





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

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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)

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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

$$y_{c} = c_{0} + \sum_{k} \left[c_{1,k} \cos(\sigma_{k}t) + c_{2,k} \sin(\sigma_{k}t) \right]$$

In matrix form: $y = Xc + \varepsilon$

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

Lasso regression (Tibshirani, 1996) $\hat{c} = \arg\min_{c_0,c} \left\{ \frac{1}{2n} \sum_{i=1}^n (y_i - c_0 - x_i^T c)^2 + \lambda \sum_{j=1}^p |c_j| \right\}$



Find optimal λ using

- Elastic Net regression (Zou and Hastie, 2005) $\hat{c} = \arg\min_{c_0,c} \left\{ \frac{1}{2n} \sum_{i=1}^n (y_i - c_0 - x_i^T c)^2 + \lambda \sum_{j=1}^p \left(\frac{1-\alpha}{2} c_j^2 + \alpha |c_j| \right) \right\}$

Methods

$$y_{a} = a_{0} + \sum_{k} A_{k} \cos(\sigma_{k}t - \alpha_{k})$$

$$y_{b} = b_{0} + \sum_{k} B_{k} \cos(\sigma_{k}t - \beta_{k})$$

$$y_{c} = c_{0} + \sum_{k} C_{k} \cos(\sigma_{k}t - \Phi_{k})$$

ETZ

- 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:

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

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



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



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Sensitivity Analysis

Sensitivity to SWOT record length and initial launch time



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



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• M₂ amplitude



• M₄ 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

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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



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