

# RAMAN SPECTROSCOPY AND FUSION CLASSIFICATION

## TO IDENTIFY PLASTIC RECYCABLES TARGETING MICROPLASTICS

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### Abstract

Identification of plastic type for microplastic particles (size range of 0.001 mm – 5 mm) is vital to understand the sources and consequences of microplastics in the environment. Fourier-transform infrared and Raman spectroscopy are two dominating techniques used to identify microplastics. The most common method to identify microplastics with spectroscopic data is library searching, a process that utilizes search algorithms against digital databases containing spectra of various plastics. Presented in this study is a new method to utilize spectroscopic data called fusion classification. Fusion classification consists of merging multiple non-optimized classification methods (classifiers) to assign samples into categories (classes). The purpose of this study is to demonstrate the applicability of fusion classification to identify microplastics..

### Objective

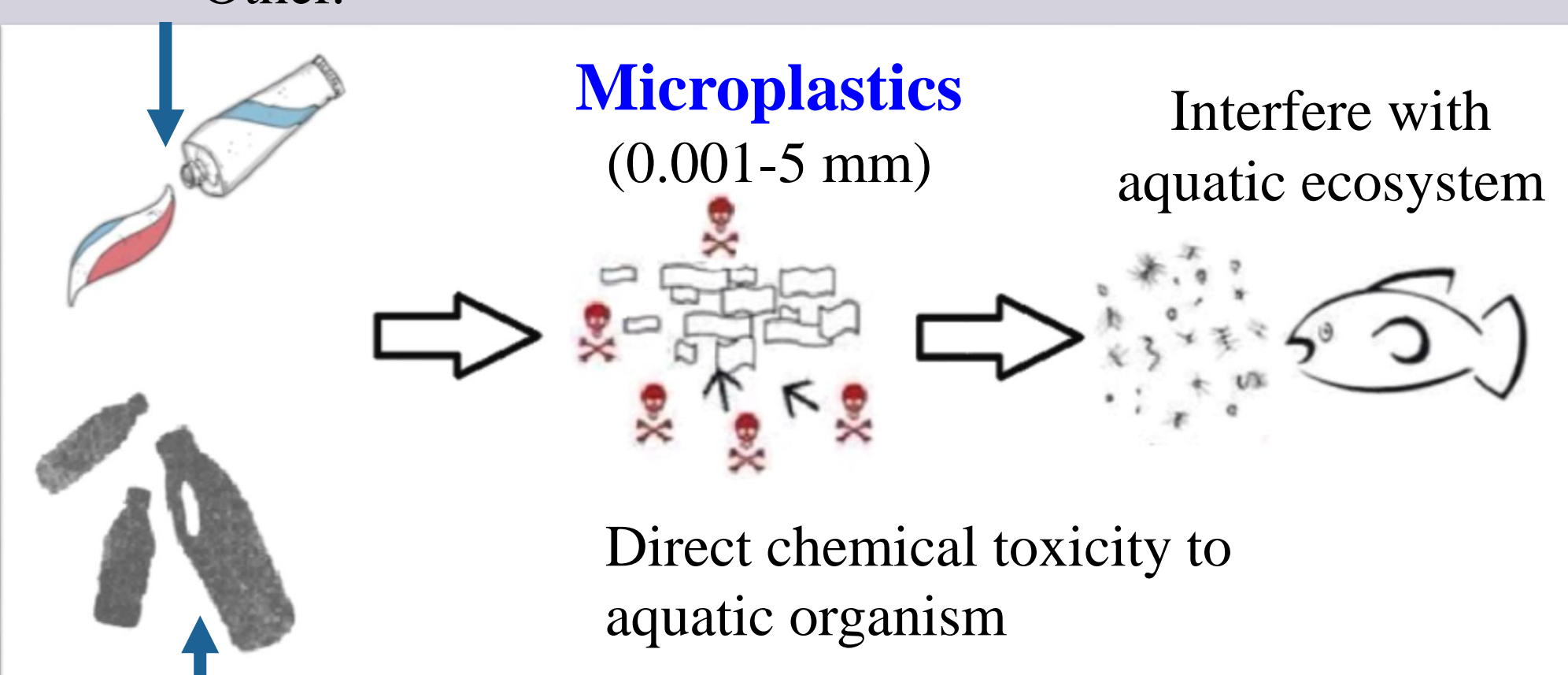
- Identify plastic recyclables using fusion classification to improve microplastic identification accuracy

### Background

- > 4.5 billion metric tons of plastic produced in 2015.
- 36.2 billion metric tons projected by 2050.
- 4.8 – 12.7 million metric tons enter the ocean annually.

### Primary Source

- Intentionally engineered:
  - Microbeads used in cosmetic products.
  - Other.

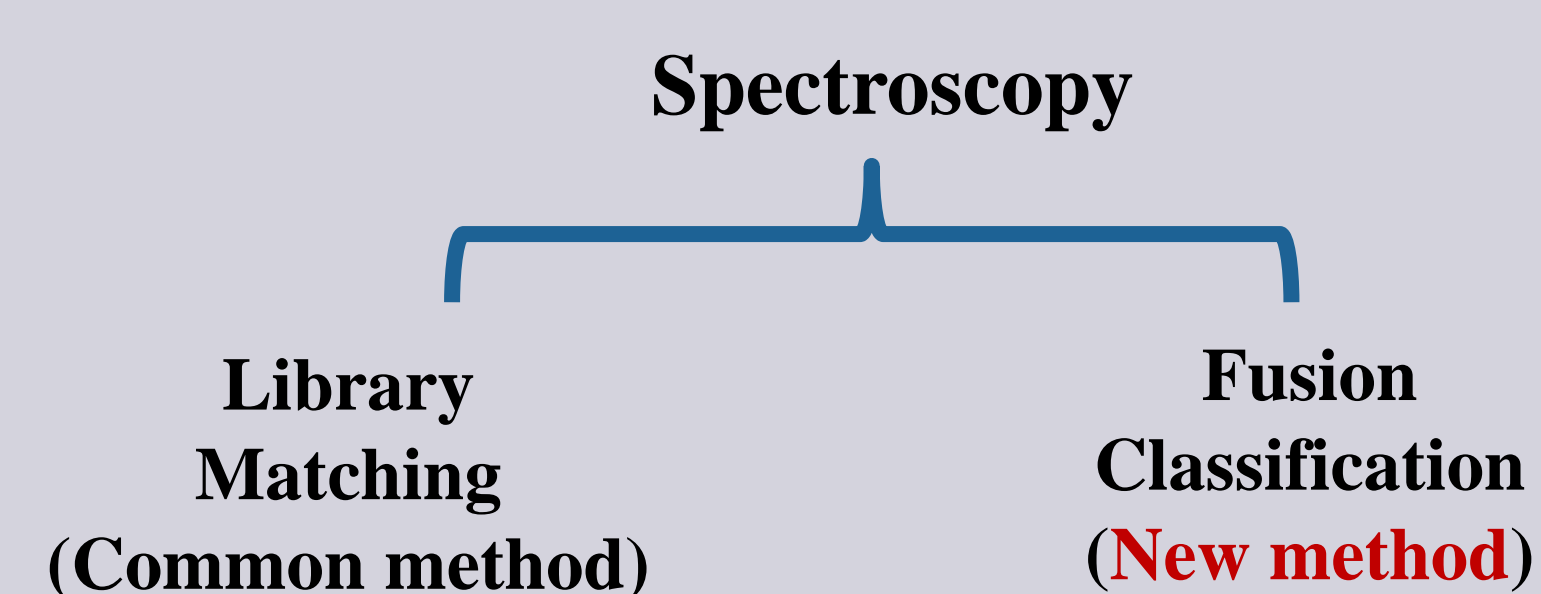


### Secondary Source

- Consequence of:
  - Photolytic, mechanical, thermal and biological degradation of any plastic goods.

### Limitations of Spectroscopic Analysis

- Interference of spectroscopic data caused by:
  - Sediments
  - Degree of degradation
  - Additives such as dyes, antioxidants, etc.



### Approach

#### Fusion Classification

- Assigning a sample to a category (class) using classification methods (classifiers).
- 17 classifier used in order to:
  - Reduce risk misidentification.
  - Improve classification accuracy.
  - Overcome limitations of stand alone classifiers.

Table 1: Classifiers

Classifiers with Tuning Parameter	Classifiers with No Tuning Parameter
<ul style="list-style-type: none"> <li>Mahalanobis distance (MD)</li> <li>Q-residual (Qres)</li> <li>Sine</li> <li>Divergence criterion (DC)</li> <li>Partial least squares discriminant analysis (PLS2-DA)</li> <li>k nearest neighbor (kNN)</li> </ul>	<ul style="list-style-type: none"> <li>Euclidean distance</li> <li>Procrustes analysis unconstrained (PA)</li> <li>Inner product correlation</li> <li>Determinant</li> <li>Procrustes Analysis constrained (PA<sup>c</sup>)</li> <li>Cosine</li> <li>Extended inverted signal correction difference (EISCD)</li> </ul>

#### Classifiers with Tuning Parameter

- Tuning parameter based on a number value:
  - PLSDA - latent variables (LVs)
  - kNN - number of nearest neighbors
  - MD, Qres, DC, and Sine – eigenvectors

#### Classifiers with No Tuning Parameter

- Determine the degree of similarity for a target sample compared to each class mean.
- Threshold selection required.

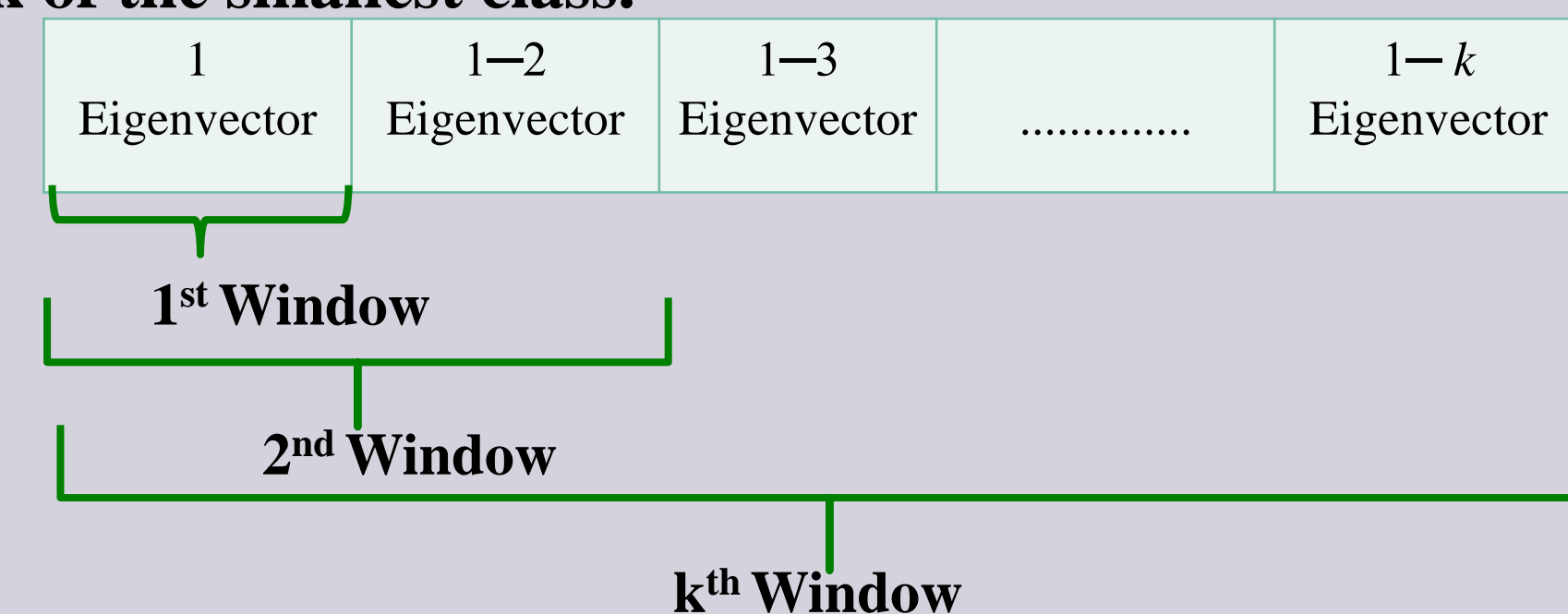
### Our Method

- No training (optimization), weights, or threshold selection of each classifier:
  - Uses raw values.
- Optimization based on a window of respective tuning parameter values:
  - Simplifies classification ensemble

#### Tuning Parameter Window Selection

- Rule of thumb:
  - 99% information of class (X) is captured.
  - LVs and eigenvectors are not excessively composed of noise.
  - Maximum window size is based on the rank (k) of smallest class

Example: Eigenvector based single classifier. Where k is the rank of the smallest class.



Brett Brownfield, Tony Lemos, and John H. Kalivas *Analytical Chemistry* 2018 90 (7), 4429-4437

### Fusion Rule: SUM

- Values normalized to unit length.
- Samples assigned to class with lowest sum.

#### Classifier

- 1–5 PLS2-DA
- 6–10 kNN
- 11–5 MD
- 16–20 Sinθ
- 21–25 Q-res
- 26–30 DC
- 30–41 Non-traditional classifiers

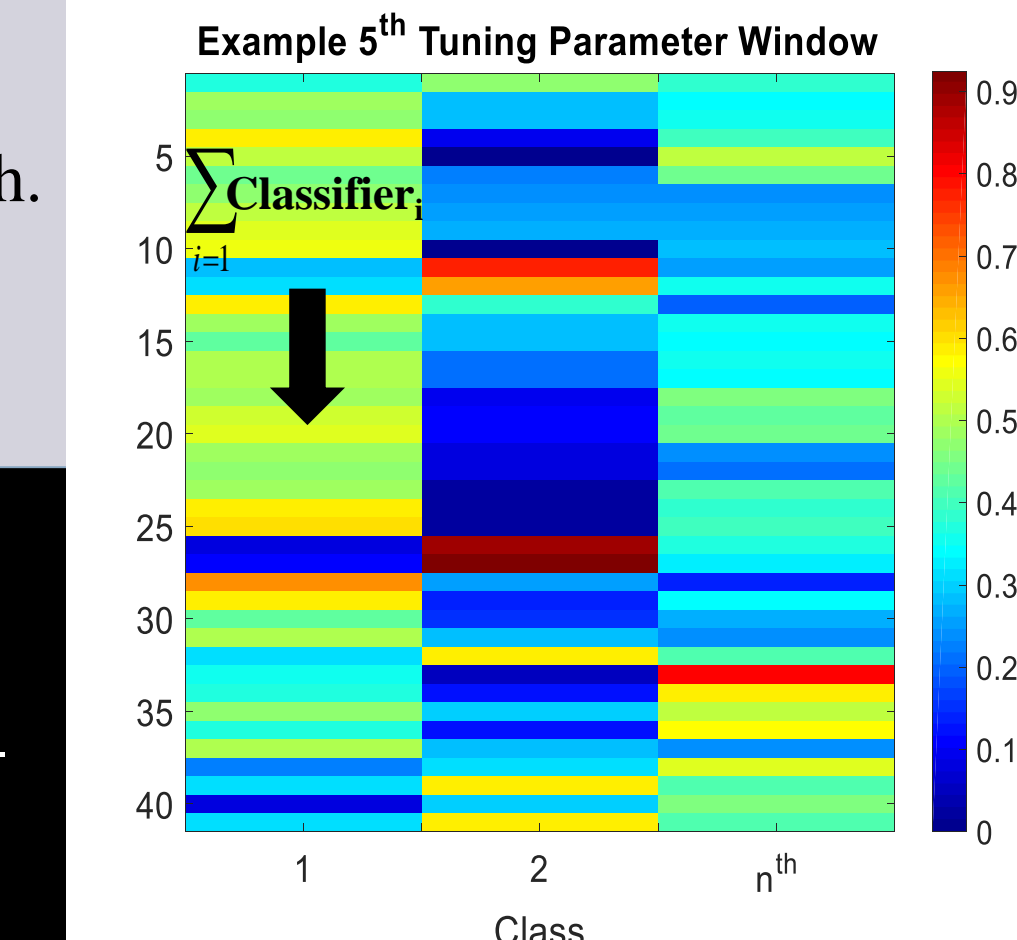


Fig. 1: Classification of a sample at the 5<sup>th</sup> tuning parameter window

### % Performance Parameters

	Belong to Class	Does not belong to Class	
Positive Result	True Positive (TP)	False Positive (FP)	% Accuracy = $\frac{TP+TN}{TP+TN+FP+FN} \times 100$
Negative Result	False Negative (FN)	True Negative (TN)	% Sensitivity = $\frac{TP}{TP+FN} \times 100$
			% Specificity = $\frac{TN}{TN+FP} \times 100$

### Experimental Design

Table 2: Sample information breakdown 103 samples and 188

Class #	Plastic Types	# of Samples	# of Spectra
1	Polyethylene Terephthalate (PET)	28	40
2	High density polyethylene (HDPE)	23	38
3	Polyvinyl chloride (PVC)	4	17
4	Low density polyethylene (LDPE)	18	28
5	Polypropylene (PP)	11	28
6	Polystyrene (PS)	19	37

Allen, V., Kalivas, J. H., & Rodriguez, R. G. (1999). Post-Consumer Plastic Identification Using Raman Spectroscopy. *Applied Spectroscopy*, 53(6), 672-681.

### Data Sets

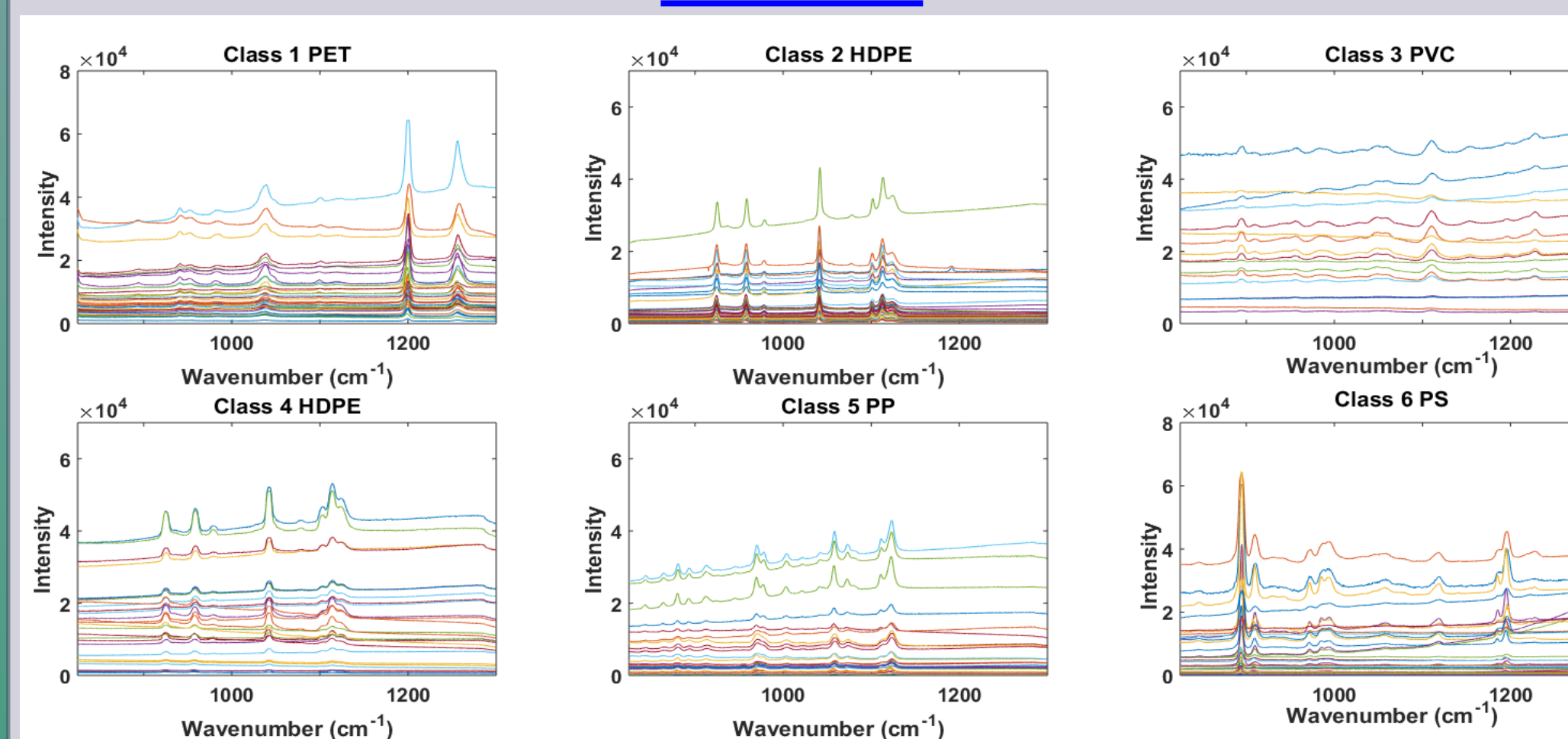


Fig. 2: Raman spectral data for each plastic type i.e. PET, HDPE, PVC, HDPE, PP and PS.

### Results

Table 3: Overall (188) library matching results

% Performance Parameter	No Threshold	Threshold Cos θ ≥			
		0.70	0.75	0.85	0.90
Accuracy	96.3	92.3	89.9	58.7	0
Sensitivity	96.4	85.8	81.7	41.6	0
Specificity	50	100	100	100	0

### Comparing fusion to frequently used stand alone classifiers

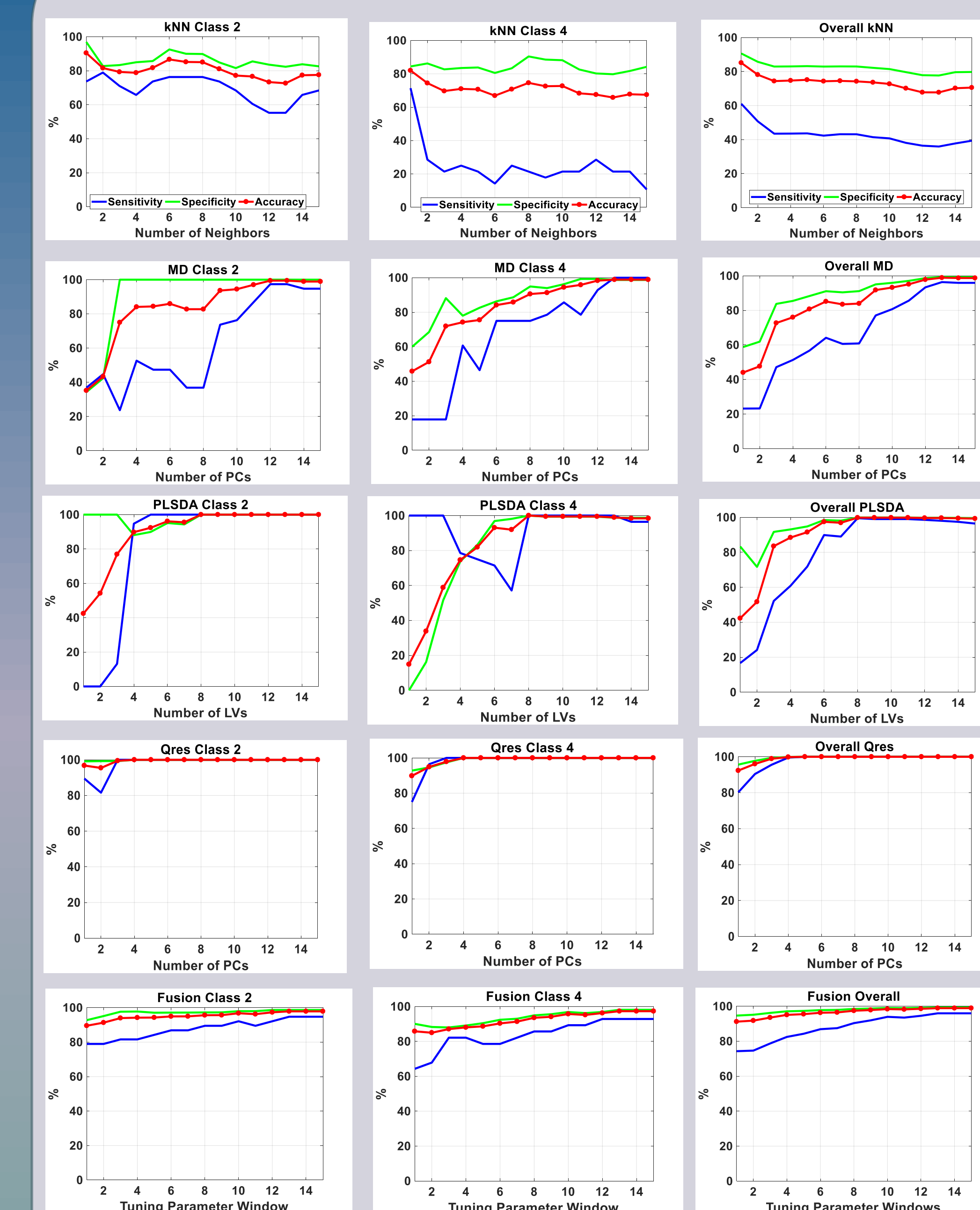


Fig.2: Each figure shows accuracy (red), sensitivity (blue) and specificity (green)

### Conclusion

#### Library Matching

- Threshold selection:
  - Value is subjective
    - Too high— risk not identifying samples.
    - Too low— risk misidentification of samples.

#### Fusion Classification

- No threshold selection for individual classifiers:
  - Simplifies classification.
  - Window size is used instead based on:
    - Class with lowest rank.
  - Higher accuracy, sensitivity and specificity than standalone classifiers:
    - Reduces the risk of misclassifying abnormal samples.
  - Identification is based on available classes.

### Future Work

- Apply fusion classification to identify:
  - Physically degraded colored microplastic using Micro-Raman and Micro-FTIR.
  - Microplastic particles in the Snake river

### Acknowledgement

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