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Mapping surface tides from SWOT data: An assessment of regularized and constrained harmonic analysis in the St. Lawrence Estuary

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SWOT 3rd ST Meeting

Montreal, Canada

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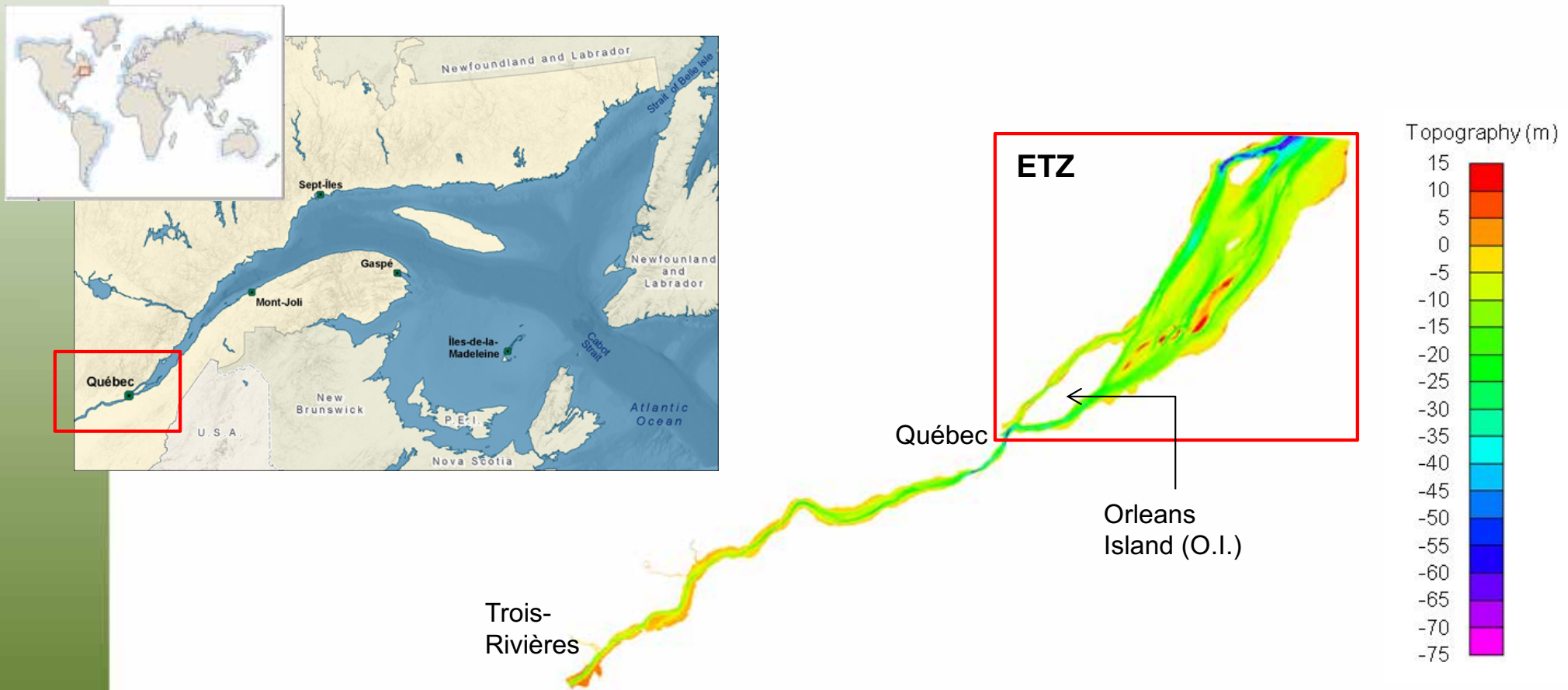
Objectives

- Evaluate the potential of SWOT data to map surface tides in estuaries
- Develop a robust method for the recovery of tidal constituent properties
 - Be robust to SWOT temporal resolution (limited tidal aliasing)
 - Include multiple tidal constituents in each tidal band (good frequency resolution)
 - Allow 2D mapping of nonlinear tides (spatially coherent)
 - Be independent of system geometry (easily transferable)



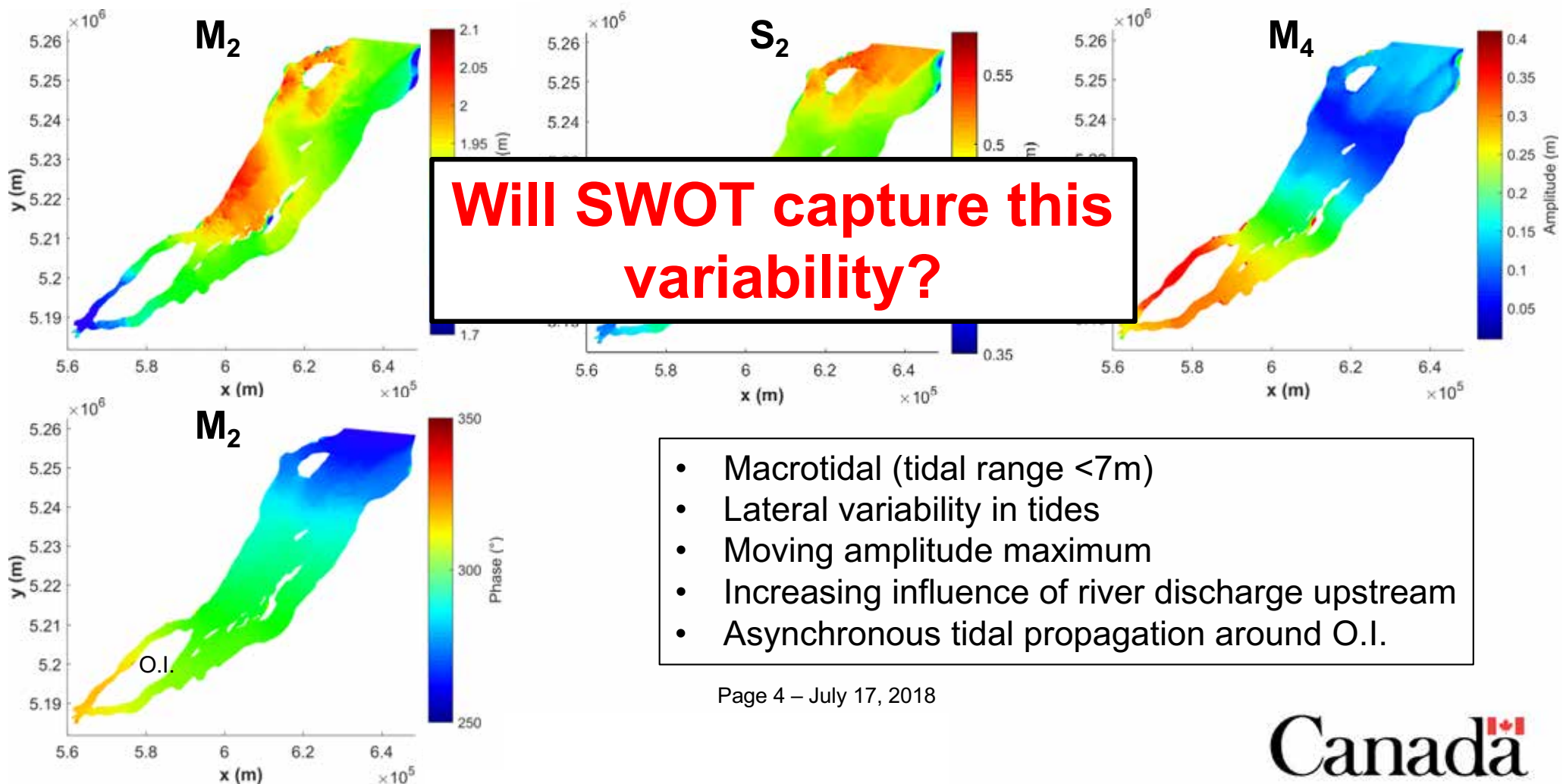
Study Site

- St. Lawrence Estuarine Transition Zone (ETZ)



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Methods

- Classical Harmonic Analysis (HA)

- Harmonic model:

$$y_c = c_0 + \sum_k C_k \cos(\sigma_k t - \Phi_k)$$

or

$$y_c = c_0 + \sum_k [c_{1,k} \cos(\sigma_k t) + c_{2,k} \sin(\sigma_k t)]$$

In matrix form: $\mathbf{y} = \mathbf{X}\mathbf{c} + \boldsymbol{\varepsilon}$

y_c :	tidal heights
c_i :	model coefficients
σ_k :	tidal frequency
t :	time
$C_k = \sqrt{c_{1,k}^2 + c_{2,k}^2}$:	amplitude
$\Phi_k = \tan^{-1}(c_{2,k}/c_{1,k})$:	phase

- Hypotheses:

- Tides are stationary and independent of time-varying riverine, oceanic and atmospheric influences.
 - There is a fixed number of tidal constituents with discrete periodicities, phase angles and amplitudes.
 - Tidal constituents are mutually independent.



Methods

- Ordinary Least-Squares (OLS)

$$\hat{c} = \arg \min_c \left\{ \sum_{i=1}^n (y_i - x_i^T c)^2 \right\} \text{ or } \hat{c} = (\mathbf{X}^T \mathbf{X})^{-1} \mathbf{X}^T \mathbf{y}$$

- Regularized Least-Squares

- Ridge regression (Hoerl and Kennard, 1970)

$$\hat{c} = \arg \min_c \left\{ \sum_{i=1}^n (y_i - x_i^T c)^2 + \lambda \sum_{j=1}^p c_j^2 \right\} \text{ or } \hat{c} = (\mathbf{X}^T \mathbf{X} + \lambda \mathbf{I})^{-1} \mathbf{X}^T \mathbf{y}$$

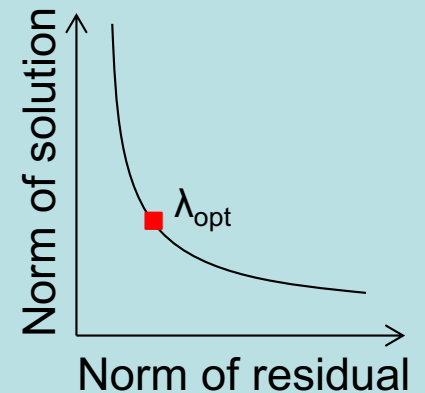
- Lasso regression (Tibshirani, 1996)

$$\hat{c} = \arg \min_{c_0, c} \left\{ \frac{1}{2n} \sum_{i=1}^n (y_i - c_0 - x_i^T c)^2 + \lambda \sum_{j=1}^p |c_j| \right\}$$

- Elastic Net regression (Zou and Hastie, 2005)

$$\hat{c} = \arg \min_{c_0, c} \left\{ \frac{1}{2n} \sum_{i=1}^n (y_i - c_0 - x_i^T c)^2 + \lambda \sum_{j=1}^p \left(\frac{1 - \alpha}{2} c_j^2 + \alpha |c_j| \right) \right\}$$

Find optimal λ using
L-curve criterion

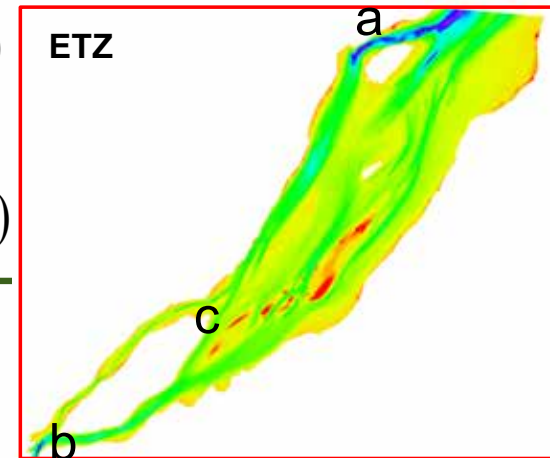


Methods

$$y_a = a_0 + \sum A_k \cos(\sigma_k t - \alpha_k)$$

$$y_b = b_0 + \sum B_k \cos(\sigma_k t - \beta_k)$$

$$y_c = c_0 + \sum C_k \cos(\sigma_k t - \Phi_k)$$



- Constrained Harmonic Analysis (ConHA)

- Model parameters at location **c** are estimated from known tidal constituent properties at two *in-situ* boundary stations **a** and **b**.
- Constrained minimization:

$$\hat{c} = \arg \min_c \left\{ \sum_{i=1}^n (y_i - x_i^T c)^2 \right\}$$

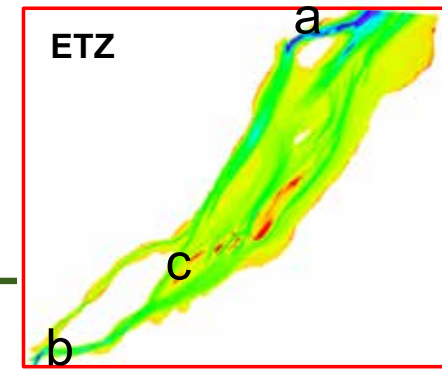
$$\text{subject to } \begin{cases} \min(a_0, b_0) < c_0 < \max(a_0, b_0) \\ (1 - f) \times \min(A_k, B_k) < C_k < (1 + f) \times \max(A_k, B_k), \\ \alpha_k < \Phi_k < \beta_k \end{cases}$$

where f is an amplification factor.

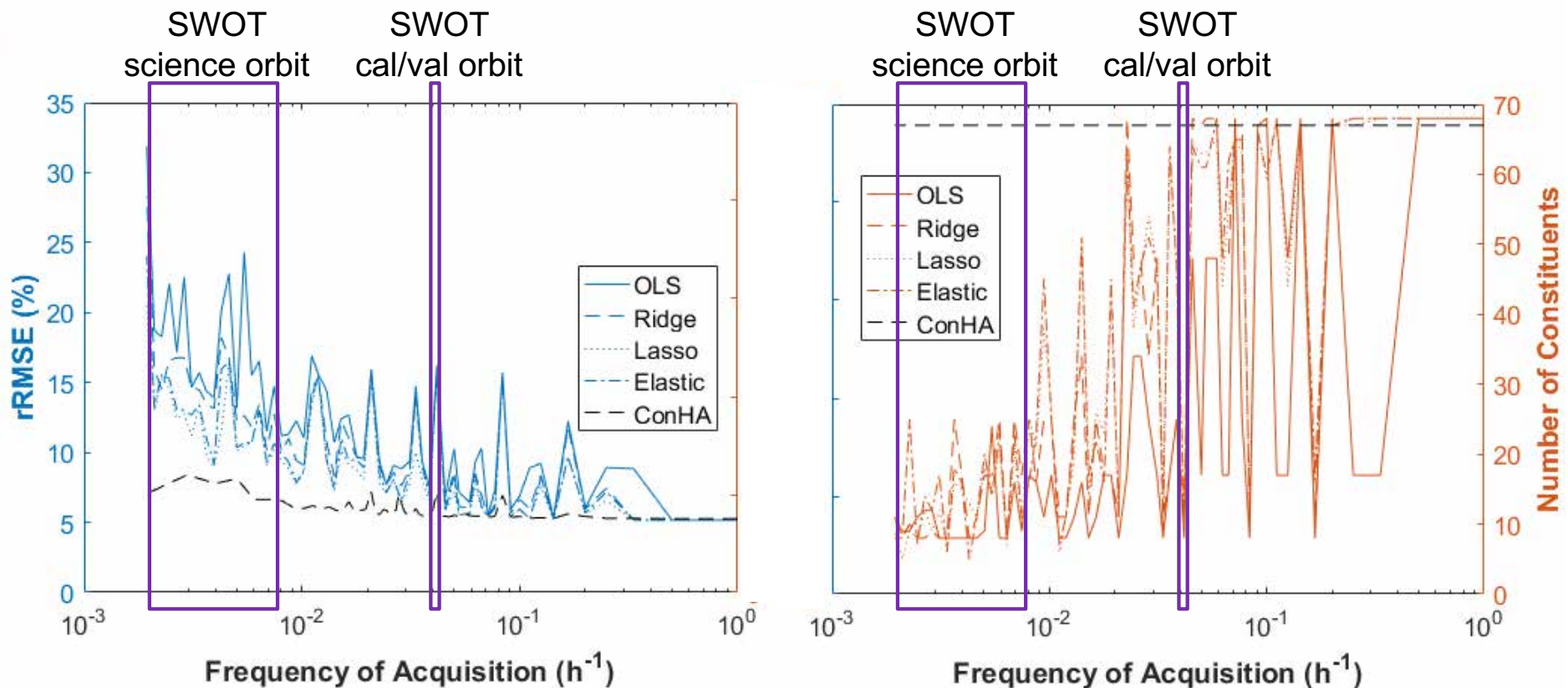
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Sensitivity Analysis

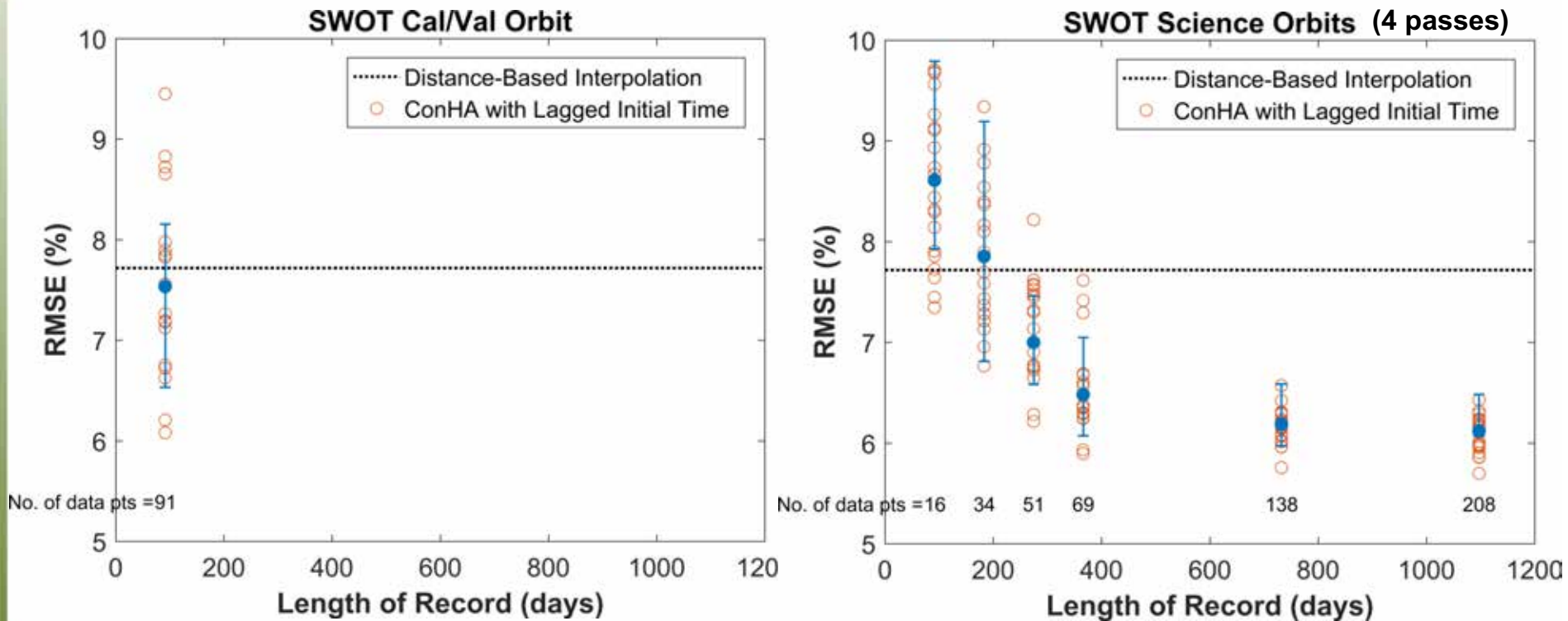


- Sensitivity to frequency of acquisition (1-year signal at station *c*)



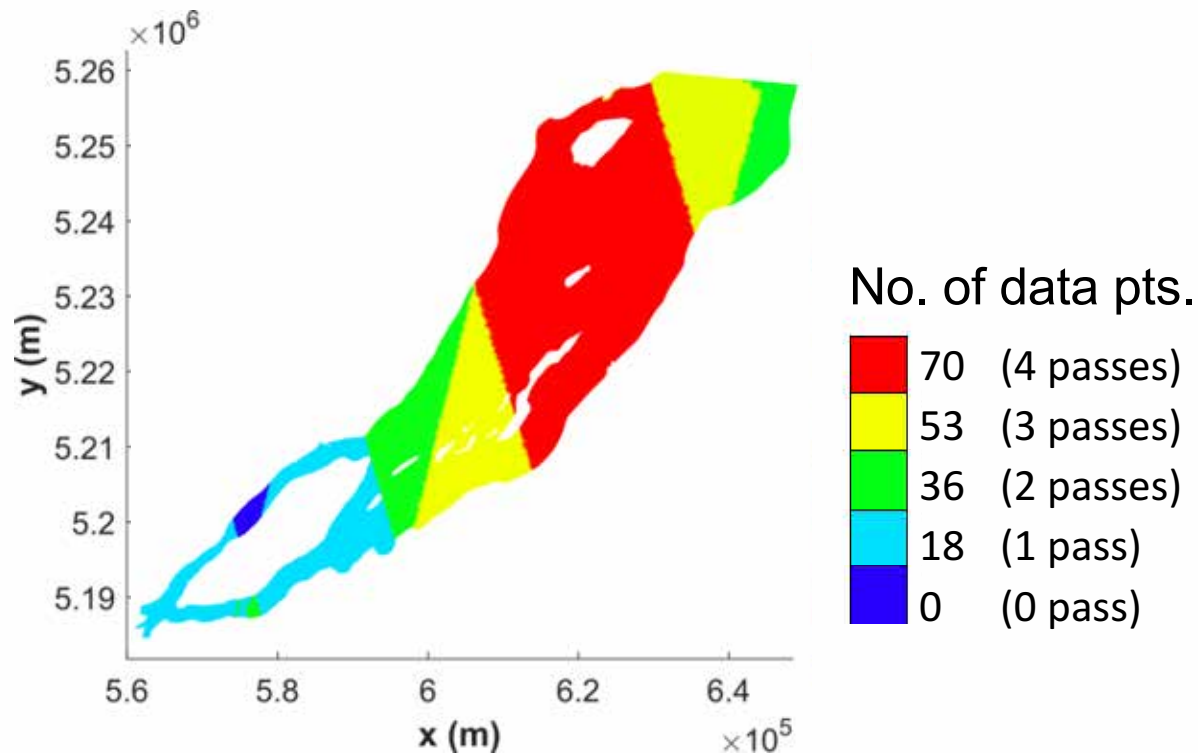
Sensitivity Analysis

- Sensitivity to SWOT record length and initial launch time



Mapping Estuarine Tides

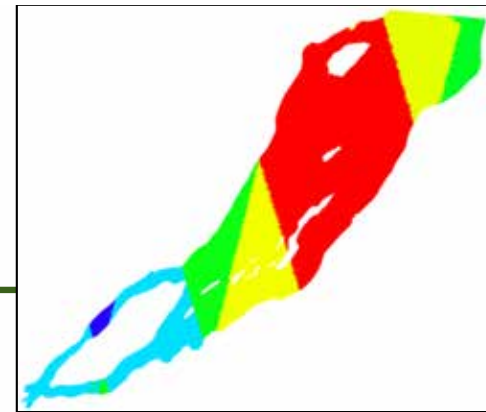
- 2D water level fields from a 1-year numerical simulation were used to produce a 2D SWOT sample



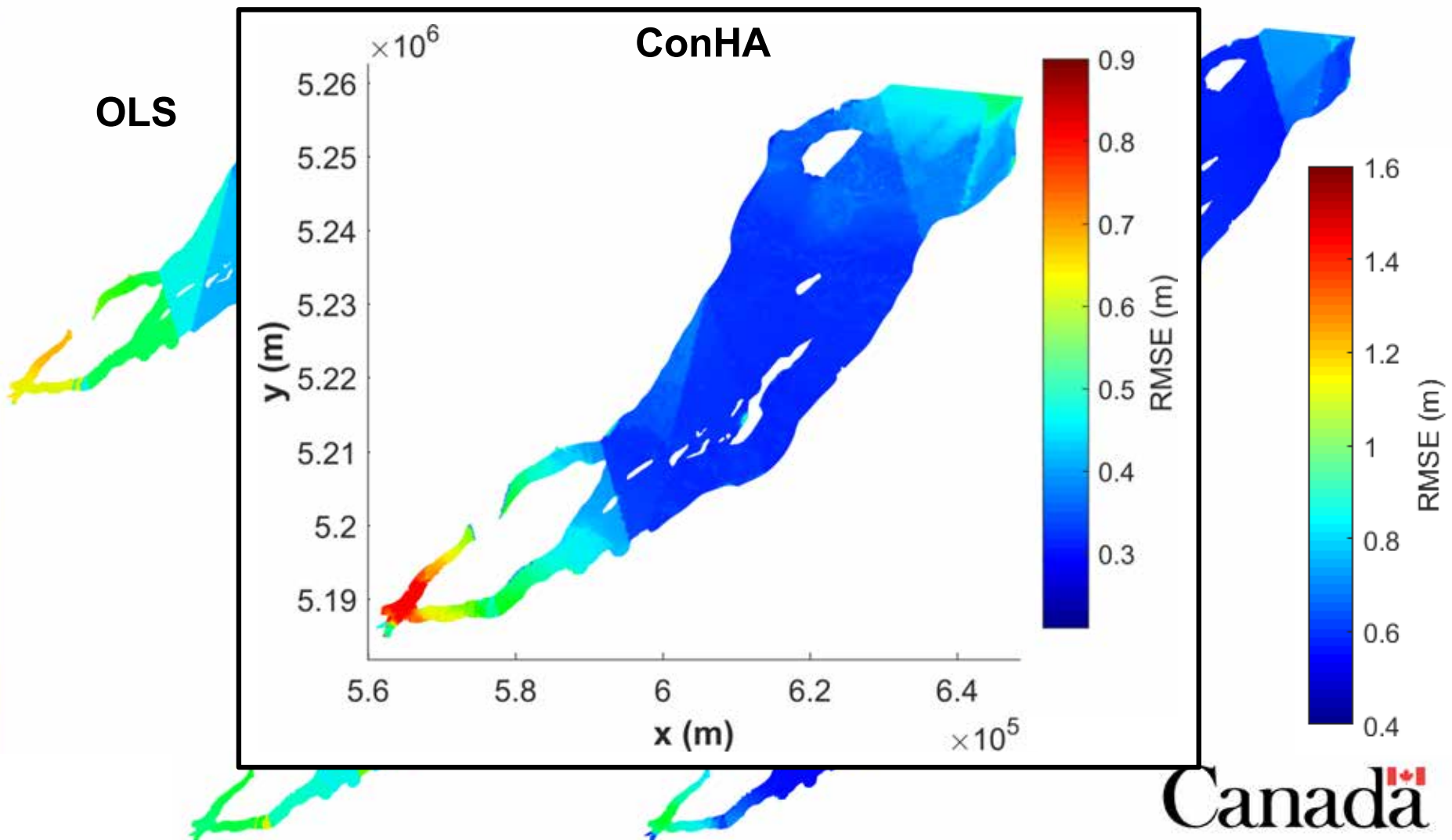
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Mapping Estuarine Tides

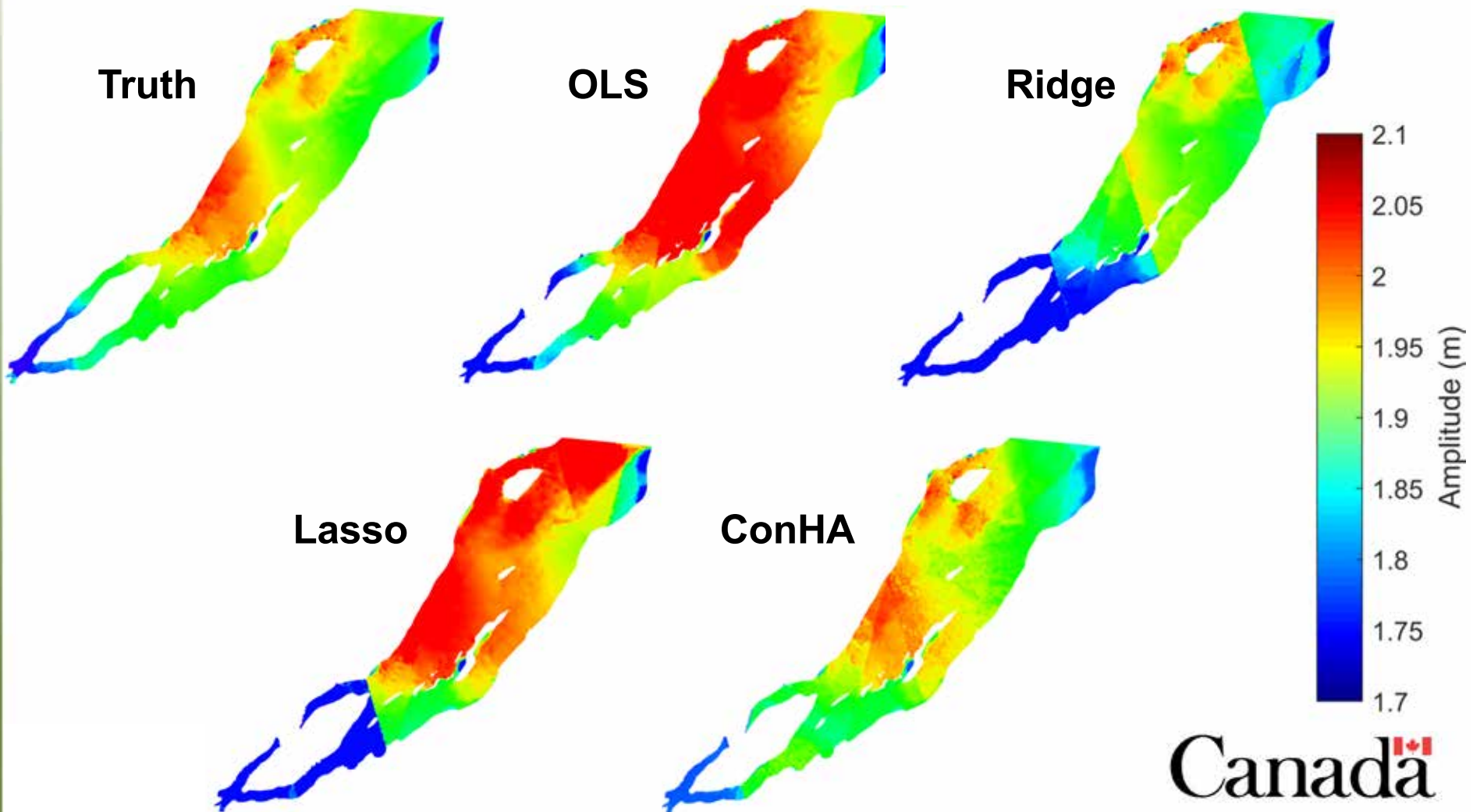
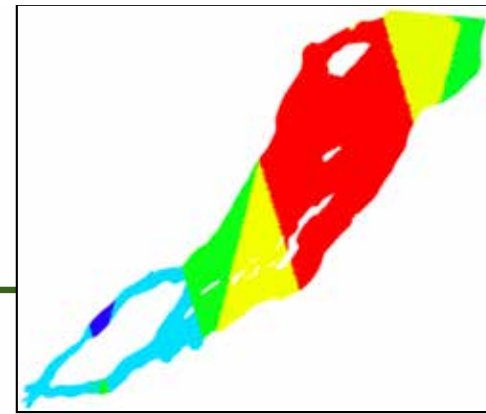


- RMSE



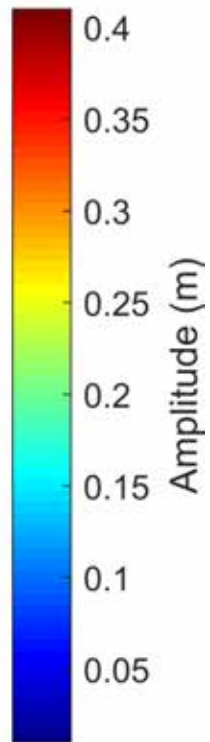
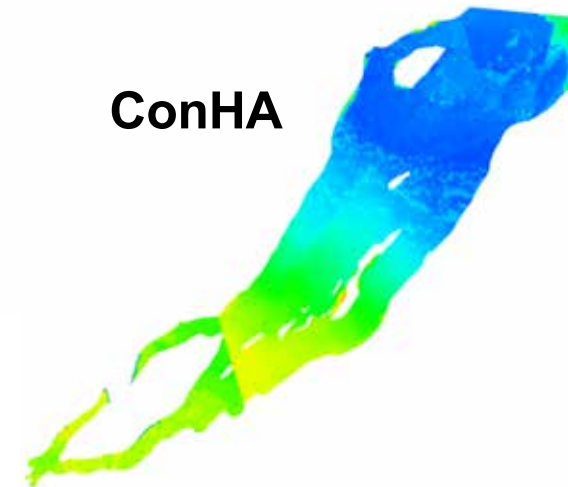
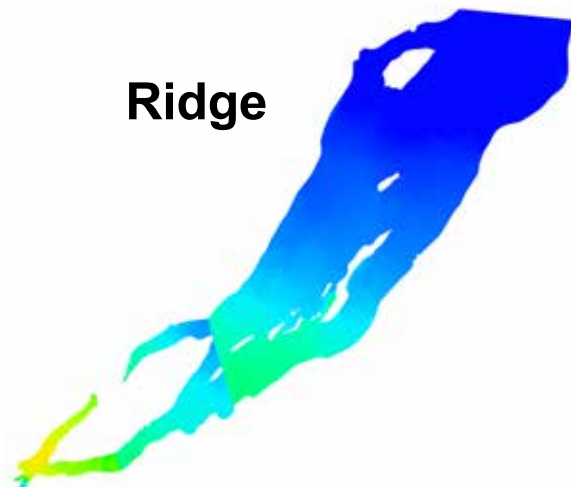
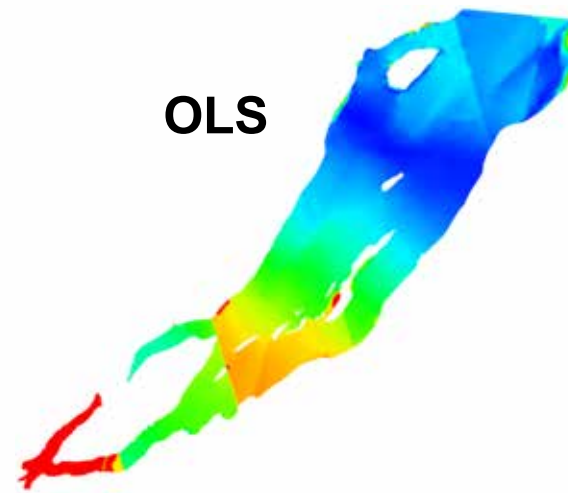
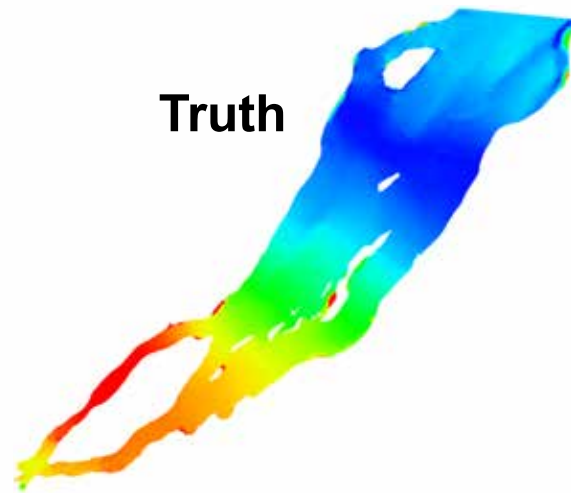
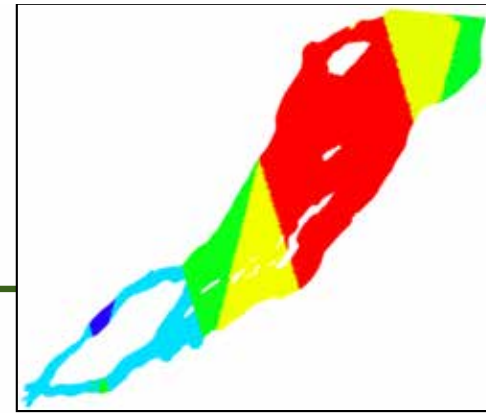
Mapping Estuarine Tides

- M_2 amplitude



Mapping Estuarine Tides

- M_4 amplitude



Summary and Highlights

- Regularized least-squares
 - More robust than OLS, but sensitive to number of constituents
 - Longer record (~3 years) needed to improve tidal prediction
 - Good alternative if no in-situ data
- Constrained harmonic analysis (new approach)
 - Requires prior knowledge on tides at boundary stations
 - Robust to downsampling and reduced record length
 - Limited tidal aliasing
 - Resolves for multiple constituents in each tidal band
 - No prior spatial function needed (easily transposable)
 - Accuracy <8% during cal/val
 - Stable accuracy (<7%) achieved after 1 year of science mission

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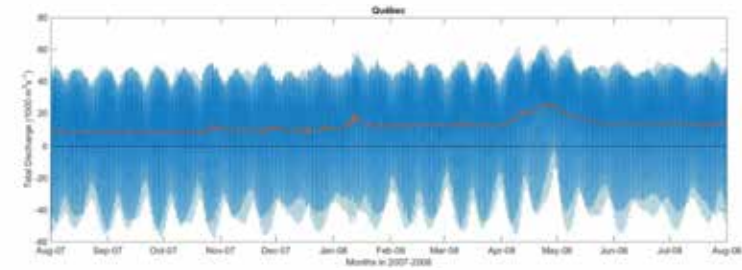


Perspective and Applications

- Future work
 - Full error analysis including SWOT errors
 - Non-stationary tides (e.g. influenced by river flow, storm surges)
 - NS_TIDE (Matte et al. 2013, *JTECH*)
- Extended uses
 - Coastal areas where tides are spatially coherent
 - Remote areas with limited field efforts
 - Model assimilation
 - Discharge estimation



Discharge Estimation



- Estimating river discharge from tides
 - Jay and Kukulka (2003), Moftakhari et al. (2013, 2016)

$$Q_R = \alpha + \beta \times TP^\gamma$$

- Cai et al. (2014): analytical approach
- Tidal discharge estimation
 - Solving the continuity equation (Matte et al. 2018, *JGR*)

$$Q_{tot} = \sum_r Q_r - \int_{\Omega} \frac{\partial h}{\partial t} d\Omega$$

- A new multi-gauge rating curve approach

$$Q = a + (bh_1 + ch_2)^d + (eh_2 + fh_3)^g$$

