## Machine Learning Based Salinity Emulator Dashboard

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# Study Objectives

Develop machine learning based tools to estimate Delta salinity, with specific goals to:

- > Explore novel deep-learning approaches beyond the conventional Multilayer Perceptron (MLP) model.
- Develop ML models in the multi-task learning (MTL) paradigm allow a single model to evaluate outputs at multiple study locations.
- > Develop a browser-based dashboard to facilitate user interaction (Focus of this presentation)







# Study Background

The Dashboard is based on the ML Architectures presented in two recent studies:

#### Multi-Location Emulation of a Process-Based Salinity Model Using Machine Learning

by 🙁 Siyu Qi <sup>1,\*</sup> 🖂 😳, 😩 Minxue He <sup>2,\*</sup> 🖂, 🙁 Zhaojun Bai <sup>3</sup> 🖂, 😩 Zhi Ding <sup>1</sup> 🖂, 😩 Prabhjot Sandhu <sup>2</sup> 🖂, A Yu Zhou <sup>2</sup> , A Peyman Namadi <sup>2</sup> , A Bradley Tom <sup>2</sup> , A Raymond Hoang <sup>2</sup> and 🙎 Jamie Anderson <sup>2</sup> 🖂

https://www.mdpi.com/2073-4441/14/13/2030

#### **Application of Multiple ML Architectures:**

- Multi-layer perceptron (MLP)
- Long-Short-Term Memory (LSTM)
- Gated Recurrent Unit (GRU)
- Residual Network (ResNet)

**Target (label)**: daily DSM2 historical simulations ("noise free") of EC from 1990-2019

#### Novel Salinity Modeling Using Deep Learning for the Sacramento–San Joaquin Delta of California

by 🙁 Siyu Qi 1,\* 🖂 💿, 🙁 Minxue He 2,\* 🖂 💿, 🙁 Zhaojun Bai 3, 🙁 Zhi Ding 1, 🙁 Prabhjot Sandhu 2, Strancis Chung<sup>2</sup>, Peyman Namadi<sup>2</sup>, Yu Zhou<sup>2</sup>, Raymond Hoang<sup>2</sup>, Revealed Bradley Tom<sup>2</sup>, 😣 Jamie Anderson <sup>2</sup> and 😣 Dong Min Roh <sup>4</sup> 💿

https://www.mdpi.com/2073-4441/14/22/3628

#### **Applied novel architectures:**





• residual long short-term memory (Res-LSTM) network • residual gated recurrent unit (Res-GRU) model

**Target (label)**: observed data from 2001–2019

# Study Domain

#### 8 Input variables

Index	Input Feature Name	Definition			
1	Northern Flow	Sum of Sacramento, Yolo Bypass, Mokelumne River, Cosumnes River, and Calaveras River flows			
2	San Joaquin River Flow	San Joaquin River at Vernalis Flow.			
3	Pumping	Sum of pumping from Banks Pumping Plant, Jones Pumping Plant, and Contra Costa Water District at Rock Slough, Old River, and Victoria Canal.			
4	Delta Cross-Channel Gate Operation	Delta Cross-Channel Gate Openings.			
5	Consumptive Use	Net Delta Consumptive use estimated by Delta Channel Depletion (DCD) and Suisun Marsh Channel Depletion (SMCD) models.			
6	Martinez Tidal Energy	Tidal energy at Martinez, calculated as the daily maximum–the daily minimum astronomical tide at Martinez.			
7	San Joaquin River EC	Electrical conductivity measured at San Joaquin River at Vernalis.			
8	Sacramento River EC	Electrical conductivity measured at Sacramento River at Greens Landing.			







#### 23 Electrical Conductivity (EC) Monitoring Stations



# Results



Exceedance probability plot and time series plot of Res-LSTM simulated versus observed salinity at daily time step.





# **Dashboard Introduction**

- A complementary browser-based Delta Salinity Dashboard was developed to serve as the front-end user interface for the DSM2 salinity emulation machine learning models described in the previous slides.
- Users can interactively explore hypothetical scenarios (e.g., by varying Delta boundary conditions including inflows, export levels, boundary salinity, etc.) and view the corresponding salinity outputs at key compliance locations during user-defined simulation periods.







## Dashboard Architecture

Models are trained on **Google Colab** using the scripts reviewed today (they can also be trained locally). Models, template input data, evaluation dashboard scripts are stored in a **GitHub** repository.



Pre-trained emulator models are hosted on **Microsoft Azure**, and *evaluation* of the models are computed on their servers.



Servers can be scaled up or out to accommodate higher machine workloads (for more complex models) or larger user volumes.

## **Dashboard Access**



## <u>dwrbdodash.azurewebsites.net</u>















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### Click tab for parameter to be scaled

ANI nputs - Input Scaler



Adjust monthly scaling sliders







Ilation Year													
08	2009	2010	2011	2012	2013	2014	2015	2016	2017	2018			





