

# ASCENT Project 10

## Aircraft Technology Modeling & Assessment

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Cost Share Partner: Boom Supersonics, Gulfstream, Georgia Institute of Technology



**Objective:** Model and assess potential evolution of commercial airline fleet due to the introduction of future supersonic aircraft and how technology development could affect the environmental impacts of aviation (e.g., fleet-level fuel burn, emissions and noise). The effort will examine *SST vehicle modeling (in support of CAEP Exploratory Study); fleet route simulation; fleet simulation, and AEDT supersonic modeling.*

**Project Benefits:** Provide an understanding of how introduction of new supersonic transports that could enter into commercial airline service and private use will affect fleet-wide fuel burn, noise and emissions.

### Research Approach:

#### SST Vehicle Modeling:

- CFD based aero shaping; installed propulsion modeling; mission analysis; emissions and LTO noise analysis
- Perform design Mach trade study for three SST classes
- Model facsimile of OEM SST for CAEP E-Study

#### Fleet Route Simulation:

- Computing potential time savings per OD pair
- Computing value of travel time savings per OD pair
- Detailed SST aircraft performance on complex mixed missions

#### AEDT SST Modeling:

- Generate SST perf. data; aero and propulsion truth models
- Construct appropriate physics-rooted regressions to model drag, thrust, and fuel-burn, fit, and validate
- Develop implementation plan to incorporate into AEDT

### Major Accomplishments (to date):

**SST Vehicle Modeling:** Completed 11 SSTs for design Mach trade study; completed 3 OEM vehicles; completed study on VRNS impact on climb out NOx; completed nvPM study

**Fleet Route Simulation:** Developed flexible route optimization tool; Completed future SST demand study where demand depends on vehicle; Support for CAEP E-Study

**AEDT SST Modeling:** Completed data generation and initial model development for ~7 SSTs; developed and implemented Python-based validation tool; setup engagement b/w OEMs & FAA

**Future Work / Schedule:** Revisit noise modeling assumptions for all vehicles based on feedback (11/2021), Investigate subspace optimization approach for aero shaping (12/2021); complete fleet simulations (12/2021); Refine SST regressions on all SSTs (12/2021); Validate regression approach for all SSTs (12/2021); Supersonic interdependency study (9/2022)

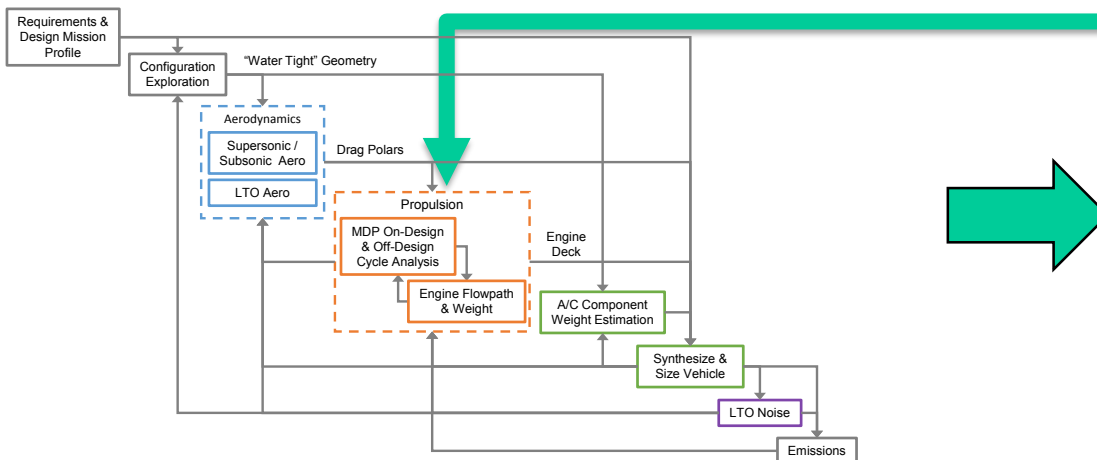
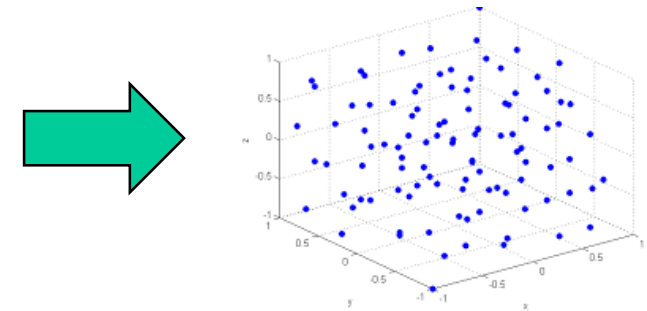
# SST Vehicle Modeling: Research Approach

For each vehicle class / Mach number combination, explored a design space of 100,000+ designs by varying the following parameters:

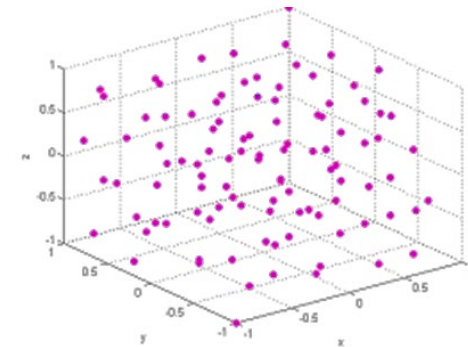
## 1. Establish design variable and ranges

Trajectory Parameters	Engine Cycle Parameters	Vehicle Design Parameters	Airframe Geometry
Takeoff De-rate % Thrust	Fan Pressure Ratio	Takeoff Thrust to Weight Ratio	Trailing edge sweep (inboard and outboard)
% Programmed Lapse Rate	HPC Pressure Ratio	Takeoff Wing Loading (weight to wing area ratio)	Twist (root, mid and tip)
Acceleration During for 2 <sup>nd</sup> Segment	Throttle Ratio (i.e., turbine rotor inlet temperature)		Taper ratio (inboard and outboard)
2 <sup>nd</sup> Segment End Altitude			Aspect Ratio
Cutback Altitude			Dihedral (inboard and outboard)
			Wing break location

## 2. Create set of configurations (design space) to be explored

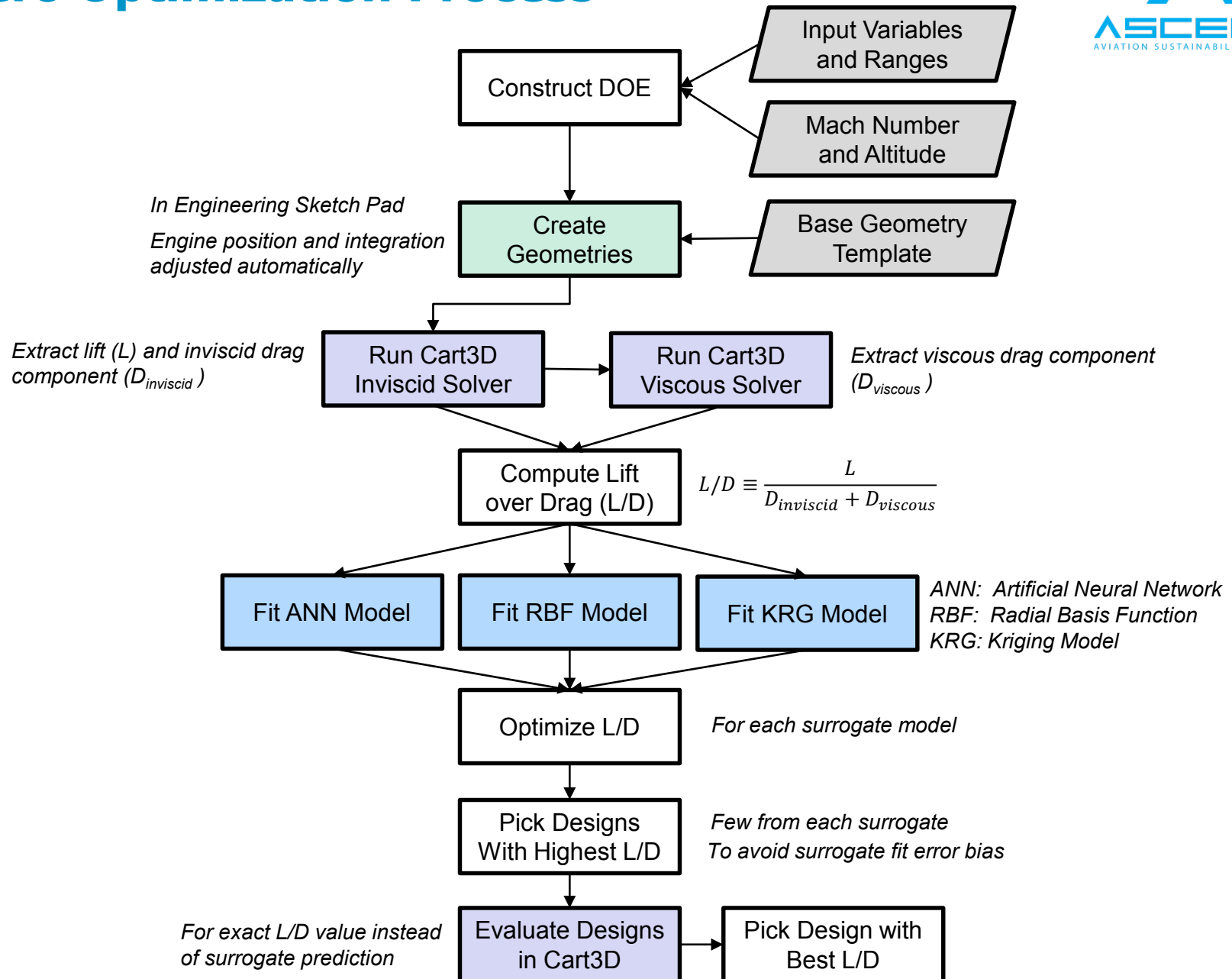


## 3. Simulate Vehicle with FASST M&S Environment

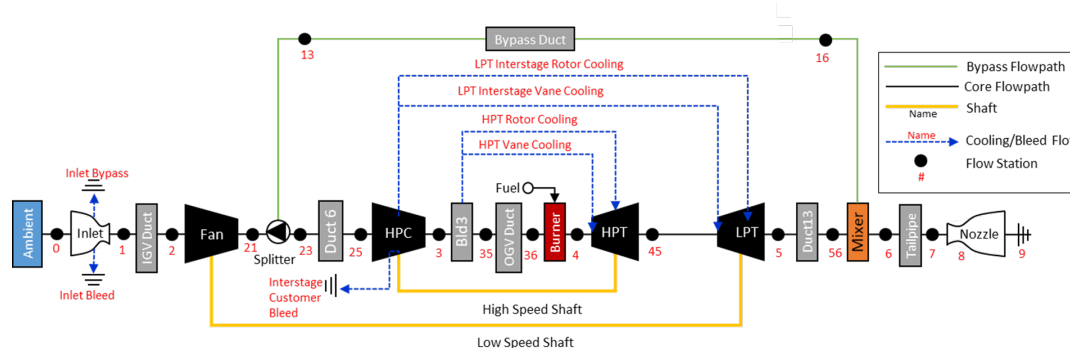


## 4. Record metrics of interest (fuel burn, noise, emissions, etc.)

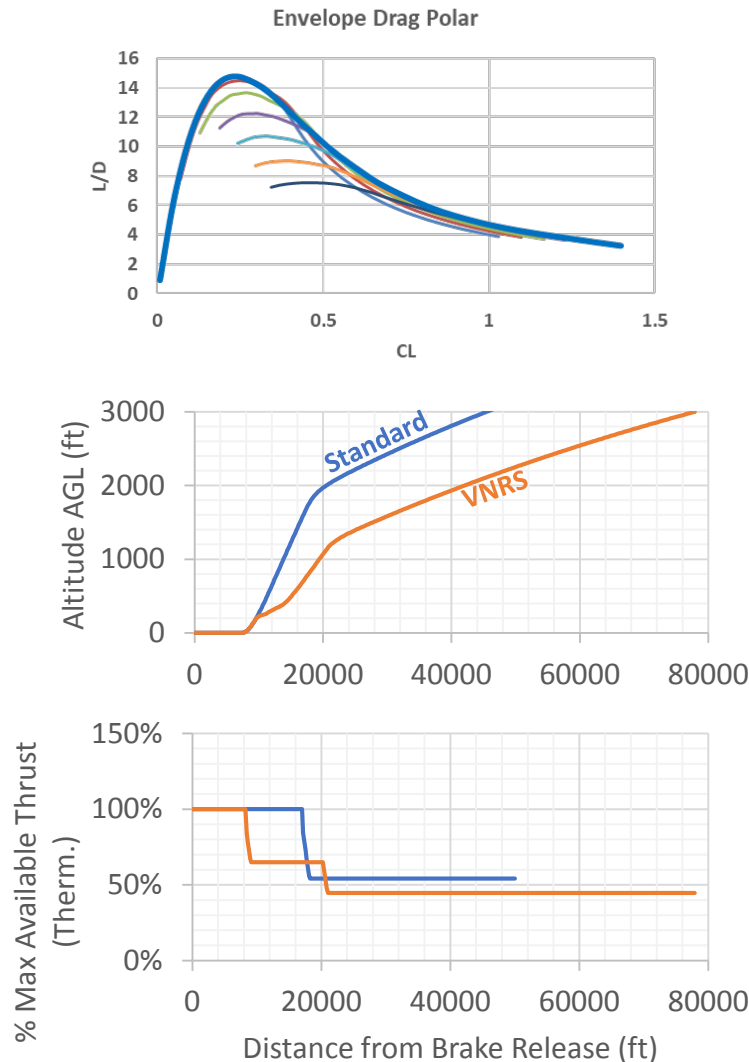
# Aero Optimization Process



- Non-afterburning, two-spool, mixed-flow turbofan
- Cycle/performance modeled with NPSS and weight modeled with WATE++
  - Parametric maps for inlet and nozzle from PIPSI/INSTAL
  - Parametric maps for fan and HPC from CMPGEN with design point computed with simple meanline code
  - Parametric maps for HPT and LPT scaled for design point computed with simple meanline code
  - Cooling flows modeled using updated CoolIt algorithm for advanced cooling tech
  - Nominal losses assumed for ducts, burner and mixer
- Multi-design point (MDP) approach was used to simultaneously meet requirements at multiple flight conditions
- Off-design power management uses both fuel flow and nozzle throat
  - Typically hold fan op-line as thrust changes
  - For LTO: reduce thrust initially along fan constant speed line to reduce noise



# VNRS Takeoff Assumptions

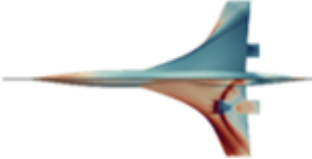
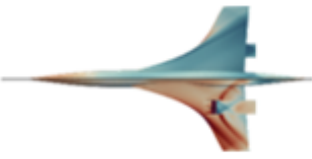
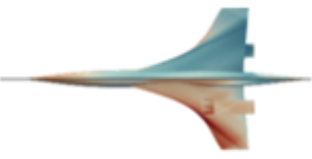
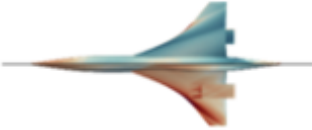


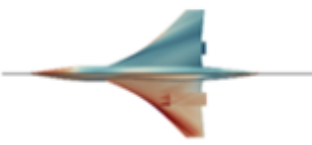


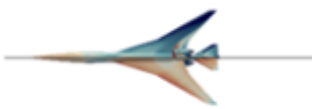
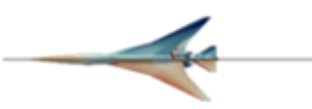


The variable noise reduction system (VNRS) includes:

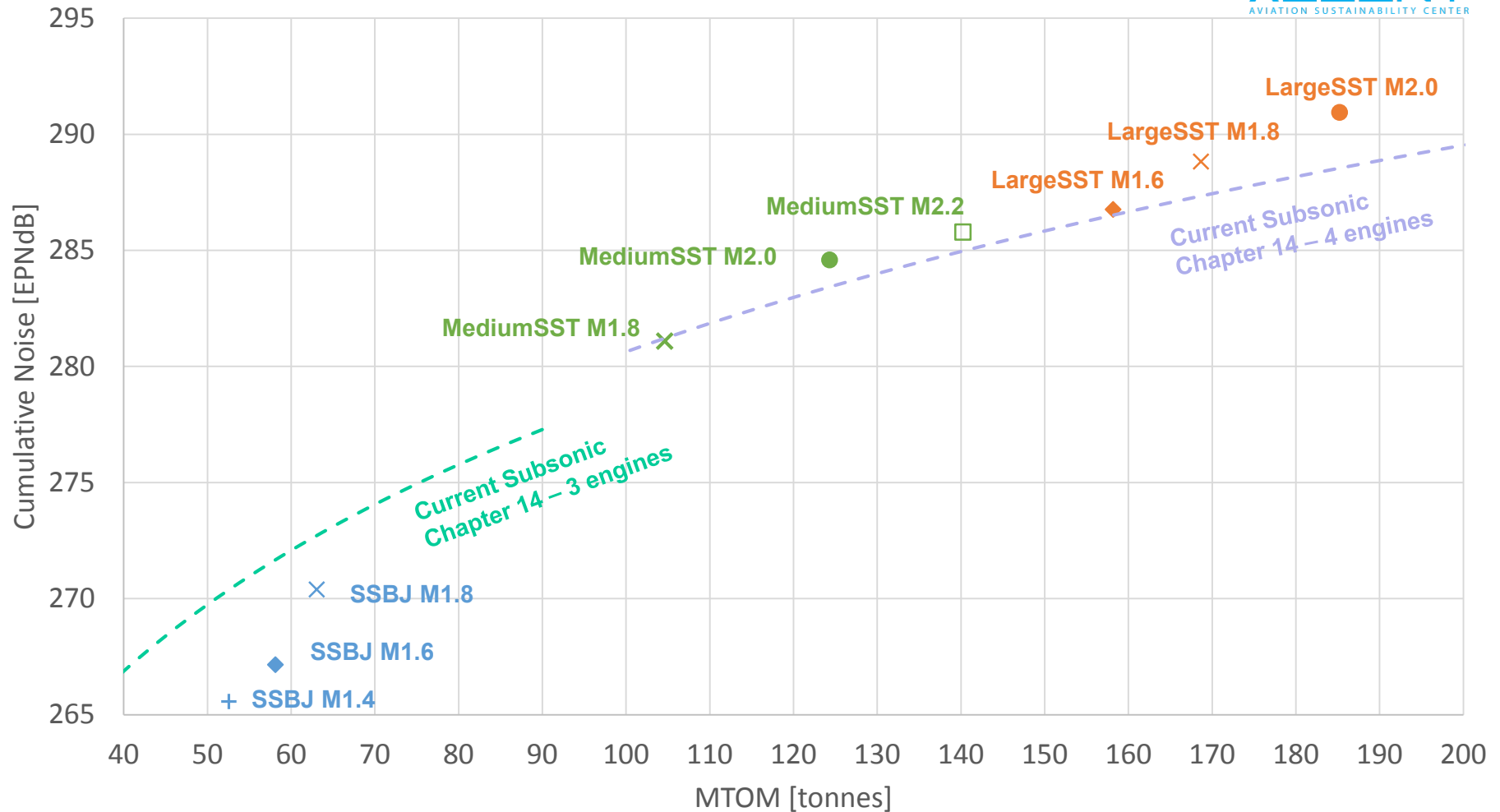
- Programmed high lift devices (PHLD) schedule optimized for  $L/D$
- Programmed thrust lapse rate (PLR) implemented right after the obstacle
- Second segment acceleration (2SA) which is broken down into
  - Initial climb angle
  - Constant speed transition altitude
  - Cutback altitude

# Summary Matrix of Airframe Designs

\*Vehicles notionally scaled by passenger class. L/D shown at 55,000 ft.

	$M_\infty = 1.4$	$M_\infty = 1.6$	$M_\infty = 1.8$	$M_\infty = 2.0$	$M_\infty = 2.2$
100 PAX	"Large SST"	 <i>Max L/D = 8.46</i>	 <i>Max L/D = 8.39</i>	 <i>Max L/D = 8.09</i>	
75 PAX					 <i>Max L/D = 7.13</i>
55 PAX	"Medium SST"		 <i>Max L/D = 7.51</i>	 <i>Max L/D = 7.26</i>	 <i>Max L/D = 7.07</i>
25 PAX		 <i>Max L/D = 7.73</i>			
SSBJ	 <i>Max L/D = 9.41</i>	 <i>Max L/D = 8.72</i>	 <i>Max L/D = 8.29</i>		"SSBJ"

# Chosen Designs: Noise vs MTOM



- Initial optimum chosen for min noise from raw data (>5000 designs per vehicle).
- Subsequent optimization utilized surrogate models (trained on the raw data) to minimize fuel constrained to less than 1dB above the raw data min noise.
- Final noise optimization for trajectory to further minimize noise.

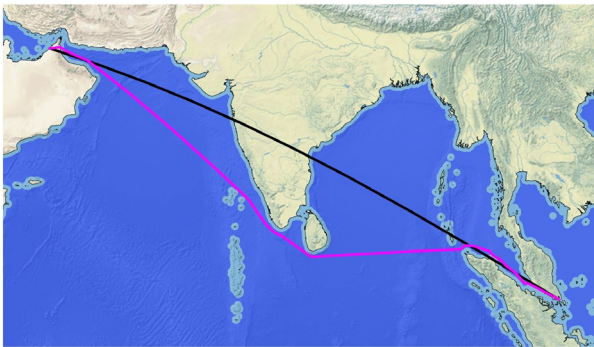


# Fleet Demand Route

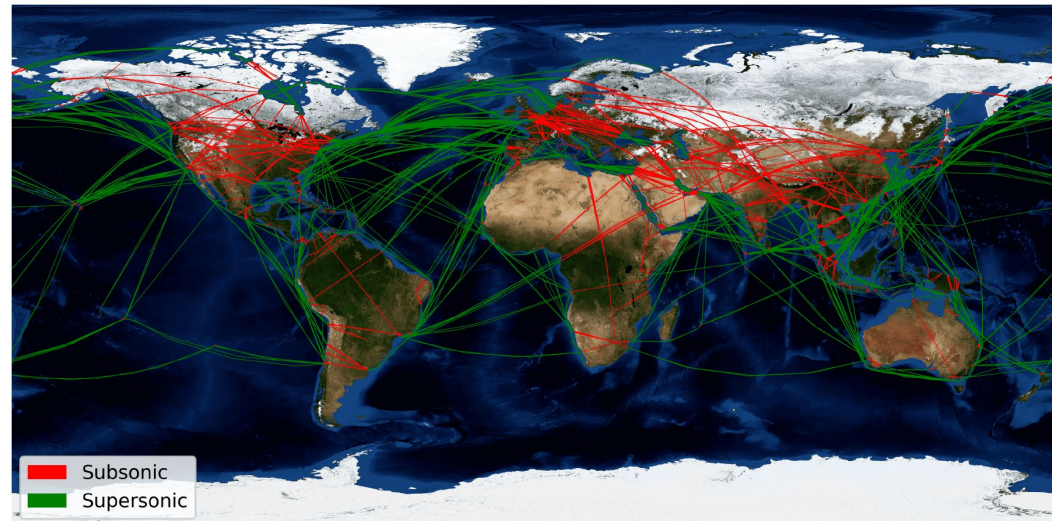
**Goal:** Determine potential SST demand globally in the future and estimate fuel burn and emissions

**Approach:** Vehicle capability based time savings and cost per route to estimate demand and aircraft performance

- Developed SST route optimization tool that takes aircraft specific performance into account and allows trading off time savings and fuel efficiency
- Support for CAEP E-Study
  - Supplied demand input to scenarios
  - Performed full flight runs for BJ and commercial vehicles
- Developed detailed global emissions dataset for Projects 22 and 58
- Use Mach Design Trade Study vehicles to define most favorable market for vehicle size and design Mach number
- Mach 2.2, 55 seat aircraft yields a maximum of 970 feasible unidirectional routes for 2035



Example: Dubai – Singapore





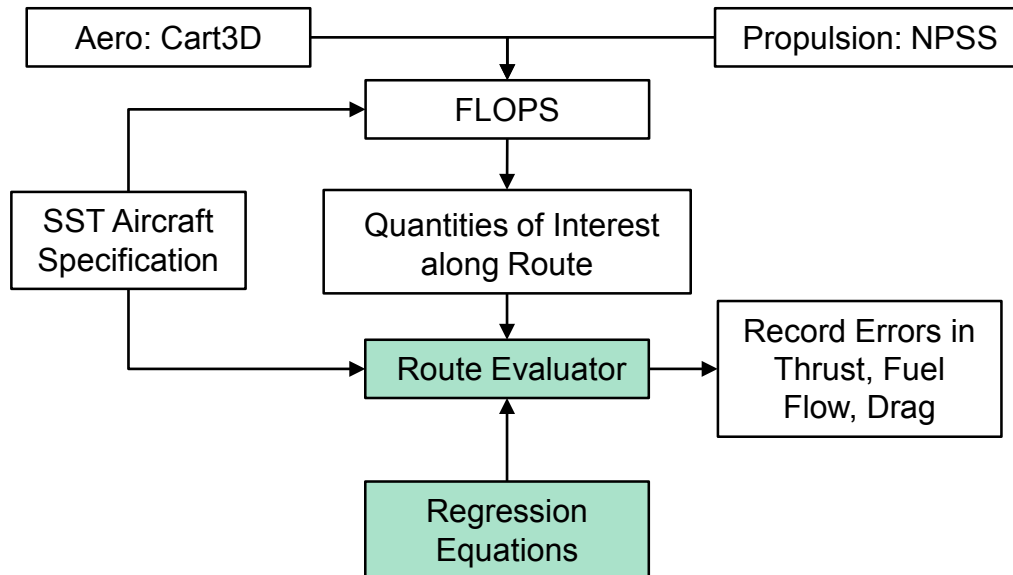
# AEDT SST Modeling

**Goal:** Develop & validate approach for modeling Supersonic Transport (SST) aerodynamics & propulsion performance within AEDT

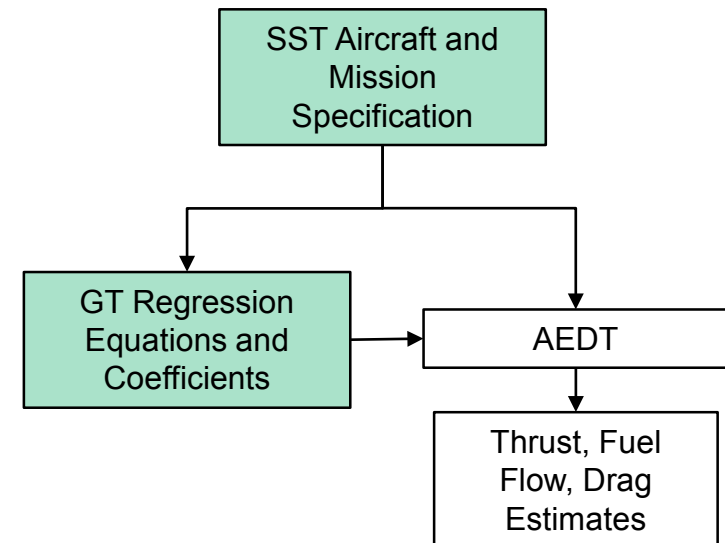
## Approach:

- Generate performance “*truth*” data for SST aircraft concepts using:
  - Aerodynamic Analysis: CART3D inviscid Computational Fluid Dynamics (CFD) with viscous correction
  - Propulsion Analysis: NPSS
- Develop new regression equations to model both supersonic and subsonic regimes
  - Physics-rooted functional forms
  - Focus on maximizing parsimony (minimal number of coefficients) while minimizing prediction errors
- Fit truth data to physics-rooted functional forms, quantify errors and propagate through aircraft mission
  - Region of validity for regressions extracted from notional SST missions
  - Fit several regressions using identical functional forms; applicable within specified envelope, i.e., region of validity
  - Assess prediction errors and propagate through simulated mission(s)
- **Validate approach on notional missions flown by SSTs**
- Provide implementation plan to incorporate models within AEDT

## Validation Process



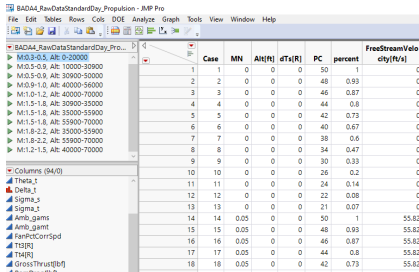
## Implementation in AEDT



# Validation Process Implementation Details

1

## Data table

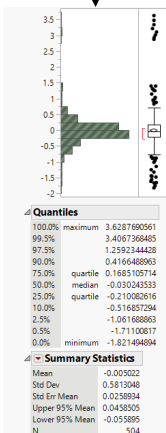


Case	MN	Alt(H)	dT(H)	PC	percent	FuelFlowVelo	city(F/H)
1	1	0	0	50	1	0	0
2	2	0	0	48	0.93	0	0
3	3	0	0	46	0.87	0	0
4	4	0	0	44	0.8	0	0
5	5	0	0	42	0.73	0	0
6	6	0	0	40	0.67	0	0
7	7	0	0	38	0.6	0	0
8	8	0	0	34	0.47	0	0
9	9	0	0	30	0.33	0	0
10	10	0	0	26	0.2	0	0
11	11	0	0	24	0.14	0	0
12	12	0	0	22	0.08	0	0
13	13	0	0	21	0.07	0	0
14	14	0.05	0	50	1	55.82	0
15	15	0.05	0	48	0.93	55.82	0
16	16	0.05	0	46	0.87	55.82	0
17	17	0.05	0	44	0.8	55.82	0
18	18	0.05	0	42	0.73	55.82	0

Analyze > Fit Model

Model Validation

2



Save Columns > Publish Prediction Formula

4

## Code window

```
Python code generated by JMP v13.2.0 ***

// Fit Least Squares - NetThrust[lbf]
// Fit Least Squares - FuelFlow[lbm/hr]

// Fit Least Squares - NetThrust[lbf]
// Fit Least Squares - FuelFlow[lbm/hr]
```

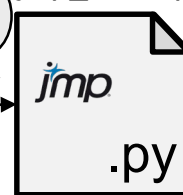
Generate Python Code

JMP source

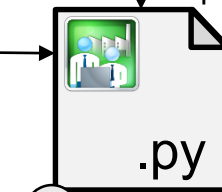


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5 jmp\_score.py



Validation Script

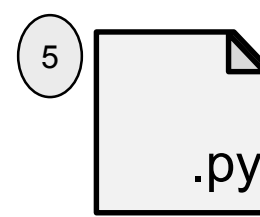


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Load data



5 <FLOPS data>.py



Results



7

