

# IMPROVING DEGRADED MICROPLASTIC IDENTIFICATION USING FTIR-BASED CNN MODELING

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## INTRODUCTION:

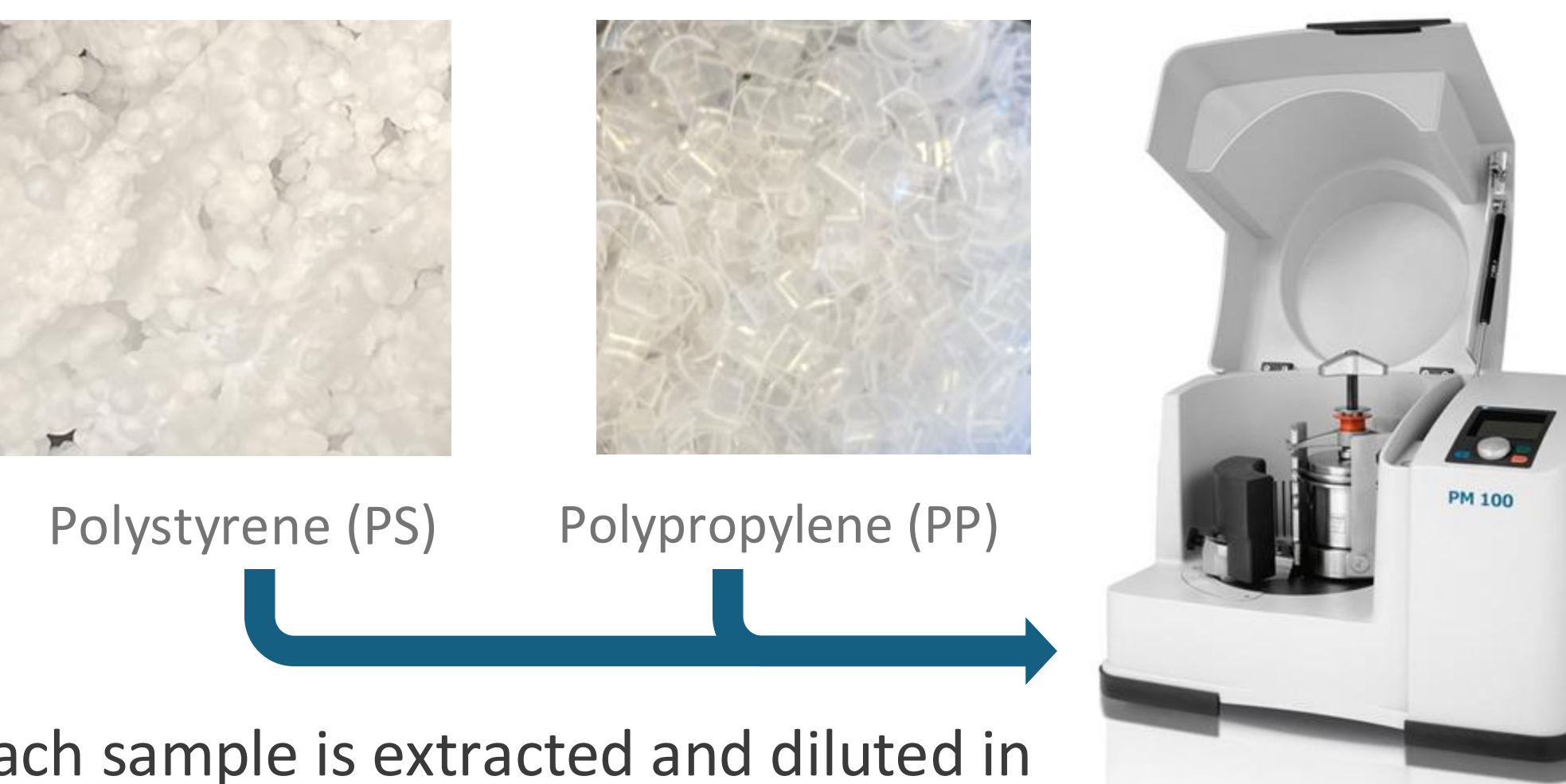
Identifying plastics in the environment is difficult as their chemical compositions change due to ultra-violet (UV) exposure. This work captures **the degradation of plastics in a controlled environment and trains a Convolutional Neural Network (CNN) model to better identify these plastics.**

- **Fourier Transform Infrared (FTIR) Spectroscopy:** Uses infrared light which is absorbed at different frequencies in the atomic bonds of a sample. This absorption is observed in spectra data and helps to identify different bonds.
- **Is more sensitive to the oxidation process polymers undergo during the degradation process.**

## METHODOLOGY

Sample Preparation

Samples are prepared into 2-5mm sample sizes and then grinded down in a Planetary Ball Mill PM 100 <sup>1</sup>.



Each sample is extracted and diluted in two beakers with 300mL of deionized water.



Samples are degraded in a UV weathering chamber, at an intensity of 340 UVA. Samples are taken out every 4 days and undergo centrifugation at 3000-3500 rpm. 3-10μL of concentrated sample is dried and scanned using FTIR Spectroscopy.

Fisher Scientific Accuspin 8C Clinical Centrifuge

Stage of the FTIR

- Additional samples are scanned including:
- polystyrene (PS)
  - polyethylene terephthalate (PET)
  - polypropylene (PP)
  - polyethylene (PE)
  - polycarbonate (PC)
  - nylon (PA)
  - polyvinyl chloride (PVC)
  - polyester

UV-light Degradation

Data Collection

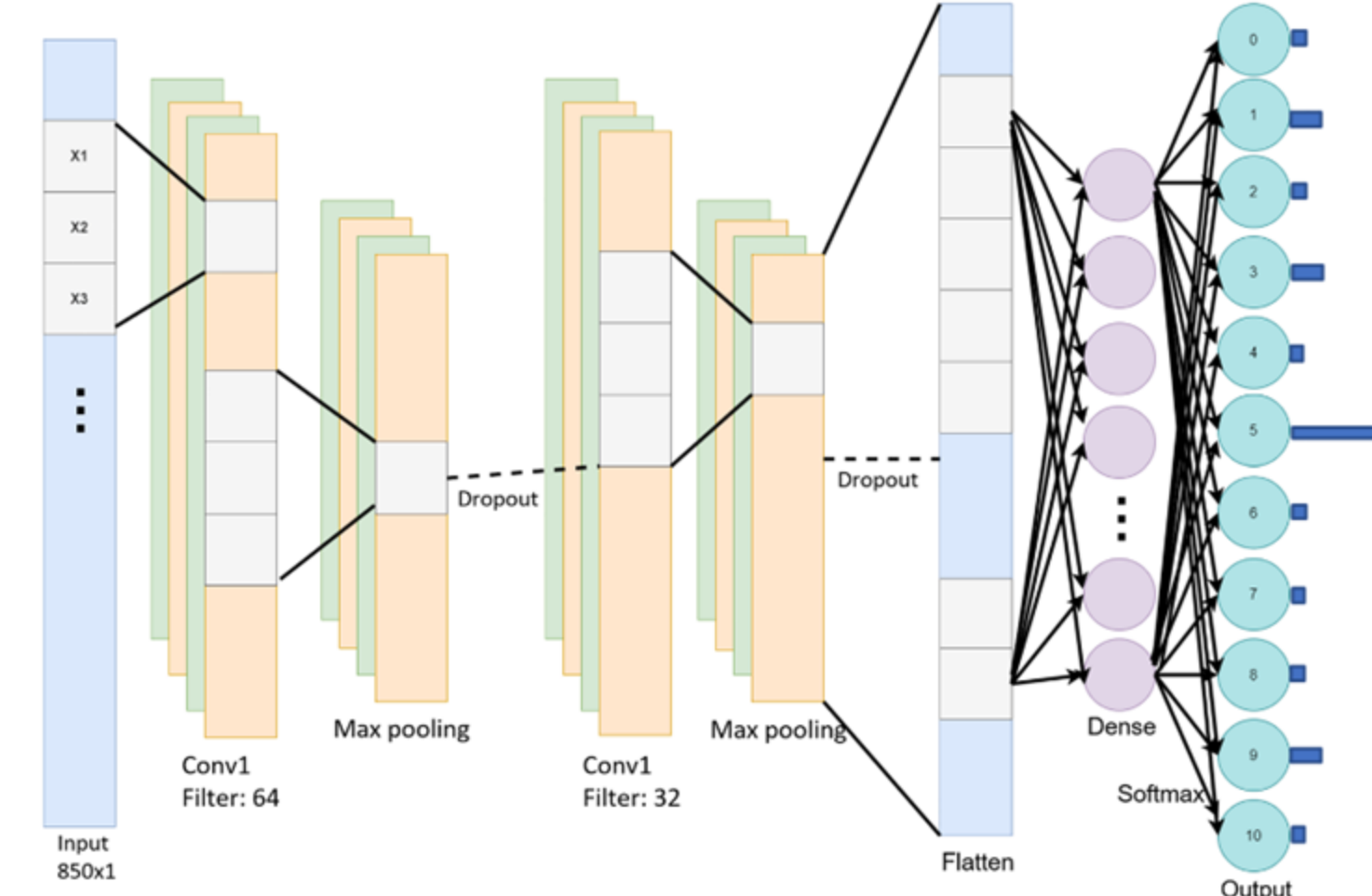
## METHODOLOGY:

Model Training

A dataset focused on these eight plastics and polyurethane is created with FTIR spectra data from the FLOPP and FLOPP-e library <sup>2</sup> and sample scans.

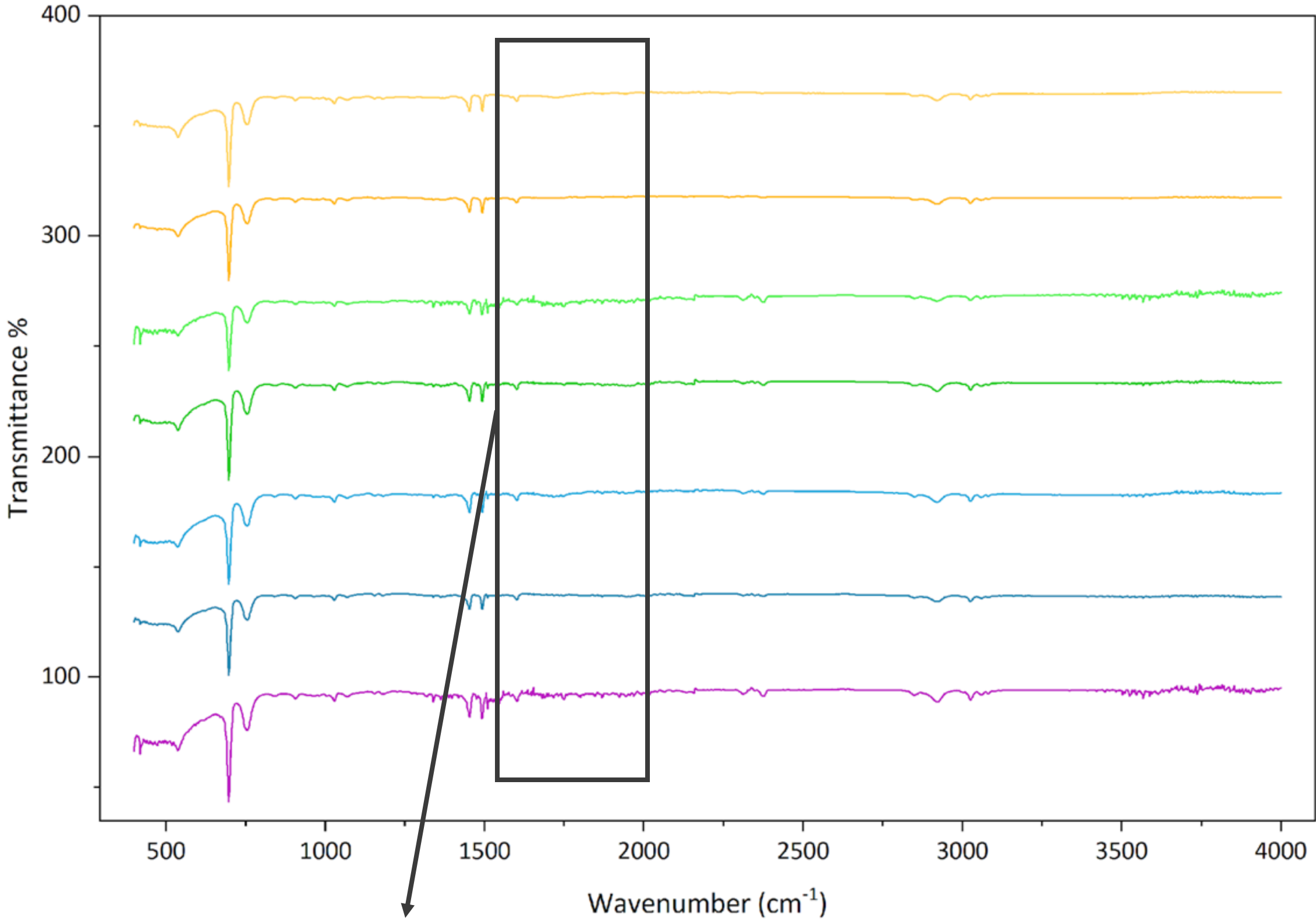
- # of spectrums used: 363
- Spectra range used: 675-4000 cm<sup>-1</sup>

This data is used to train a 1-D CNN model. The CNN structure used is shown above <sup>3</sup>.



## RESULTS:

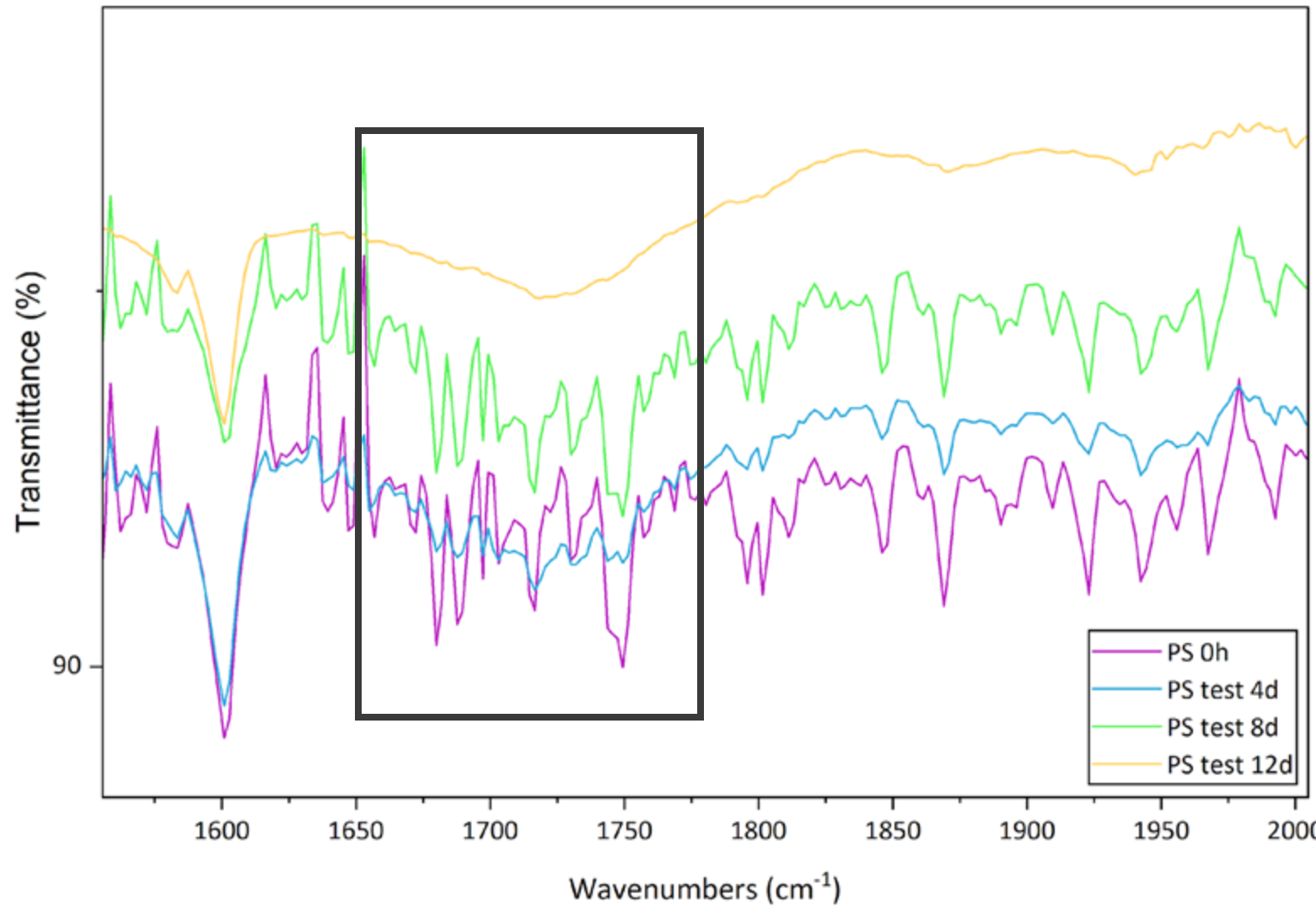
Polystyrene During UV Degradation



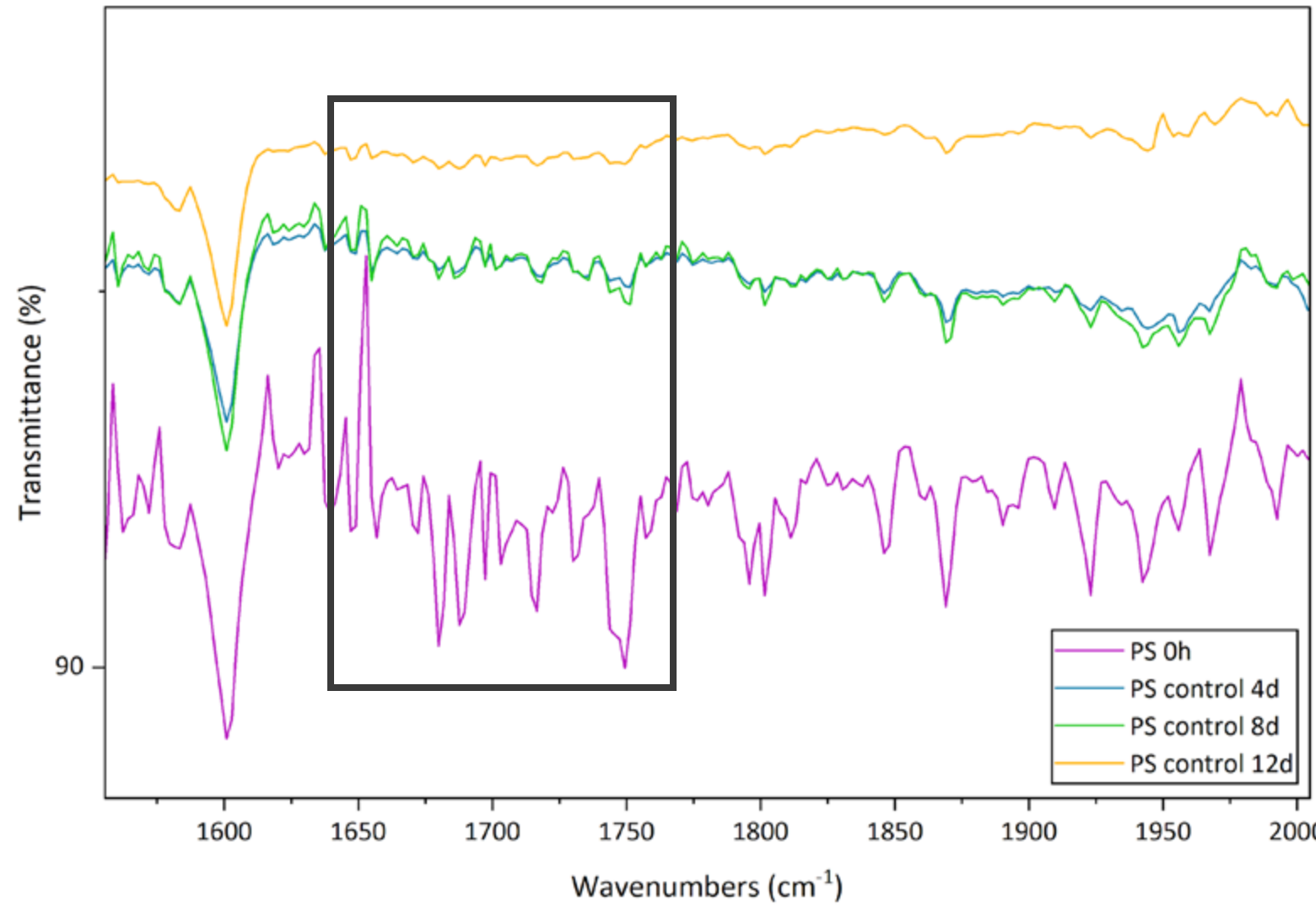
FTIR spectra of polystyrene UV degradation of both groups during a 12-day period.

- PS 0h
- PS control 4d
- PS test 4d
- PS control 8d
- PS test 8d
- PS control 12d
- PS test 12d

Polystyrene Test Group During UV Degradation



Polystyrene Control Group During UV Degradation



The peak at ~1725 cm<sup>-1</sup> indicates a carbonyl bond is present, a sign of oxidation which is not present in the control group. This agrees with other literature where carbon-oxygen bonds occur during UV exposure <sup>4</sup> and validates the methodology used.

## ACKNOWLEDGEMENTS:

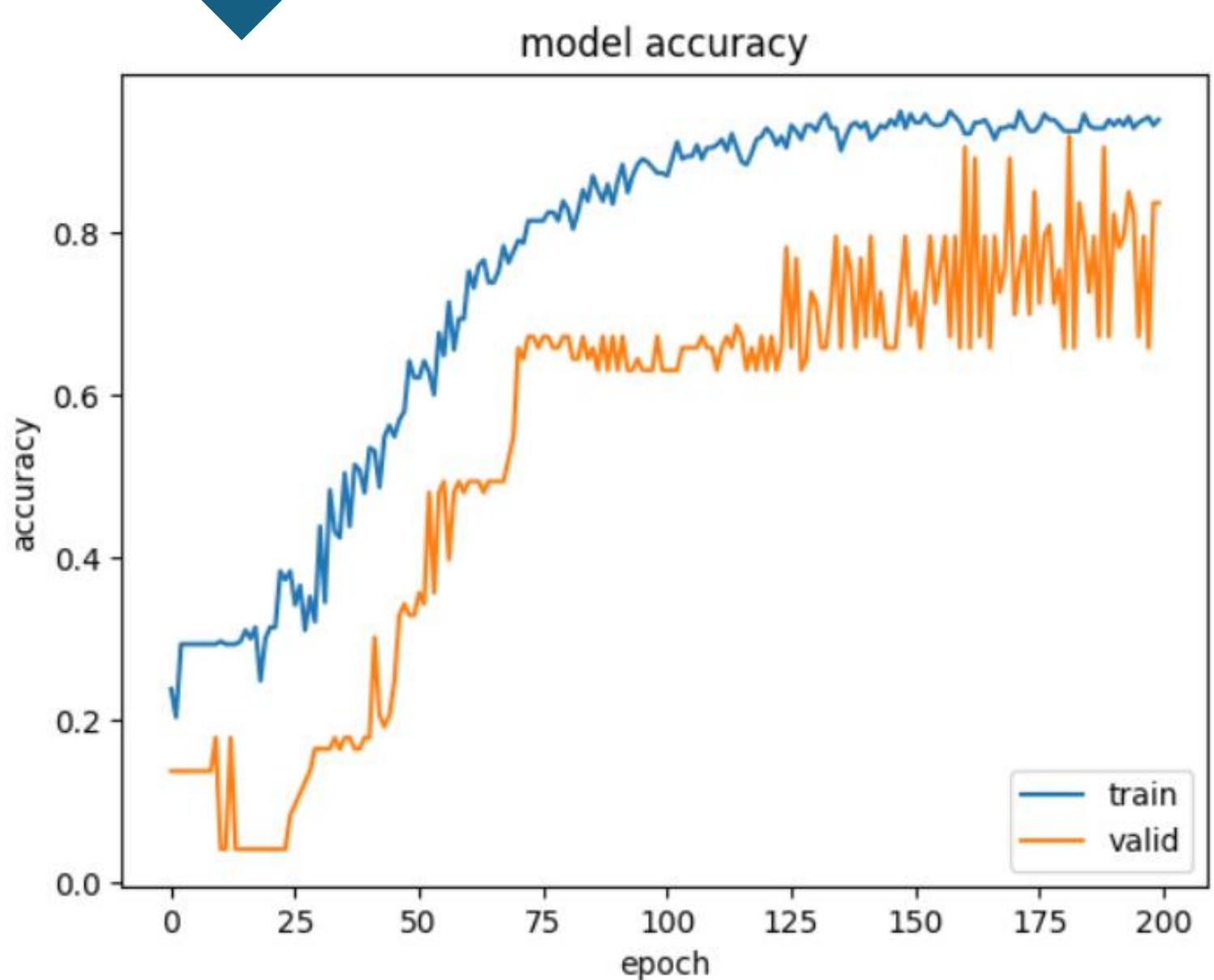
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## RESULTS:

Below showcases the split in validation (20%) and training sets.

```
history=model.fit(x_train, y_train, batch_size=64, epochs=200, validation_split = 0.2, verbose=1)
```

Below are the training and validation accuracy curves graphed.



**Training:** How well the model performs on training data.

**Validation:** How well the model performs on new data.

**Epochs:** How many times the model looks at the data.

Training Accuracy: 94.6%

Validation Accuracy: 83.6%

## FUTURE WORK:



- Conduct data augmentation to further improve accuracy
- Collect more data on different plastics (polyurethane)
- Further degrade existing samples and different plastics
- Use a saline solution (3.5% sodium chloride in a water solvent) and study the UV degradation of plastic samples in saltwater environments (left).

## REFERENCES:

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