





# MULTI-OMICS PROFILING OF PLASMA FOR BREAST CANCER DISCRIMINATION BY ATR-FTIR SPECTROSCOPY AND MACHINE LEARNING

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#### INTRODUCTION

Early and accurate detection of breast cancer (BC) remains a clinical challenge, driving the search for sensitive and non-invasive diagnostic strategies.

ATR-FTIR spectroscopy



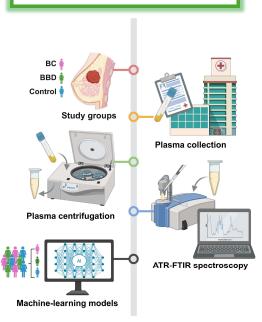
Machine learning (ML)

promising multi-omics tool for analyzing biofluids, which provides a minimally invasive source of tumor biomarkers

#### **OBJECTIVES**

To evaluate the diagnostic potential of ATR-FTIR + ML algorithms in distinguishing healthy controls and benign breast diseases (BBD) from early-stage BC patients using plasma.

## **METHODS**



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### **RESULTS**

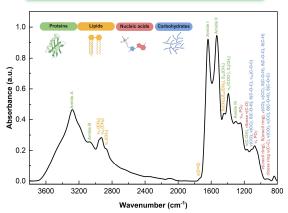


Figure 1. Representative ATR-FTIR spectrum of plasma from the BC group. Tentative of biomolecular peak assignments, where  $\nu$  = stretching vibrations,  $\delta$  = bending vibrations, s = symmetric vibrations and as = asymmetric vibrations. Created with BioRender.com.

Table 1. Machine learning performance using ATR-FTIR spectra of plasma for differentiating the study groups. The table shows the spectral regions for the best algorithm and the relative metrics. The best performance is in bold. BC= Breast cancer; BBD= Benign breast disease.

Region	Algorithm	ACC	SEN	SPE	AUC
CONTROL vs. BC					
3600-3000 cm <sup>-1</sup>	AdaBoost	0.8	0.833	0.769	0.801
3000-2800 cm <sup>-1</sup>	Naive Bayes	0.88	0.917	0.846	0.859
1800-800 cm <sup>-1</sup>	AdaBoost	0.88	0.833	0.923	0.878
3000-2800 + 1800-800 cm <sup>-1</sup>	Naive Bayes	0.88	0.917	0.846	0.897
1700-1480 cm <sup>-1</sup>	Gradient Boosting	0.92	0.917	0.923	0.846
1200-800 cm <sup>-1</sup>	Naive Bayes	0.8	0.833	0.769	0.811
BBD vs. BC					
3600-3000 cm <sup>-1</sup>	Gradient Boosting	0.778	0.75	0.8	0.767
3000-2800 cm <sup>-1</sup>	Gradient Boosting	0.667	0.667	0.667	0.644
1800-800 cm <sup>-1</sup>	Gradient Boosting	0.741	0.75	0.733	0.728
3000-2800 + 1800-800 cm <sup>-1</sup>	AdaBoost	0.778	0.75	0.8	0.775
1700-1480 cm <sup>-1</sup>	AdaBoost	0.741	0.75	0.733	0.742
1200-800 cm <sup>-1</sup>	AdaBoost	0.889	0.917	0.867	0.892
CONTROL vs. BBD					
3600-3000 cm <sup>-1</sup>	Neural Network	0.821	0.8	0.846	0.892
3000-2800 cm <sup>-1</sup>	Gradient Boosting	0.821	0.8	0.846	0.785
1800-800 cm <sup>-1</sup>	Naive Bayes	0.893	0.867	0.923	0.928
3000-2800 + 1800-800 cm <sup>-1</sup>	AdaBoost	0.964	1	0.923	0.962
1700-1480 cm <sup>-1</sup>	Neural Network	0.929	0.867	1	0.938
1200-800 cm <sup>-1</sup>	Neural Network	0.893	0.867	0.923	0.969

# CONCLUSION

This study reinforces the potential of plasma as an informative source for liquid biopsy, which may contribute to earlier BC detection, improve clinical decision-making and reduce unnecessary invasive procedures in both benign and malignant breast alterations.





**ACKNOWLEDGMENTS** 





