

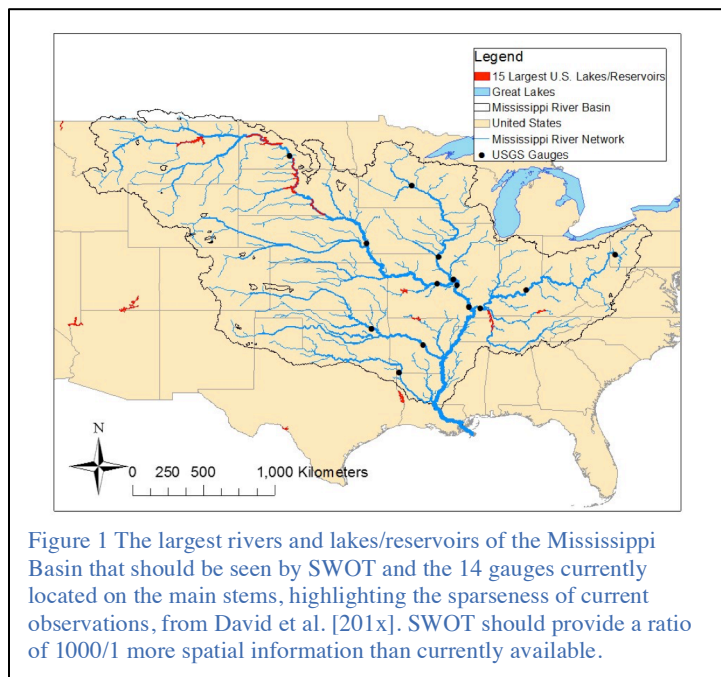
Integrating SWOT Measurements Into Global Hydrologic and Hydraulic Models

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1. Introduction & Objectives

1.1. Introduction

The Surface Water and Ocean Topography (SWOT) Mission [Alsdorf *et al.*, 2007a, 2007b] is tentatively scheduled for launch in 2021 and is expected to provide unprecedented observations of width, height and slope for the largest terrestrial water bodies, from space. The characteristics of the main SWOT instrument have been well studied [e.g. Enjolras and Rodriguez, 2009] and the relative contribution to terrestrial hydrology of SWOT never-seen observations compared to *in situ* observations of lakes [Biancamaria *et al.*, 2010] and rivers [Pavelsky *et al.*, 2014] have been quantified. However, the waterscape features that SWOT should unveil at sub-monthly resolution and continental to global scales remain a mystery. Worse, we yet have to fully characterize these variations at such large scales (Figure 1). Particularly, current observing capabilities are not able to reveal the spatial features of flow wave propagation [David *et al.*, 2011].



A new generation of continental to global scale hydrologic and hydrodynamic (“hydro”) models has been under development for the past two decades [Bates and De Roo, 2000; Beighley *et al.*, 2009; Kouwen, 2010; David *et al.*, 2011; Yamazaki *et al.*, 2011; Decharme *et al.*, 2012]. Such models can help us describe, before launch, the waxing and waning of surface water bodies that we expect to see with SWOT. Given the lack of existing SWOT-like observations at continental to global scales, macroscale hydro models are currently our best way of estimating the spatiotemporal variations of surface water features that should be observed by SWOT at continental scale and at sub-monthly time steps. The SWOT

Hydrology Simulator could then be used to characterize the measurement capabilities, although such large-scale applications of the simulator have yet to be implemented.

Additionally, one could hope that the unprecedented observations will help improving our understanding and modeling capabilities for the global terrestrial water cycle and of the climate system. However, while the direct observations of SWOT over global landmasses should provide never-seen terrestrial water gradients at sub-monthly and at 10-100m resolutions, there will remain spatiotemporal gaps between SWOT retrievals. The [ability of SWOT land measurements](#) to directly [contribute](#) to our understanding of [terrestrial hydrology](#) and of the [climate system](#) will therefore likely be demonstrated through their [integration within models](#). Of particular interest are global climate models, and specifically their land hydrology components.

We argue in this proposal that, before launch, the SWOT Science team ought to be able to describe the continental scale sub-monthly waterscape gradients that SWOT should see on land, characterize the measurement capabilities of such features, and recommend on best ways to assimilate data into models.

We plan to build on ongoing research in several institutions to establish an international collaboration that will focus on understanding the best integration methods between expected SWOT terrestrial retrievals and existing global hydrologic/hydrodynamic models.

1.2. Objectives

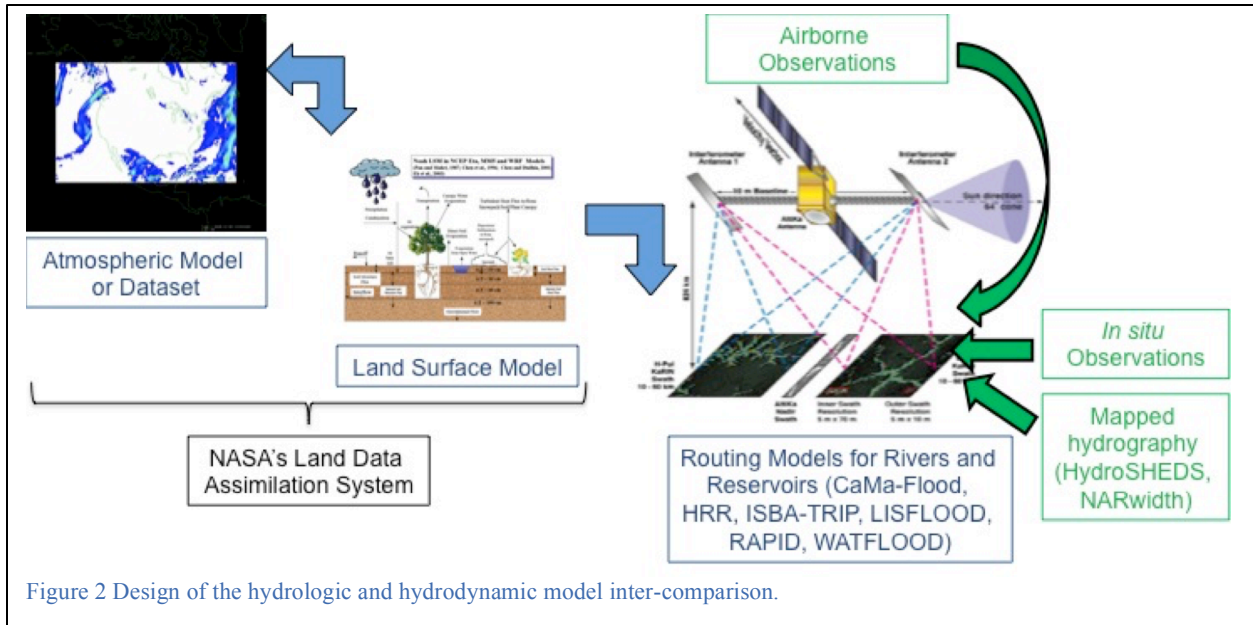
The primary objectives of this proposal are to shed light on the expected SWOT observations at continental to global scale and to investigate the integration of SWOT measurements into a series of global hydrologic models. We will establish a SWOT-friendly continental-to-global scale modeling framework and focus our analysis using six hydro models applied over four continental-scale river basins spanning various climate zones and driving hydrologic processes. Our research will characterize the ability of SWOT terrestrial observations to contribute to the understanding of the terrestrial hydrologic cycle (with impacts on the climate system) and expose the details of how much SWOT data and hydrologic/hydrodynamic models can accomplish when working hand in hand. The proposed research will focus on answering the following two sets of related fundamental science questions:

1. [How can we best prepare for the expected SWOT continental to global measurements before SWOT even flies?](#) That is, how can we understand the relationships between existing surface water variations and expected SWOT large-scale observing capabilities? The combination of outputs from the SWOT hydrology simulator at continental scale (or alternatives, see Section 1.3.1) and the *inter*-model, *intra*-model and multi-forcing variability within hydrologic/hydrodynamic models will provide guidance.

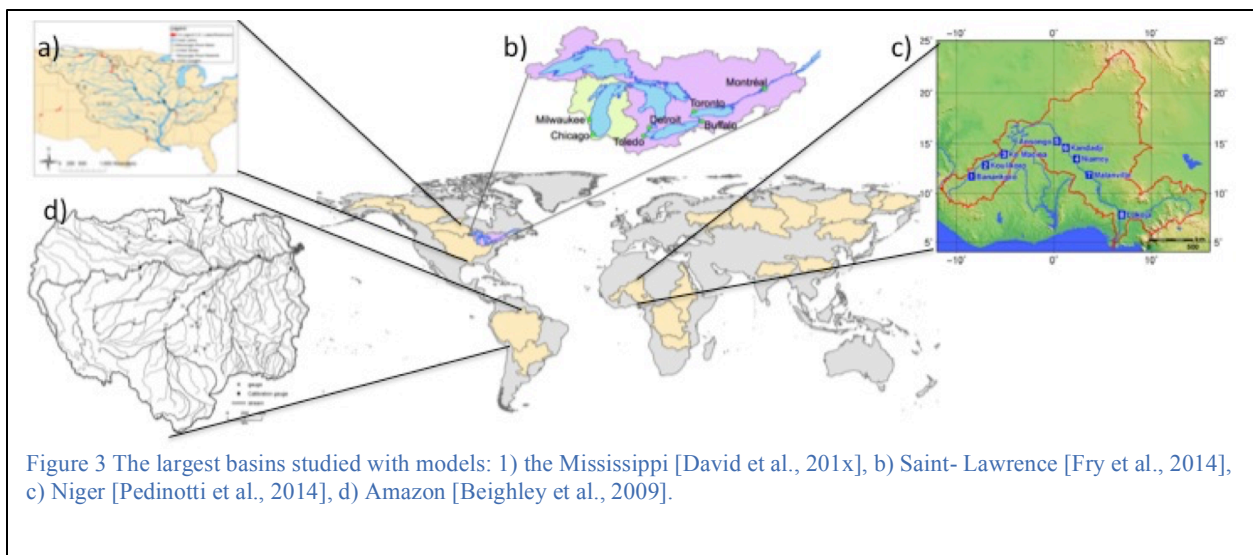
2. [What is the added value of including SWOT terrestrial measurements into global hydro models](#) for enhancing our understanding of the terrestrial water cycle and the climate system? Are current global hydrologic models ready to ingest expected SWOT data? What SWOT variable(s) or SWOT-derived product(s) offer the best promise for integration and for data assimilation? The comparison and the integration of simulations from the SWOT hydrology simulator (or alternatives) into large-scale hydro models will inform on such capabilities.

2. Approach

We will combine an [inter-comparison](#) framework (Figure 2) consisting of a series of [six horizontal water transfer schemes](#): [CaMa-Flood](#) [Yamazaki *et al.*, 2011], [HRR](#) [Beighley *et al.*, 2009], [ISBA-TRIP](#) [Oki and Sud, 1998; Decharme *et al.*, 2012], [LISFLOOD-FP](#) [De Roo *et al.*, 2000; Neal *et al.*, 2012], [RAPID](#) [David *et al.*, 2011], and [WATFLOOD](#) [Kouwen, 2010]. These models will be fed by runoff produced by the four land surface models of NASA's Global Land Data Assimilation System [Rodell *et al.*, 2004].



Our analysis will focus on [four continental-scale river basins located across climatic zones, with distinct hydrological challenges, and subject to various levels of anthropogenic influence](#) (Figure 3). The river basins will be that of the Amazon, Mississippi, Niger, and Saint Lawrence. The [Mississippi](#) Basin spans climate zones varying from cold to wet to semi-arid and contains 8 of the 15 largest US lakes and reservoirs. The [Amazon](#) Basin is hot and humid and subject to regular flooding. The [Niger](#) River flows through a large semi-arid region and is also subject to annual flooding within the complex inner-delta region. The [Saint-Lawrence](#) Basin is located in a cold climate and includes the five American Great Lakes. These basins are selected for their hydrological challenges, but also in order to leverage past and ongoing studies by team members, hence minimizing the burden of data gathering and pre-processing (e.g. basin delineation, downloading and formatting observations, preparing river networks). The inter-comparison will be informed by a series of existing studies. Co-I Beighley and Collaborator Yamazaki both have experience with the Amazon Basin [*Beighley et al., 2009; Yamazaki et al., 2012*]. PI David has three studies of the Mississippi Basin [*Cai et al., 2014; David et al., 2015;*



Tavakoly et al., 2016]. Collaborator Boone directed studies of the Niger Basin [*Pedinotti et al.*, 2012, 2014]. Collaborator Fortin guided a study of the Lake Michigan Basin [*Fry et al.*, 2014] that is currently being extended to the Saint Lawrence Basin.

We expect that such [inter-comparison](#) of horizontal hydrologic/hydrodynamic models could set the foundation for a multi-model framework of surface water simulations and become [as valuable for SWOT as GLDAS has been for GRACE](#) [*Rodell et al.*, 2009; *Famiglietti et al.*, 2011].

3. Analysis and anticipated results

3.1. Inter-comparison deliverables 1: the impact of modeling paradigms within hydro models on simulations of expected SWOT retrievals

The modeling approaches vary in equation complexity (advection, diffusion or both), process representation (with/without representation of flooding, with/without surface routing prior to river routing), spatial decomposition (catchment-based or grid-based), spatial resolution (varying between 5 and 50 km in computational element size), temporal resolution (daily to sub-daily), and inclusion of man-made infrastructures (i.e. reservoirs). We expect to [shed light on the impact of these various modeling paradigms on simulations of width, height, slope and flow](#) within water bodies in order to [prepare for implications on SWOT retrievals](#). Our analysis will focus on answering the following questions:

- How much difference is due to the [type and resolution of the modeling elements \(grid/catchment\)](#)?
- What is the influence of the [river routing method](#) (kinematic, Muskingum, Manning) and of the [surface routing method](#) (lumped, kinematic) on simulated flow hydrographs?
- What is the impact of the [flooding representation](#) (w-d relationships, cross section data, topographic index) on simulated widths and heights?
- How important is the accounting for [anthropogenic influence](#) (e.g. reservoirs)?
Additionally we will assess the necessity of the following processes that are currently either poorly represented or completely missing from continental to global scale hydro models:
 - River/aquifer interactions
 - Human withdrawals

3.2. Inter-comparison deliverables 2: inform on what to expect from SWOT

One of the main mysteries that SWOT should be able to shed light on is the sub-monthly variability of spatial gradients of surface water width, height, slope and flow. In order to help inform on what we should expect from SWOT retrievals, [the inter-comparison will produce](#) the following [flow statistics \(for width, height, slope and flow\)](#) for each of our [study basins, hydro models](#), and [GLDAS runoff inputs](#). We will hence characterize the expected direct SWOT observations (width, height and slope) and SWOT-derived observations (flow) and quantify the greatest errors or sources of uncertainty through the following questions:

- How large is the [intra-model variability](#) (mean, median, maximum, minimum, standard deviation) [in flow quantities](#) (width, height, slope, flow) at the continental to global scale?
- How large is the [inter-model variability](#) (mean, median, maximum, minimum, standard deviation) [in flow quantities](#) (width, height, slope, flow) at the continental to global

scale?

- How large is the [multi-runoff variability](#)? That is, what is the influence of the runoff products from GLDAS models on *intra*- and *inter*-model variability?
- How do [these variations compare with currently-available](#) [e.g. *Andreadis et al.*, 2013; *Gleason and Smith*, 2014; *Yamazaki et al.*, 2014; *Allen and Pavelsky*, 2015]?

We plan to analyze [the impact of the variability in expected observations](#) (width, height, slope) [and derived quantities](#) (flow) on the potential retrievals of SWOT by feeding the simulations into the [SWOT hydrology simulator](#). Hence, the proposed work will help preparing for SWOT by [exposing what SWOT should see from space](#). Given that each model currently produces data in varying formats, we will review each model output approach and revise the source codes to ensure consistency of files used in the inter-comparison.

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