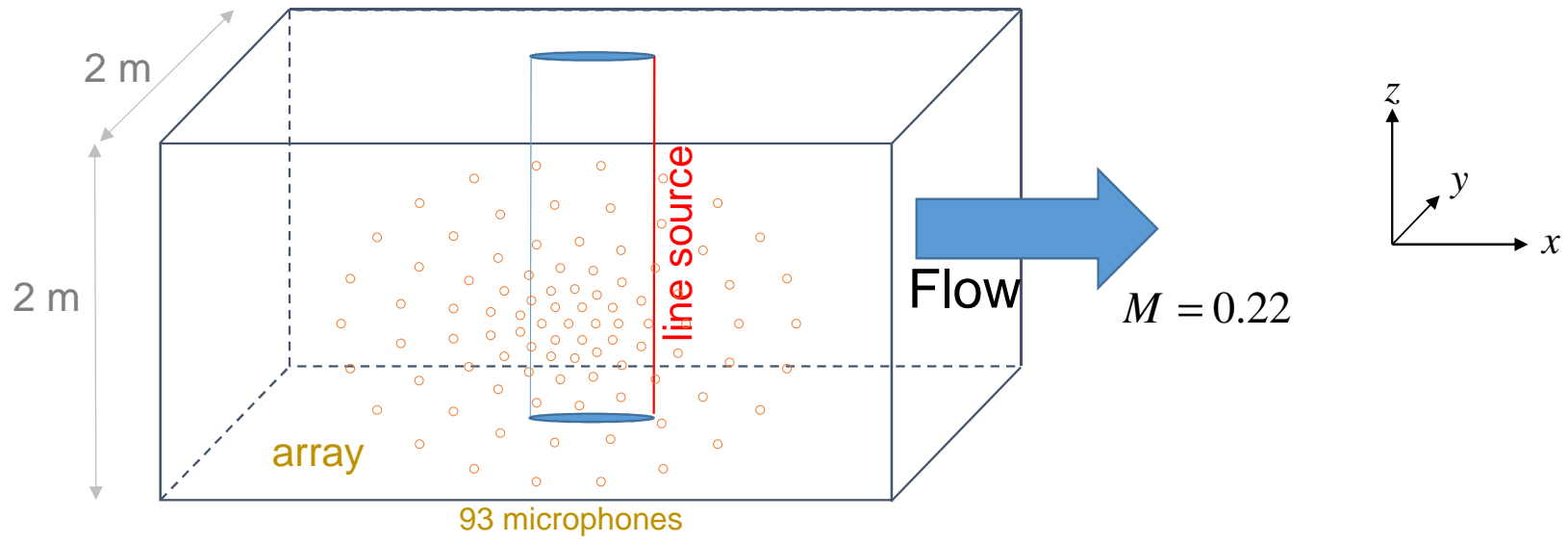


# Analytical Benchmark 1

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*PSA3*

# Analytical Benchmark 1



# Cross-Spectral Matrix

$$\mathbf{C} = \mathbf{P} + \mathbf{N}$$

Signal: line source:  
large number of incoherent  
monopoles of equal strengths

Noise:  
60 s Gaussian white noise; 51200 Hz;  
incoherent between microphones;  
FFT blocksize = 1024;  
Hanning + 50% overlap;

# Cross-Spectral Matrix

$$\begin{aligned}
 \mathbf{C} &= \left\langle \left( \sum_{k=1}^K \mathbf{p}_k + \mathbf{n} \right) \left( \sum_{k=1}^K \mathbf{p}_k + \mathbf{n} \right)^* \right\rangle = \\
 &= \sum_{k=1}^K \sum_{l=1}^K \langle \mathbf{p}_k \mathbf{p}_l^* \rangle + \sum_{k=1}^K \langle \mathbf{p}_k \mathbf{n}^* \rangle + \sum_{k=1}^K \langle \mathbf{n} \mathbf{p}_k^* \rangle + \langle \mathbf{n} \mathbf{n}^* \rangle \\
 &= \sum_{k=1}^K \langle \mathbf{p}_k \mathbf{p}_k^* \rangle + \sum_{k=1}^K \sum_{\substack{l=1 \\ k \neq l}}^K \langle \mathbf{p}_k \mathbf{p}_l^* \rangle + \sum_{k=1}^K \langle \mathbf{p}_k \mathbf{n}^* \rangle + \sum_{k=1}^K \langle \mathbf{n} \mathbf{p}_k^* \rangle + \langle \mathbf{n} \mathbf{n}^* \rangle \\
 &\rightarrow \sum_{k=1}^K \mathbf{p}_k \mathbf{p}_k^* + \langle \mathbf{n} \mathbf{n}^* \rangle = A \sum_{k=1}^K \mathbf{g}_k \mathbf{g}_k^* + \langle \mathbf{n} \mathbf{n}^* \rangle
 \end{aligned}$$

# Auto-spectra

Signal: 
$$P_{mm} = \frac{A}{(4\pi)^2 (1-M^2) \sqrt{\frac{x_m^2}{1-M^2} + \left(\frac{W}{2}\right)^2}} \left\{ \arctan \left[ \frac{\frac{H}{2} - z_m}{\sqrt{\frac{x_m^2}{1-M^2} + \left(\frac{W}{2}\right)^2}} \right] + \arctan \left[ \frac{\frac{H}{2} + z_m}{\sqrt{\frac{x_m^2}{1-M^2} + \left(\frac{W}{2}\right)^2}} \right] \right\}$$

Signal at array centre: 
$$P_{11} = \frac{A}{16\pi(1-M^2)W}$$

Noise: 
$$\langle N_{mm} \rangle^{1/2} = 10 \text{ Pa}$$

(86.89 dB per frequency bin)

# Challenge

$$N_{mm} \gg S_{mm}$$

$$\text{obtain: } S_{11} = \frac{A}{16\pi(1-M^2)W}$$

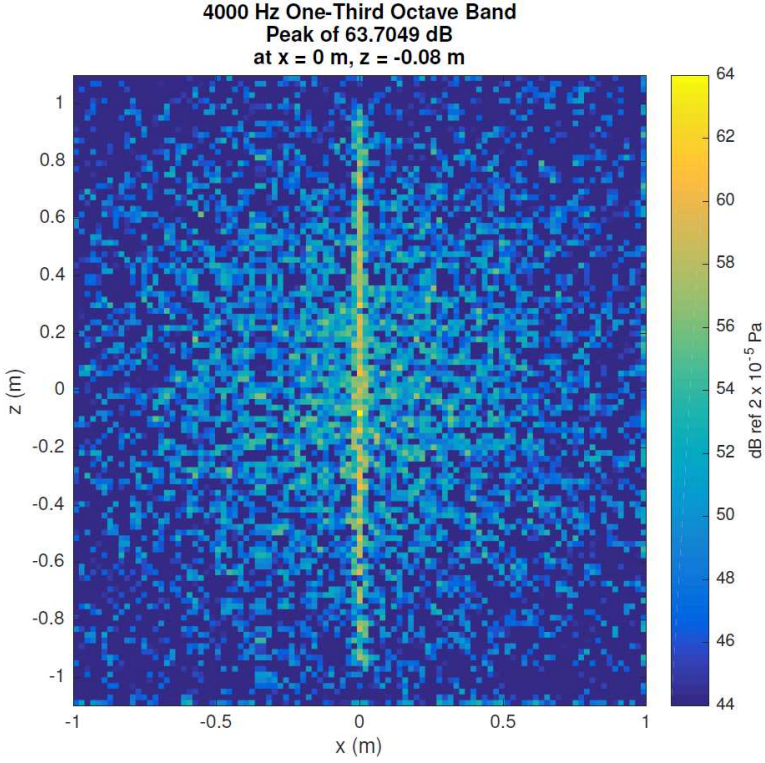
# Participants

- Ennes Sarradj, BTU
- Chris Bahr, NASA
- Ric Porteous et al, Adelaide & UNSW

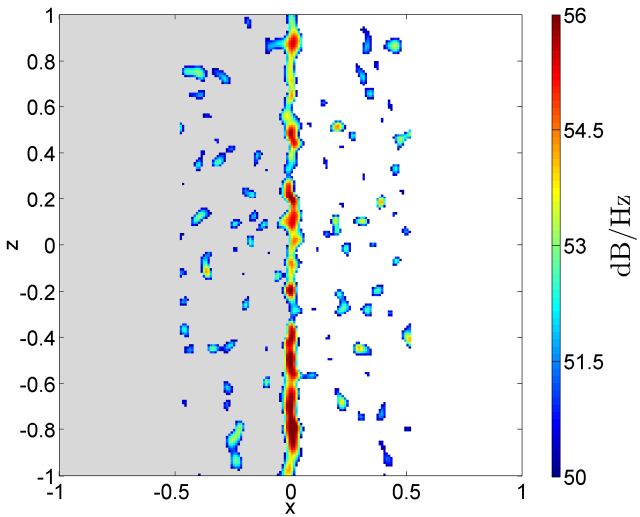
**b.tu** Brandenburg  
University of Technology  
Cottbus - Senftenberg



# Example images



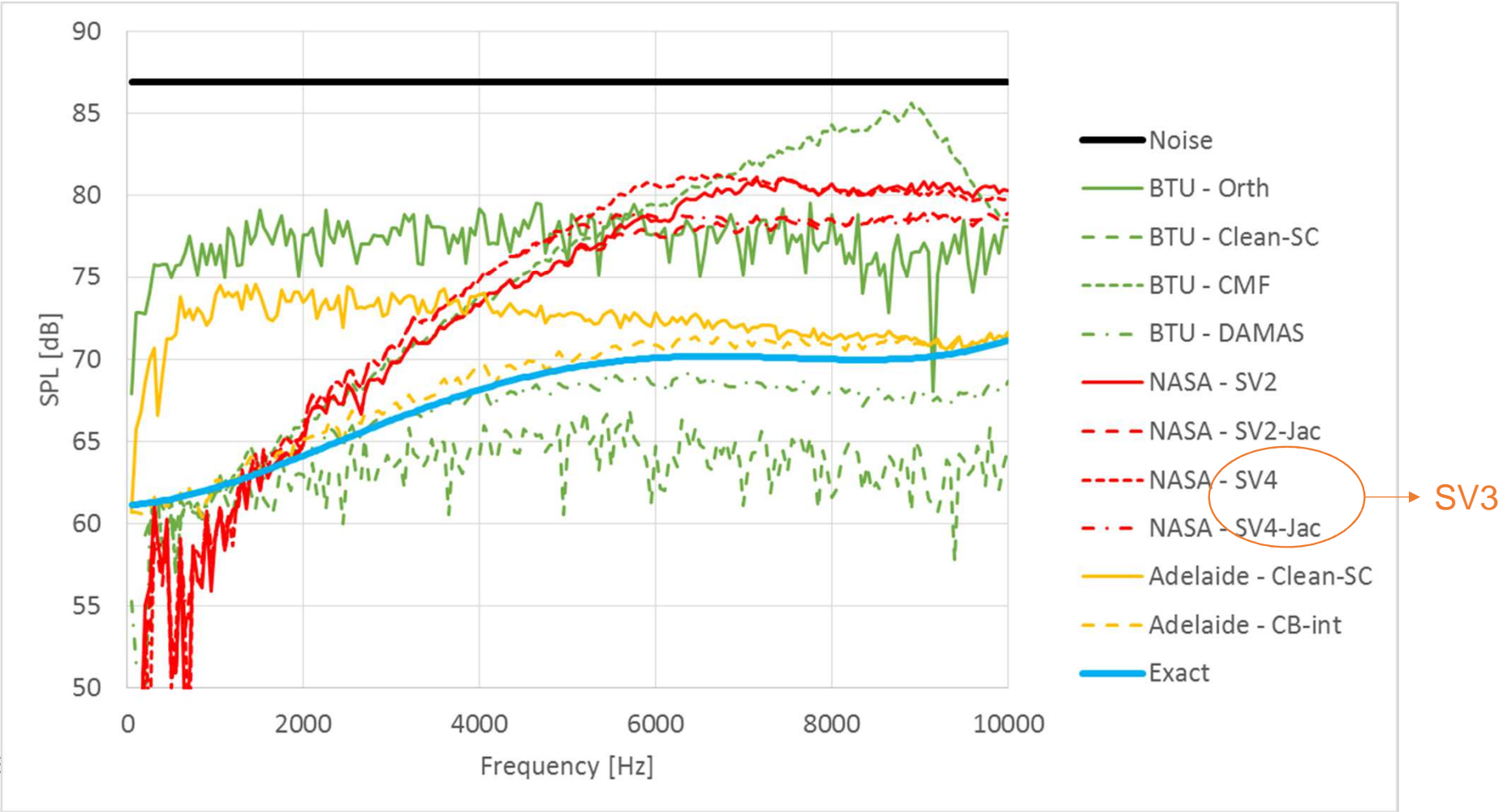
NASA



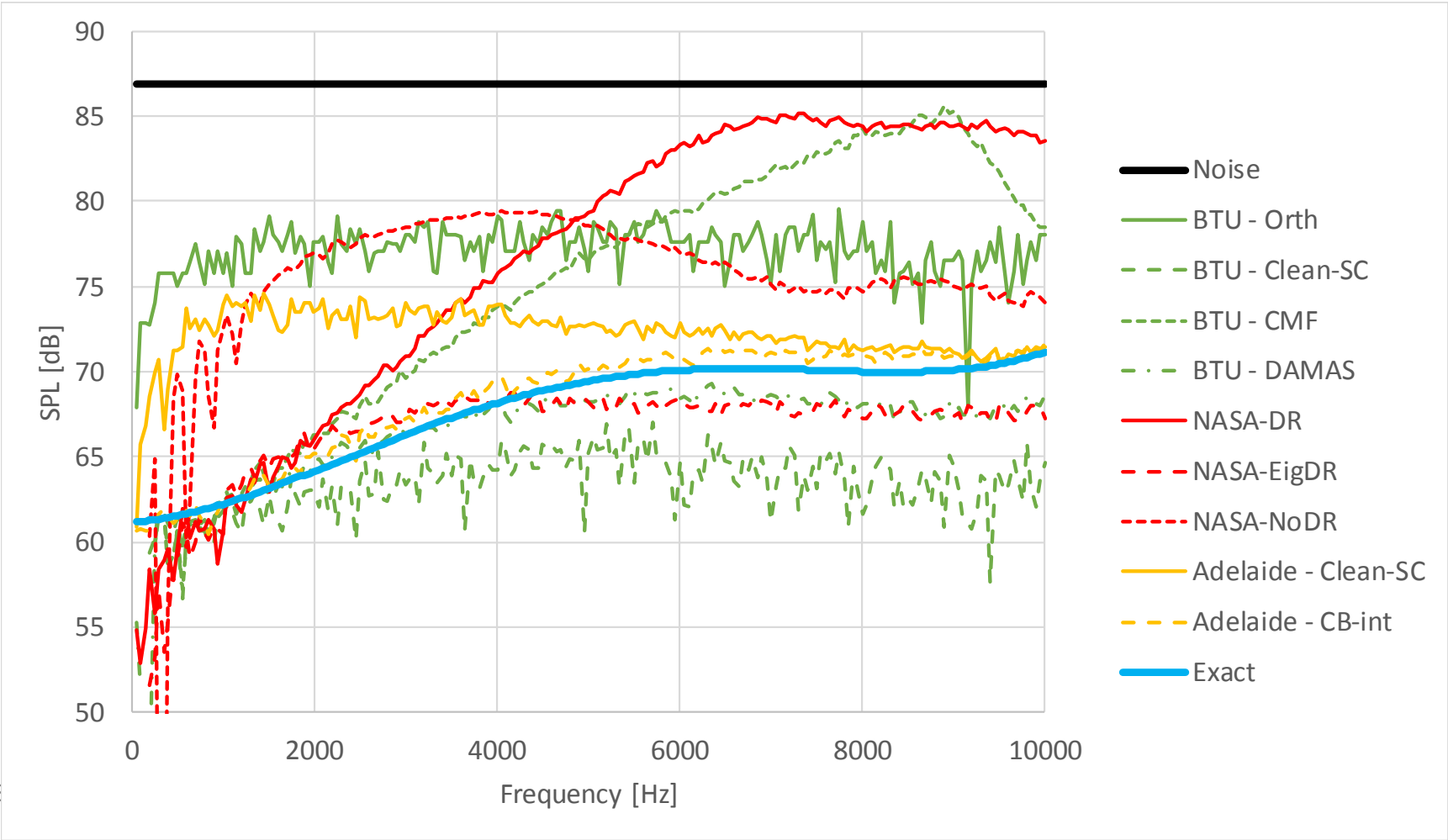
Adelaide



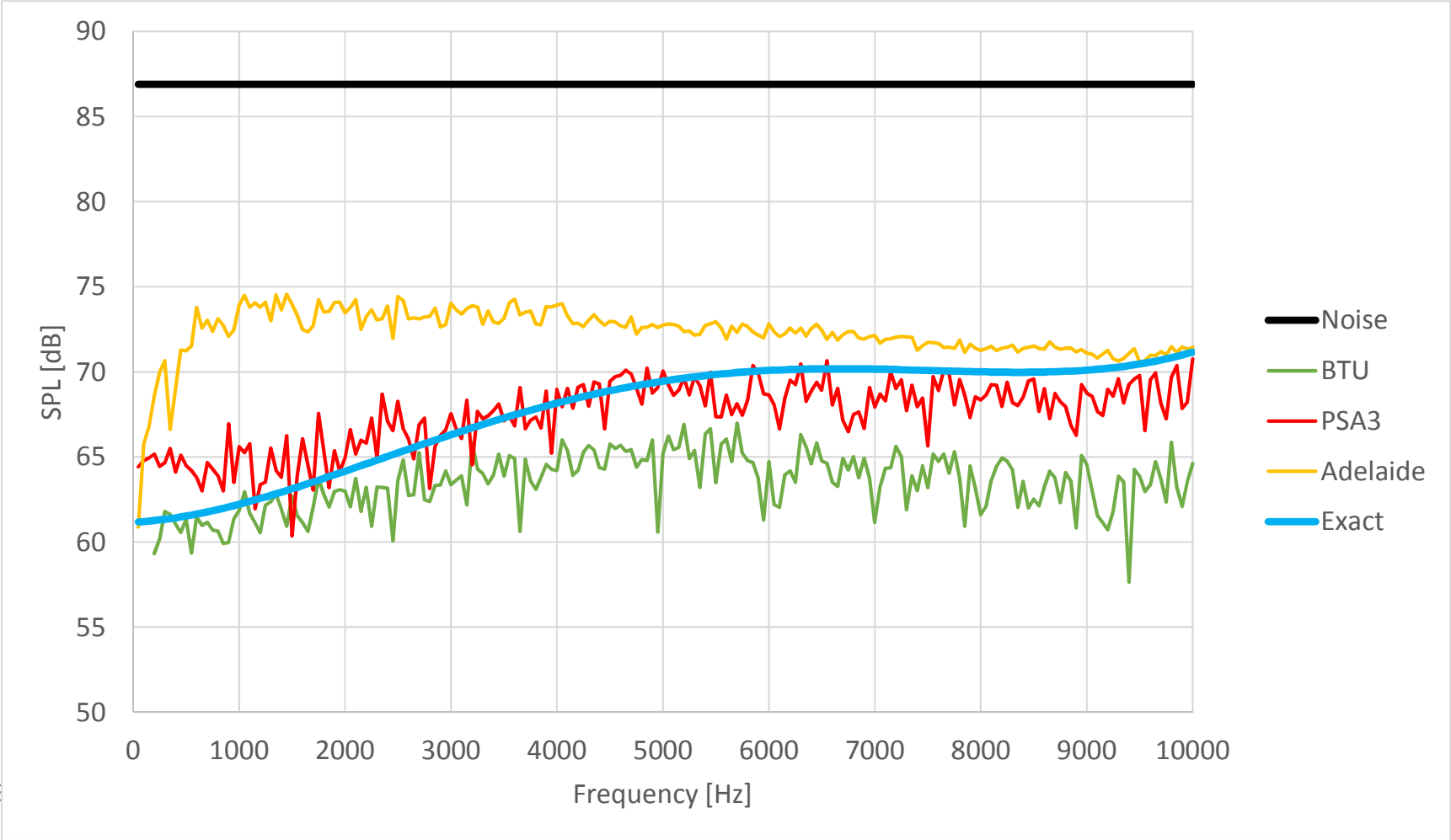
# Results (June, 2015)



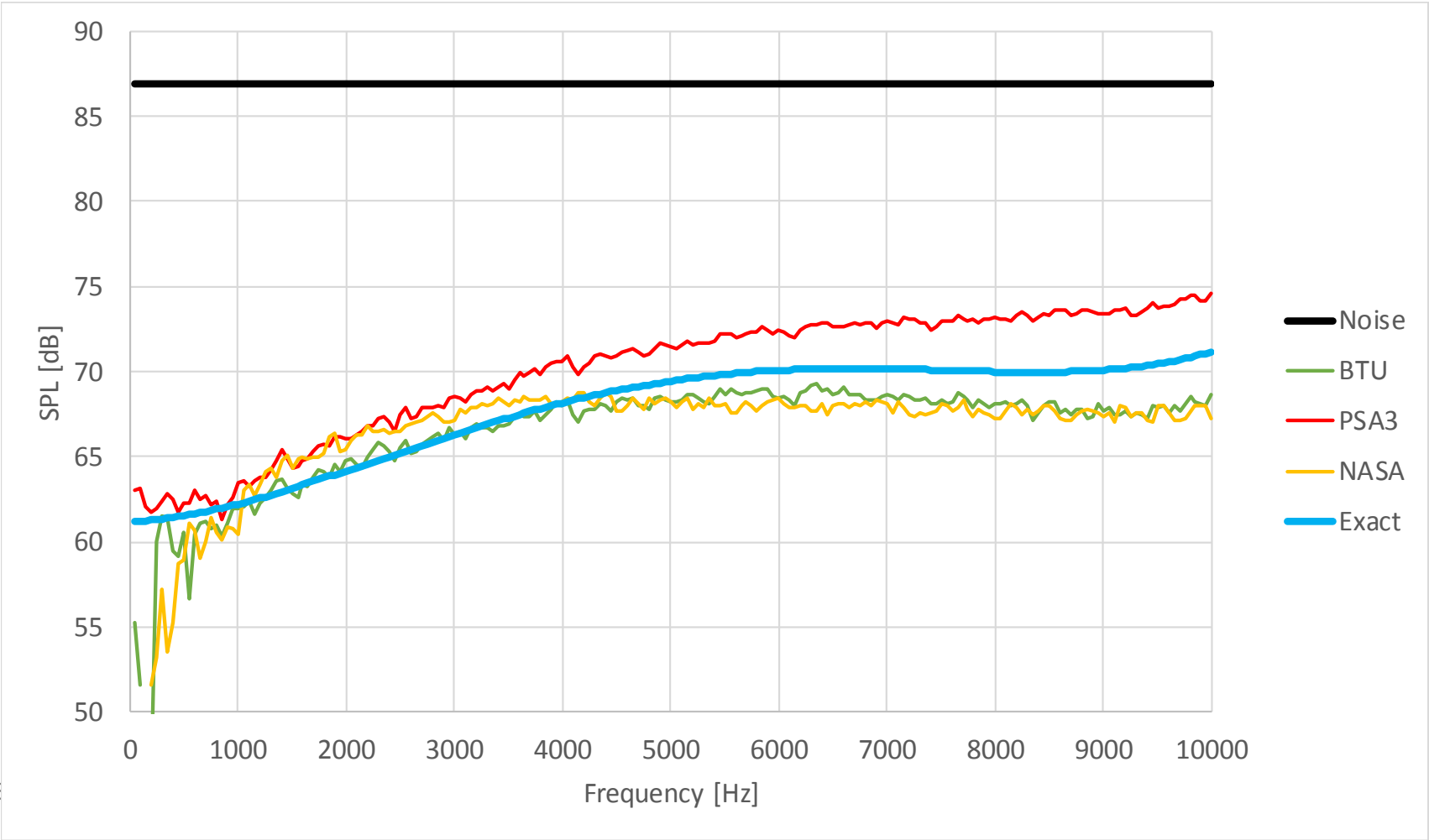
# Results (February, 2016)



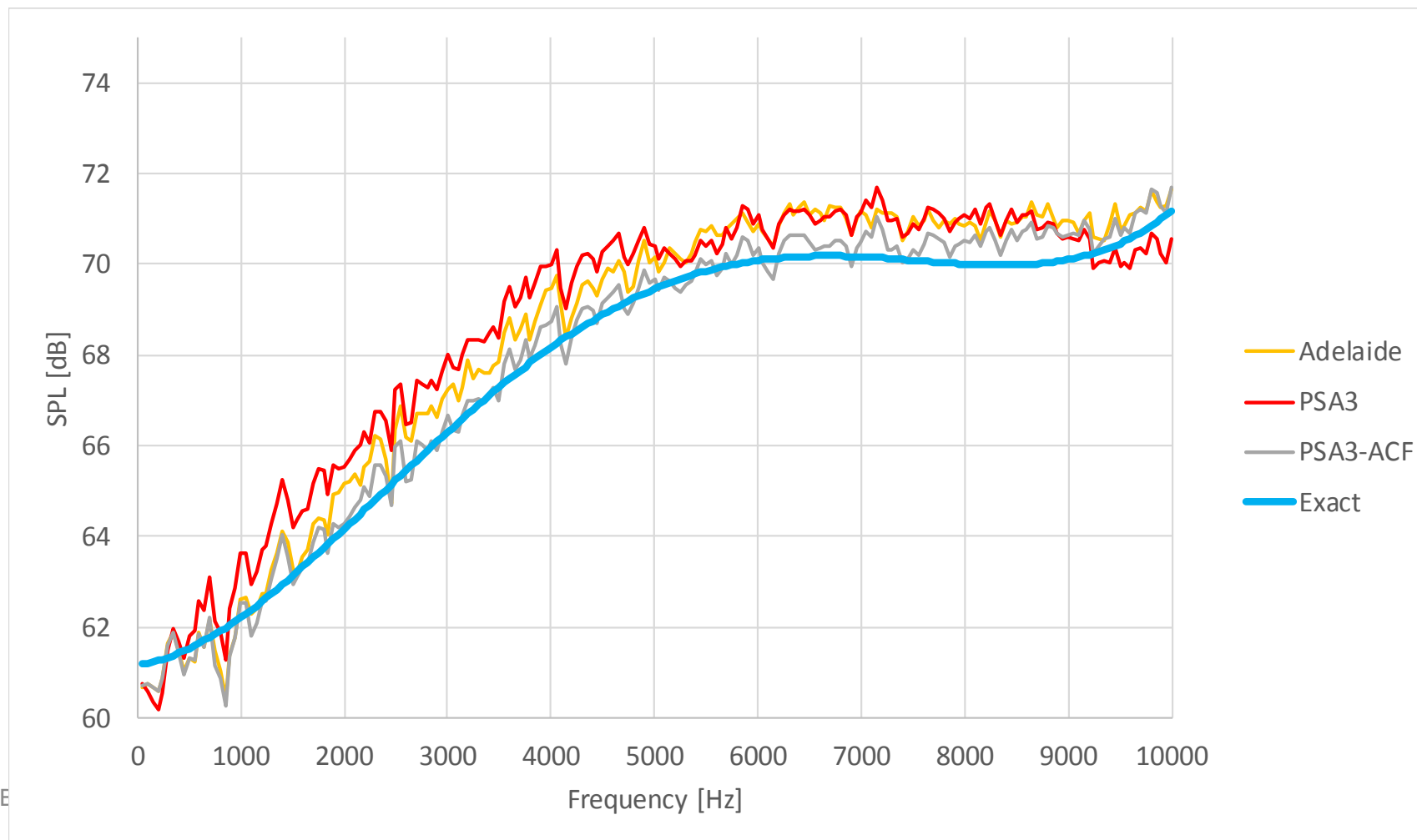
# Results: CLEAN-SC



# Results: DAMAS



# Results: Source Power Integration



# Source Power Integration

Traditional: 
$$A = \frac{\sum_{j=1}^J (\mathbf{w}_j^* \bar{\mathbf{C}} \mathbf{w}_j) A_0}{\sum_{j=1}^J (\mathbf{w}_j^* (\overline{\mathbf{g}_0 \mathbf{g}_0^*}) \mathbf{w}_j)}$$

For line source: 
$$A = \frac{\sum_{j=1}^J (\mathbf{w}_j^* \bar{\mathbf{C}} \mathbf{w}_j) A_{\text{line}}}{\sum_{j=1}^J \left( \mathbf{w}_j^* \left( \sum_{k=1}^K \overline{\mathbf{g}_k \mathbf{g}_k^*} \right) \mathbf{w}_j \right)}$$

Array calibration function

# Full line least squares

minimize  $\|\bar{\mathbf{C}} - A\bar{\mathbf{L}}\|^2 = \sum_{(m,n) \in S} |C_{mn} - AL_{mn}|^2$        $\mathbf{L}$  is CSM due to line source of unit strength

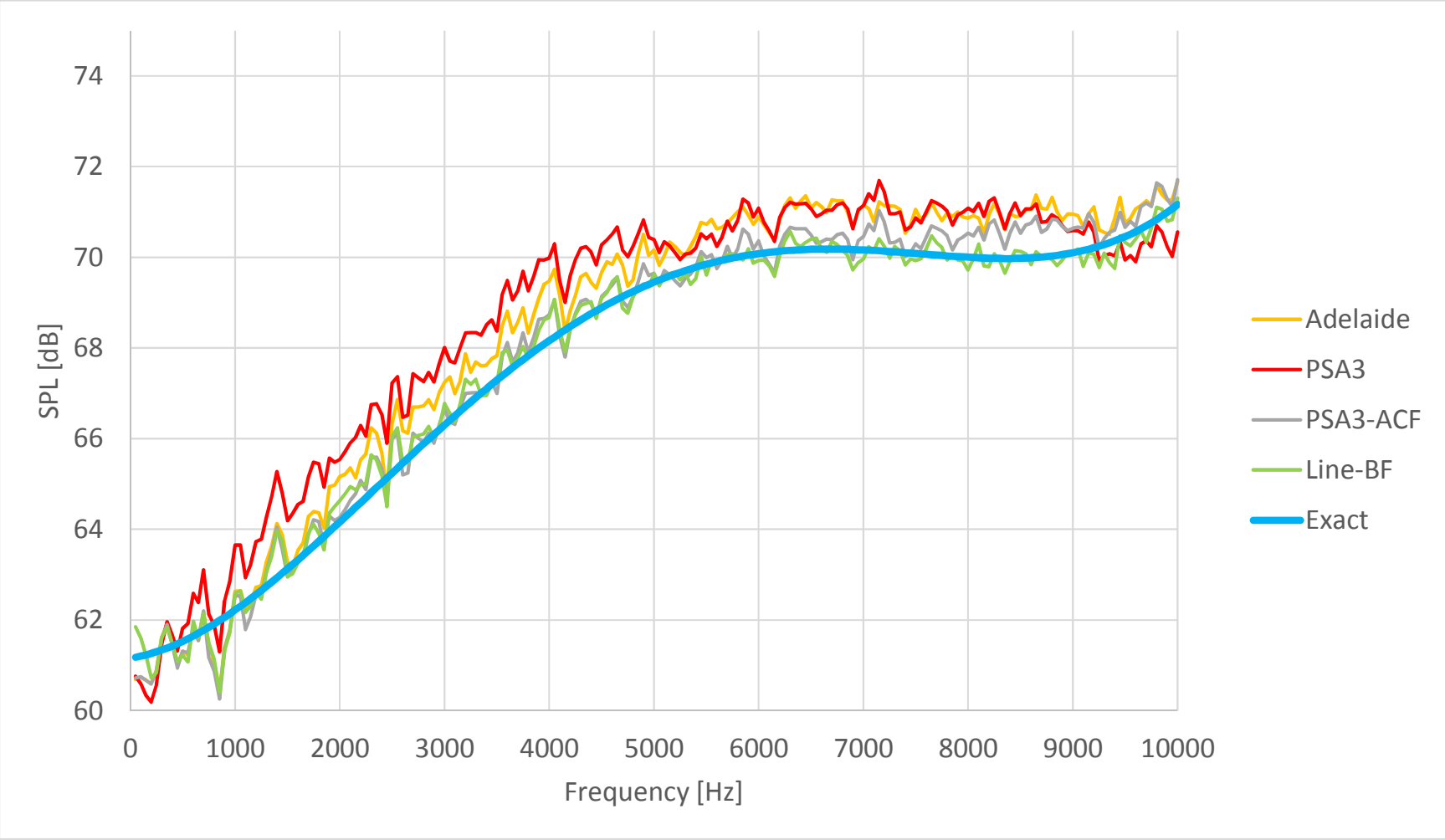
$$\text{solution: } A = \frac{\sum_{(m,n) \in S} L_{mn}^* C_{mn}}{\sum_{(m,n) \in S} |L_{mn}|^2} = \frac{\sum_{j=1}^K \left( \sum_{(m,n) \in S} \mathbf{g}_{j,m}^* C_{mn} \mathbf{g}_{j,n} \right)}{\sum_{(m,n) \in S} \left( \sum_{j=1}^K \mathbf{g}_{j,m}^* \mathbf{g}_{j,n} \sum_{k=1}^K \mathbf{g}_{k,m} \mathbf{g}_{k,n}^* \right)}$$

$$= \frac{\sum_{j=1}^K \sum_{(m,n) \in S} \left( \mathbf{g}_{j,m}^* C_{mn} \mathbf{g}_{j,n} \right)}{\sum_{j=1}^K \left( \mathbf{g}_j^* \bar{\mathbf{C}} \mathbf{g}_j \right)}$$

$$= \frac{\sum_{j=1}^K \left( \sum_{(m,n) \in S} \mathbf{g}_{j,m}^* \left( \sum_{k=1}^K \mathbf{g}_{k,m} \mathbf{g}_{k,n}^* \right) \mathbf{g}_{j,n} \right)}{\sum_{j=1}^K \left( \mathbf{g}_j^* \left( \sum_{k=1}^K \overline{\mathbf{g}_k \mathbf{g}_k^*} \right) \mathbf{g}_j \right)}$$

$$L_{mn} = \sum_{k=1}^K \mathbf{g}_{k,m} \mathbf{g}_{k,n}^*$$

# Full line least squares





# Conclusions

- Large spreading of results
- Significant differences in DAMAS and CLEAN-SC results
  - Strongly dependent on scan area
- DAMAS benefits from eigenvalue subtraction
- Power integration gives the best result
  - Correct array calibration function required
  - Similar to solving direct minimization problem
  - Remaining differences due to integration threshold