Urban Air Mobility Noise Reduction Modeling Project 49

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Project 49

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Penn State

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Cost Share Partner(s): Continuum Dynamics, Inc; Blue Ridge Research and Consulting, LLC; Sikorsky, a Lockheed Martin Company; Supernal

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Research Approach:

- Build on success of PSU Noise Prediction System (PSU-NPS) developed in ASCENT Projects 6 & 38:
 - Couple flight simulation, aerodynamic modeling (CDI's CHARM), and PSU-WOPWOP
- Tailor approach to unique characteristics of UAM by modeling flight dynamics of distributed electric propulsion vehicles including multiple propellers and rotors with DEPSim
- Develop low noise UAM trim strategies

Objective:

- Develop a physics-based noise modeling system for future UAM aircraft with varied configurations
- Produce reference noise database for notional UAM configurations for hover, transition, cruise
- Identify configuration changes and operational strategies that minimize acoustic impacts

Project Benefits:

- Physics-based UAM noise predictions and reference noise database
- Better understanding of UAM noise characteristics
- Guide design of maneuvers with focus on noise, aircraft controllability, performance
- Development of semi-empirical models for rapid operational noise prediction

Major Accomplishments (to date):

- Dataset of measurements by Brooks, Pope & Marcolini (BPM) digitized for analysis
- Novel turbulent boundary layer trailing edge noise trends observed
- Background research on adapting Pegg helicopter noise model for Distributed Electric Propulsion (DEP) / multirotor aircraft

Future Work / Schedule:

- Retune constants of BPM semiempirical airfoil self-noise model based on angle of attack dependent scaling trends observed
- Understand trends in parameters tuned for Pegg model for distributed electric propulsion aircraft configurations

Presentation Outline

- Introduction, Objectives, Approach, Outcomes
- PSU-WOPWOP broadband noise prediction
 - Enhancement of the Brooks, Pope and Marcolini trailing edge noise model
 - Rederivation and tuning of the Pegg broadband noise model
- Summary:
 - Accomplishments
 - Future work





Motivation

- Rapidly growing interest in the development and use of Urban Air Mobility (UAM) aircraft
- Noise is widely recognized as one of the barriers to public acceptance of UAM operations
- UAM are a new category of aircraft, and their acoustic characteristics are not well understood: new important physics
- Information about UAM noise is needed to:
 - Design quiet configurations
 - Understand how to operate UAM quietly
 - Inform the approach to noise certification
 - Understand the impact on communities



Wisk Cora



BETA Ava XC





Objectives

Near term:

- Develop a physics-based noise modeling system for future UAM aircraft with varied configurations
 - PSU Distributed Electric Propulsion Simulation (DEPSIM)
 - Continuum Dynamics, Inc's CHARM rotor analysis
 - PSU-WOPWOP noise prediction

Long term:

- Produce a database of noise predictions for notional UAM configurations across a wide range of operating modes: hover, transition, cruise
- Identify configuration changes and operational strategies that minimize acoustic impacts





Approach

- Build on success of helicopter PSU-NPS developed under ASCENT Projects 6 & 38
- Tailor approach to unique characteristics of UAM:
 - Use PSU Distributed Electric Propulsion Simulation (DEPSim) to model flight state of multiple propellers and rotors
 - Model unsteady aerodynamic loading using CDI's CHARM
 - Couple DEPSim with PSU-WOPWOP
 - Improve PSU-WOPWOP for better computational efficiency with large numbers of rotors or propellers
 - Update and generalize broadband noise modeling capabilities of diverse fidelity
- Develop low noise UAM trim strategies
 - Assess tradeoffs between noise, safety & performance
 - Evaluate practicality and "flyability" of non-unique trim strategies using Penn State flight simulation facilities





Outcomes and Practical Applications

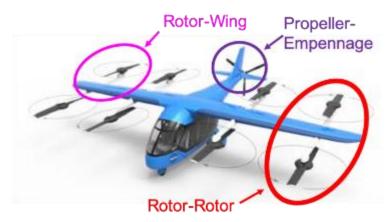
Outcomes:

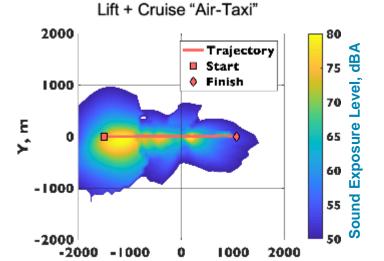
- Develop initial capability to predict UAM acoustics
- Improved understanding of UAM noise characteristics
 - Levels, directivity, spectral content
 - Variations between configurations
 - Variations between operating modes
- Identification of noise reduction opportunities
 - Low noise configurations
 - Noise abatement flight operations

Practical Applications:

- Low noise design tool for the UAM industry
- Data to support the development of certification procedures
- Initial set of representative UAM noise data for integration with FAA/DOT tools

NASA UAM Reference Vehicle





X, m





Broadband Noise Prediction





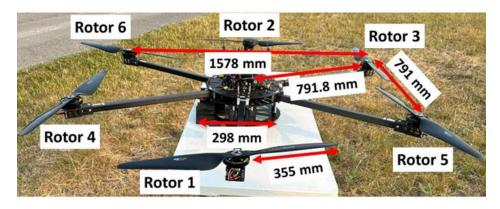
- Low-tip Mach number rotors: high-frequency broadband noise dominant (annoying to humans)
- Several sources of broadband noise: rotor-wake interactions, atmospheric turbulence, self-noise
- Contribution of sources dependent on flight conditions
 - UAV speed, angle of climb/descent, etc.
- Important low-altitude flight condition: hover
 - Expected during landing, takeoff, package drop etc.
 - High noise contribution with decreasing distance from humans on-ground
 - Self-noise expected to be dominant source of broadband noise (in *absence* of other contributions)



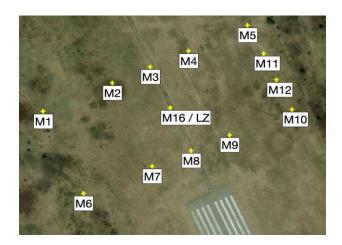


Uncrewed Aerial Vehicle (UAV) Noise Prediction

- Noise from small UAVs characterized by high-frequency broadband noise
 - Airfoil self-noise is dominant high-frequency
 (>1000 Hz) source in hover
 - BWI (blade-wake interaction) might also be present (usually around 500 1000 Hz)
- Broadband noise predictions using BPM model for UAV do not match well with outdoor data measured
- Reconfigurable hexacopter at PSU
 - Well instrumented: aircraft & rotor states measured (including RPM)
 - Large outdoor microphone array at Mid-State Regional Airport



Reconfigurable Hex



Microphone locations





BPM Self-Noise Model Review

Equations in BPM are based on first-principles scaling functions

$$SPL_{TBLTE} = 10 \log_{10}(\frac{\delta^* M^5 L \overline{D}_h}{r_e^2}) + A(\delta^*, St)$$

- M, δ^*, L, r_e are the flow Mach number, boundary layer displacement thickness, airfoil span and distance from trailing edge to observer, St is the Strouhal number (fL/U)
- $A(\delta^*, St)$ is tuned with measured data
- \overline{D}_h is normalized directivity function: Brooks et al. (1989) states directivity function by assuming acoustic compactness: chord/wavelength > 1
- Reasonable assumption for helicopter blades but not for UAV blades

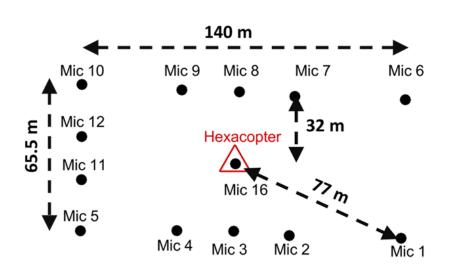
Acoustic Frequency (Hz)	Acoustic Wavelength (m)	Chord/Wavelength	
		PSU Hex rotor (chord ≈ 0.05 m)	UH-60 rotor (chord ≈ 0.53 m)
1000	0.343	0.14	1.54
2000	0.1715	0.29	3.09
3000	0.1143	0.43	4.63

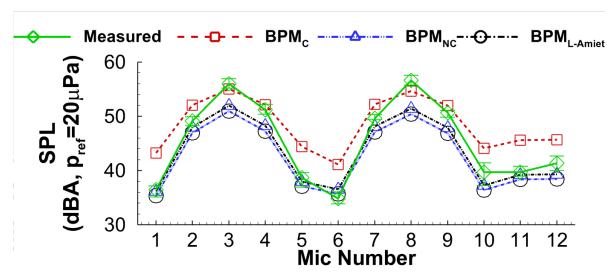




Results: PSU Hexacopter Hover 40 ft (over Mic 16)

- BPM_C: Original BPM model with compact directivity function
- BPM_{NC}: Modified BPM model with non-compact directivity function
- BPM_{L-Amiet}: Modified BPM model with numerical directivity function (valid for all blades)
- Accounting for acoustic non-compactness improves predictions
 TE noise model should not overpredict since other broadband noise sources are unaccounted for





Mic grid layout

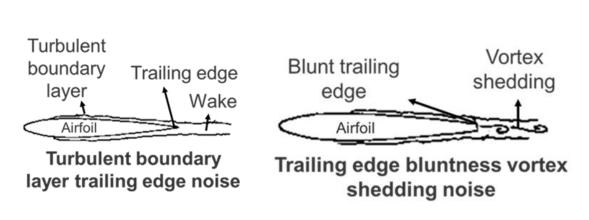
(Mukherjee 2025) OASPL variation vs mics 1 - 12



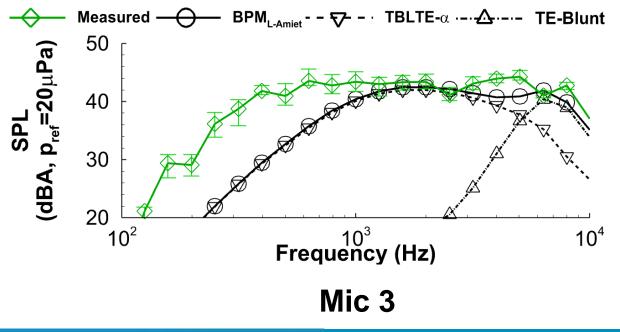


Results: PSU Hexacopter Hover 40 ft (over Mic 16)

- Improvements in spectral match at several positions
- ≈ 1000 Hz peak is due to turbulent boundary layer trailing edge (TBLTE) noise
- ≈ 7000 Hz peak due to bluntness of trailing edge (TE-Blunt)



Airfoil self-noise mechanisms (adapted from Brooks et al. 1989)

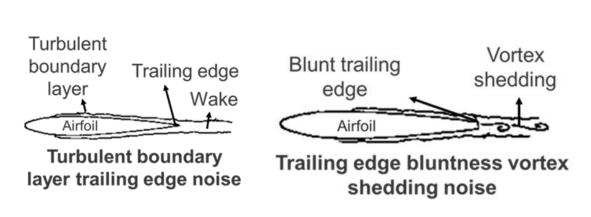




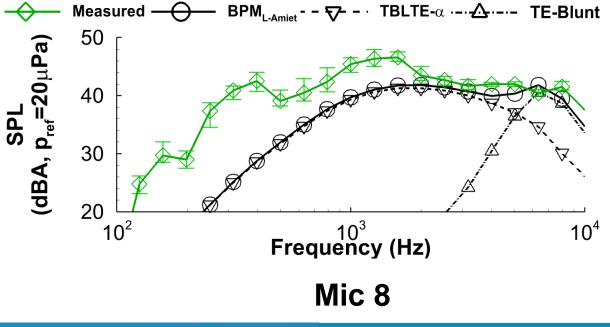


Results: PSU Hexacopter Hover 40 ft (over Mic 16)

- Differences in some locations remain
- ≈ 1000 Hz peak missed by nearly 6 dBA
- TBLTE model needs further improvements



Airfoil self-noise mechanisms (adapted from Brooks et al. 1989)





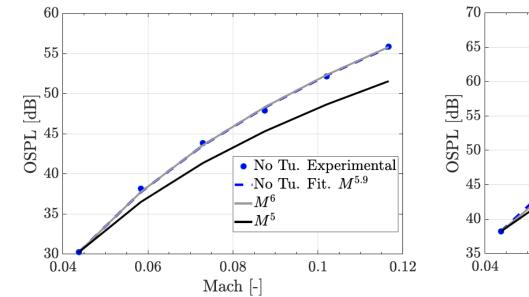


Modifying TBLTE Scaling Law

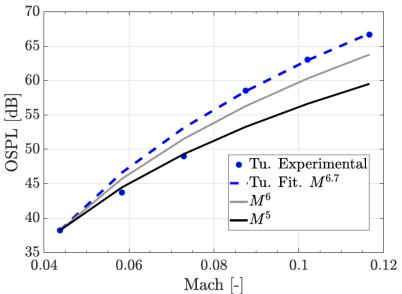
• Original M^5 law was analytically derived for flat plate with no external turbulence in flow

$$SPL_{TBLTE} = 10 \log_{10}(\frac{\delta^* M^n L \overline{D}_h}{r_e^2}) + A(\delta^*, St)$$

• Santos et al. (2024) measured trends in noise from airfoil in wind tunnel with flow turbulence -n=6.7 depending on turbulent intensity (Reynolds number = 3e+5, angle of attack = 0°)



0.08% turbulent intensity



12.5% turbulent intensity





TBLTE Scaling in data by Brooks et al. (1989)

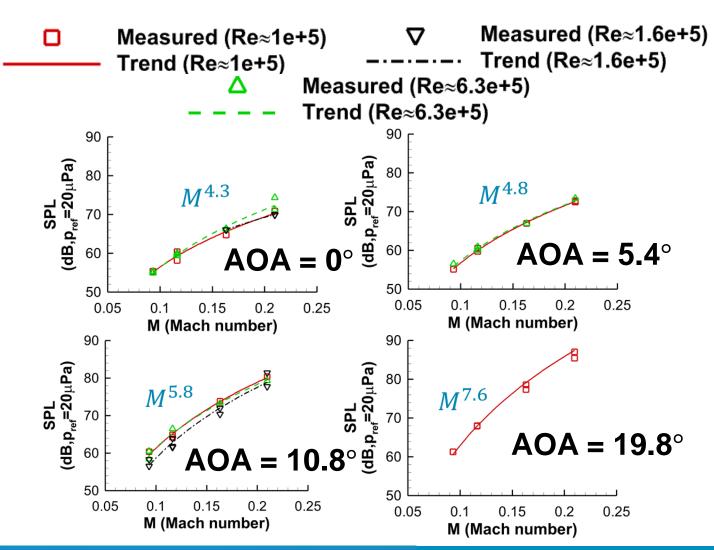
- Brooks et al. (1989) did *not* report any analysis of total sound pressure level (SPL)
- They provide experimental 3rd octave spectra for nearly 104 flow conditions
- Variation in flow Mach number, airfoil angle of attack (AOA), Reynolds number
- Data were digitized from the original report and processed to estimate total SPL





TBLTE Scaling in data by Brooks et al. (1989)

- Mach number (Mⁿ) scaling estimated for different angle of attack (AOA) & Reynolds number (Re)
- For a given AOA, Re has little impact on Mⁿ scaling
- Mach scaling exponent "n" increases with angle of attack
 - Turbulence in flow increases with angle of attack
- Should BPM model be recalibrated to account for this variation ?







Pegg Broadband Noise Prediction Model

Originally developed for predicting helicopter broadband noise (rotor-wake interaction and trailing edge noise)

$$SPL_{1/3} = 20 \log \left(\frac{V_T}{c_0}\right)^3 + 10 \log_{10} \left[\frac{A_B}{r^2} (\cos^2(\theta_1) + 0.1)\right] + S_j + f(\overline{C}_L) + \mathbf{130}$$

Rederiving the Pegg model revealed the following expression:

$$SPL_{1/3} = 20 \log \left(\frac{V_T}{c_0}\right)^3 + 10 \log_{10} \left[\frac{A_B}{r^2} (\cos^2(\theta) + 0.1)\right] + S_j + f(\overline{C}_L) + 10 \log_{10} \left[\frac{Kr_K^2 c_0^6 10^{16.29}}{\cos \theta_K}\right]$$

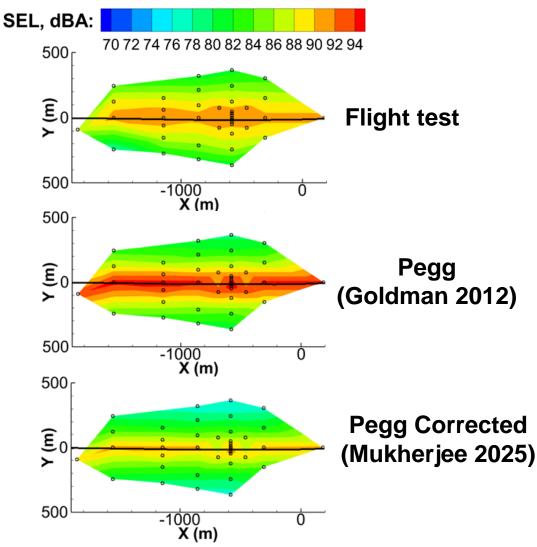
- Parameter "K" is tuned based on noise measured at a microphone, distance r_K and polar angle $heta_K$
- Advantage of Pegg model: Simple, details of rotor geometry and aerodynamics not required (unlike BPM model)
 - Can be used to develop more empirical models for tools such as AEDT





Pegg Model: PSU-WOPWOP Correction

- Error in expression for the mean lift coefficient of the rotor \overline{C}_L
 - Expression by Goldman (2012): $\overline{C}_L = \frac{6T}{\rho A_B V_T^2}$
 - Pegg's expression: $\overline{C}_L = \frac{T}{\frac{1}{2}\rho V_{0.7}^2 A_B} = \frac{4.081T}{\rho A_B V_T^2}$
 - Not documented in original report
- Overprediction up to 13 dBA for heavily loaded rotor
- Simulation of Bell 206L in 80 knots level flyover



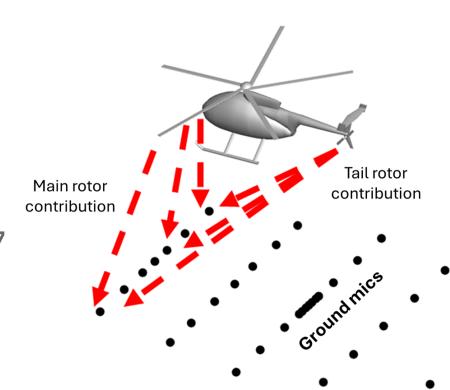




Pegg Model: Retuning With FAA/NASA 2017 Flight Test Data

$$SPL_{1/3} = 20 \log \left(\frac{V_T}{c_0}\right)^3 + 10 \log_{10} \left[\frac{A_B}{r^2} (\cos^2(\theta) + 0.1)\right] + S_j + f(\overline{C}_L) + 10 \log_{10} \left[\frac{Kr_K^2 c_0^6 10^{16.29}}{\cos \theta_K}\right]$$

- Several unknowns in the general equation need to be determined
 - K: constant that helps match the total SPL
 - S_i: spectral weights for octave bands
 - Pegg provides values for 13 octave bands: constant regardless of rotor geometry, flight condition
- Total unknowns for main & tail rotor: 13+13+1=27
 - Minimum of 3 mics required to obtain an answer
 - Additional mic data should be used to obtain more robust values that correct for limitations of analytical terms



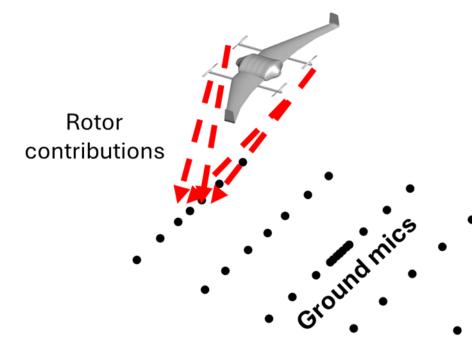




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- Several unknowns in the general equation need to be determined
 - K: constant that helps match the total SPL
 - S_i: spectral weights for octave bands
 - Pegg provides values for 13 octave bands: constant regardless of rotor geometry, flight condition
- Distributed Electric Propulsion (DEP) aircraft:
 - How many mics needed ?
 - Placement of mics ?







Summary

Major Accomplishments:

- Including acoustic non-compact effects improves sound pressure level and spectral match
- Turbulent boundary layer trailing edge noise underpredicted at some microphone locations
- Large dataset published by Brooks et al. digitized for analysis
- New scaling laws derived from data were found to depend upon angle of attack, but not flow Reynolds number
- Calibration of Pegg model for multirotor aircraft

Future Work:

- Recalibrate BPM model based on Mach scaling laws observed in data
- Compare calibrated model with flight test data for PSU Hexacopter
- Calibration of Pegg model for multirotor aircraft
 - Investigate robustness of tuned parameters using flight test data
 - Compare trends with helicopter broadband noise





References

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