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This work demonstrates that the prediction of Hi-C data from chromatin accessibility is a promising strategy to replace the use of genomic distance thresholds.

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