

WHOSE PARTICLE IS IT?

PARTICLE FORENSICS IN THE MOUNT MAUNGANUI AIRSHED

Cherie Tollemache^{1,2A}, Anushka Elangasinghe^{1,2B}, Robert Deas^{2A}, Paul Baynham¹,
Hamesh Patel^{1,3,4}, Daniel Lynch¹, Alexander Elliot^{2A}, Joel Rindelaub^{2A},
Duncan McGillivray^{2A}, David Williams^{1,2A}, Jennifer Salmond^{2B} and Brett Wells¹

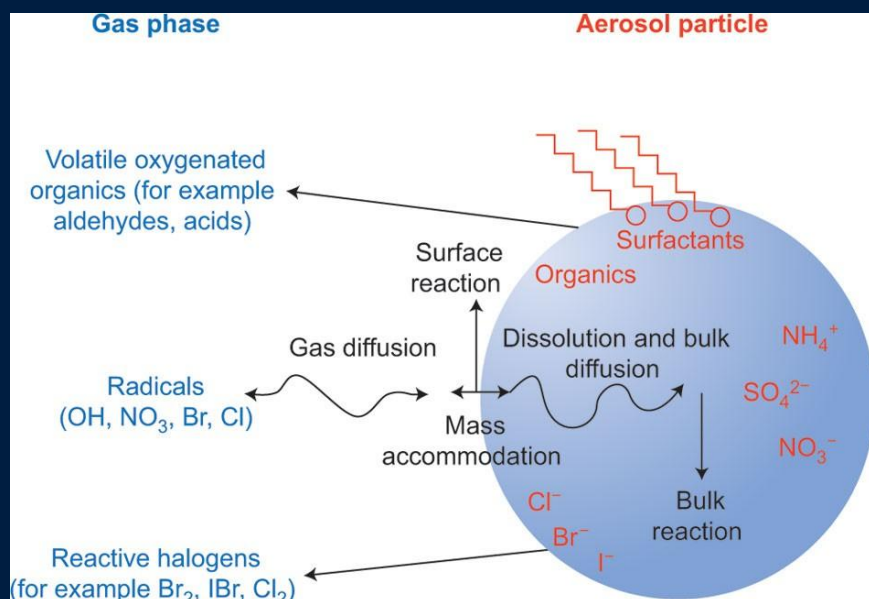
¹ Mote Limited, 40A George St, Mount Eden, Auckland, 1024, New Zealand

² University of Auckland, ^A School of Chemical Sciences and ^B School of Environment, Auckland 1011, New Zealand

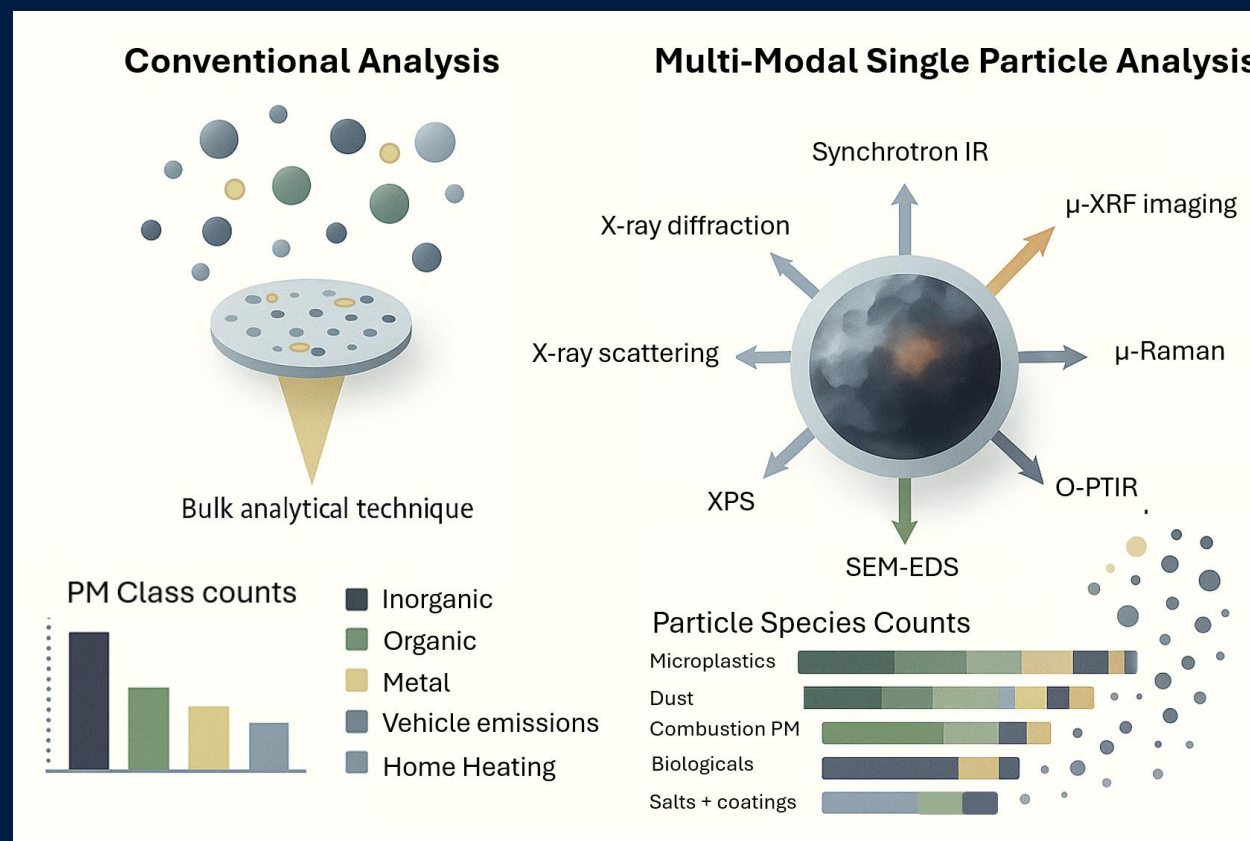
³ Queensland University of Technology, International Laboratory for Air Quality and Health, Brisbane, Australia

⁴ Mote Limited, Level 3, 240 Queen Street, Brisbane, Australia

The Chemistry of the particle matters!



Core Chemistry tells you where it has come from
 Surface chemistry tells you where it has been and
 how it will interact



'Heartbreaking' reality of Mount Maunganui's air quality

Alisha Evans, Local Democracy Reporter | Local Democracy Reporting
June 26, 2024 - 3:22pm

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Whareroa marae's environmental spokesperson Joel Ngatuere implored council candidates to understand the issues around air quality in Mount Maunganui.



Whareroa Marae = Ngāi Te Rangī iwi

Whareroa Marae & Community Impact

- Continuous air quality concerns raised since 2015
- Formal call for a managed retreat of heavy industry in 2020
- Progress stalled by existing industry user rights under the resource management act
- Averages ~500 pollution hotline calls to the regional council annually

Mount Maunganui

a coastal industrial zone classified as polluted in 2019



Mount air quality impacting health-review confirms



The review confirmed poor air quality at Mount Maunganui was creating significant health risks and premature deaths. Photo: Supplied.

Social & Health Costs: An estimated \$22 million annual economic burden caused by localized air pollution impacts.



Commission chair Anne Tolley said there was a "critical need" to find space for emitting businesses. Bay of Plenty Times Photo / Alex Cairns.

Industrial Land Scarcity: A severe shortage of alternative, appropriately zoned land to facilitate a managed retreat.



Mount Maunganui Industry Environmental Accord



signed by 33 businesses in the industrial zone, committing them to take further action to improve their environmental footprint.

Priority One reports the signatories have committed to 251 actions

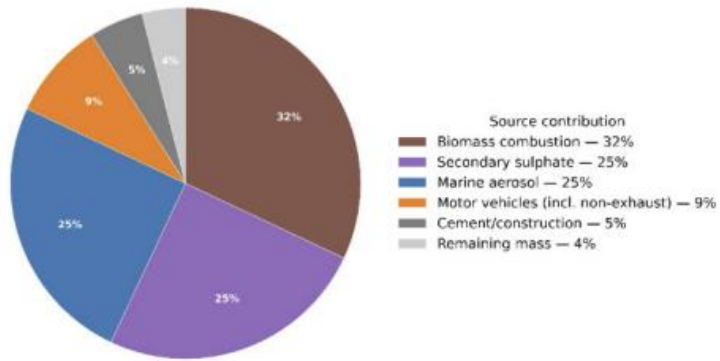
- 122 complete
- 120 underway or scheduled.

Community advocates ‘Clear the Air’ point out that these metrics represent voluntary, un-audited self-reporting calling for independently verified reductions in emissions and exposure

GNS – 2 year long XRF-PMF campaign

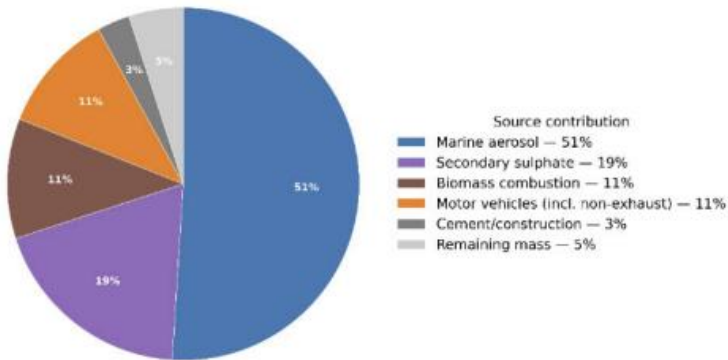
Davy, P. K., & Trompetter, W. J. (2023–2025). *Average source mass contributions to PM10 at the Mount Maunganui Library site*

Ranch Road - PM2.5 (average source contributions)

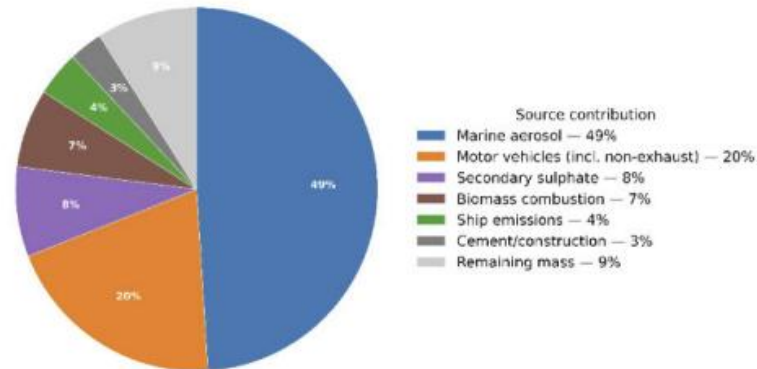


- The airshed-scale mass balance is established
- XRF provides elemental analysis of the whole particle population
- XRF does not provide molecular chemistry (organics) or oxidation state chemistry (inorganics) – models make a lot of assumptions

Ranch Road - PM10 (average source contributions)



Mount Maunganui Library - PM10 (average source contributions)

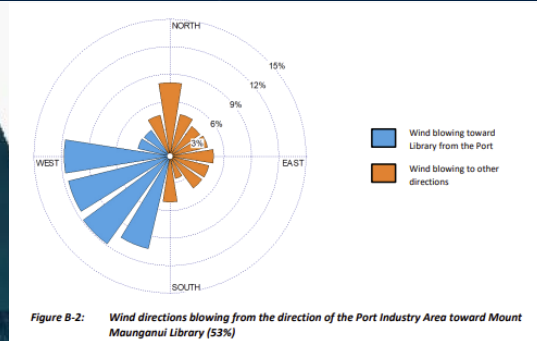


Broad source factors do not identify individual emitting activities or resolve the particle chemistry

Who is contributing most to the PM population?
Where in the operational processes do they come from?

Area Sources in Mount Maunganui

Marine aerosol
Sea salt, marine organic matter and biological material



Regional soil and windblown dust
Natural crustal minerals from exposed surfaces and surrounding land

Mount Maunganui Airshed

Microplastics
multiple diffuse and local sources
Not attributable to one activity
without polymer identification.

Biological aerosol
Pollen, mould spores, bacteria, plant fragments
Widespread but seasonally and spatially variable

Area Sources – Transport and Combustion

Expected PM:

- Exhaust soot - organic carbon
- Heavy fuel oil soot
- Aviation fuel soot
- Brake, tyre, rail and wheel wear
- Resuspended road and yard dust
- Secondary sulfate and nitrate
- Ultra-fine combustion metals



Point Sources

Asphalt production

- Aggregate minerals
- Combustion Soot
- Bitumen Associated organics

Concrete production and cutting

- Calcium rich mineral dust
- Silicate aggregates

Metalworks and Surface treatments

- Iron-rich steel particles
- Fe-Cr-Ni alloy particles
- Welding-fume metal oxides
- Abrasive blasting media
- Paint and coating fragments

Fertiliser industry

- Phosphate- and calcium-rich fertiliser dust
- Sulphate-bearing particles
- Potassium and chloride salts

Chemical and Resin Processing

- Pine-derived resinous PM
- Synthetic resin-associated PM
- Condensable organic aerosol
- Process-dependent secondary particles

Waste Management NZ

- Municipal refuse & organic dust
- Polymer & plastic fragments
- Condensed organic aerosols (COA)
- Polycyclic aromatic hydrocarbons (PAHs)

Salt Processing

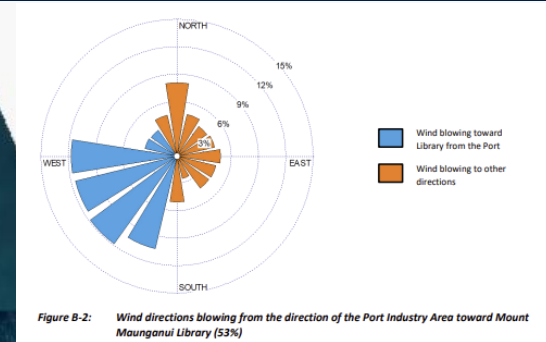
- NaCl
- Coarse Mineral Dust
- Boiler Combustion Soot
- Additive Chemical Particulates

Solid Product Handling and Storage

- Palm Kernel Extract
- Feed Wheat
- Soya meal
- Other plant derived particle

Logging

- Pine bark
- Pine wood
- Soil
- Road dust



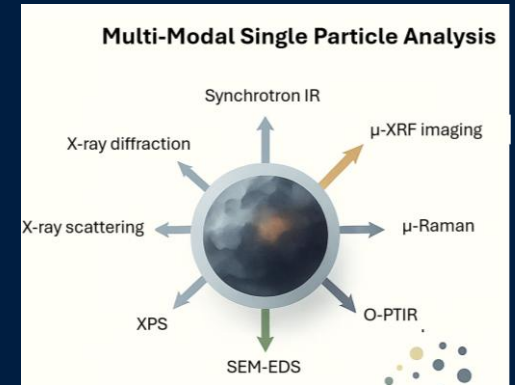
Mount Maunganui Airshed

Environmental Samples

Auckland – Coastal Urban

- TSP
- PM10
- Mt Eden (near CBD)
- Kyber Pass (heavy traffic)

Pilot Study – only one or two samples collected over 24 hours in a single location



Clean Marine Air – Coastal clean

- TSP
- PM10
- West Coast onshore winds

Mount Maunganui – Coastal Industrial Zone

- TSP
- PM10
- Three flow rates tested



Reference Samples – case study driven

Salts

- Sodium Chloride
- Sodium Sulphate
- Sodium Phosphate
- Sodium nitrate
- Magnesium sulphate
- Magnesium Phosphate
- Magnesium Nitrate
- Potassium Sulphate
- Potassium Nitrate

Combustion Soot

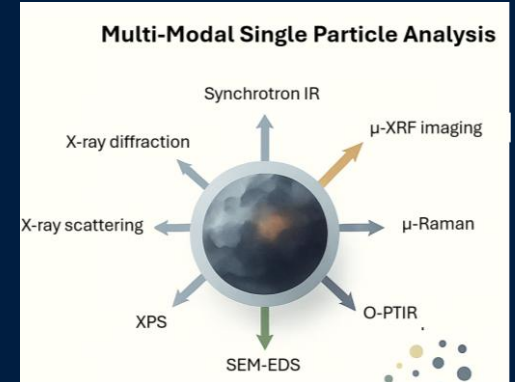
- Heavy Fuel Oil
- Diesel
- Coal
- Pine Wood
- Pallet (treated wood)
- Sparklers (fireworks)

Biologicals

- Sea Water
- Pollen
- Mould spores

Microplastics

- PET
- PMMA
- PE
- PP
- PLA
- PVC
- Polyurethane
- Polystyrene
- ABS
- Styrene butadiene



Animal Feeds

- Palm Kernel Extract
- Poultry Feed

Mineral Dusts:

- Unsealed Road Dust
- Dried River Bed Dust
- Fine Quarry Dust

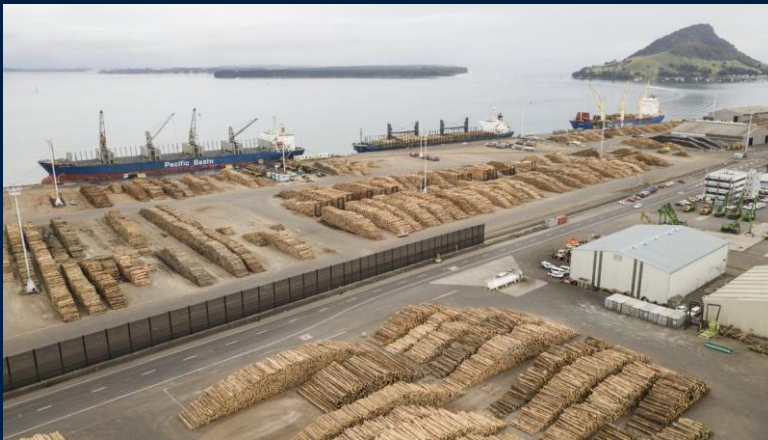
Vehicle emissions

- Tyre rubber
- Brake dust
- Diesel vehicle exhaust
- Petrol Vehicle Exhaust

Logging Emissions

- Pine Bark
- Debris from debarking row
- Material before logs loading
- Material after log loading
- Material after the sweeper

Organic PM – Logging Emissions or PKE??



What they share

- Plant-derived, lignocellulosic materials
- Generate fugitive dust during transfer, storage and transport
- Heterogeneous mixtures - not single chemical classes
- Overlapping carbohydrate, C–H and carbonyl IR signatures



What is Pine dust?

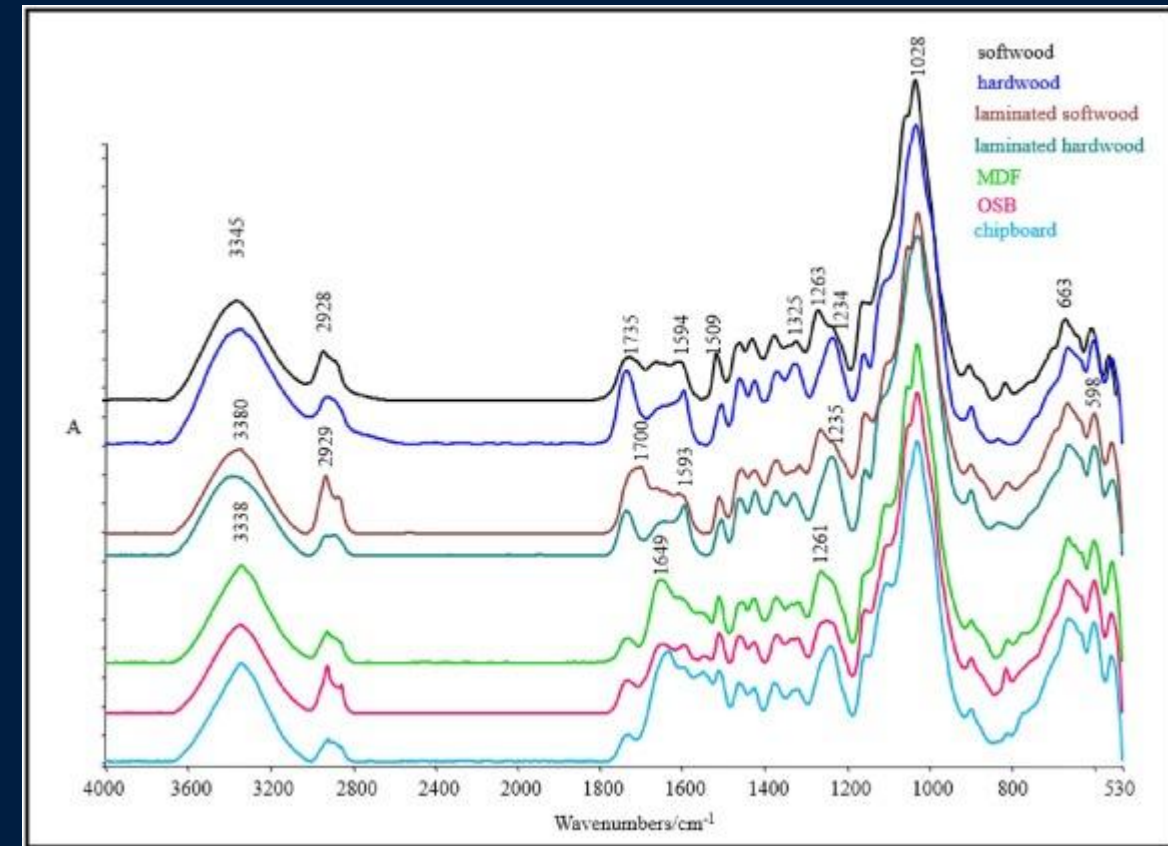
Core diagnostic windows:

- $\sim 1730\text{--}1740\text{ cm}^{-1}$ — carbonyl (hemicellulose / oxidation)
- $\sim 1505\text{--}1520\text{ cm}^{-1}$ — aromatic lignin
- $\sim 1600\text{ cm}^{-1}$ — aromatic ring vibrations (supporting)
- $\sim 1260\text{--}1270\text{ cm}^{-1}$ — guaiacyl lignin (softwood-relevant)
- $\sim 1155\text{--}1165\text{ cm}^{-1}$ — polysaccharide C–O–C
- $\sim 1100\text{--}1120\text{ cm}^{-1}$ — cellulose/hemicellulose

Known sources of variability

- Species and genetic provenance
- Juvenile vs mature wood
- Earlywood vs latewood
- Geographic and environmental growth conditions
- Post-harvest aging and oxidation
- Milling and dust formation effects
- Particle size, morphology, and surface sensitivity

Bulk IR measurements of woods



What is PKE dust?

Core diagnostic window:

- $\sim 1735\text{--}1750\text{ cm}^{-1}$ — ester carbonyl, residual palm-kernel oil
- $\sim 1640\text{--}1660\text{ cm}^{-1}$ — amide I; protein, bound-water contribution
- $\sim 1530\text{--}1550\text{ cm}^{-1}$ — amide II; protein
- $\sim 1150\text{--}1170\text{ cm}^{-1}$ — glycosidic C–O–C; mannan/cellulose
- $\sim 1020\text{--}1060\text{ cm}^{-1}$ — carbohydrate C–O; mannan-rich fibre

Known sources of variability

- Palm-kernel provenance
- Palm Kernel processing conditions
- Efficiency of oil extraction and residual lipid content
- Variable protein-to-fibre composition
- Kernel shell and other fibrous material
- Drying, storage, moisture and oxidation
- Particle tissue, size and morphology
- Localised protein-, lipid- and carbohydrate-rich particles

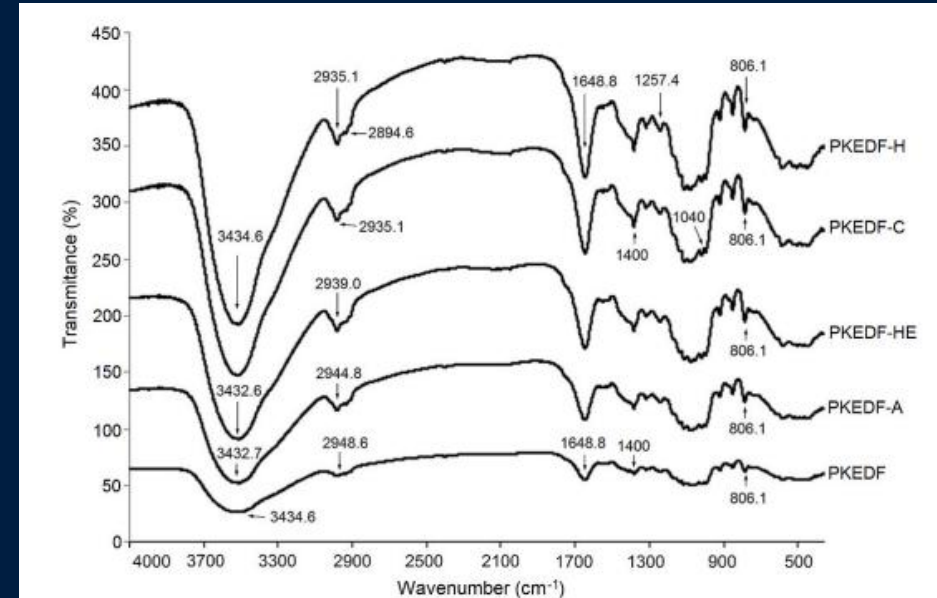
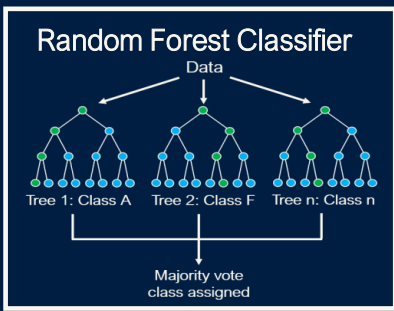


Fig. 2. Fourier-transformed infrared spectroscopy of PKEDF, PKEDF-A, PKEDF-HE, PKEDF-C and PKEDF-H.

Figure shows effects of different solvent treatments for extraction of PKE on FTIR Spectral signatures



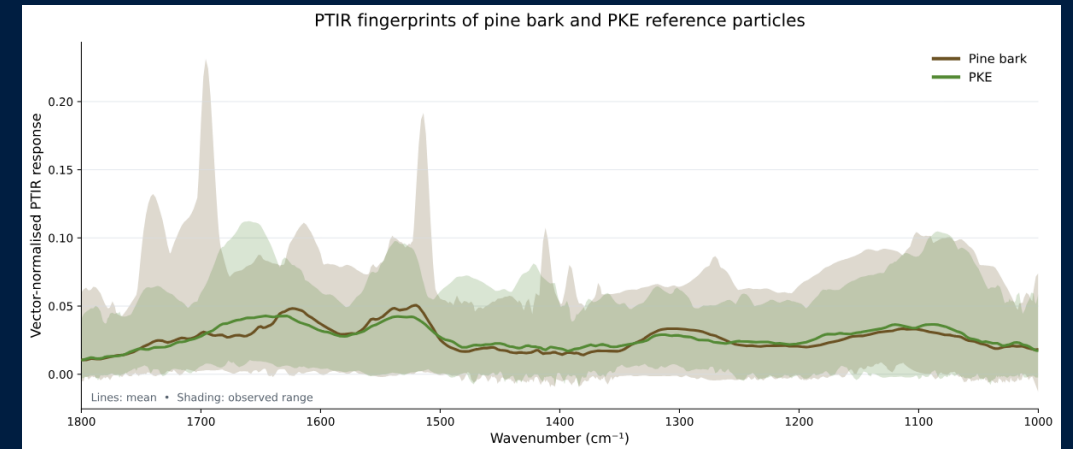
Particle Classification – Two Source Model

The binary model correctly assigned 82% of pine particles and 70% of PKE particles.

Two-source classification

		Predicted	
		PKE	Pine
Actual	PKE	42 70%	18
	Pine	21	98 82%
78% overall accuracy		76% balanced accuracy	Confused 39 spectra cross classes

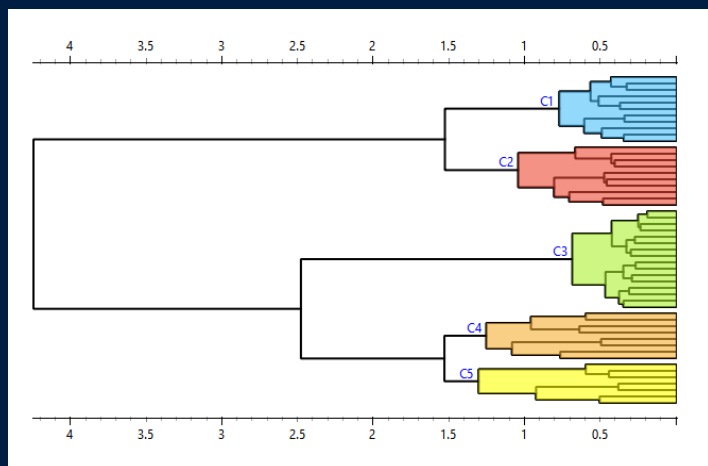
- Pine and PKE are partially separable from their PTIR fingerprints.
- A forced binary decision overstates certainty for chemically ambiguous particles.



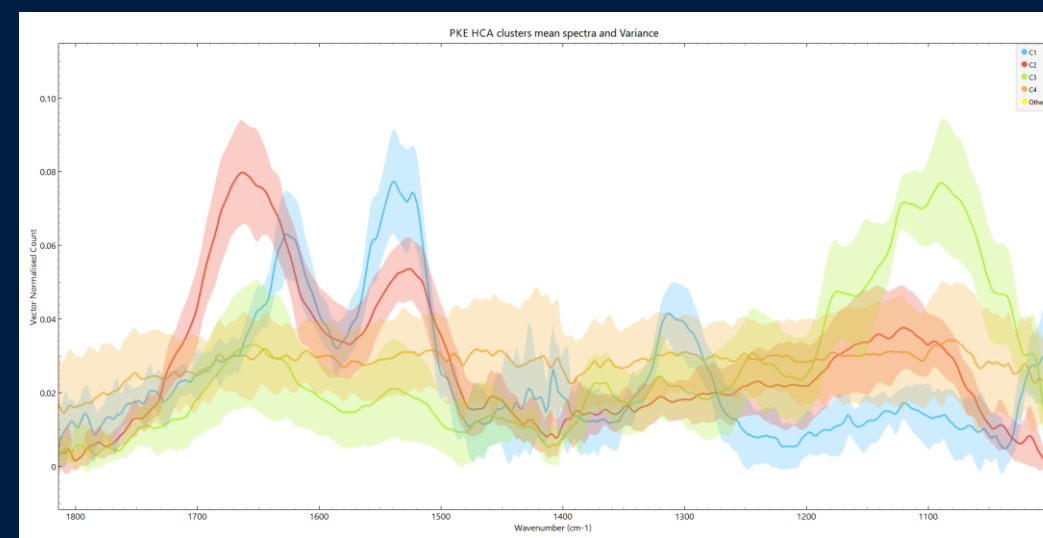
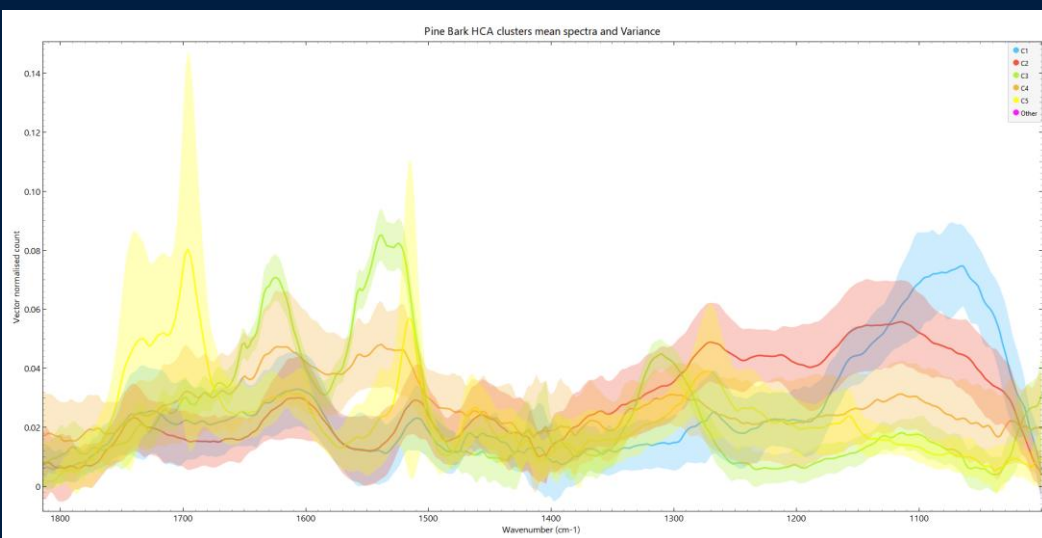
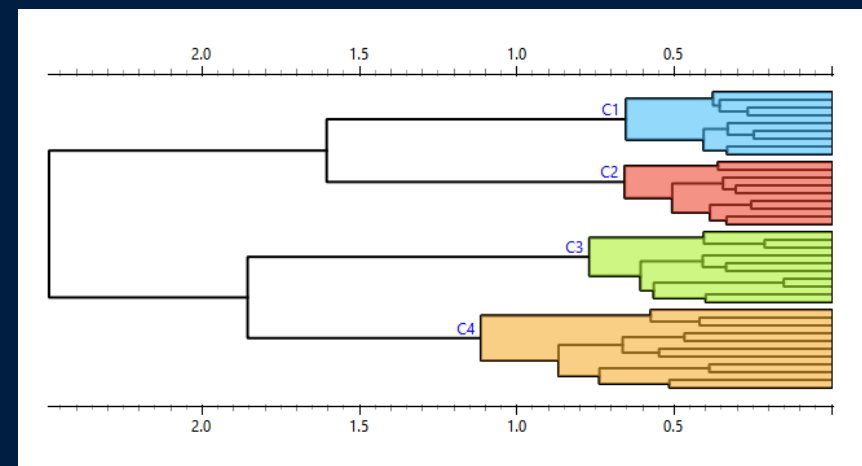
Neither source is represented by a single spectral fingerprint

Both Sources are Chemically Heterogenous

Pine Bark HCA – 5 Sub Groups

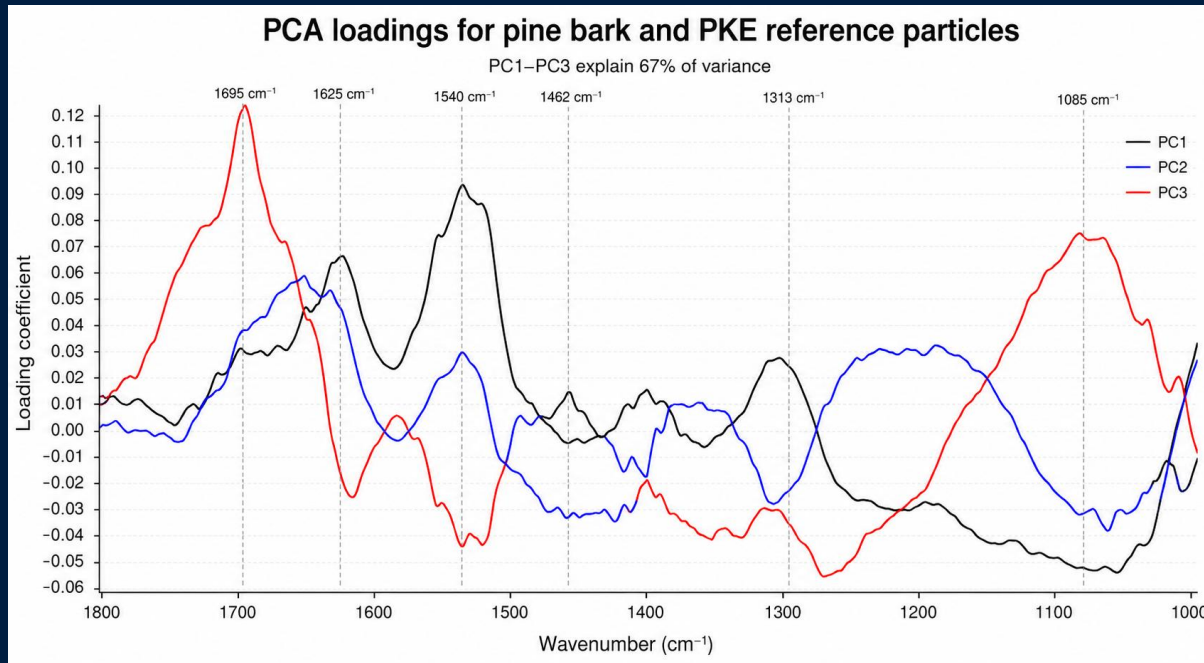


PKE HCA – 4 Sub Groups



What chemical differences drive the particle groupings?

Principal Components Analysis (PCA) identifies the combinations of spectral features that vary most strongly between individual particles.



PC1 — protein/lignin balance

Differences around 1625, 1540 and 1313 cm^{-1} distinguish particles with stronger protein- or lignin-associated features.

PC2 — carbohydrate structure

Variation across approximately 1200–1085 cm^{-1} reflects differences in cellulose, hemicellulose and other plant carbohydrates.

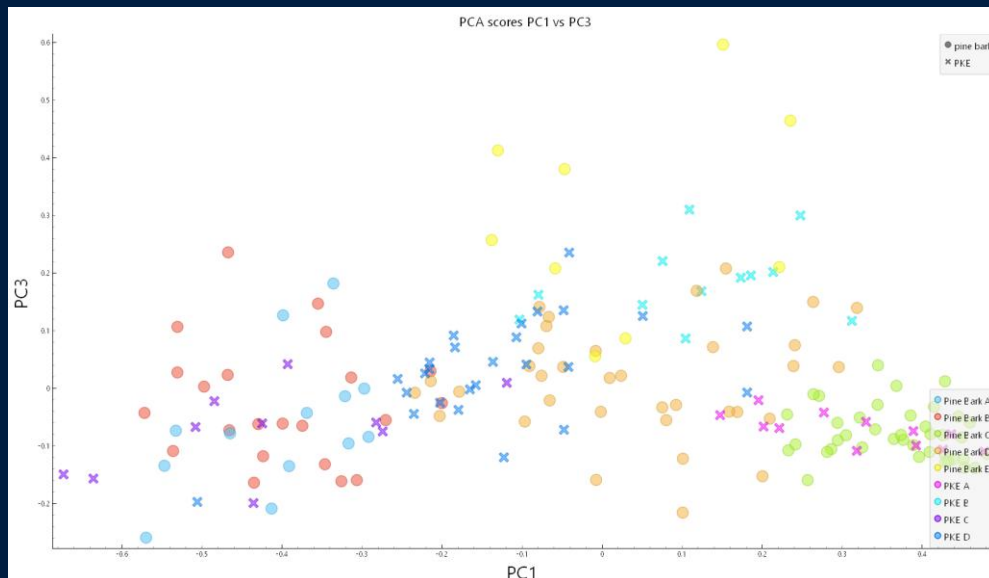
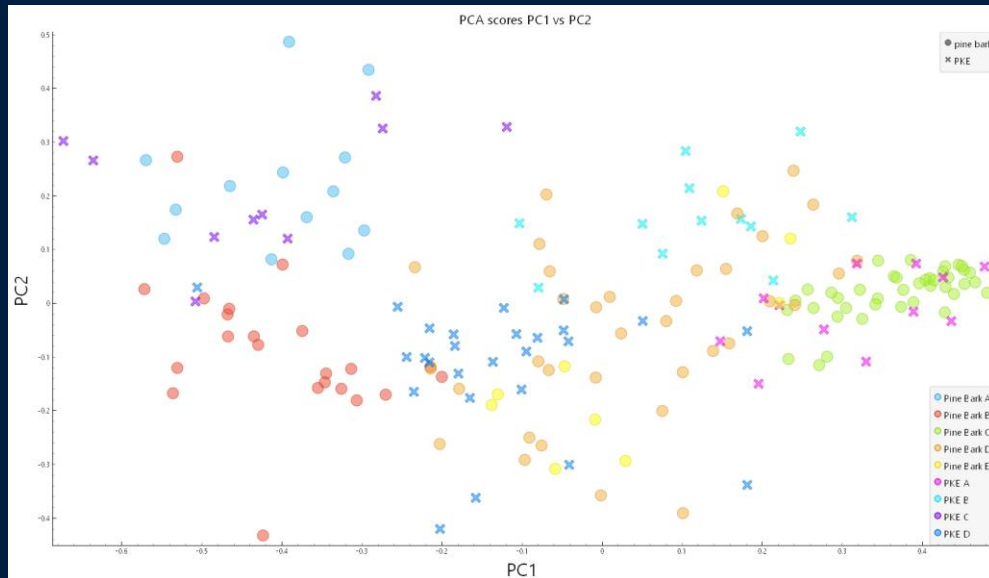
PC3 — carbonyl and oxidation chemistry

The strong contribution near 1695 cm^{-1} reflects differences in carbonyl-containing material associated with lipids, oxidation and processed plant matter.

The model is not finding one unique pine or PKE marker.

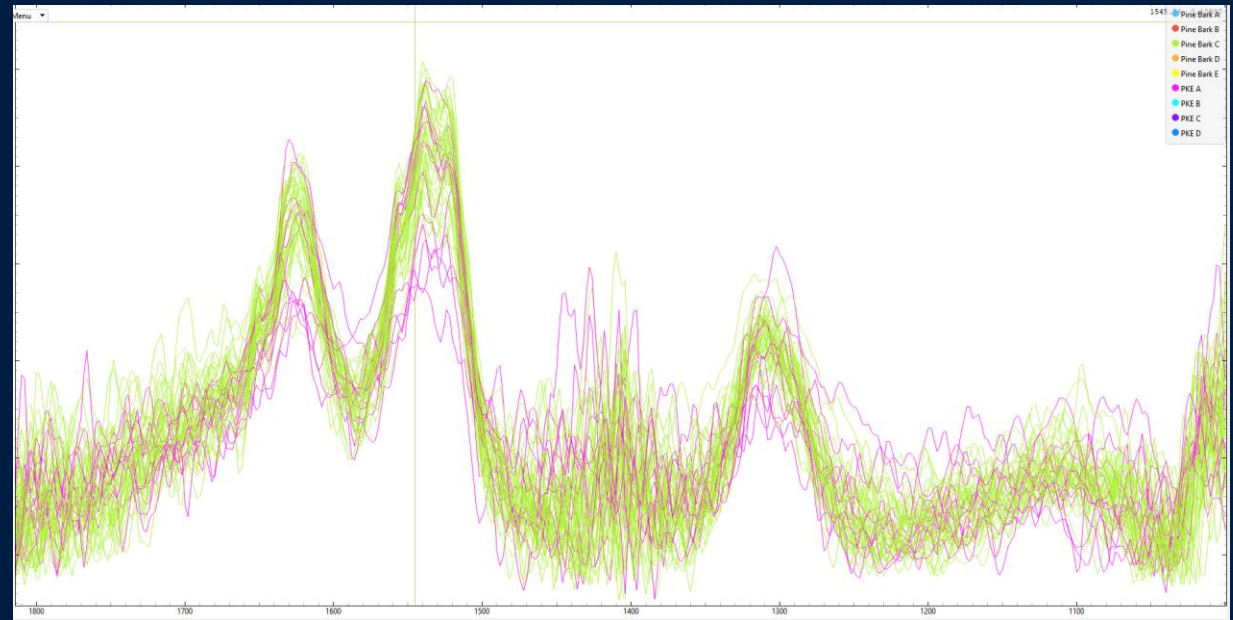
It is separating particles using combinations of protein/lignin balance, carbohydrate structure and carbonyl chemistry.

Principal Components Analysis of Subgroups



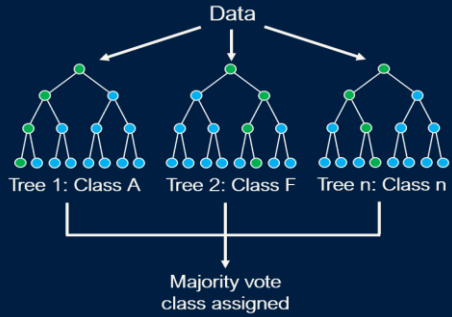
PCA Scores agree with the binary confusion matrix

- Some subgroups of Pine Bark and PKE overlap in PCA Scores
- Pine Bark C and PKE A overlap particularly strongly in both scores plots



Clear plant-derived chemistry but lack sufficient source-specific information to distinguish pine from PKE.

Random Forest Classifier



Classification by Subgroup

Nine-subgroup classification

		Predicted										
		Pine Bark A	Pine Bark B	Pine Bark C	Pine Bark D	Pine Bark E	PKE A	PKE B	PKE C	PKE D	Σ	
Actual	Pine Bark A	9	0	0	0	0	0	0	4	0	13	
	Pine Bark B	1	18	0	0	0	0	0	1	2	22	
	Pine Bark C	0	0	40	0	0	0	0	0	0	40	
	Pine Bark D	0	1	3	17	1	1	1	0	11	35	
	Pine Bark E	0	0	0	1	7	0	1	0	0	9	
	PKE A	0	0	7	3	0	2	0	0	0	12	
	PKE B	0	0	0	1	0	0	11	0	0	12	
	PKE C	2	1	0	0	0	0	1	6	0	10	
	PKE D	0	1	0	4	1	0	0	2	18	26	
Σ		12	21	50	26	9	3	14	13	31	179	

Best recognised

Pine C 100% • PKE B 92% • Pine B 82%

Main source-crossing confusion

Pine D ↔ PKE D

72%

subgroup accuracy

68%

balanced accuracy

41

particles crossed sources

Removed from future training models

Subgrouping explains the uncertainty but does not completely resolve source attribution.

Logging Debris Samples

positive control for pine dust, negative control for PKE



Logging Emissions

- Pine Bark
- Debris from debarking row
- Material before logs loading
- Material after log loading
- Material after the sweeper

Mineral Dusts:

- Unsealed Road Dust
- Dried River Bed Dust
- Fine Quarry Dust

Expected outcome:

- No classification of PKE
- Debarking debris: highest proportion of pine-like spectra.
- Ground sweepings before and after log loading: lower pine-like proportion and more rejected spectra After mechanical sweeping: predominantly rejected, expect more mineral-like spectra.

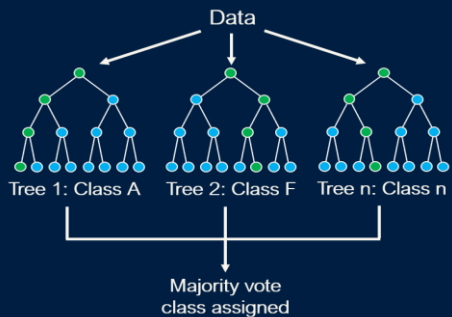
Limitations

- Mineral samples are not from the same region (Canterbury)
- Samples were measured at a different facility and had previously been exposed to synchrotron radiation

Solutions:

- Prepare fresh samples of logging debris and test model performance
- Prepare high purity mineral samples using local sources

Random Forest Classifier



Recognising ambiguity improves the model

- Adding mineral references ambiguous plant, low-signal and mixed organic–mineral classes reduced forced source assignments.
- Remaining errors were concentrated among low-signal, mixed and ambiguous spectra.
- These output classes represent evidence quality as well as predicted chemistry.

10- Class Uncertainty aware model

Actual ↓ Predicted →	Pine A	Pine B	Ambiguous plant	Low signal	Pine E	PKE B	PKE C	Mineral A	Mineral B	Mixed organic–mineral	Σ
Pine A	7	0	0	1	0	0	2	0	3	0	13
Pine B	1	19	0	1	0	0	0	0	1	0	22
Ambiguous plant	0	0	49	3	0	0	0	0	0	0	52
Low signal	0	1	5	43	3	1	1	0	1	6	61
Pine E	0	0	0	1	7	1	0	0	0	0	9
PKE B	0	0	0	1	0	11	0	0	0	0	12
PKE C	1	2	0	0	0	1	4	0	2	0	10
Mineral A	0	0	0	0	0	0	0	7	0	0	7
Mineral B	4	1	0	2	0	0	1	1	8	1	18
Mixed organic–mineral	0	0	0	4	0	0	0	0	0	77	81
Σ	13	23	54	56	10	14	8	8	15	84	285

Best recognised

Mixed 95% • Ambiguous plant 94% • PKE B 92%

Remaining overlap

Low signal ↔ mixed / ambiguous

81%

overall accuracy

75%

balanced accuracy

94%

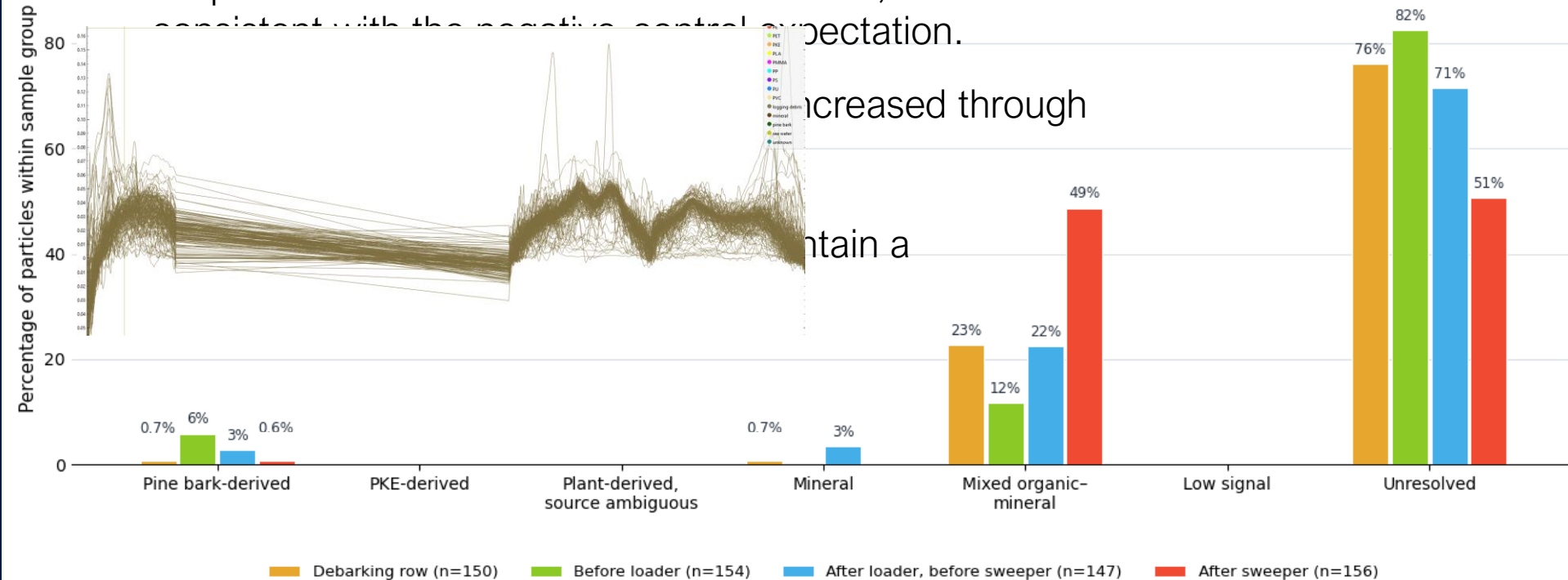
ambiguous plant recall

Logging Debris Samples

Classification captures increasing mineral contribution through the sweeping process

Thresholded classifications in logging debris

- No particles were classified as PKE-derived, consistent with the negative control expectation.



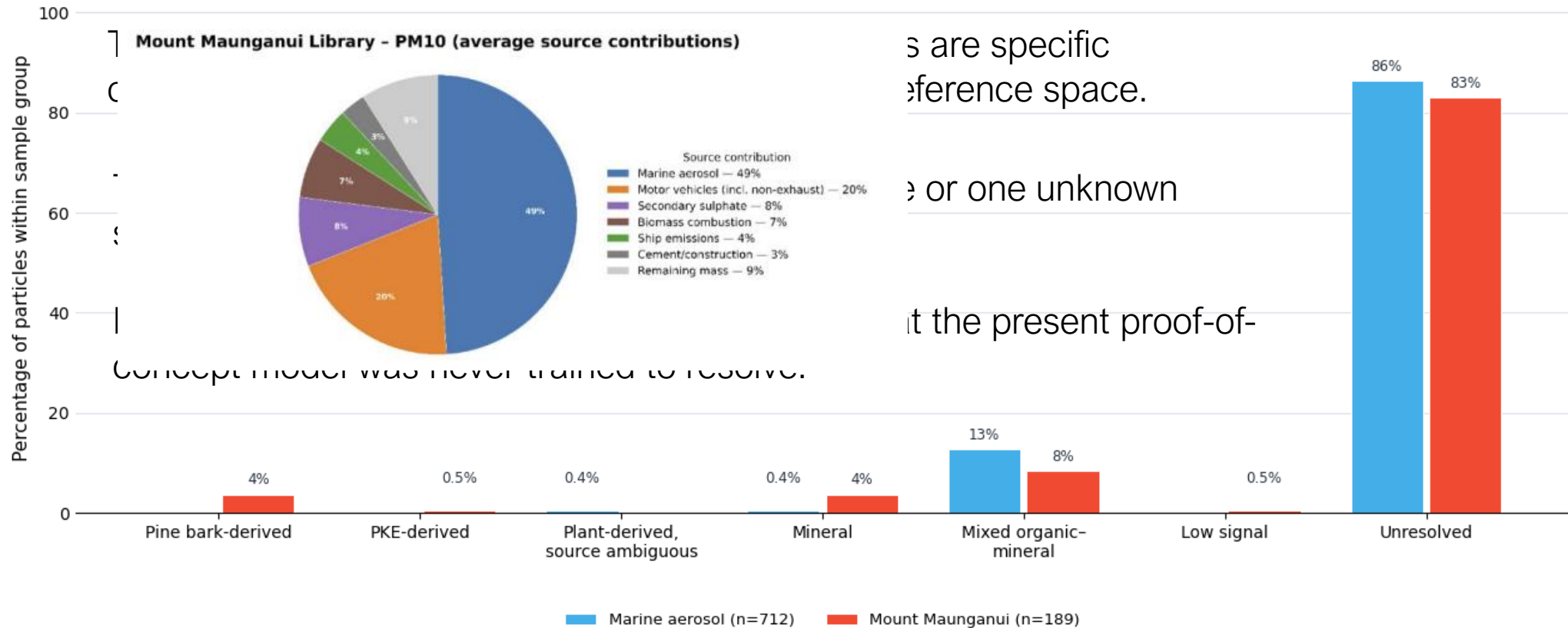
Two working hypotheses

1. Missing stem-wood reference chemistry in the model
2. Synchrotron-associated alteration of organic particles.

Environmental Samples

Mount Maunganui and Clean Marine Air

Thresholded classifications in environmental particles



Point Sources

Asphalt production

- Aggregate minerals
- Combustion Soot
- Bitumen Associated organics

Concrete production and cutting

- Calcium rich mineral dust
- Silicate aggregates

Metalworks and Surface treatments

- Iron-rich steel particles
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Salt Processing

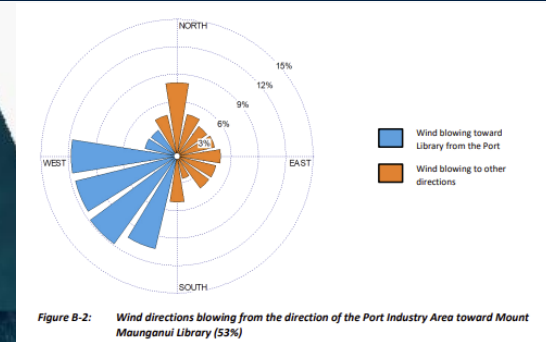
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Solid Product Handling and Storage

- Palm Kernel Extract
- Feed Wheat
- Soya meal
- Other plant derived particle

Logging

- Pine bark
- Pine wood
- Soil
- Road dust



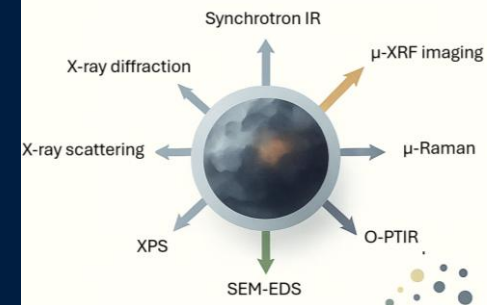
Mount Maunganui Airshed

Conclusions

- The logging controls recovered the expected increase in mineral-associated particles through the loading and sweeping sequence.
- Mount Maunganui contained candidate pine-bark-like and mineral-like signals, but most particles remain outside the current reference space.
- The present model is a proof of concept - not yet a complete quantitative description of the particles in airshed.
 - We have spent 2 years developing reliable single particle measurements across the range of particle types in complex industrial environments
 - We have developed preliminary models for the different techniques and reference particle types that are very promising for particle-by-particle source attribution in environmental samples.

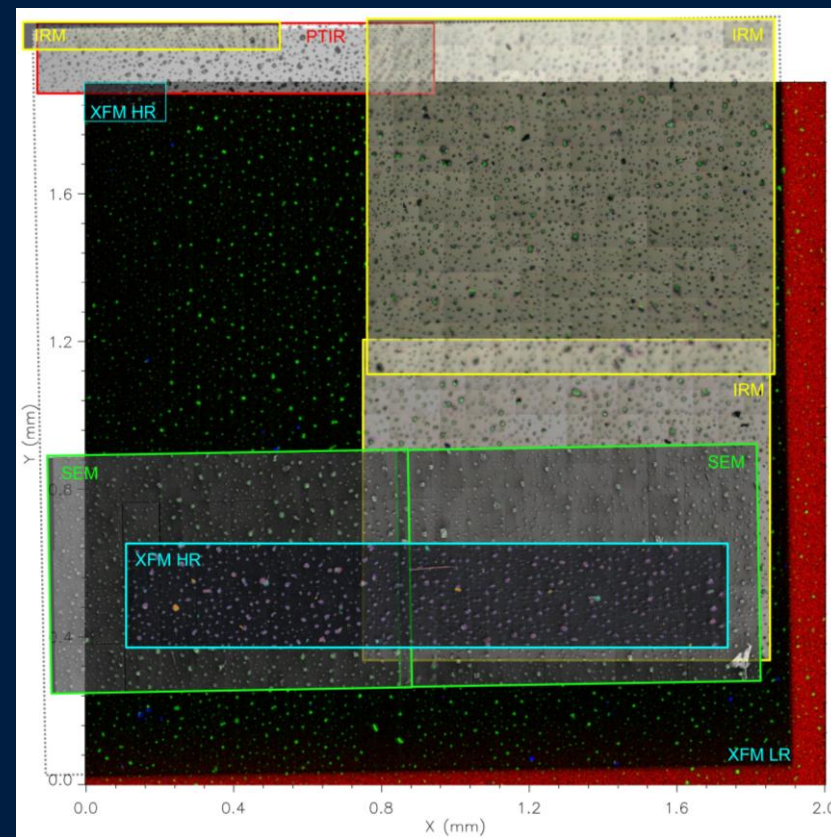
Reliable attribution requires wider range source references and multiple, independent lines of particle-level evidence.

Multi-Modal Single Particle Analysis



Multimodal Fusion – Future Frontier

- Define diagnostic features from each mode: molecular bands and ratios, elemental composition, morphology, size and spatial associations.
- Co-register features at the individual-particle level and combine them into a shared particle-by-feature matrix.
- Build multimodal reference definitions that capture both characteristic source features and within-source variability.
- Test environmental particles against the combined reference space, retaining uncertainty when evidence is incomplete or conflicting.
- Apply validated classifications to environmental samples across space and time to identify emission patterns and evaluate mitigation.



Acknowledgements

This research was partially undertaken on the XFM and IRM beamlines at the Australian Synchrotron and Supported by beamline scientists Dr Andrew Langendam (XFM) and Dr Annaleise Klein (IRM)



We acknowledge the merit beamline grants and travel/access funding provided by the New Zealand Synchrotron Group (NZSG) and ANSTO.



Photothermal measurements were undertaken at Sydney Analytical - the University of Sydney and Monash University with method development support across 2025 from Photothermal spectroscopy Corp. Eoghan Dillon and Mustafa Kanzis



Sampling program and SEM-EDS measurements were undertaken at Mote Limited. All experimental costs outside of the synchrotron were funded by Mote Ltd as part of their particle speciation development program