

# Deconvolution of Post-Detonation Mixtures of Soot

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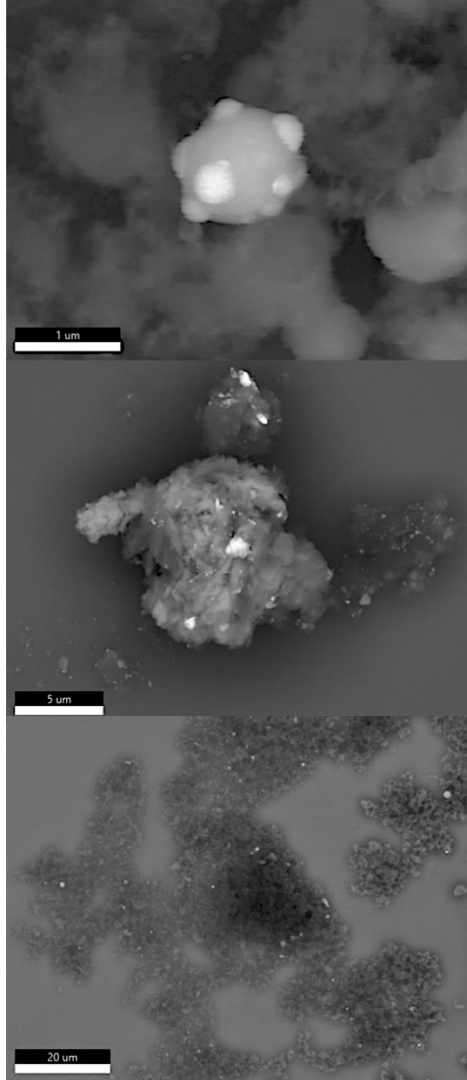
# Sources of Soot

- **Atmospheric Soot**

- Formed through incomplete combustion
- Submicron Particles

- **Detonation Soot**

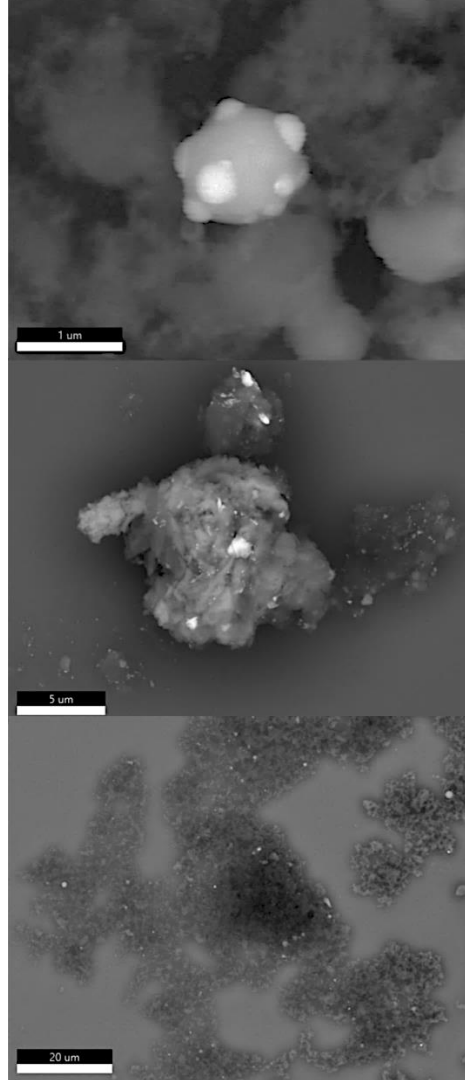
- Produced by shock-driven decomposition
- Much higher temperature and pressure regimes
- Chemical properties dependent on the fuel, detonation conditions, and environment



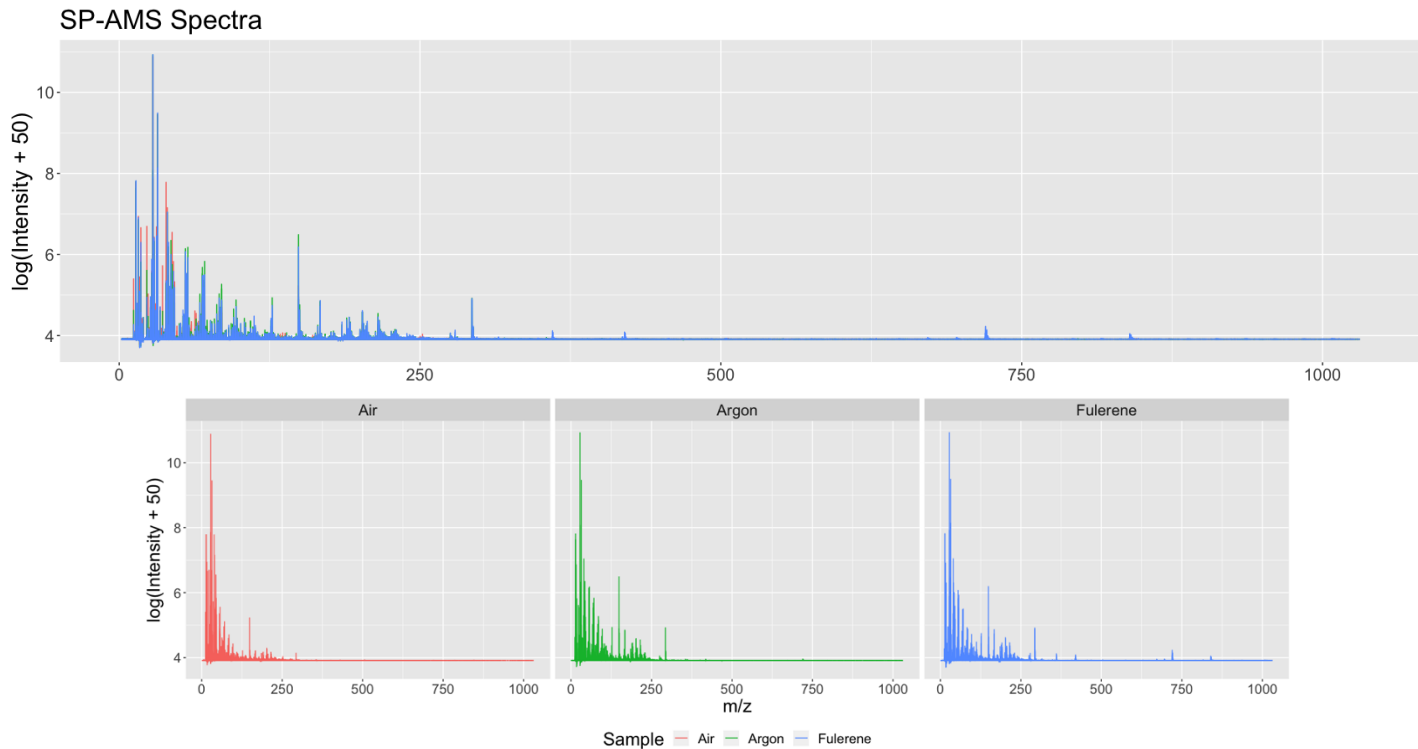
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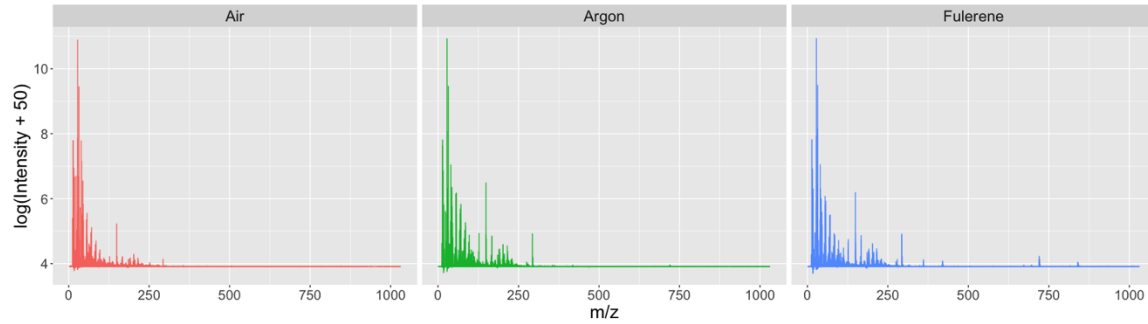
These factors are useful for defining source-dependent features



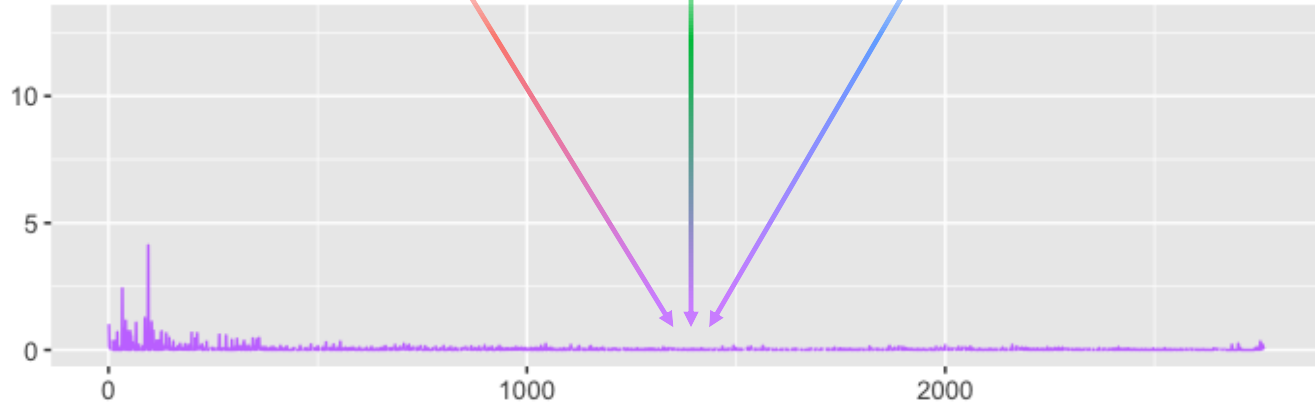
# Source-Dependent Signatures



# Mixture of Signatures

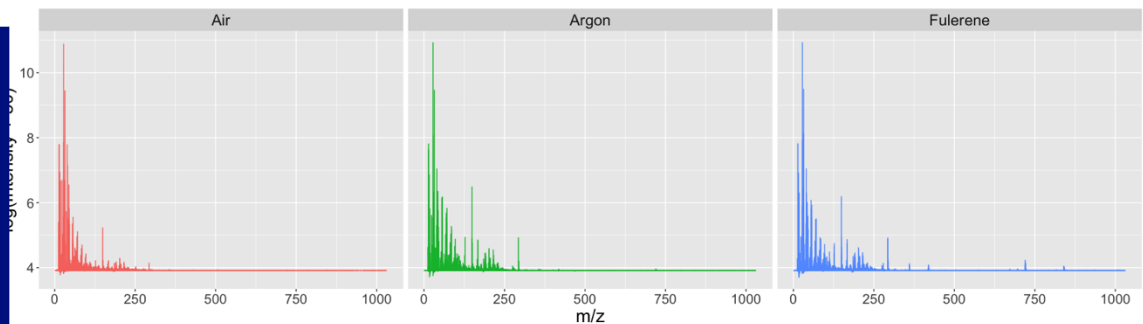


## Mixture of Detonation Soots

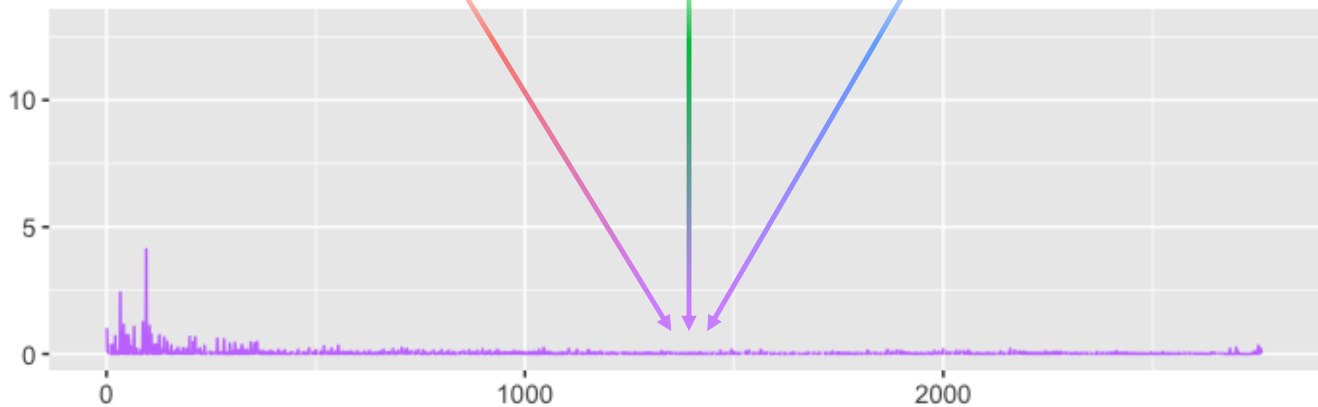


# Mixture of Signatures

Is there another way we can represent this data?



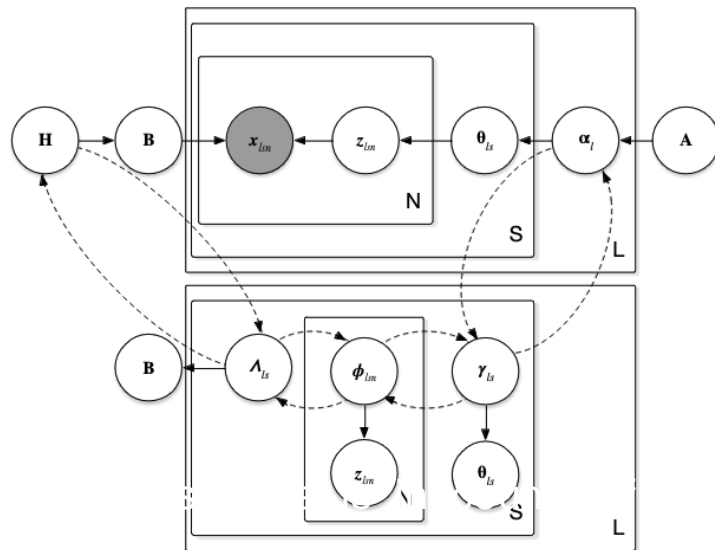
Mixture of Detonation Soots



# Latent Dirichlet Allocation

## Key Points

- Statistical technique
- Specifically designed for **mixtures**
- Originally designed for **topic modelling**
  - **Topics**, characterized by **words**;
  - **Soot** characterized by **moiety families**;

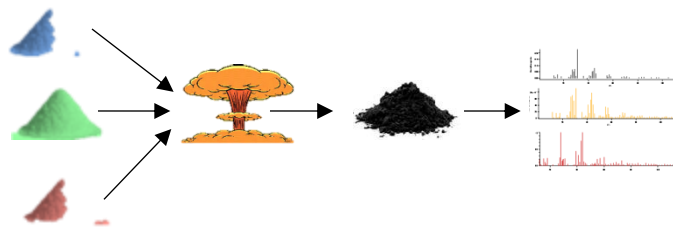


$$q(\Theta, \mathbf{B}, \mathbf{Z} | \Gamma, \Lambda, \Phi) = \prod_{l=1}^L \prod_{s=1}^S \prod_{n=1}^N \left[ q_{ls}(\theta_{ls} | \gamma_{ls}) \prod_{r=1}^R q_{ls}(z_{lrs} | \phi_{lrs}) \prod_{m=1}^M q_m(\beta_m | \lambda_{lsm}) \right]$$

# Latent Dirichlet Allocation

## Key Points

- Statistical technique
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  - **Topics**, characterized by **words**;
  - **Soot** characterized by **moiety families**;
- Extension to **mixtures of mixtures**:
  - Originally designed for various types of dust particles (data is in counts of **dust particle types** associated with different **locations**)
  - Extension to soot: use peak heights as “proportions”, and use these to sample “particles” (**moiety families**) from a given “location” (**detonation soot**)
  - Multiple soot types, each defined by their own mixture of moiety families





# Convert Magnitudes of Spectra into Count Data

	Air1	Air2	Air3	Air4	Air5	Air6	Air7	Ar1	Ar2	Ar3	Ar4	Ar5	Ar6	Ful1	Ful2	TR1	TR2	TR3
HRBC_FullereneC	1	1	1	1	1	1	1	44	42	55	34	42	43	299	230	211	293	294
HRBC_LowC	471	470	473	524	451	479	433	502	448	382	450	525	529	47	35	125	130	140
HRBC_MidC	3	3	4	4	6	4	3	5	0	3	0	0	0	0	0	0	1	1
HRCh1	7	8	8	8	8	6	8	4	3	3	2	2	2	16	19	16	17	17
HRMxM_familyOther	626	615	620	621	628	619	634	530	544	560	534	501	510	35	39	142	136	135
HRNH4	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
HRNO3	10	10	8	9	9	9	10	4	5	0	2	8	2	15	16	12	13	12
HROrg_CH	343	390	388	385	394	433	419	1062	1071	997	1028	1150	1166	2769	3152	2629	2532	2527
HROrg_CHN	24	35	26	29	35	27	29	91	13	48	35	7	30	259	112	94	116	121
HROrg_CHO1	302	312	298	255	238	247	215	246	207	234	264	247	246	1277	1093	997	985	983
HROrg_CHOgt1	108	112	100	99	108	114	101	0	0	0	38	18	5	144	140	114	123	123
HRSO4	3	2	2	2	3	2	2	3	22	7	7	6	5	4	6	3	4	4
HRmetal	3102	3042	3072	3063	3119	3060	3144	2511	2659	2717	2607	2493	2462	137	160	656	651	643

1. Partition the data into moiety families;
2. Use peak heights as proportions used to sample particles
3. Get Particle Counts

# A Case Study

Suppose that...

- We have a library of three types of detonation soots:

1. Composition B detonated in Air
2. Composition B detonated in Argon
3. Fullerene

} this is what we use  
to **train our model**

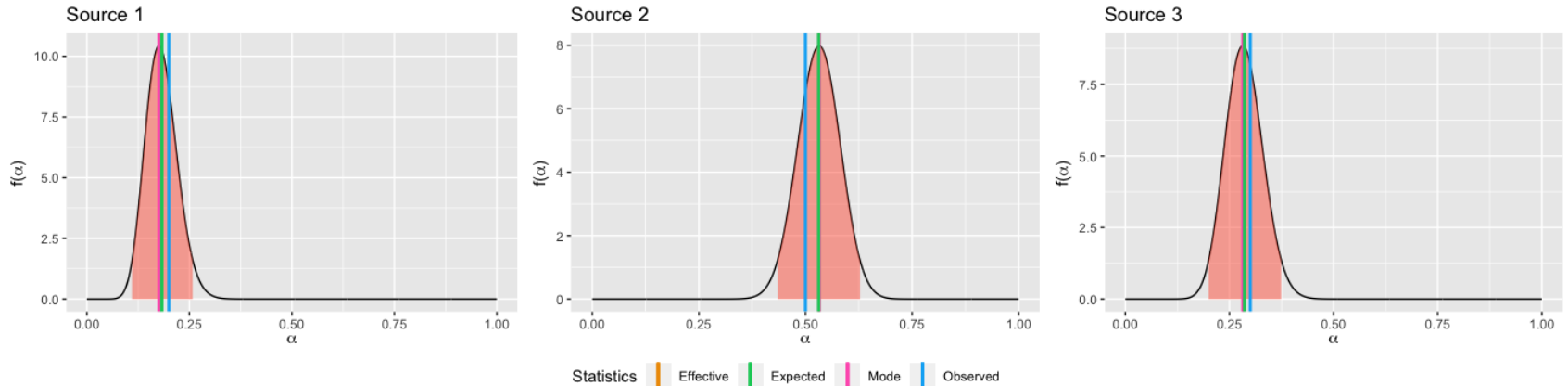
- We can only observe **pure samples** of potential sources (from our library)
- We have observed three samples of trace soot (from the same detonation)
- Our trace source is a **mixture** of the three sources

- 20% Composition B in Air
- 50% Composition B in Argon
- 30% Fullerene

} we don't know this...  
this is what we **want**  
**to learn!**

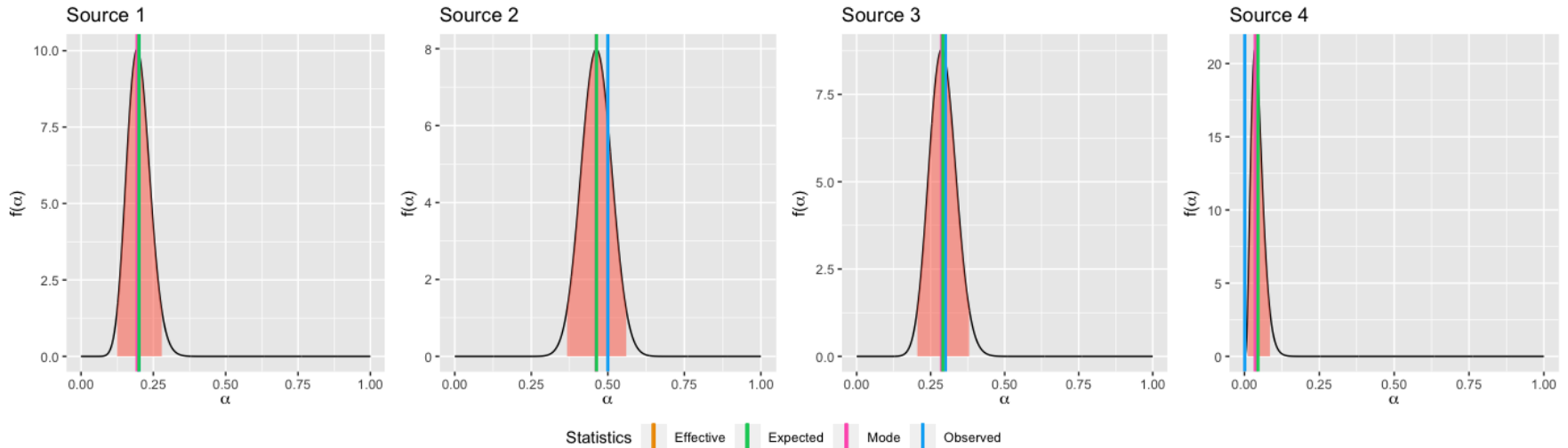
# A Case Study

When we train on all three sources:



# A Case Study

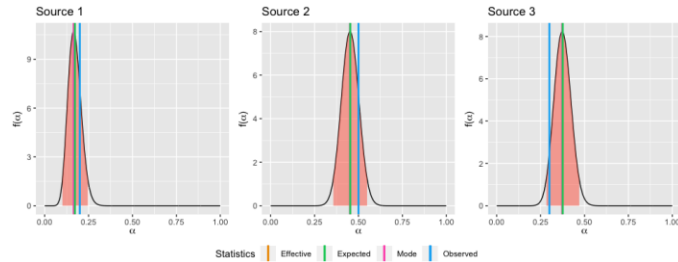
When we give the model a little more flexibility  
(What if our sample has a source that our library doesn't know about?):



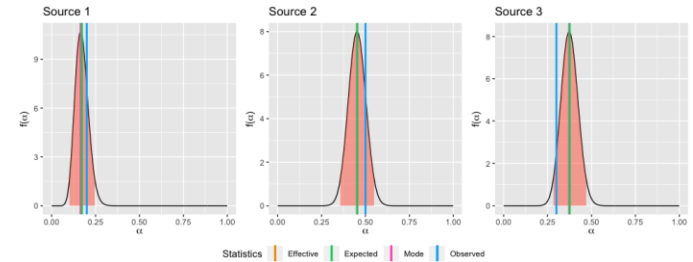
# A Case Study

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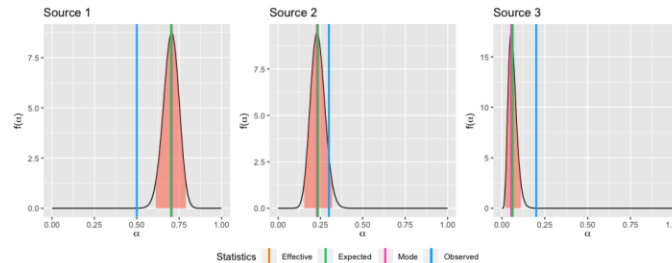
(What if our sample has a source that our library doesn't know about?):



Fullerene not in library



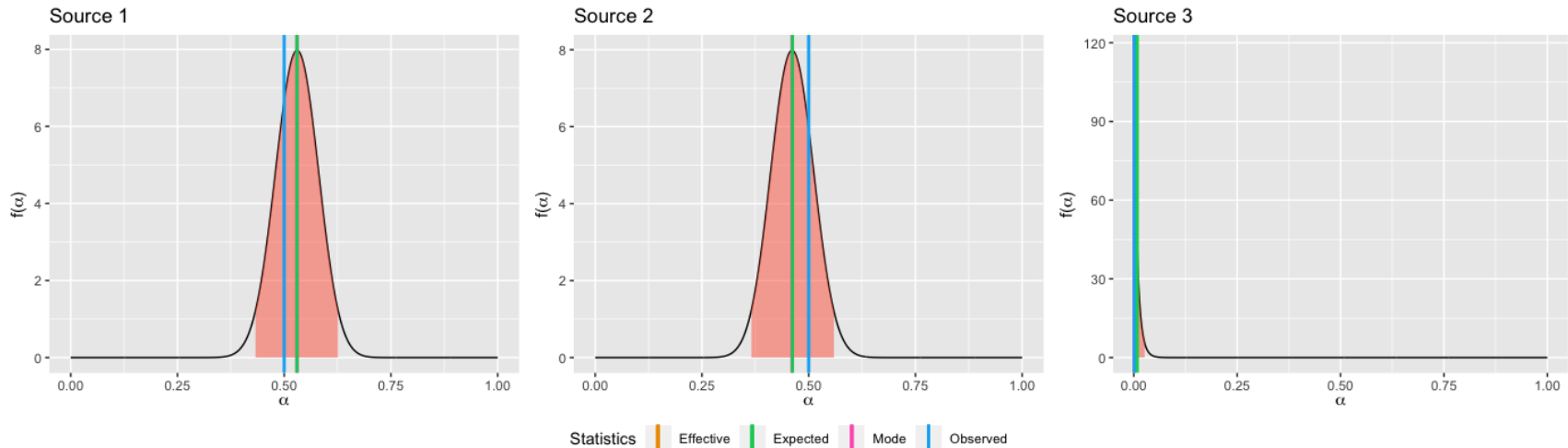
Ar not in library



Air not in library

# A Case Study

When not all of the sources are in the sample  
(What if our sample doesn't contain all of the sources in our library?):



# Takeaways:

- Given a library of potential sources, we can identify which sources are present (and in which proportions) in a mixture of soot
  - When all sources are present in the mixture
  - When only some sources are present in the mixture
- Even if a source is missing from that library, we can identify which sources are present (and in which proportions) in a mixture of soot