

Promoter-anchored chromatin interactions predicted from genetic analysis of epigenomic data

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Introduction

- Promoter-anchored chromatin interaction (PAI) is essential to understand the transcriptional regulation and the genetic regulatory mechanisms underpinning complex trait variation¹.
- Current high-throughput assays for detecting PAI are not scalable to population-based cohorts because of the complexity of generating a DNA library and extremely high sequencing depth to achieve high detection resolution².

Materials and Methods

- We proposed an analytical approach to predict chromatin interaction using summary level methylation quantitative trait locus (mQTL) data.
- Our analytical approach relies on the SMR (summary-data-based Mendelian randomization) & HEIDI (heterogeneity in dependent instruments) method³ to detect the association between DNAm levels of two CpG sites due to the same set of genetic variants.
- The summary-level blood mQTL data were from a meta-analysis of LBC and BSGS cohort ($n=1,980$).

Results

- We identified 34,797 PAIs for the promoter regions of 4,617 genes.
- The predicted PAIs show strong overlap with the chromatin contacts identified by previous experimental assays.

PAI analysis

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Results

- The promoter-interacting DNAm sites are enriched in enhancers or near expression QTLs.
- Genes whose promoters are involved in PAIs are more actively expressed, and gene pairs with promoter-promoter interactions are enriched for co-expression.

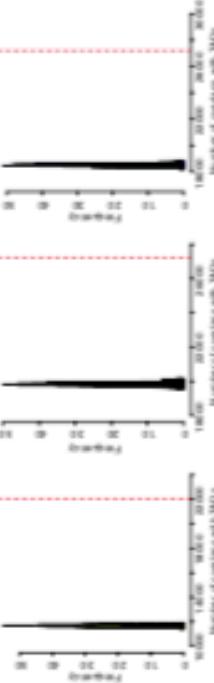


Figure 1 Schematic of the PAI analysis. PAI (promoter-promoter) interactions are detected by two probes anchored to the promoter of Gene A. Gene B is used as a control gene. Gene C is indicated as a control gene.

- The mean Pearson correlation of expression levels for gene pairs whose promoters were involved in promoter-promoter interaction and proportion of promoter-PAI genes in gene activity groups.



Figure 2 Overlaps of the predicted PAIs with topologically associating domains (TADs) identified by a) Rao et al.⁴, b) Dixon et al.⁵, and c) Javie-Carrasco et al.⁶ using Hi-C and promoter-t-captured Hi-C (PCHi-C).

- The predicted interactions from our method are more enriched in the PCHi-C loops⁶ compared to other prediction methods.



Figure 3 Enrichment of the predicted interactions in the significant PCHi-C loops⁶ defined based on a range of P-value thresholds. The fold enrichment value was computed by a 2×2 contingency table and the horizontal red dashed line indicates no enrichment.

Conclusions

- We propose an analytical approach that uses the sum nary-level data from cohort-based mQTL studies to predict PAIs.
- The predicted PAIs show strong overlap with the chromatin contacts identified by experimental assays and outperformed other prediction methods.
- This study demonstrates the use of mQTL data to predict PAIs and provides insights into the role of PAIs in complex trait variation.

Publication

References

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Contact



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