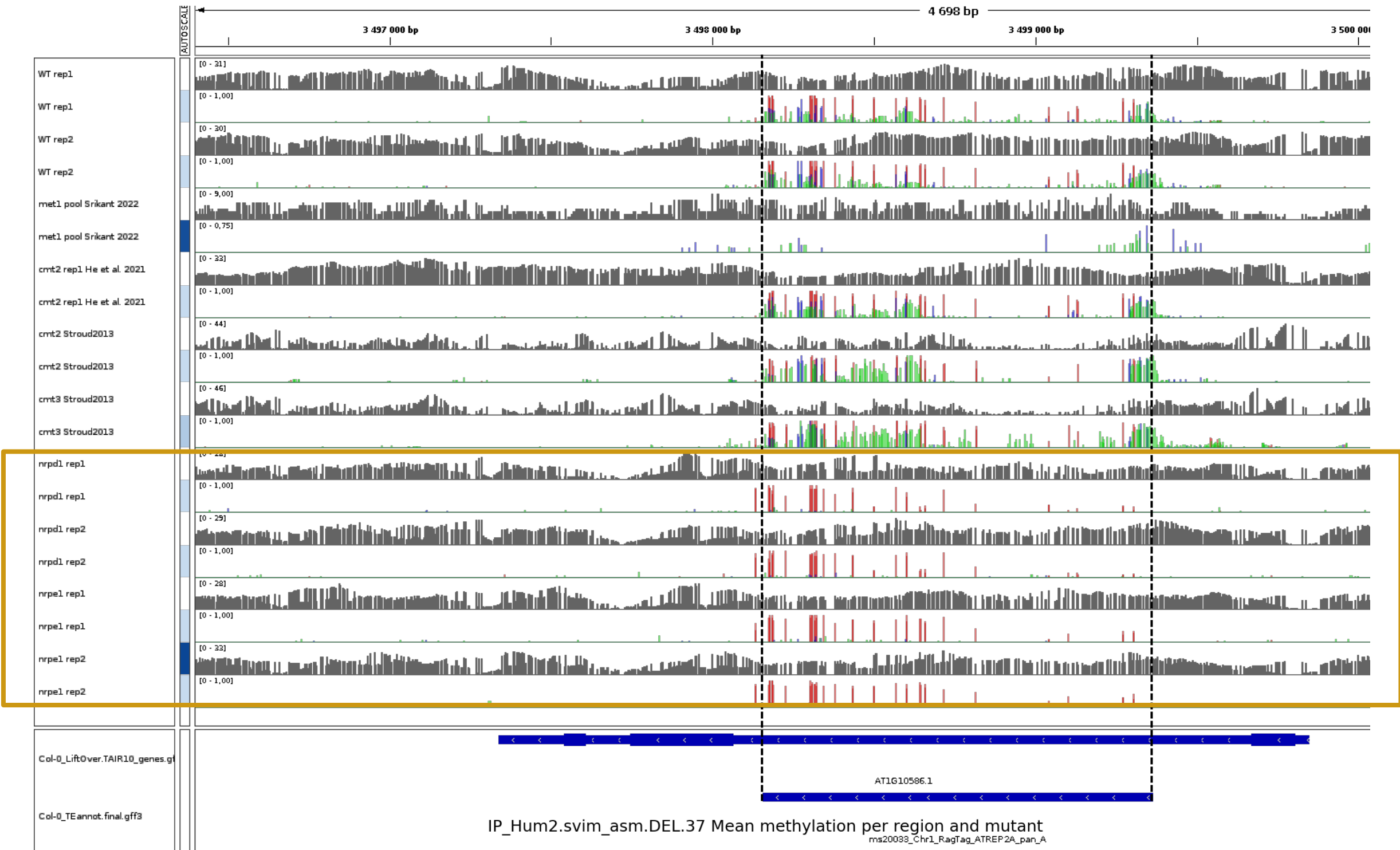
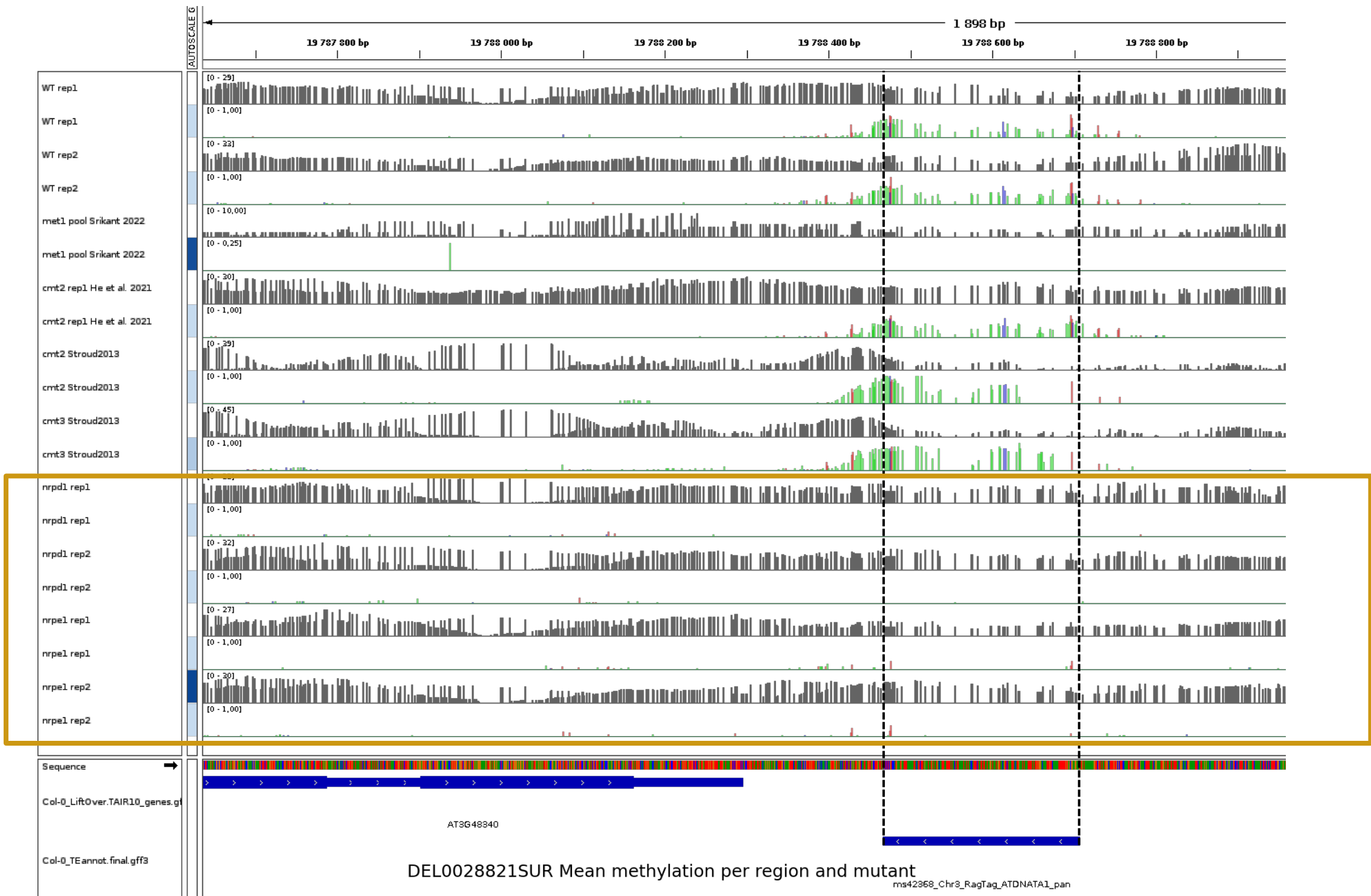
DELO028821SUR



Biological confirmation

DEL0028821SUR

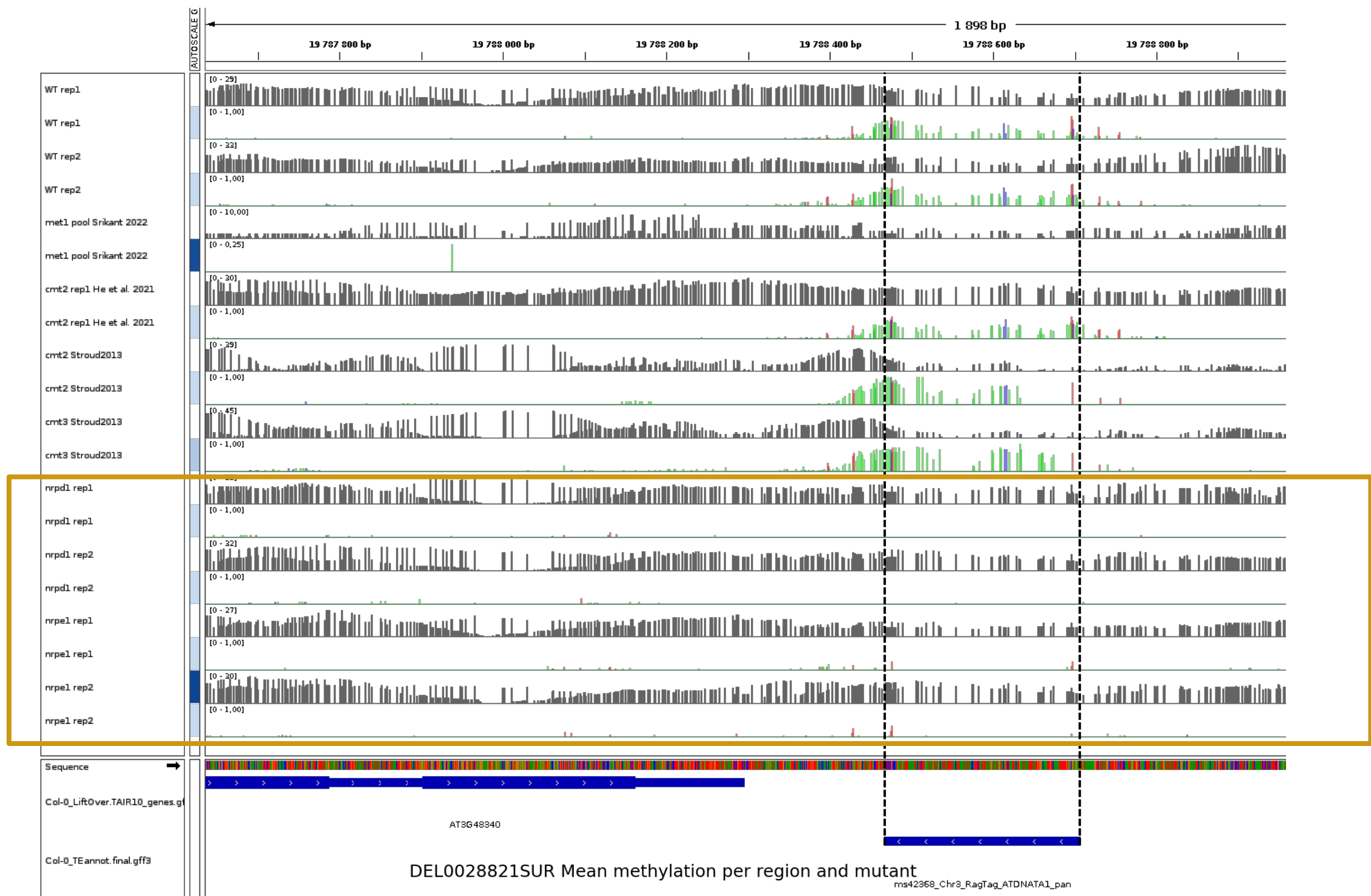
IP_Hum2.svim_asm.DEL.37



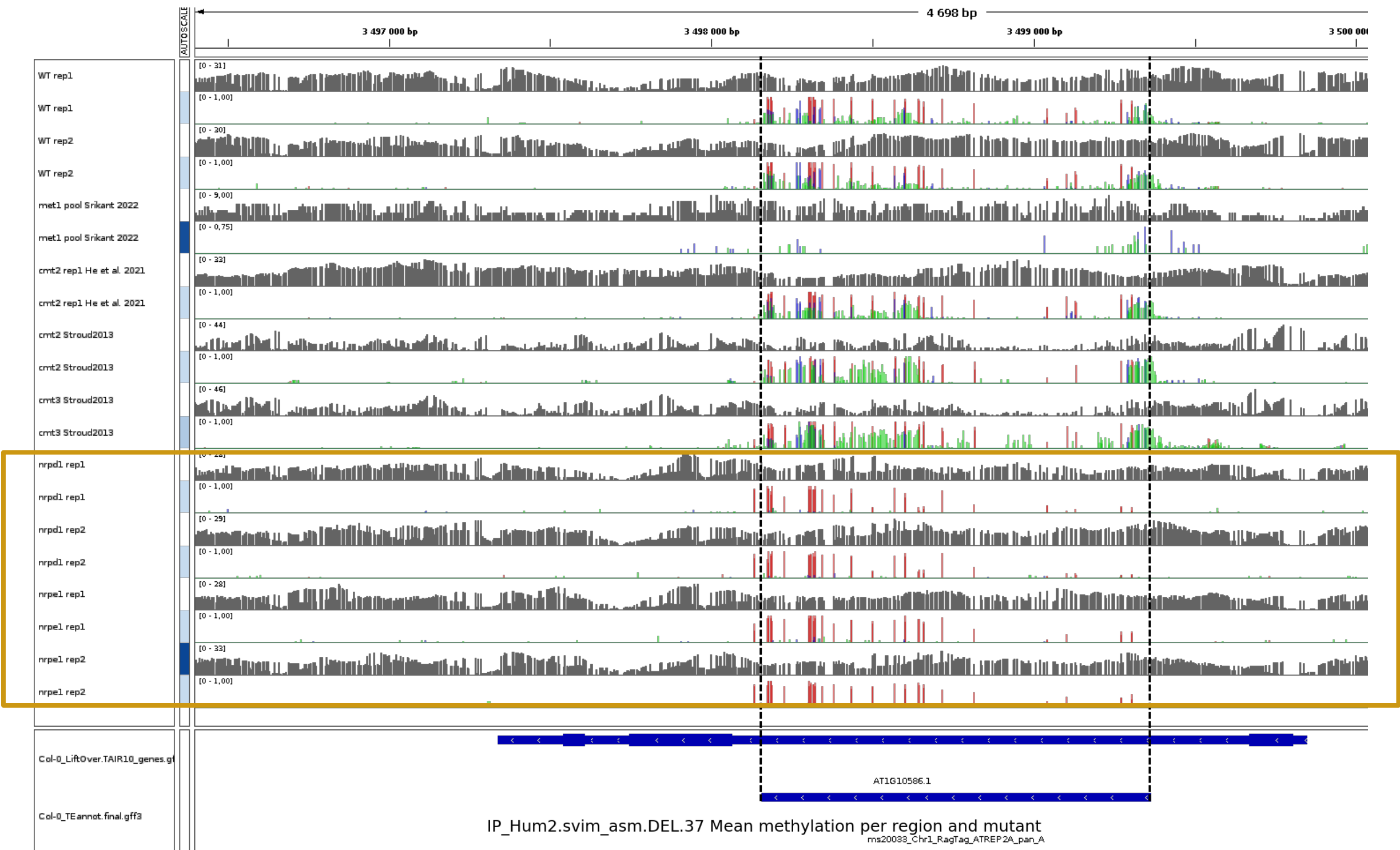
Biological confirmation

DEL0028821SUR

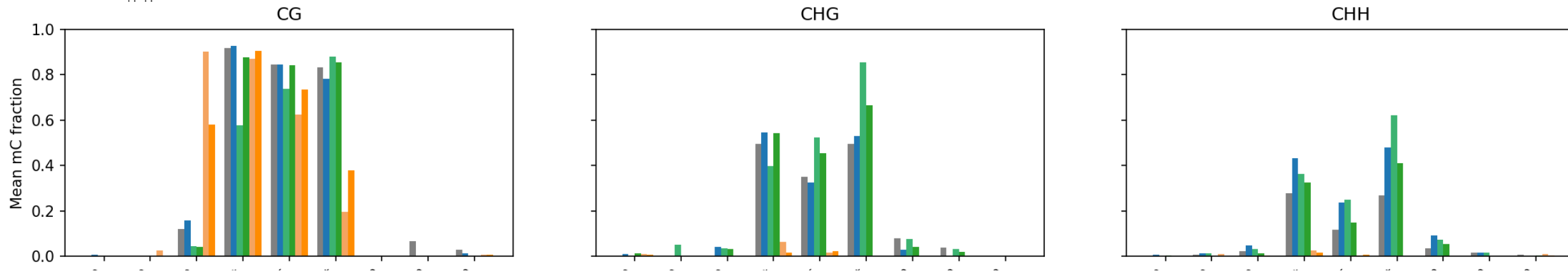
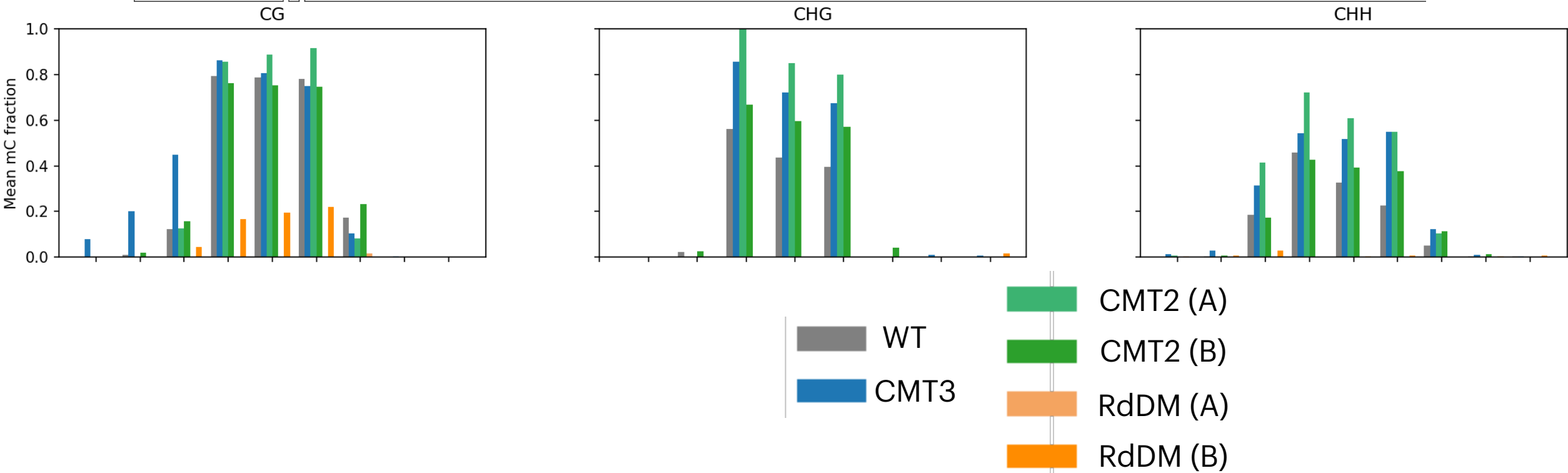
IP_Hum2.svim_asm.DEL.37



DEL0028821SUR Mean methylation per region and mutant



IP_Hum2.svim_asm.DEL.37 Mean methylation per region and mutant



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

Conclusions

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

- 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**

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