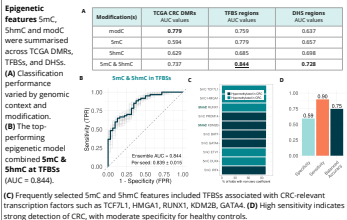


1. Introduction

Cell-free DNA (cfDNA) contains complementary genetic, epigenetic and fragmentomic signals that can inform cancer detection. However, these biomarkers are typically measured using separate assays, increasing sample input requirements, analytical complexity, and inter-assay variability.

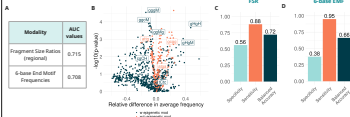
duet 6-base mosaic and **duet evoC** can simultaneously profile 5mC, 5hmC, genetic variants and fragmentomic features from a single cfDNA sample¹. Using cfDNA in plasma from healthy controls and Stage I-IV colorectal cancer (CRC) patients, we derived multiple molecular features using **duet evoC** and evaluated their individual classification performance. Integrating modality-specific signal consistently improved CRC detection, demonstrating the value of multimodal liquid biopsy with 6-base sequencing.

3. Resolving 5mC & 5hmC enhances CRC detection



Together, these results highlight that 5mC & 5hmC profiling at regulatory regions provides a informative and biologically relevant biomarkers for cDNA-based CRC detection.

5. Methylation-aware end motifs strengthen CRC classification



Fragmentomics features. 6-base sequencing enabled estimation of regional fragment size ratios (FSR) and quantification of 6-base end-motif frequencies (EMF), incorporating cytosine modification states: unmodified (C), modC (C), 5mC (M), and 5hmC (H).

(A) Both fragmentomics features distinguished CRC from healthy controls, with FSR showing the highest performance (AUC=0.715), followed by 6-base EMF (AUC=0.708).

(B) Top 6-base EMF features included canonical 4-base motifs (ggcg, atga, atgg) and modification-aware motifs (gggM, gHgH, ggMg) highlighting contributions from 5mC and 5hmC.

(C) FSR showed high sensitivity with moderate specificity.

(D) The 6-base EMF classifier achieved high sensitivity, but lower specificity, reflecting better performance for the identification of cancer than controls.

7. Multimodal cfDNA integration improves CRC classification



Epigenetic, genetic and fragmentomic features were combined in a **late-fusion framework** using weighted prediction probabilities. Together, these results show that combining complementary 6-base cDNA signals improves CRC classification beyond individual modalities.

(A) Weights optimised performance (AUC 0.821 → 0.890).

(B) Key contributors included modC in TCGA DMRs, 5mC & 5hmC in TFBs and DHSs, 6-base EMF and FSR.

(C) Multimodal integration produced stepwise gains in classification performance.

(D) The multimodal model correctly reclassified 3 CRC samples missed by the best modC model.

2. Study design & analysis

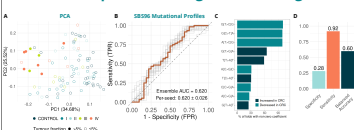
CRC cohort: 92 individuals; 32 healthy donors, 60 CRC patients.²

Features:

- (i) **epigenetic signals** (5mC, 5hmC, modC) at CRC-associated TCGA DMR (differentially methylated regions)³, TFBs (transcription factor binding sites)⁴ and DHSs (DNase I hypersensitive sites)⁵.
- (ii) **genetic signals** represented by SB596 mutational patterns⁶
- (iii) **fragmentomic signals** including fragment size ratios⁷, end motif frequencies⁸, and nucleosome occupancy profiles⁹.

Multimodal integration: Modality-specific classifiers were trained and evaluated using leave-one-out cross-validation followed by weighted integration of predictions to identify optimised biomarker combinations for CRC classification.

4. Mutational profiles distinguish later-stage CRC



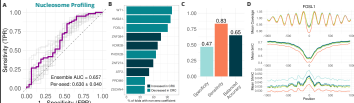
Genetic features (SB596 mutational profiles) were derived from high-confidence somatic variants identified using the biomodal duet software. Later-stage CRC samples with higher tumour fractions showed distinct mutational profiles, while early-stage CRC more closely resembled controls.

(A) PCA shows separation enriched for high tumour fraction Stage III/IV samples (filled circles).

(B) The SB596 classifier achieved an AUC of 0.620 (C) Top substitutions included G(T)→C(G), G(C)→T(A), and A(T)→C(G).

(D) High sensitivity but low specificity indicates limited discrimination of healthy controls. Overall, genetic features alone showed modest predictive power but may provide complementary information when integrated with other modalities.

6. Nucleosome profiling reveals CRC-associated TF activity



Nucleosome profile features were derived from 6-base fragment-level coverage at TFBs, providing a read-out of nucleosome occupancy and chromatin accessibility; lower central coverage indicates higher accessibility.

(A) TFBs nucleosome profiles distinguished CRC from healthy controls (AUC = 0.657).

(B) Top features included CRC-associated transcription factors WT1, HMGAI1, FOSL1, and KDM2B.

(C) The classifier had high sensitivity with moderate specificity.

(D) FOSL1 showed reduced central coverage in Stage IV CRC versus control, consistent with increased accessibility and accompanied by changes in methylation, reduced 5mC and 5hmC, that are expected to drive the opening of chromatin.

Together, nucleosome profiling provides a complementary fragmentomic readout of CRC-associated regulatory changes in cfDNA with intriguing complementarity with 5mC and 5hmC changes that would be invisible with a traditional modC readout.

8. Conclusions

6-base sequencing data enables simultaneous interrogation of epigenetic, genetic and fragmentomic biomarkers from a single cfDNA sample. Each modality captures distinct complementary aspects of tumour biology, with epigenetic features providing the strongest individual signal. Integrating modalities consistently improves CRC classification, achieving an AUC of 0.890. Together, these results demonstrate the potential of a unified multi-modal workflow to maximize liquid biopsy performance while reducing assay complexity.

9. References

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2. Puddu, et al. 6-base cDNA profiling reveals complementary genetic, epigenetic and fragmentomic biomarkers for cancer detection. bioRxiv preprint doi: <https://doi.org/10.1101/2023.08.15.555555>; this version posted August 15, 2023. The copyright holder for this preprint (which was not certified by peer review) is the author/funder, who has granted bioRxiv a license to display the preprint in perpetuity. It is made available under aCC-BY-NC-ND 4.0 International license.
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