

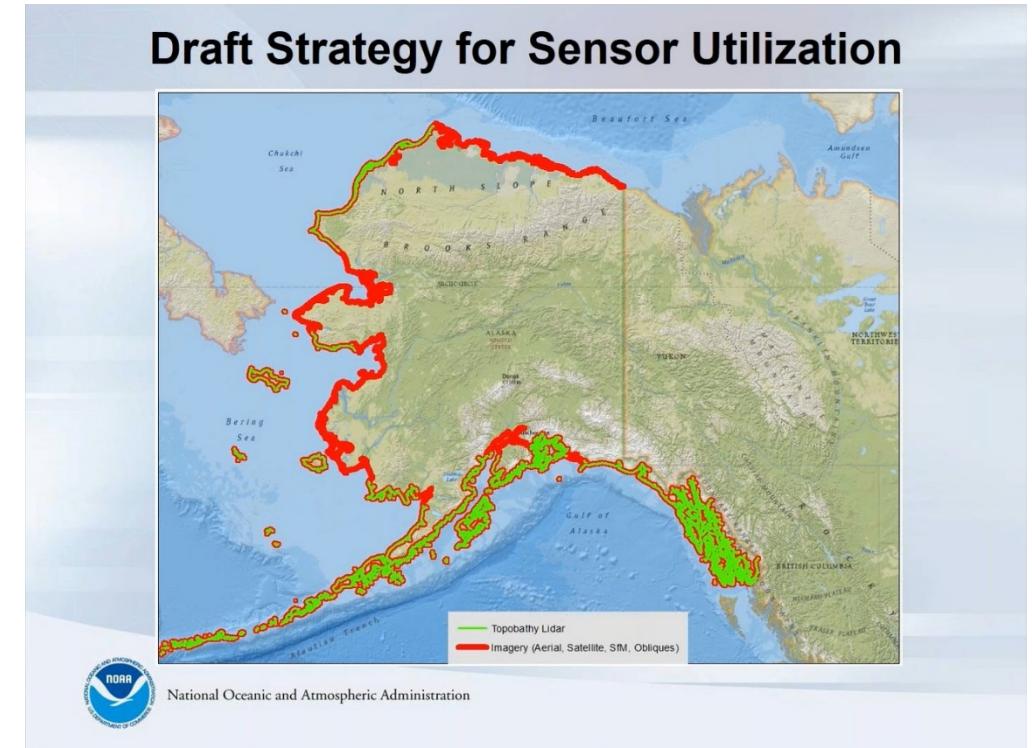
Near Real-Time Airborne Total Propagated Uncertainty Framework Bathy Lidar



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Nathan Hopper Ph.D.
Woolpert Maritime Research*

Outline

- Why Total Propagate Uncertainty?
- Some History
- Calculation Framework
- Modeling and Simulation

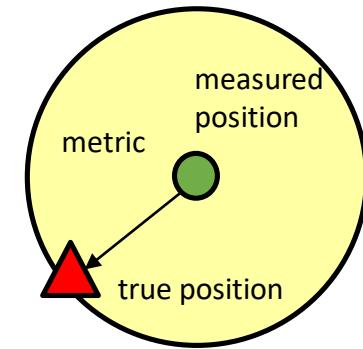
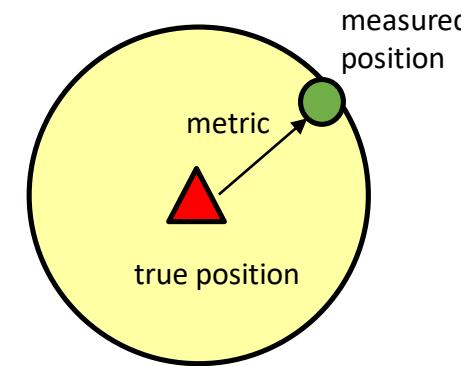


NOAA NGS November 12, 2020

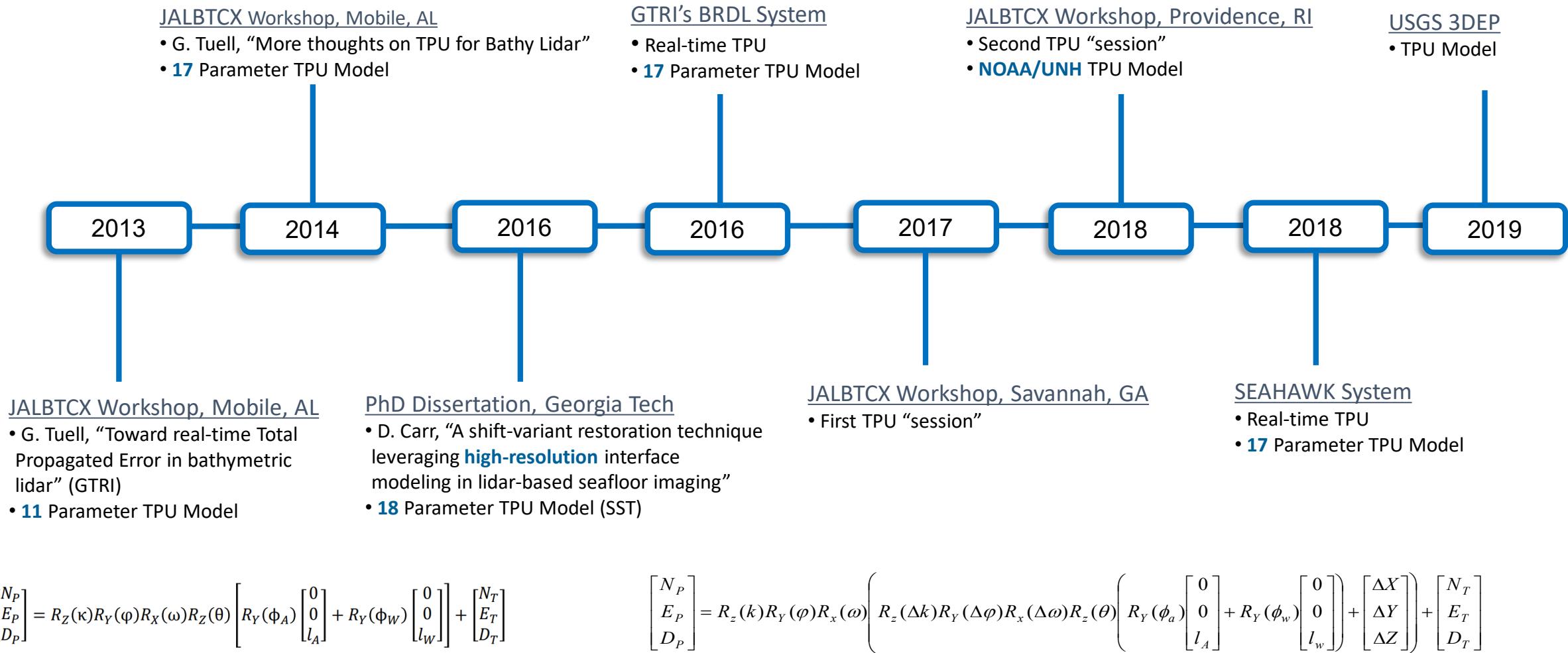


Why Uncertainty?

- Any **dataset** is only a representation of reality, therefore contains **uncertainty** about the nature of the real world being represented.
(De, Goodchild and Longley, 2007)
- Bathymetric Lidar Sources of Uncertainty
 - Trajectory, boresight, sensor measurements, water characteristics and ranging values with variances.
 - Total Propagated Uncertainty (TPU)
- How much is acceptable?
 - Surveying / Mapping
 - Specification / Standard



Recent TPU History



TPU Computational Framework

- Nested Approach
 - MATLAB
 - Implementation in C++ /CPU
 - GPU optimization in CUDA/C++
 - **Compute Unified Device Architecture (CUDA)** is a new approach to solve complex problems by transforming the GPU into a massive parallel processor
 - **140,000** TPU values per second
 - **Hybrid** CPU/GPU solution create a live point cloud with TPU in real time

Near Real-Time Computation

* CUDA implementation process 5,000 to 20,000 shots at a time.

GPU Optimization
CUDA/C++.

Hybrid
CPU/GPU
Solution

Post Processing

* MATLAB & C++ computations on a per shot basis

CPU

C++

MATLAB

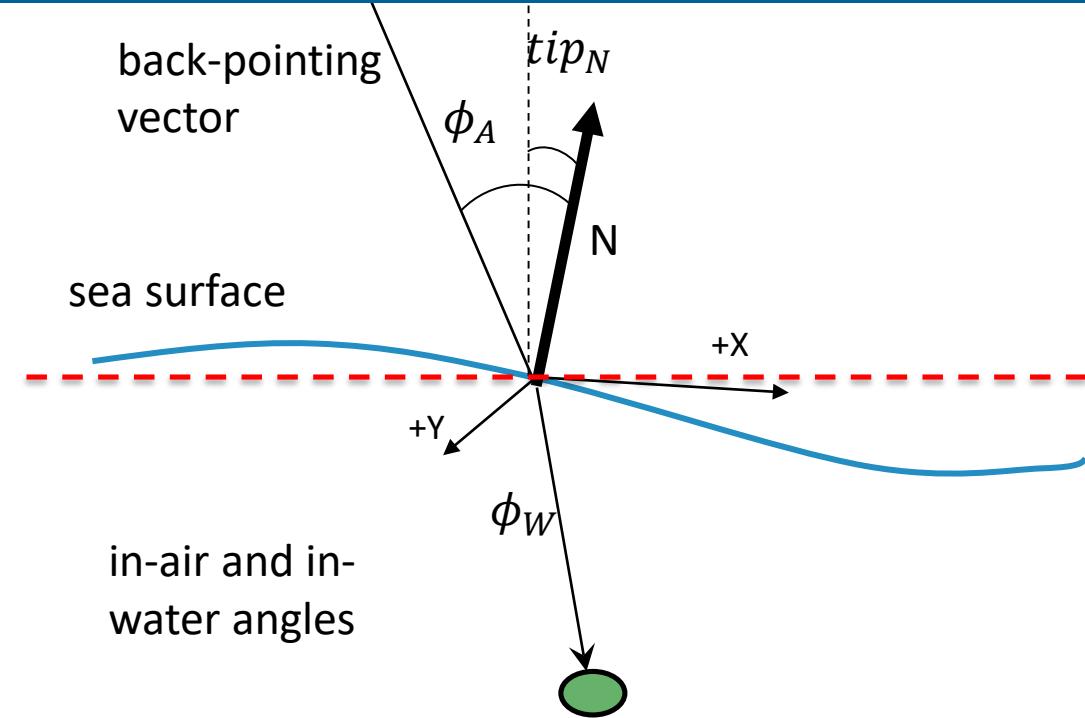
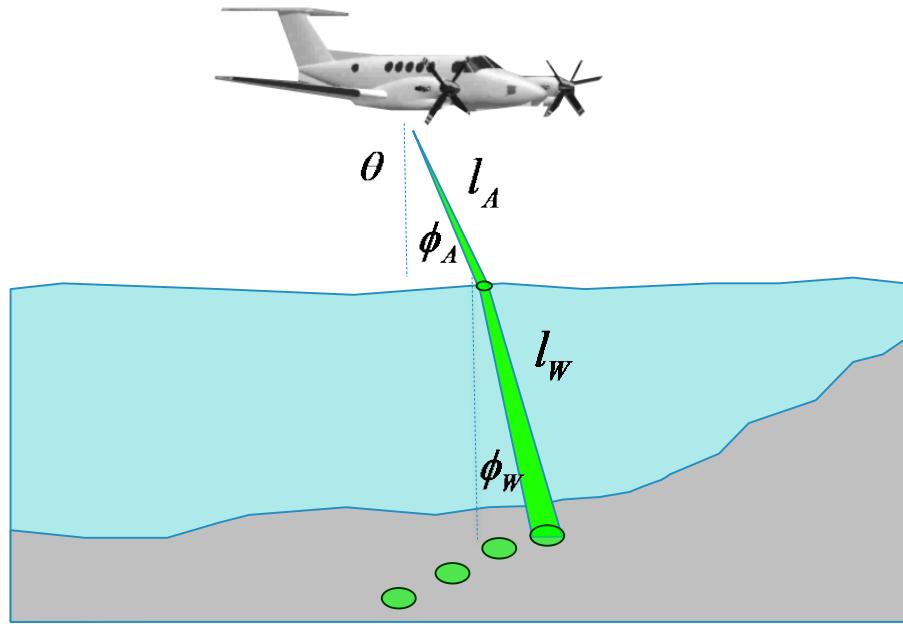
Simulations
Algorithm Development

Sensor Models
Digital Twins



TraditionallyTPU is computed using 17-parameter model

$$\begin{bmatrix} X & Y & Z & \omega & \varphi & \kappa \end{bmatrix}$$



surface often assumed to be horizontally flat

$$\begin{bmatrix} N_P \\ E_P \\ D_P \end{bmatrix} = R_z(k)R_Y(\varphi)R_x(\omega) \left(R_z(\Delta k)R_Y(\Delta\varphi)R_x(\Delta\omega)R_z(\theta) \left(R_Y(\phi_a) \begin{bmatrix} 0 \\ 0 \\ l_A \end{bmatrix} + R_Y(\phi_w) \begin{bmatrix} 0 \\ 0 \\ l_w \end{bmatrix} \right) + \begin{bmatrix} \Delta X \\ \Delta Y \\ \Delta Z \end{bmatrix} \right) + \begin{bmatrix} N_T \\ E_T \\ D_T \end{bmatrix}$$

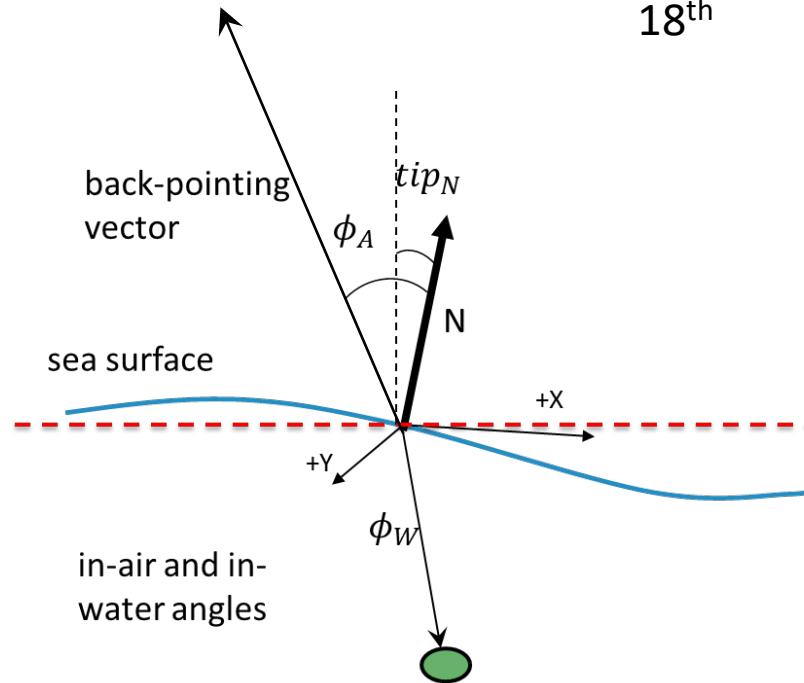
18 Parameter TPU Model

- For every seafloor coordinate

$$\begin{bmatrix} E \\ N \\ U \end{bmatrix}_M = R_Z(-h)R_x(p)R_y(r) \left[R_Z(-\Delta h)R_x(\Delta p)R_y(\Delta r)R_Z(-\theta - \Delta\theta) \left(R_x(\phi_A) \begin{bmatrix} 0 \\ 0 \\ -l_A \end{bmatrix} + R_x(\phi_W)R_y(\omega_W) \begin{bmatrix} 0 \\ 0 \\ -l_W \end{bmatrix} \right) + \begin{bmatrix} \Delta Y \\ \Delta X \\ -\Delta Z \end{bmatrix} \right] + \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

18th

- 18 Parameters
 - Trajectory
 - Boresight
 - Sensor measurements
 - Water characteristics
 - Ranging values with variances
 - **Sea Surface Topography**



TPU Computation

- A series of matrix operations compute the partial and inner derivatives

$\sigma_{N_T}^2$	$\sigma_{N_TE_T}$	$\sigma_{N_TD_T}$	σ_{N_Tk}	$\sigma_{N_T\varphi}$	$\sigma_{N_T\omega}$	$\sigma_{N_T\Delta k}$	$\sigma_{N_T\Delta\varphi}$	$\sigma_{N_T\Delta\omega}$	$\sigma_{N_T\Delta X}$	$\sigma_{N_T\Delta Y}$	$\sigma_{N_T\Delta Z}$	$\sigma_{N_T\theta}$	$\sigma_{N_T\phi_A}$	$\sigma_{N_Tl_A}$	$\sigma_{N_Tl_w}$	$\sigma_{N_T\phi_w}$	$\sigma_{N_T\omega_w}$
$\sigma_{E_T}^2$	$\sigma_{E_TD_T}$	σ_{E_Tk}	$\sigma_{E_T\varphi}$	$\sigma_{E_T\omega}$	$\sigma_{E_T\Delta k}$	$\sigma_{E_T\Delta\varphi}$	$\sigma_{E_T\Delta\omega}$	$\sigma_{E_T\Delta X}$	$\sigma_{E_T\Delta Y}$	$\sigma_{E_T\Delta Z}$		$\sigma_{E_T\theta}$	$\sigma_{E_T\phi_A}$	$\sigma_{E_Tl_A}$	$\sigma_{E_Tl_w}$	$\sigma_{E_T\phi_w}$	$\sigma_{E_T\omega_w}$
\cdot	$\sigma_{D_T}^2$	σ_{D_Tk}	$\sigma_{D_T\varphi}$	$\sigma_{D_T\omega}$	$\sigma_{D_T\Delta k}$	$\sigma_{D_T\Delta\varphi}$	$\sigma_{D_T\Delta\omega}$	$\sigma_{D_T\Delta X}$	$\sigma_{D_T\Delta Y}$	$\sigma_{D_T\Delta Z}$		$\sigma_{D_T\theta}$	$\sigma_{D_T\phi_A}$	$\sigma_{D_Tl_A}$	$\sigma_{D_Tl_w}$	$\sigma_{D_T\phi_w}$	$\sigma_{D_T\omega_w}$
\cdot	\cdot	σ_k^2	$\sigma_{k\varphi}$	$\sigma_{k\omega}$	$\sigma_{k\Delta k}$	$\sigma_{k\Delta\varphi}$	$\sigma_{k\Delta\omega}$	$\sigma_{k\Delta X}$	$\sigma_{k\Delta Y}$	$\sigma_{k\Delta Z}$		$\sigma_{k\theta}$	$\sigma_{k\phi_A}$	σ_{kl_A}	σ_{kl_w}	$\sigma_{k\phi_w}$	$\sigma_{k\omega_w}$
\cdot	\cdot	\cdot	σ_φ^2	$\sigma_{\varphi\omega}$	$\sigma_{\varphi\Delta k}$	$\sigma_{\varphi\Delta\varphi}$	$\sigma_{\varphi\Delta\omega}$	$\sigma_{\varphi\Delta X}$	$\sigma_{\varphi\Delta Y}$	$\sigma_{\varphi\Delta Z}$		$\sigma_{\varphi\theta}$	$\sigma_{\varphi\phi_A}$	$\sigma_{\varphi l_A}$	$\sigma_{\varphi l_w}$	$\sigma_{\varphi\phi_w}$	$\sigma_{\varphi\omega_w}$
\cdot	\cdot	\cdot	\cdot	σ_ω^2	$\sigma_{\omega\Delta k}$	$\sigma_{\omega\Delta\varphi}$	$\sigma_{\omega\Delta\omega}$	$\sigma_{\omega\Delta X}$	$\sigma_{\omega\Delta Y}$	$\sigma_{\omega\Delta Z}$		$\sigma_{\omega\theta}$	$\sigma_{\omega\phi_A}$	$\sigma_{\omega l_A}$	$\sigma_{\omega l_w}$	$\sigma_{\omega\phi_w}$	$\sigma_{\omega\omega_w}$
\cdot	\cdot	\cdot	\cdot	\cdot	$\sigma_{\Delta k}^2$	$\sigma_{\Delta k\Delta\varphi}$	$\sigma_{\Delta k\Delta\omega}$	$\sigma_{\Delta k\Delta X}$	$\sigma_{\Delta k\Delta Y}$	$\sigma_{\Delta k\Delta Z}$		$\sigma_{\Delta k\theta}$	$\sigma_{\Delta k\phi_A}$	$\sigma_{\Delta k l_A}$	$\sigma_{\Delta k l_w}$	$\sigma_{\Delta k\phi_w}$	$\sigma_{\Delta k\omega_w}$
\cdot	\cdot	\cdot	\cdot	\cdot	\cdot	$\sigma_{\Delta\varphi}^2$	$\sigma_{\Delta\varphi\Delta\omega}$	$\sigma_{\Delta\varphi\Delta X}$	$\sigma_{\Delta\varphi\Delta Y}$	$\sigma_{\Delta\varphi\Delta Z}$		$\sigma_{\Delta\varphi\theta}$	$\sigma_{\Delta\varphi\phi_A}$	$\sigma_{\Delta\varphi l_A}$	$\sigma_{\Delta\varphi l_w}$	$\sigma_{\Delta\varphi\phi_w}$	$\sigma_{\Delta\varphi\omega_w}$
\cdot	\cdot	\cdot	\cdot	\cdot	\cdot	\cdot	$\sigma_{\Delta\omega}^2$	$\sigma_{\Delta\omega\Delta X}$	$\sigma_{\Delta\omega\Delta Y}$	$\sigma_{\Delta\omega\Delta Z}$		$\sigma_{\Delta\omega\theta}$	$\sigma_{\Delta\omega\phi_A}$	$\sigma_{\Delta\omega l_A}$	$\sigma_{\Delta\omega l_w}$	$\sigma_{\Delta\omega\phi_w}$	$\sigma_{\Delta\omega\omega_w}$
\cdot	\cdot	\cdot	\cdot	\cdot	\cdot	\cdot	\cdot	$\sigma_{\Delta X}^2$	$\sigma_{\Delta X\Delta Y}$	$\sigma_{\Delta X\Delta Z}$		$\sigma_{\Delta X\theta}$	$\sigma_{\Delta X\phi_A}$	$\sigma_{\Delta X l_A}$	$\sigma_{\Delta X l_w}$	$\sigma_{\Delta X\phi_w}$	$\sigma_{\Delta X\omega_w}$
\cdot	\cdot	\cdot	\cdot	\cdot	\cdot	\cdot	\cdot	\cdot	$\sigma_{\Delta Y}^2$	$\sigma_{\Delta Y\Delta Z}$		$\sigma_{\Delta Y\theta}$	$\sigma_{\Delta Y\phi_A}$	$\sigma_{\Delta Y l_A}$	$\sigma_{\Delta Y l_w}$	$\sigma_{\Delta Y\phi_w}$	$\sigma_{\Delta Y\omega_w}$
\cdot	\cdot	\cdot	\cdot	\cdot	\cdot	\cdot	\cdot	\cdot	$\sigma_{\Delta Z}^2$			$\sigma_{\Delta Z\theta}$	$\sigma_{\Delta Z\phi_A}$	$\sigma_{\Delta Z l_A}$	$\sigma_{\Delta Z l_w}$	$\sigma_{\Delta Z\phi_w}$	$\sigma_{\Delta Z\omega_w}$
(sym)																	
σ_θ^2												$\sigma_{\theta\phi_A}^2$		$\sigma_{\theta l_A}$	$\sigma_{\theta l_w}$	$\sigma_{\theta\phi_w}$	$\sigma_{\theta\omega_w}$
\cdot	$\sigma_{\phi_A}^2$											$\sigma_{\phi_A l_A}$		$\sigma_{\phi_A l_w}$	$\sigma_{\phi_A\phi_w}$	$\sigma_{\phi_A\omega_w}$	
\cdot	\cdot	$\sigma_{l_A}^2$										$\sigma_{l_A l_w}$		$\sigma_{l_A l_w}$	$\sigma_{l_A\phi_w}$	$\sigma_{l_A\omega_w}$	
\cdot	\cdot	\cdot	$\sigma_{l_w}^2$									$\sigma_{l_w}^2$		σ_{l_w}	$\sigma_{l_w\phi_w}$	$\sigma_{l_w\omega_w}$	
\cdot	\cdot	\cdot	\cdot	$\sigma_{\phi_w}^2$								$\sigma_{\phi_w\phi_w}$		$\sigma_{\phi_w\omega_w}$	$\sigma_{\phi_w\omega_w}$		
\cdot	\cdot	\cdot	\cdot	\cdot	$\sigma_{\phi_w\omega_w}^2$							$\sigma_{\phi_w\omega_w}$		$\sigma_{\phi_w\omega_w}$	$\sigma_{\phi_w\omega_w}$		

$\tilde{\Sigma}_M(1:16;1:16) = \Sigma_M(1:16;1:16)$
From Tuell (2014)

Navigation

Pointing Ranging Performance
Calibration

Sea surface:
2 variables



TPU Computation

- 3x3 matrix of variance – covariance

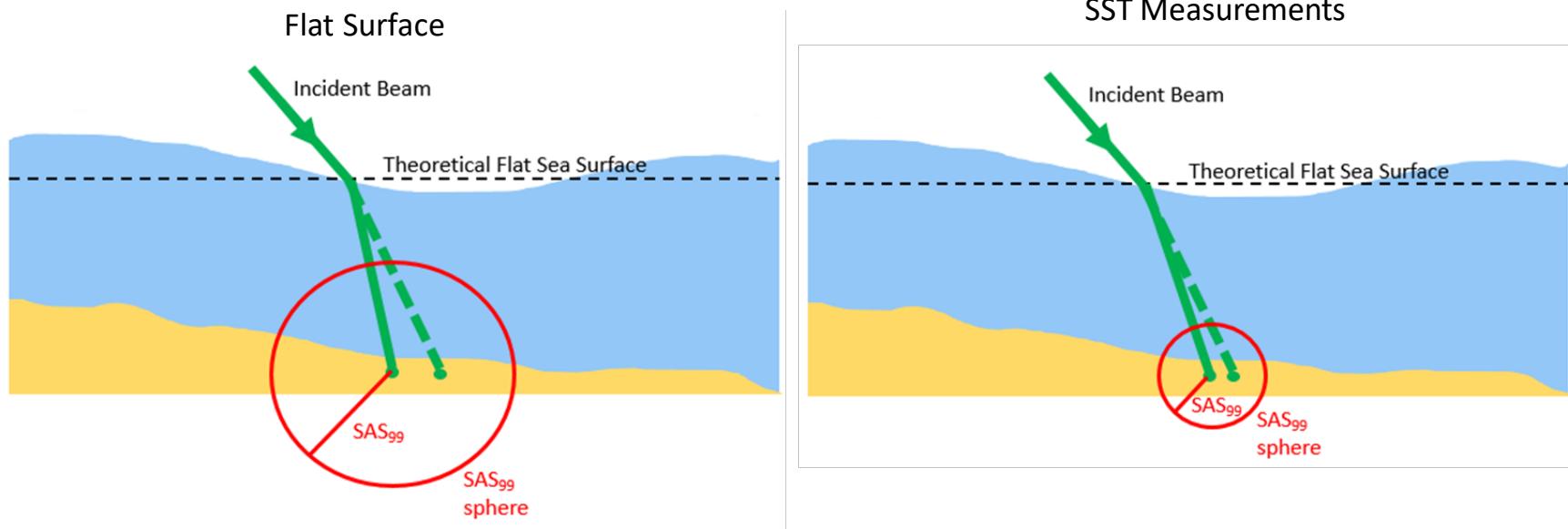
$$\Sigma_V = \begin{bmatrix} \sigma_E^2 & \sigma_{EN} & \sigma_{EU} \\ \sigma_{NE} & \sigma_N^2 & \sigma_{NU} \\ \sigma_{UE} & \sigma_{UN} & \sigma_U^2 \end{bmatrix} = J \Sigma_M J^T$$

3 variances representing ENU
6 representing off-axis components

- Single value representing accuracy

$$SAS_{99} = 1.122(\sigma_N + \sigma_E + \sigma_D)$$

Spherical Accuracy Standard 99%

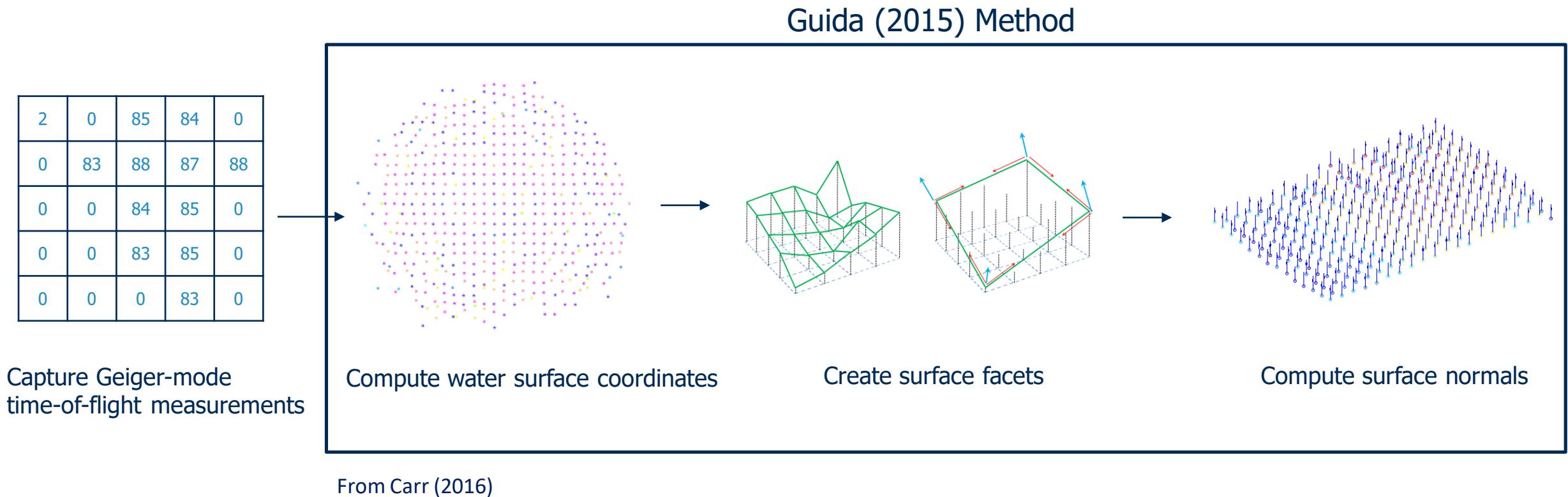


Geo location error and uncertainty are amplified when a flat surface is assumed.

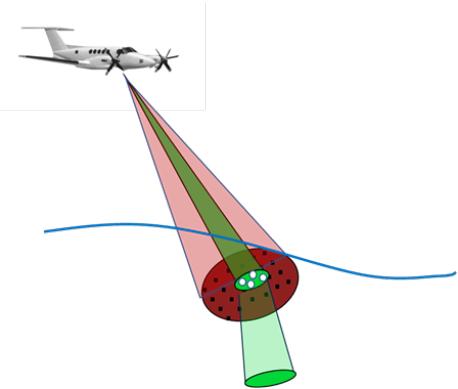


High Resolution SST

- Segmented detectors on Co-located Green and IR.



Development Design and Testing

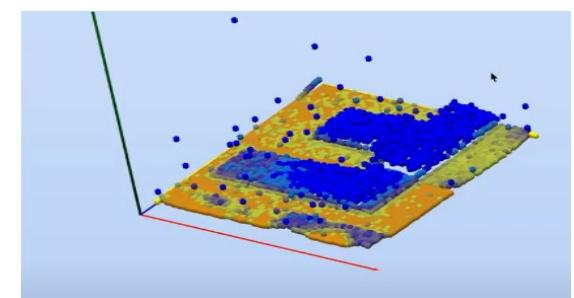
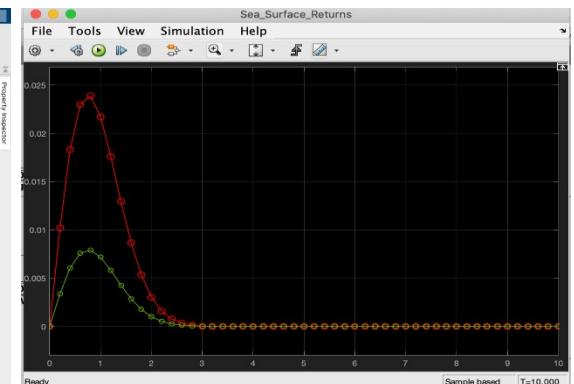
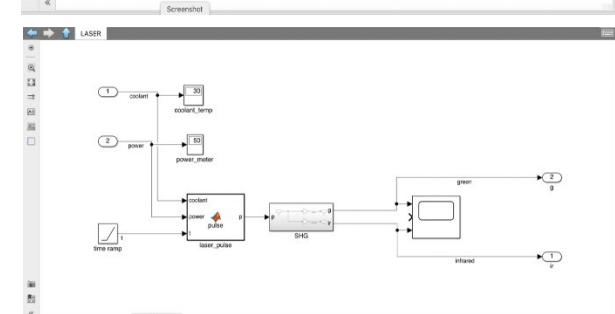
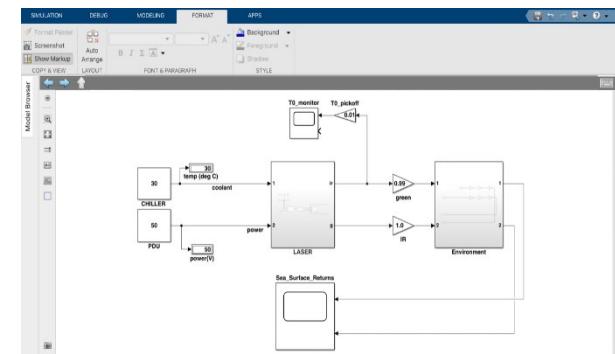
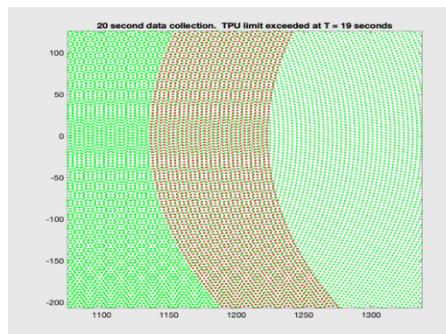
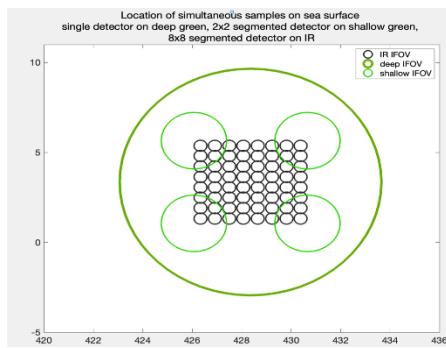
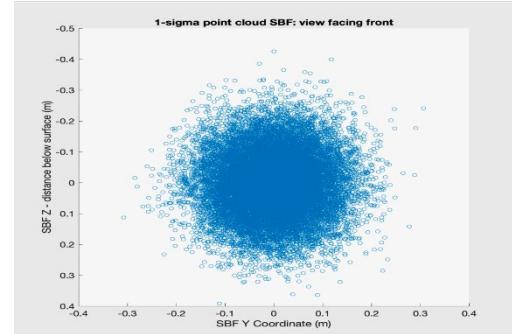


MATLAB + Simulink

Simulations
Algorithm Development

Sensor Models
Digital Twins

```
Editor - /Users/3d-ideas_macbookpro/Documents/0_3d-IDEAS/4_projects/201801_Wo... x
bulldog_TPL_NU_vector_raytrace_sea_surface_2020124.m + environment_indexwater;
green_natural_aversion = 0.25; % seconds ANSI Z136.1 section 4.4.4.2.3.2
% Vs 5. Enter Lidar measurements and their sigmas .....
nA = environment.indexAir;
nW = environment.indexWater;
pA = system.incidenceAir;
t = 0.9;
LA = 3078.0;
pw = environment.incidenceWater; % in-water incidence angle in degrees
ow = 0.0; % sea surface roll angle in degrees
lw = 0.0; % optical path length in water in meters
spa = 0.05; % sigma of pA orientation in degrees
st = 0.05; % sigma of t orientation in degrees
sLA = 0.1; % sigma of LA in degrees
spW = 5.0; % sigma of pw in degrees
sow = 0.0; % sigma of ow in degrees
slw = 0.1; % sigma of lw in degrees
Vs 6. Compute LGF coordinate vector of deep green at sea surface, (V) .....
A = [cosd(-h) -sind(-h) 0.0; sind(-h) cosd(-h) 0.0; 0.0 0.0 1.0];
R = [1 0 0; 0 1 0; 0 0 1]; % rotation matrix
R = R * R'; % rotate coordinate
```



SUMMARY

- A GPU accelerated TPU calculation from multiple channels, each creating thousands of waveforms per second is achievable, reliable, and useful.
- Near real-time TPU will be especially beneficial for assessing data quality when performing airborne surveying operations in remote locations.



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- Eric Cahoon



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