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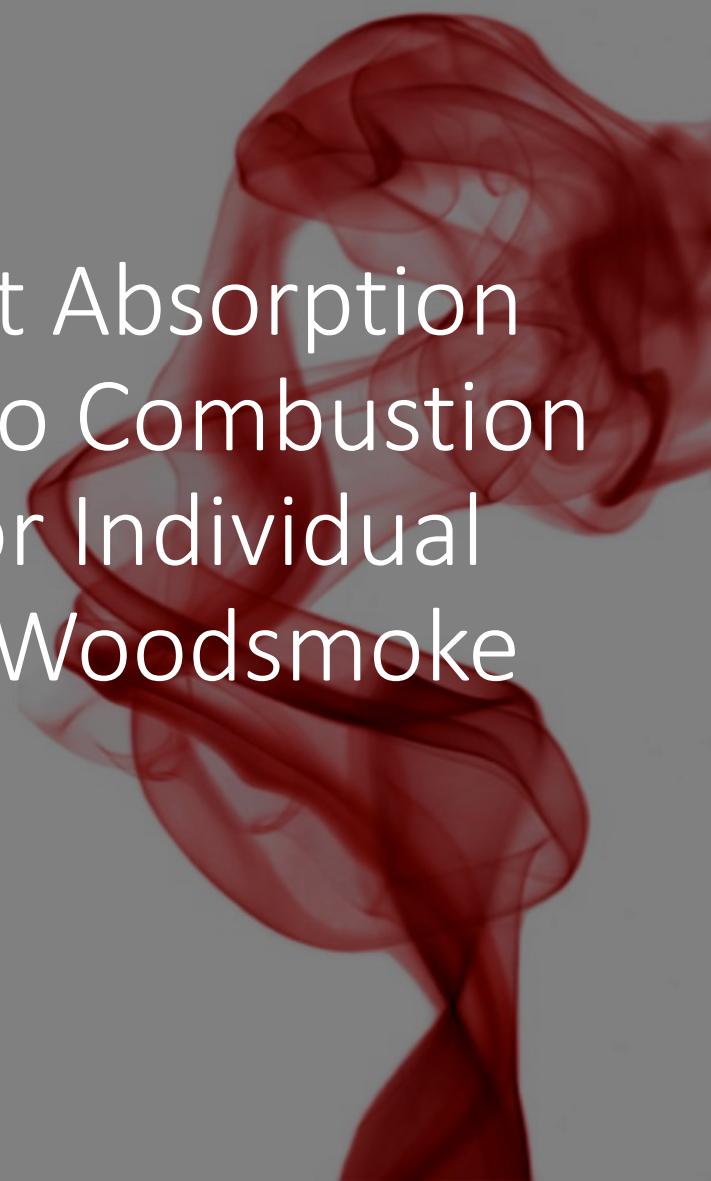
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An abstract graphic of swirling, translucent red smoke or liquid, positioned on the right side of the slide. It has a soft, flowing texture and a darker red hue against the dark gray background.

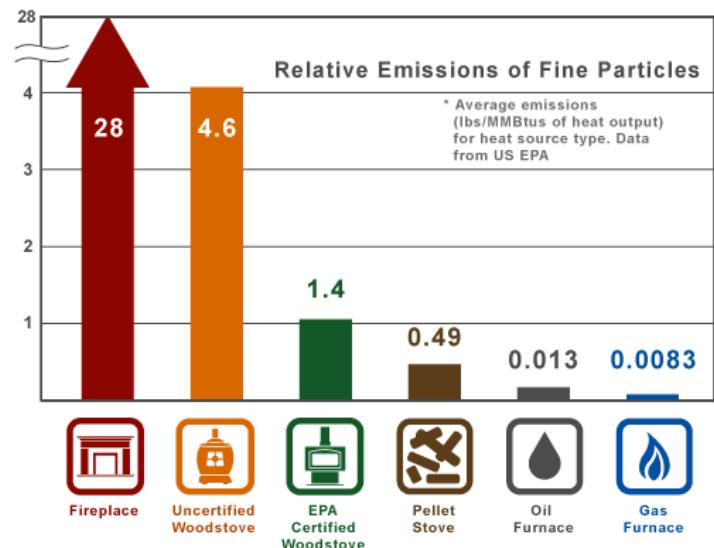
# Linking Light Absorption Properties to Combustion Efficiency for Individual Residential Woodsmoke Sources

# Woodsmoke from residential wood combustion

- Worldwide air pollution problem
- Local air pollution issue
  - Worst in densely populated areas
- Challenge: heterogeneous combustion
  - Real world emissions differ from lab tests
- No convenient monitoring method

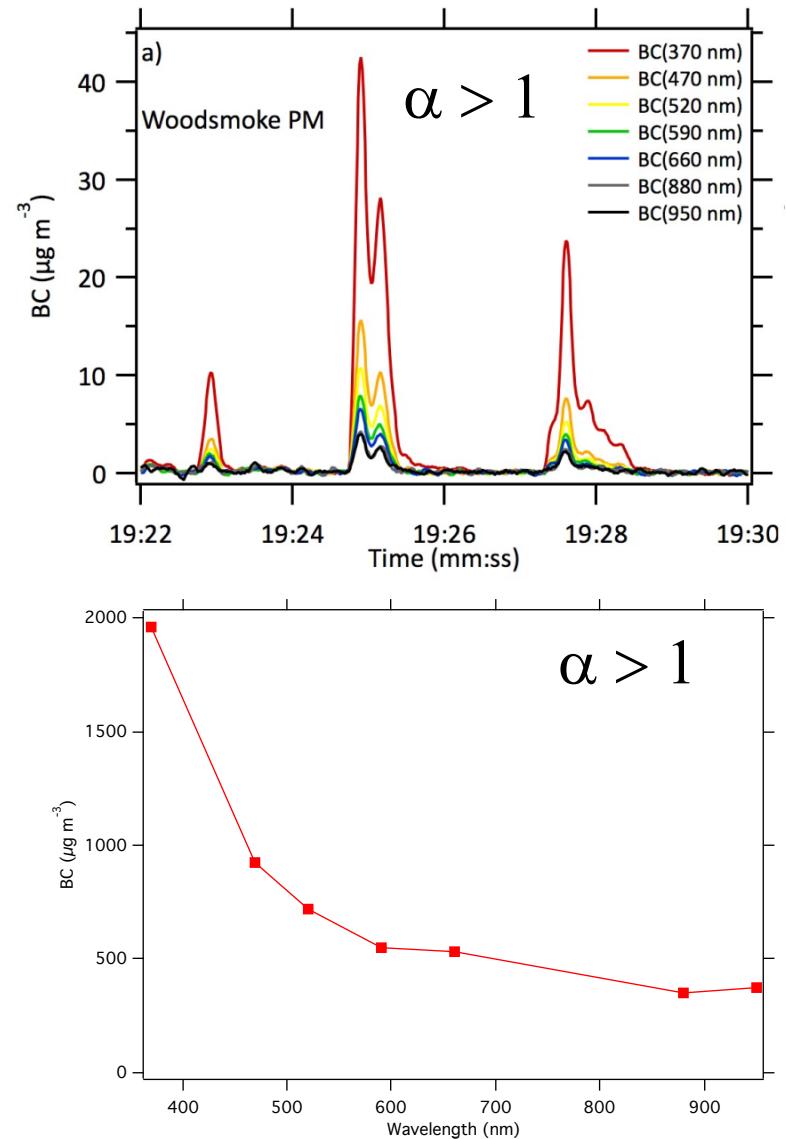
Contribution to PM<sub>2.5</sub> emissions

	US	UK
Woodsmoke	14%	35%
Vehicle	5%	13%



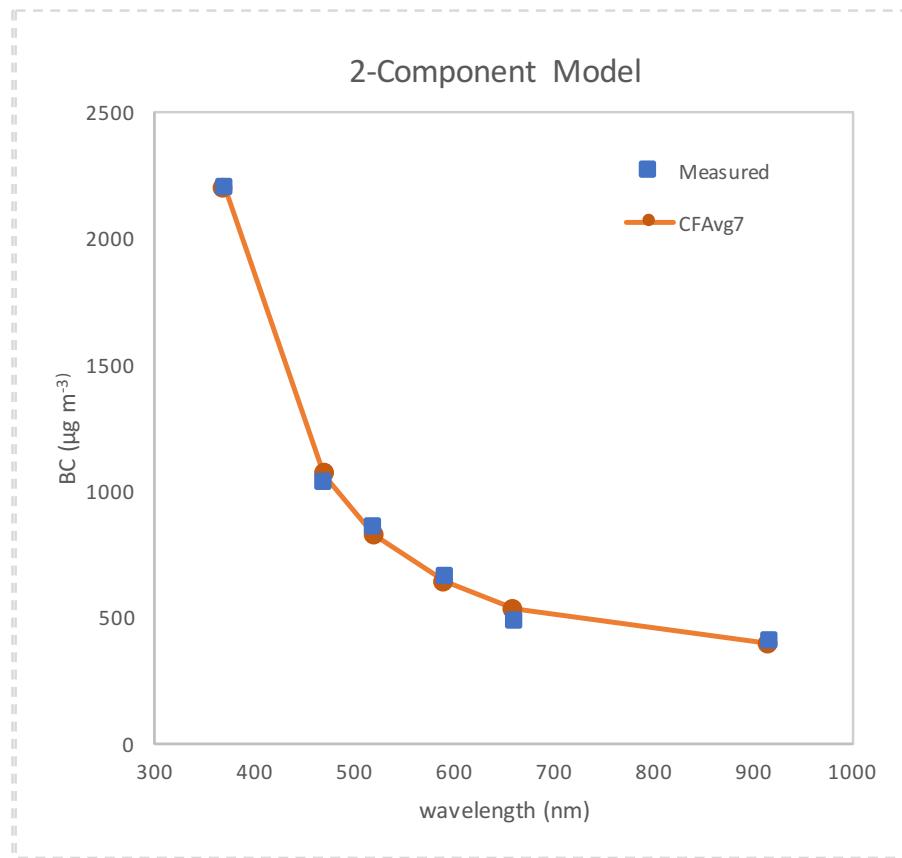
# Aerosol light-absorption properties

- Woodsmoke contains mostly **Brown Carbon (BrC)**,
  - absorb more light at the low visible and near-UV wavelengths
- **Absorption Angstrom exponents (AAE, or  $\alpha$ )**
  - power law exponent
  - approximates the change in measured absorption as wavelength changes
- $\alpha$  is commonly used optical property
  - atmospheric warming
  - source apportionment



# Deriving $\alpha_{\text{BrC}}$

- Two-component (BC and BrC) model:
  - Also used in Chen et al. (2015) and Ran et al. (2016)
  - $B(\lambda) = [\text{BC}] + [\text{BrC}] \lambda^{-\alpha_{\text{BrC}}}$
  - Use multiple wavelengths
- Tested different curve fitting parameters
  - including 370nm
  - averaging 880nm and 950nm
- Chosen fitting method
  - fit all 7 wavelengths
  - average 880nm and 950nm



# Project Overview

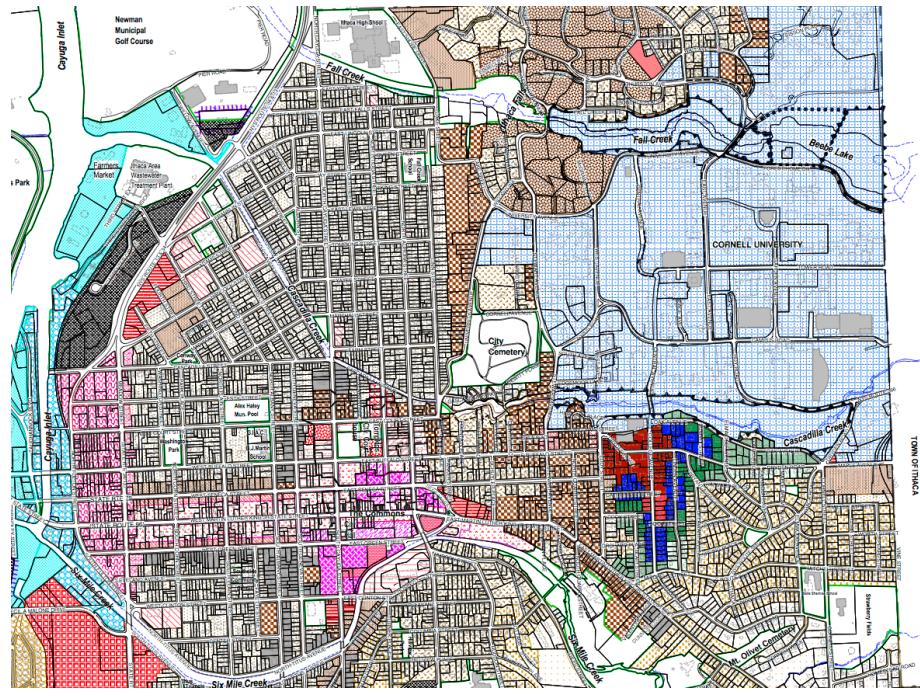
- Understand local air quality impact of residential wood combustion (**RWC**)
  - Identified a large number of sources
  - Real time measurements
- Explored the linkage between optical properties and combustion efficiency

## Highlights

First to report  
real-time,  
source-specific  
measurements

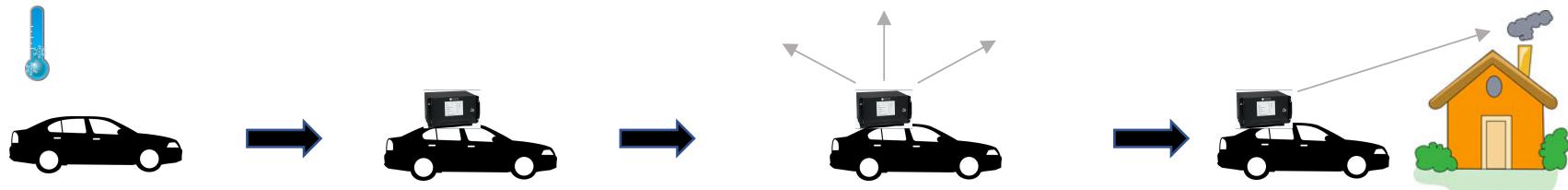
Large collection  
of RWC sources

# City of Ithaca, NY



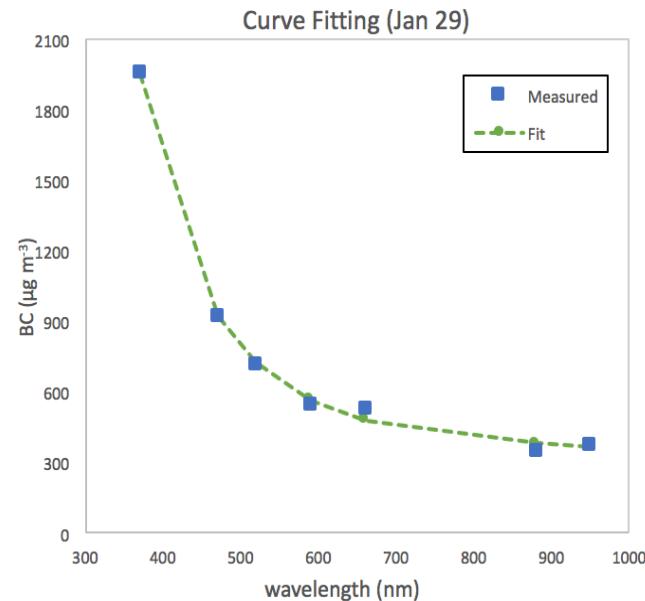
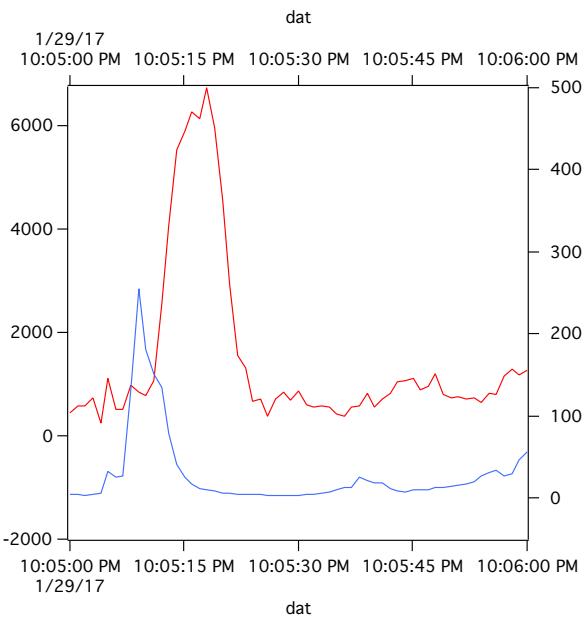
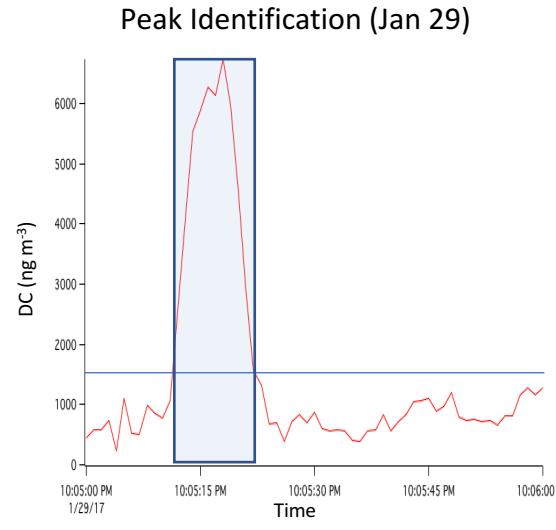
# Integrated plume sampling

- Cold, low-wind speed winter nights in Ithaca
  - No other emission sources
  - High ground-level impact expected
- A hybrid electric vehicle instrumented with:
  - Aethalometer (AE-33) - light absorption at 7 wavelengths
  - pDR-1500 - PM<sub>2.5</sub>
  - Vaisala GMP343 - CO<sub>2</sub>
  - GPS
- **Mobile** monitoring: Search for woodsmoke plumes
- **Stationary** monitoring: Stay by the source until ample plumes have been measured



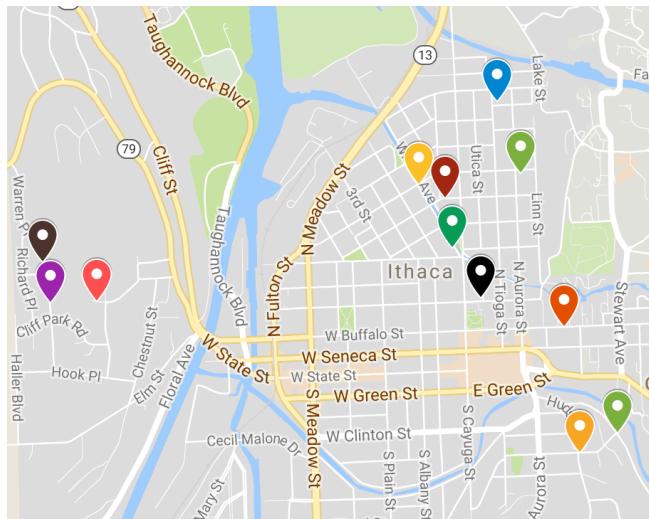
# Data Processing Software

- Peak Identification → Data Alignment → Data Analysis
- Identifies woodsmoke plumes
  - Searches for elevated DC
  - Fixes instrumental time delay differences
  - Align DC and PM<sub>2.5</sub>
  - Average data over identified peaks
  - Find  $\alpha_{BrC}$ , average PM<sub>2.5</sub>

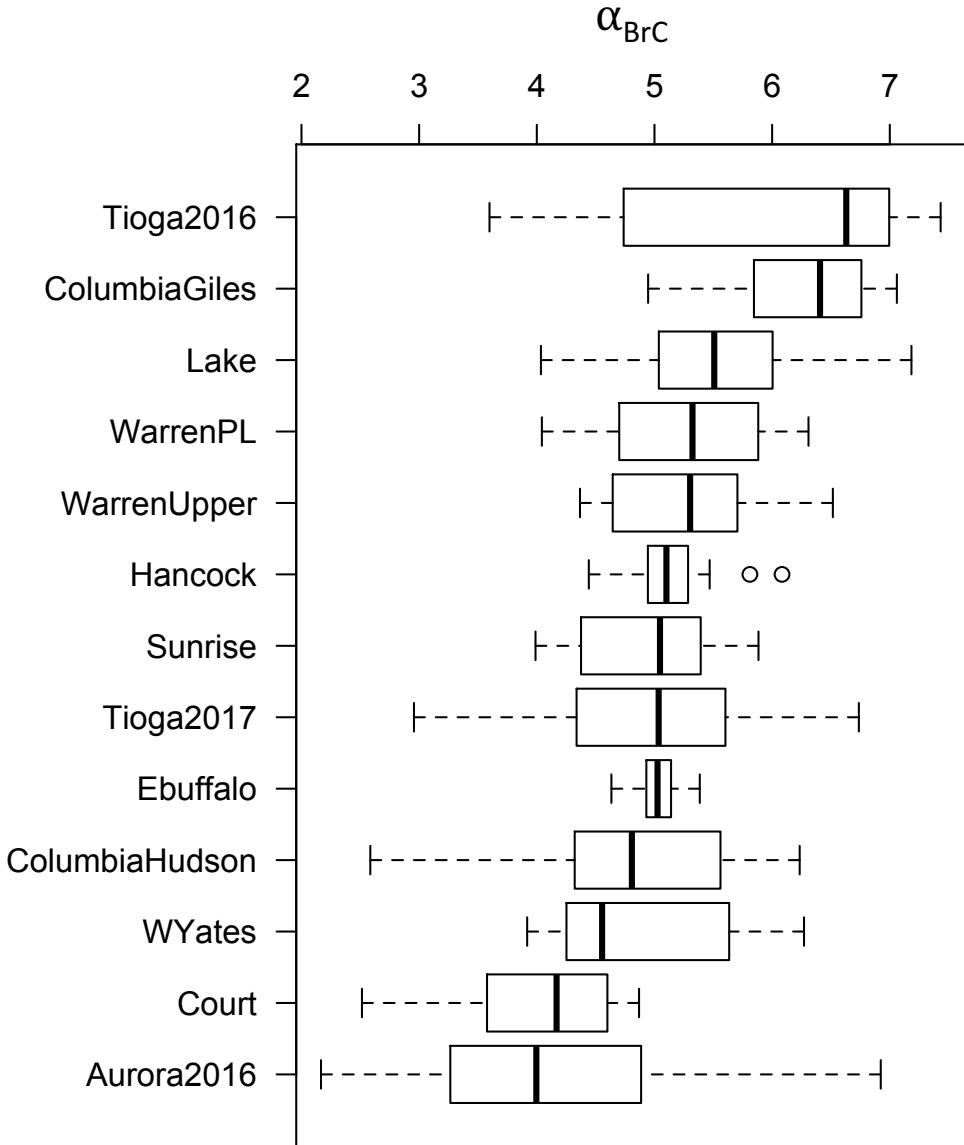


# $\alpha_{BrC}$ by Individual Sources

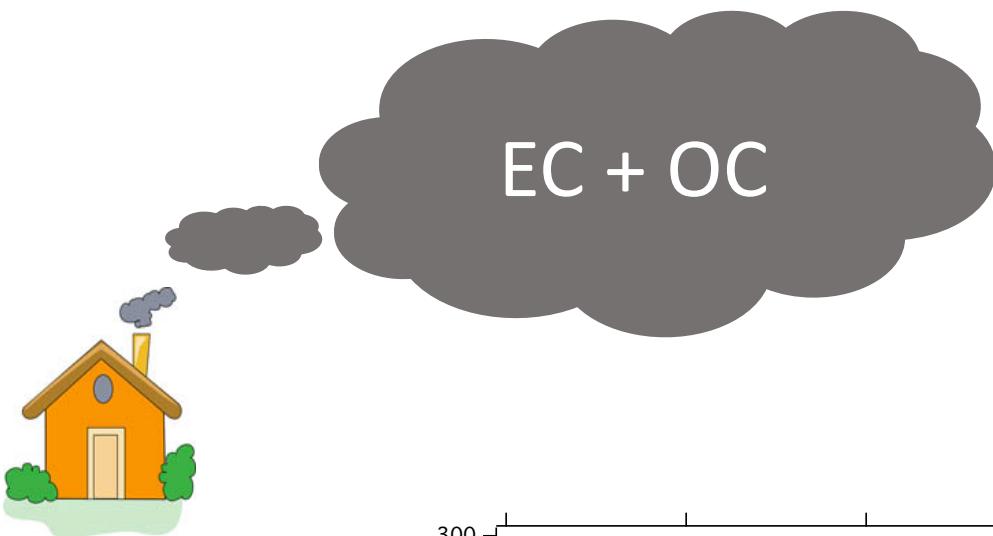
- $\alpha_{BrC}$  is in a range
- Medians between 4 and 6
- Not a constant



Woodsmoke source locations

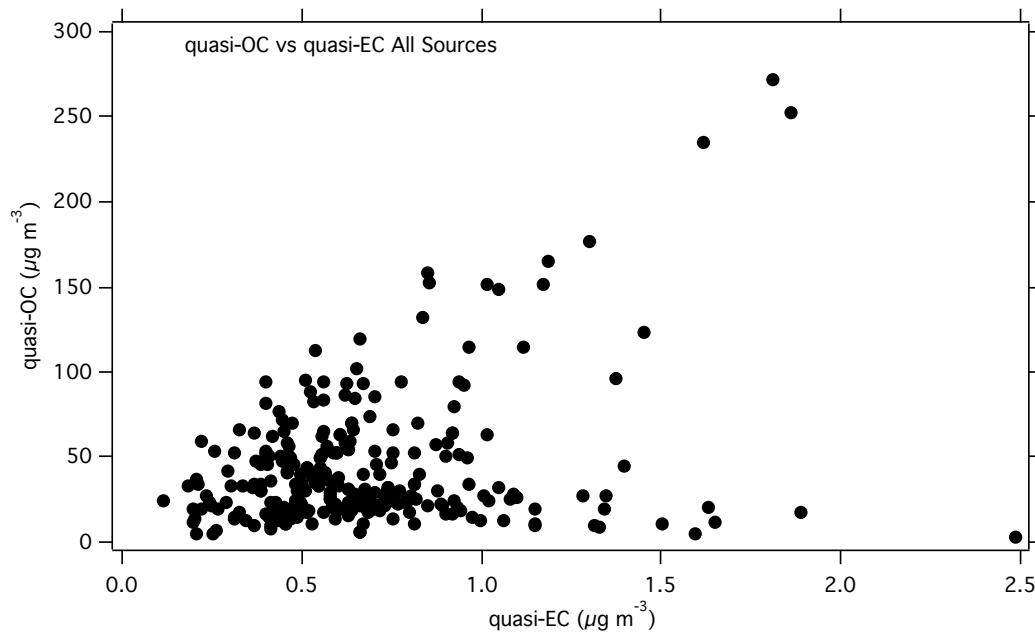


# quasi-OC vs quasi-EC



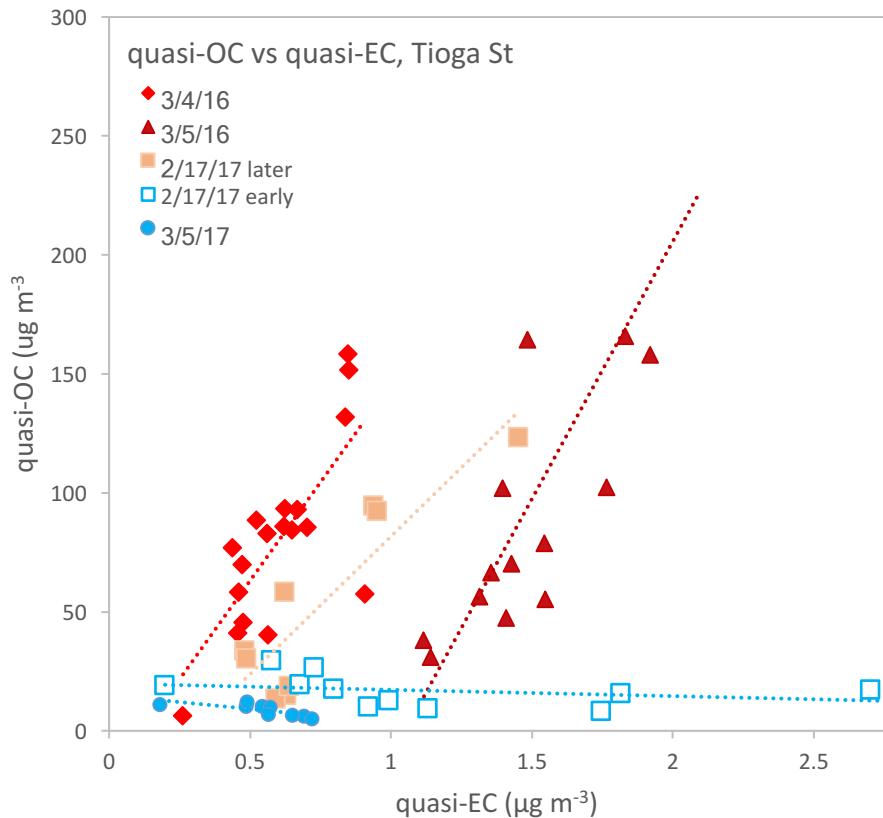
quasi-EC = BC(880)

quasi-OC = PM<sub>2.5</sub> – quasi-EC



# Results: Intervention at Tioga St

- Homeowner used improperly seasoned wood in 2016
- We helped Tioga St's homeowner improve burning practice in 2017



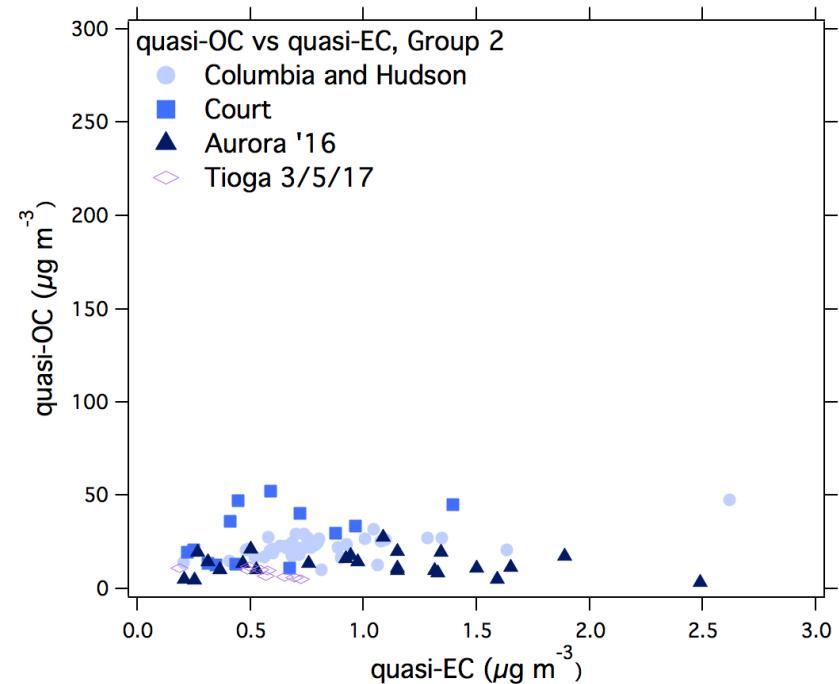
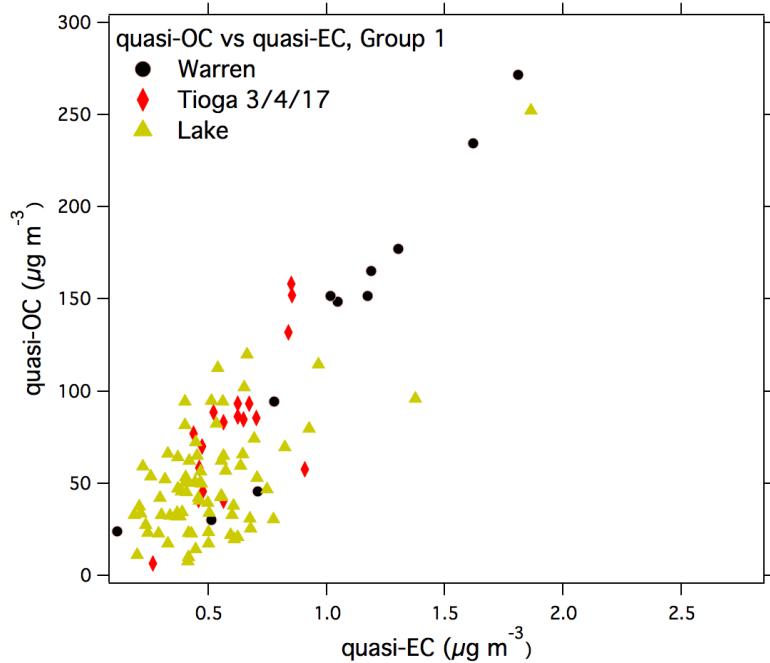
## Group 1 - Dirty burning

- Steep trend lines
- 2016 was very dirty and noticeable
- Feb 2017 was slightly cleaner

## Group 2 - Clean burning

- Homeowner adopted new burning practices
- Fewer plumes detected
- Woodsmoke was hot and rose quickly

# Results: Classification of sources



## Group 1 – Dirty Burning:

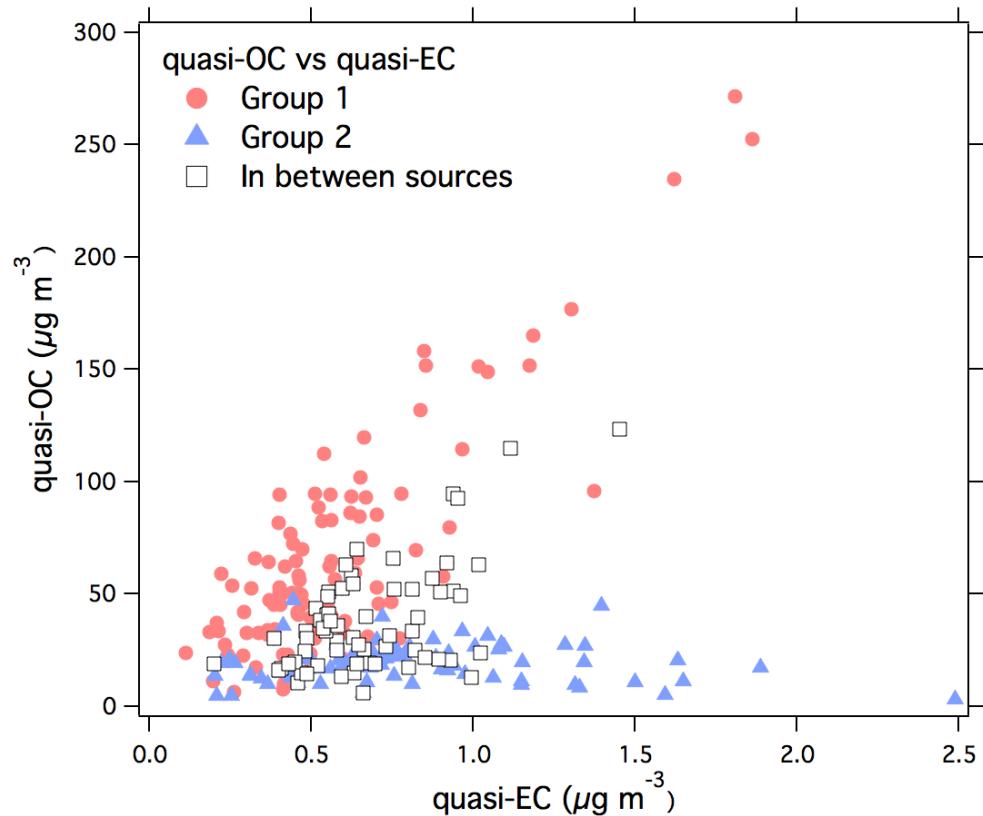
- Steep slope, large OC/EC ratio
- Tight pattern, large  $r^2$  correlation

## Group 2 – Clean Burning:

- Slope close to 0, small OC/EC ratio
- OC remains similar while EC increases

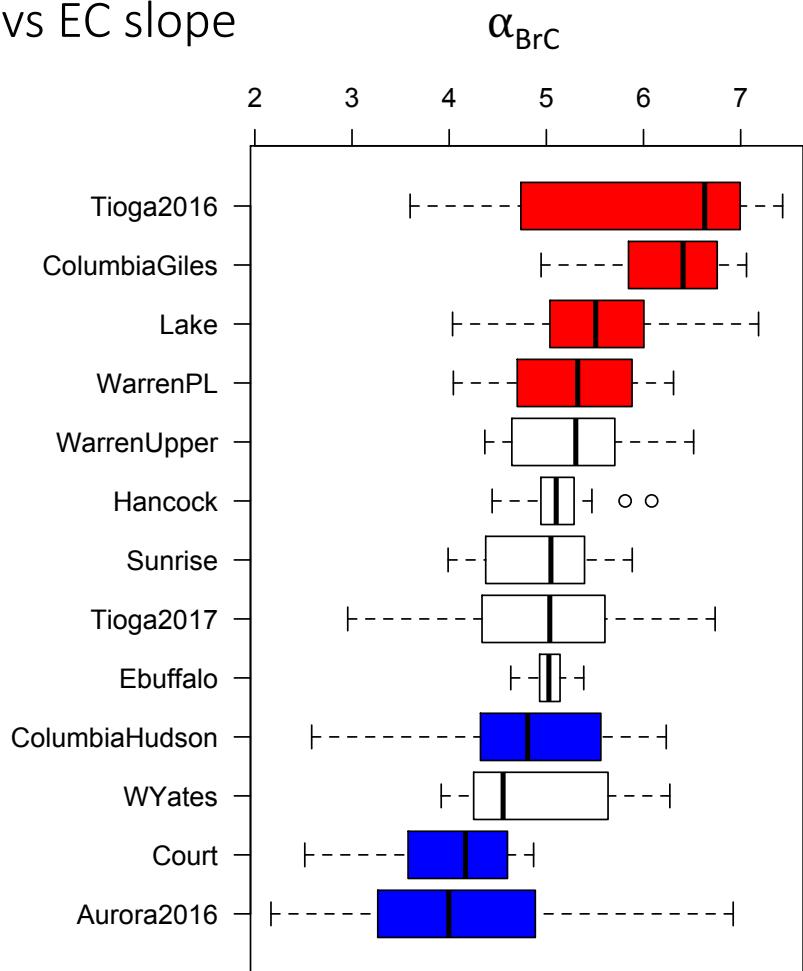
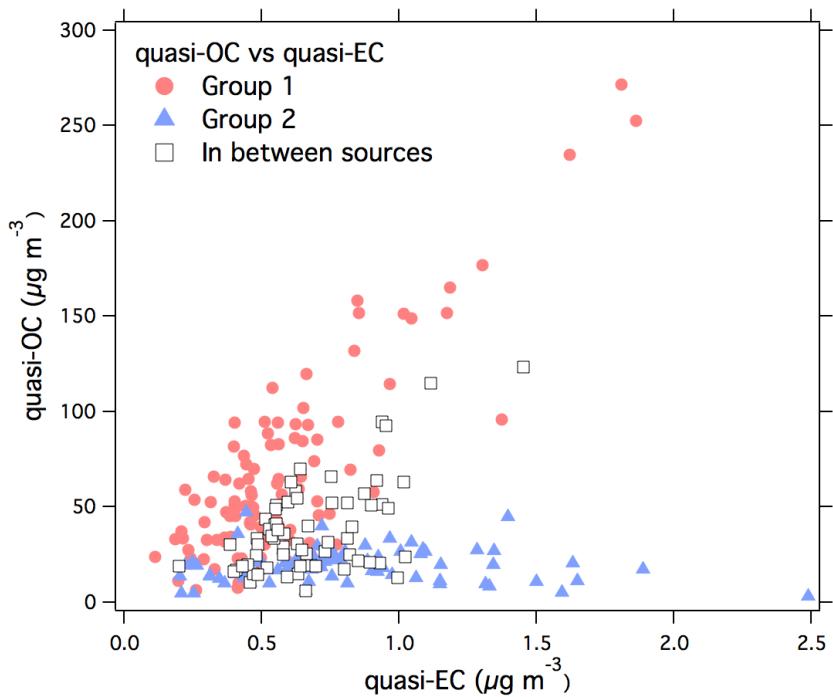
# Results: Classification of sources

- Group 1 – “upper limit”
- Group 2 – “lower limit”
- Many in-between sources
  - depends on operation
  - data has larger range
- Similar to  $\alpha_{\text{Brc}}$ 
  - OC vs EC is a range not a constant
  - Different sources have different average slope values



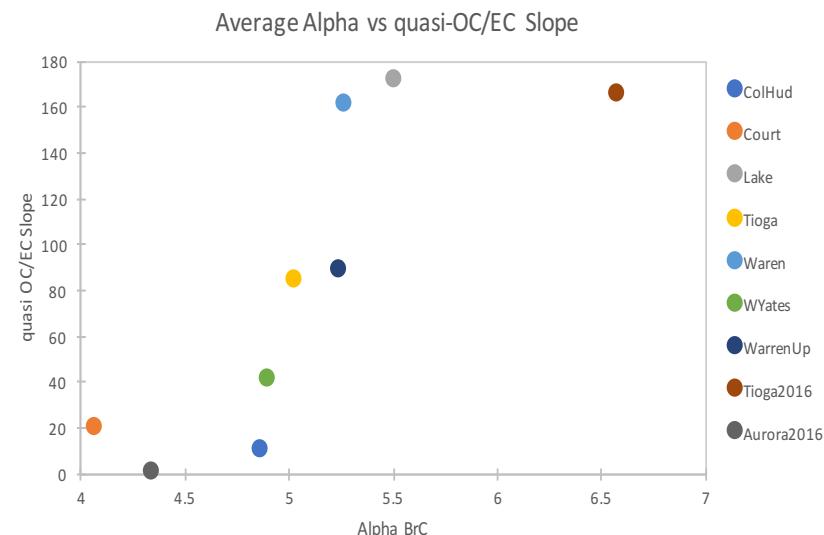
# Results: $\alpha_{\text{BrC}}$ and OC vs EC

- Positive correlation between  $\alpha_{\text{BrC}}$  and OC vs EC slope



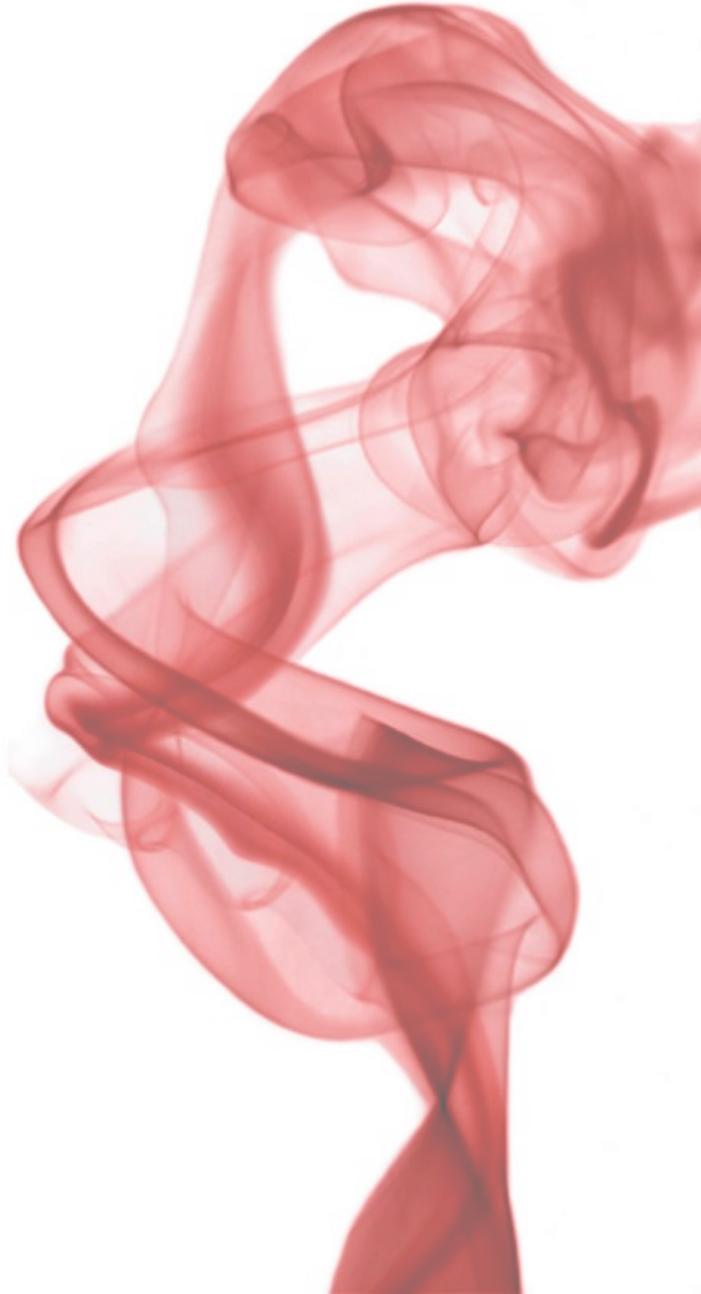
# Summary

- First reported real-time measurements of real-world, source-specific RWC sources
- Large collection of real-world RWC sources
- Sources can be differentiated by  $\alpha_{BrC}$  ranging from 4-6
  - Most literature assumes  $\alpha_{BrC}$  is a constant (around 1.6-2.2)
  - $\alpha_{BrC}$  similar to that found in Chen, 2015 lab studies
- Sources can also be differentiated by OC vs EC plots
  - Dirty sources have high OC vs EC slopes
  - Correlates with  $\alpha_{BrC}$
  - This can be done relatively inexpensively
- Larger  $\alpha_{BrC}$  and larger OC vs EC slopes indicate dirtier burning



# Acknowledgements

- Jiajun Gu, Ye Lin Kim, Ye Xie and Qikun Wang at Cornell University for conducting the field measurements
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T. HANKS  
Questions?