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# **Raman Spectroscopy for Detection of the Boundary Zone of Tumor in Colorectal Cancer**

## **A Precision Tool for Real-Time Definition of Surgical Margins**

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# 1. Background-Aim

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## **Colorectal cancer (CRC)**

is the fourth most common and the third deadliest cancer worldwide. Colon cancer treatment often involves open surgical resection as the primary treatment for localized disease. Accurate detection of tumor-positive margins is crucial to minimize the removal of healthy tissue during surgery.

## **Raman spectroscopy (RS)**

is an advanced analytical optical technique that has shown significant promise in the detection and diagnosis of colorectal cancer (CRC). Raman spectra can be used as a fingerprint for colorectal tissues, with patterns indicating anatomical and molecular changes. Machine Learning (ML) algorithms are commonly used in the analysis of spectral data for cancer prediction.

## **This study focuses on**

detecting the tumor boundary zone of CRC using Raman spectroscopy, with the goal of making RS a future clinical tool for surgeons.

## 2. Materials & Methods

### Patients & Samples:

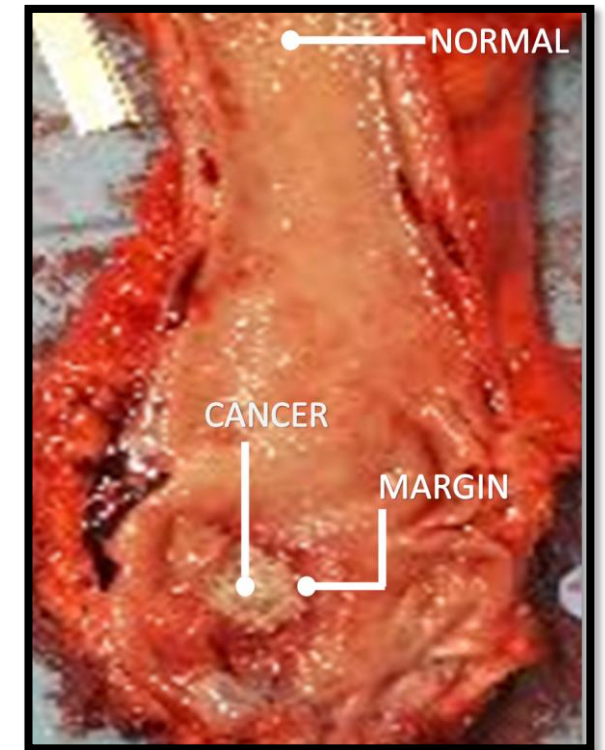
- In this work, human colorectal specimens were collected from 23 patients who underwent open surgery.
- Three different types of samples were used: normal colon area, cancerous area, and tumor boundaries.

### Raman spectroscopic measurements:

Micro-Raman spectra were recorded using a Renishaw Invia micro-Raman system operating at 785 nm.

### Data analysis:

The extracted data were analyzed spectroscopically and utilized for a classification task using machine learning algorithms. Specifically, we used Partial least squares-discriminant analysis (PLS-DA) that is a supervised dimensionality reduction method.



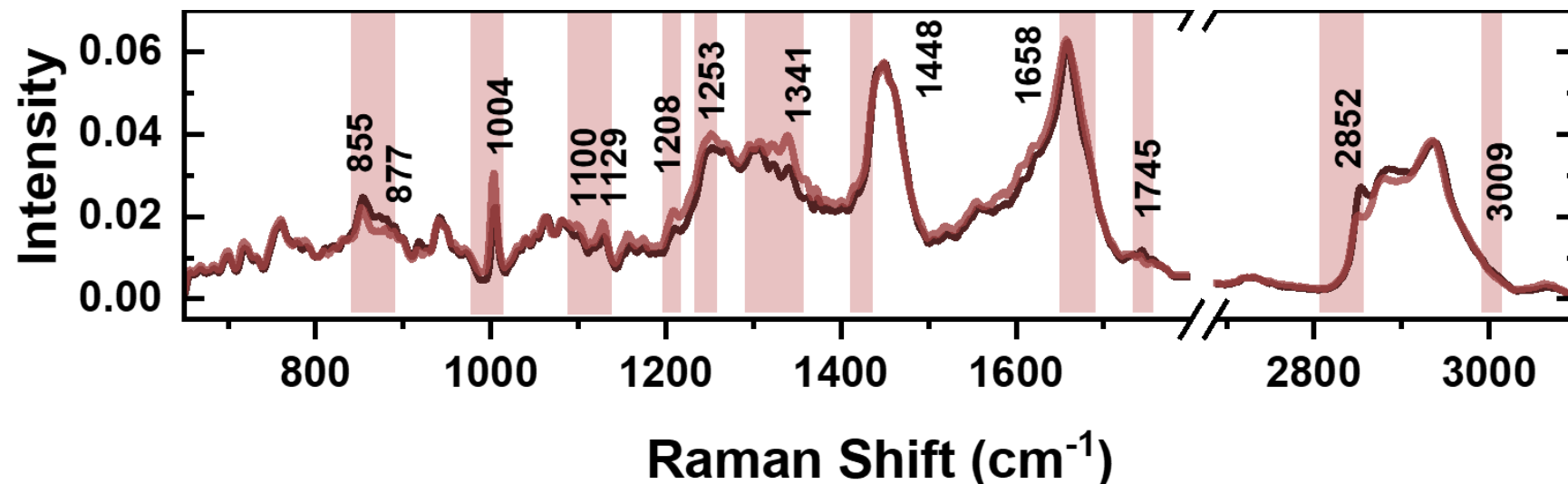
### 3. Results

## Comparison study between normal and cancerous spectral data

- Numerous significant differences between normal (black line) and cancerous (red line) spectra.
- Collagen and lipid spectral features could serve as markers for the spectroscopic detection of cancer.
- Raman spectra reveal protein overexpression in cancer tissues.
- The table includes the attributions of the significant Raman bands.

Frequencies	Assignments of Raman bands
855 $\text{cm}^{-1}$	Tyrosine, collagen (proline)
877 $\text{cm}^{-1}$	Tryptophan, hydroxyproline
1004 $\text{cm}^{-1}$	Phenylalanine
1100 $\text{cm}^{-1}$	Lipids, DNA
1129 $\text{cm}^{-1}$	Proteins
1208 $\text{cm}^{-1}$	Proteins (phenylalanine)
1253 $\text{cm}^{-1}$	Proteins (Amide III), collagen
1341 $\text{cm}^{-1}$	Lipids, proteins (aliphatic amino acids, tryptophan)
1448 $\text{cm}^{-1}$	Lipids, proteins (aliphatic amino acids)
1657 $\text{cm}^{-1}$	Proteins (Amide I), lipids (unsaturated fatty acids)
1745 $\text{cm}^{-1}$	Lipids (esters)
2852 $\text{cm}^{-1}$	Lipids, fatty acids
3009 $\text{cm}^{-1}$	Lipids

#### MEAN SPECTRA:



Shaded regions of the spectrum indicate statistically significant differences. The frequencies of the corresponding peaks are also marked above the spectrum.

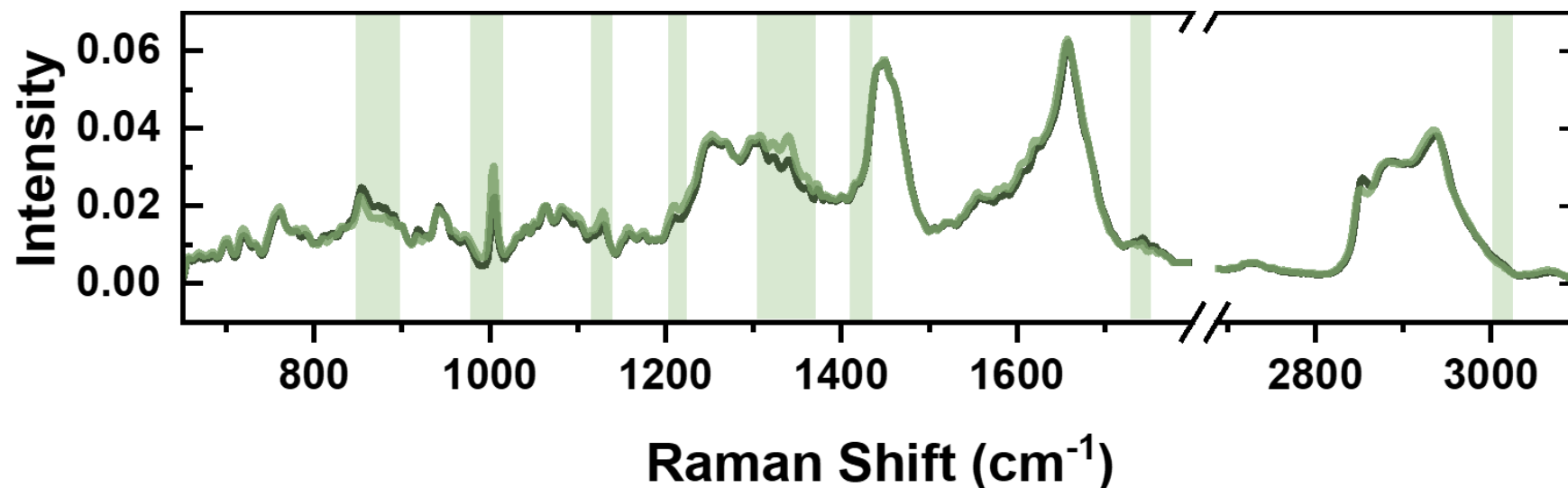
### 3. Results

## Comparison study between spectra of normal tissues and tumor boundaries

- The samples of boundaries (green line) also present several significant differences in comparison with the normal samples (black line).
- Fewer differences than in the comparison between normal and cancerous spectrum: In the boundary zone of the tumor, malignant and non-malignant cells coexist.
- The table shows the types of differences:  
↓: reduction of peak intensity, ↑: increase of peak intensity, ↔: peak broadening, →: peak shift, -: not a significant difference.

Frequency positions (cm <sup>-1</sup> )	NORMAL vs CANCER	NORMAL vs MARGIN
855	↓	↓
877	↓	↓
1004	↑	↑
1100	↑	-
1129	↑	↑
1208	↑	↑
1253	↑	-
1341	↑	↑
1448	→	→
1658	↔	-
1746	↓	↓
2852	↓	-
3009	↓	↓

#### MEAN SPECTRA:



Shaded regions of the spectrum indicate statistically significant differences.

### 3. Results

## Machine learning analysis on cancer and margin spectral datasets

- Two classification tasks were performed, the first for cancer discrimination and the second for tumor boundary discrimination.
- The experiments yielded an accuracy of 86% for the cancer data and 84% for the tumor boundary data, which is comparable to other state-of-the-art results.
- The classification accuracy for the tumor boundary data was found to be similar to that of the cancer data, suggesting that Raman spectroscopy can efficiently discriminate even heterogeneous tissue regions.

<b>Classification Metrics for PLS Model</b>			
	Accuracy	Sensitivity	Specificity
<b>NORMAL vs CANCER</b> [cancer detection]	86%	87%	86%
<b>NORMAL vs MARGIN</b> [margin detection]	84%	74%	89%

## 4. Conclusions

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### Summary:

- Spectral differences between cancerous and healthy tissues were identified.
- The study demonstrated that Raman spectroscopy can effectively differentiate between healthy and cancerous tissues, and the statistical classification results were highly satisfactory (Accuracy: 86%, Sensitivity: 87%, Specificity: 86%).
- The coexistence of healthy and cancerous cells in the marginal region of the tumor is evident in the spectral results.
- The classification accuracy between normal spectra and spectra derived from boundaries achieved 84%, demonstrating the technique's capability to distinguish cancer cells within heterogeneous regions.

### Conclusion:

This study demonstrates that RS is a technique that could not only be utilized in CRC diagnosis but also shows potential as a valuable tool for confirming clear margins in CRC in real-time during open surgery.

## 5. References

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