

Generalized Delta Conservative Constituent Modeling using Artificial Neural Networks

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Funding: State Water Contractors

Presentation to the Real Time Data Forecasting Committee
July 22, 2014

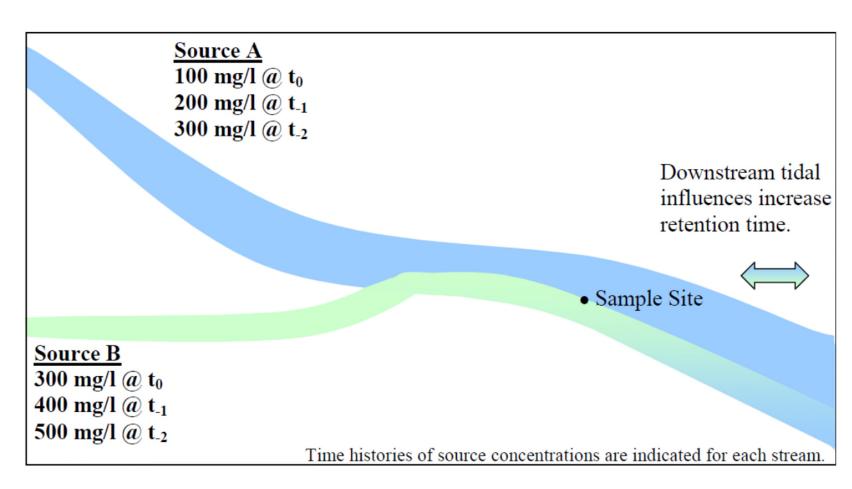


Overview

- DSM2 model was used to simulate volumetric contribution from boundary sources across multiple stations in the Delta
- DSM2 used in the fingerprint mode can be used to relate the sources of water and any conservative constituent at a given location
- In this work we have developed an emulator for DSM2 using an Artificial Neural Network approach
- Goal is to have a simplified tool that performs calculations similar to those performed by DSM2
- The ANN-based model for individual stations was then used to compute the concentrations of three conservative constituents (salinity, bromide, and DOC) given boundary values, thus emulating the DSM2 processes



Fingerprint Model Conceptualization

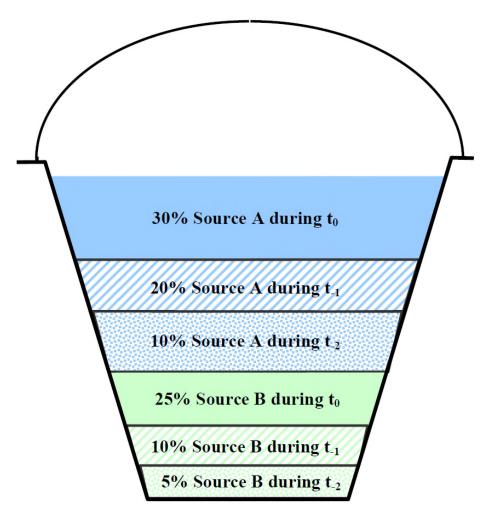


Source: Jamie Anderson, 2002, DSM2 Fingerprinting Methodology

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Concentrations at a Given Location



Source: Jamie Anderson, 2002, DSM2 Fingerprinting Methodology

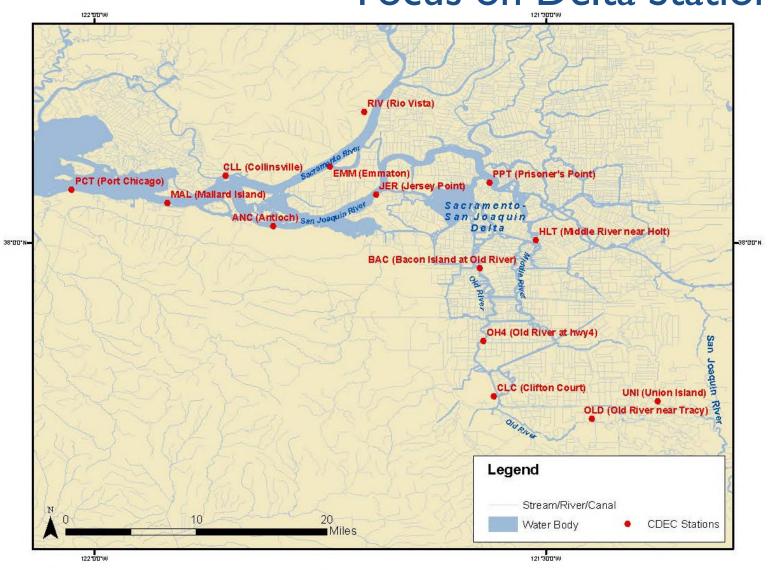


First Step: Does the DSM2 Fingerprinting Model Work as Expected?

- DSM2 finger printing results, along with EC at boundaries, were used to predict EC at validation locations within Delta
- Predicted EC values from the fingerprint model were compared to DSM2-simulated EC at six validation locations:
 - Jones Pumping Plant
 - Banks Pumping Plant
 - Old River at Bacon Island
 - San Joaquin River at Jersey Point
 - Sacramento River at Mallard Island and
 - Old River at Highway 4

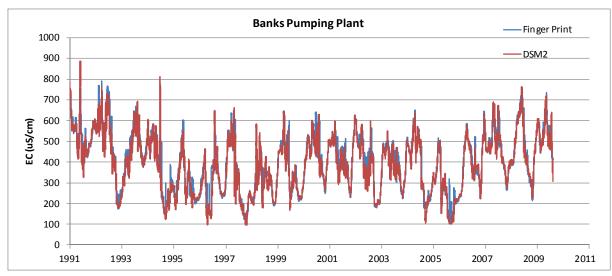


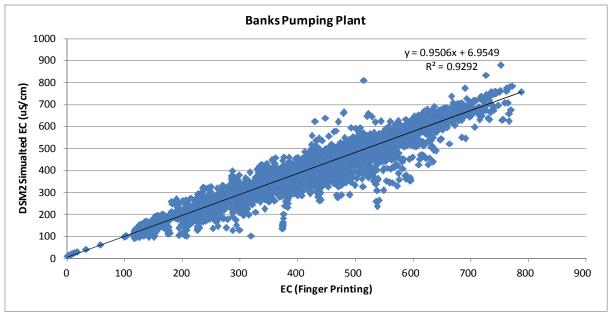
Focus on Delta Stations





Fingerprint Model Validation: Banks Pumping Plant

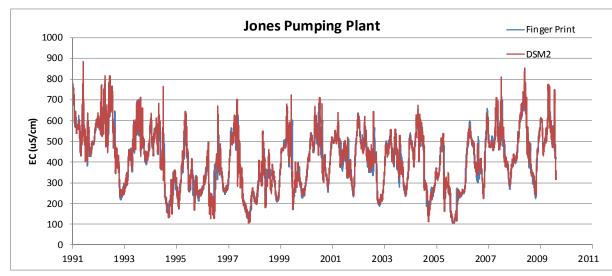


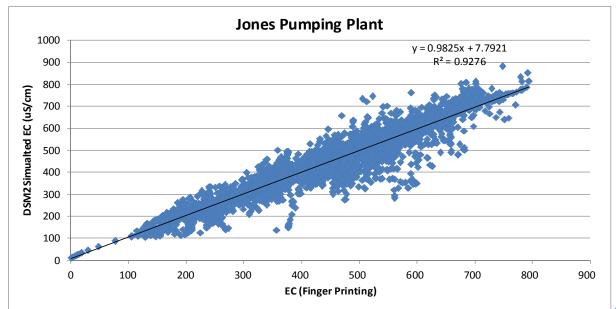


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Off Ramp Validation: Jones Pumping

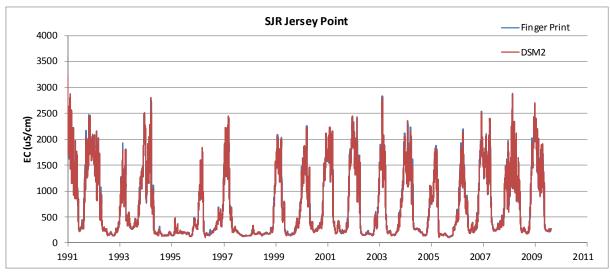


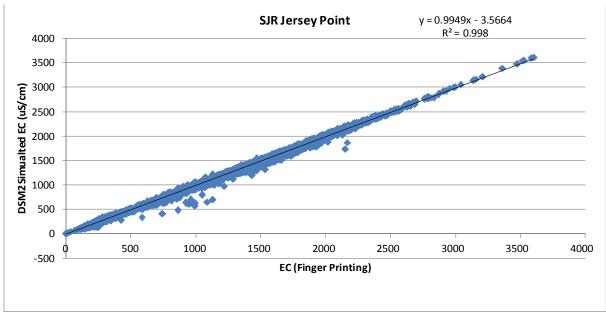


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Off Ramp Validation: SJR Jersey Point





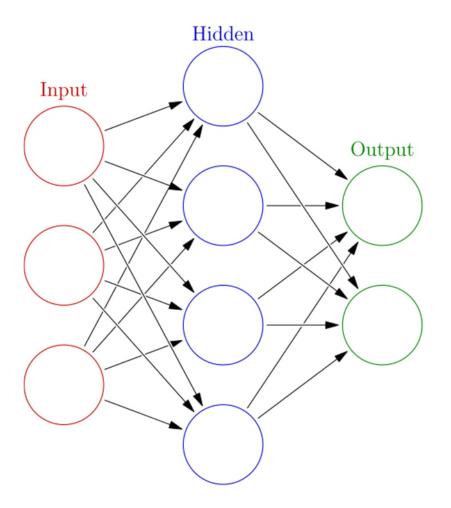


Off-Ramp Summary

 The DSM2 fingerprinting validation was considered adequate to use for ANN development



What are Artificial Neural Networks (ANNs)?



- Mathematical models inspired by biological neural networks
- Similar to natural systems, they can be "trained" using real-world data. In the context of ANNs, this includes varying inputs, hidden layers, and weights associated with individual nodes. This is analogous to model calibration.
- Over the last two decades, ANNs have been used for modeling complex relationships between inputs and outputs or to find patterns in data
- In a mathematical sense, often called universal emulators

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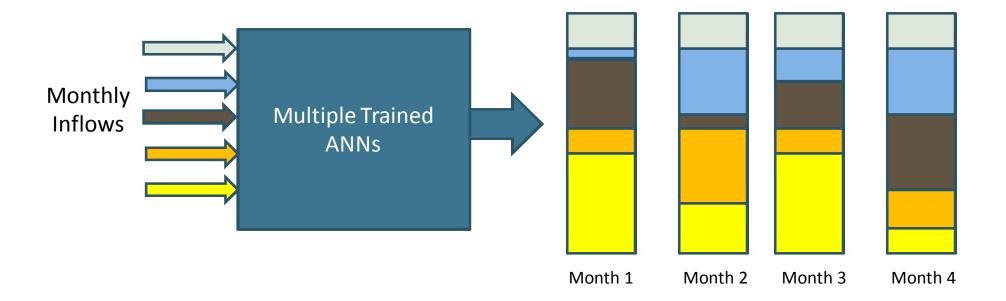
Why develop an ANN model to emulate the fingerprint model?

- A generalized tool that can be applied without the expertise needed to run DSM2
- ANNs (developed in Matlab) can be exported to Excel or called as DLLs from other modeling platforms
- Can be embedded within other tools, that seek to optimize some planning objective, especially where the model needs to be called multiple times (say, 100's of calls). This is the current application of the salinity ANN within CALSIM.



What did we implement here?

Volumetric Mix of Inflows at a Given Station





Boundary Flows Used in ANN Development

- Sacramento River at Freeport
- San Joaquin River at Vernalis
- Mokelumne River at Benson's Ferry
- Calaveras River
- Yolo Bypass
- Martinez
- Delta agricultural return flows (divided into 9 subregions)
- Fifteen (15) possible inflows, we term the first 6 the primary flows, and the rest the DICU flows
- DCC gate status (open or closed) for selected stations



Output Locations

	Model Output Location	Key Boundary Influence(s) Anticipated	
1	Port Chicago	Martinez	
2	Mallard/Chipps Island	Martinez	
3	Collinsville	Martinez	
4	Antioch	Martinez	
5	Emmaton	Martinez; Freeport; Yolo	
6	Rio Vista	Freeport; Yolo	
7	Jersey Point	Martinez; Freeport; Vernalis; Ag Returns	
8	Old River @ Bacon Island	Martinez; Freeport; Vernalis; Ag Returns	
9	Old River @ Highway 4	Martinez; Freeport; Vernalis; Ag Returns	
10	CCF Intake	Martinez; Freeport; Vernalis; Ag Returns	
11	Jones Pumping Plant	Martinez; Freeport; Vernalis; Ag Returns	
12	SJR @ Prisoner's Point	Martinez; Freeport; Vernalis; Mokelumne; Calaveras; Ag Returns	
13	Middle River @ Holt	Martinez; Freeport; Vernalis; Mokelumne; Calaveras; Ag Returns	
14	Middle River @ Victoria	Martinez; Freeport; Vernalis; Mokelumne; Ag Returns	
15	Old River @ Tracy Rd Bridge	Vernalis; Ag Returns	
16	Middle River @ Union Island	Vernalis; Ag Returns	
17	SJR @ Hw∳ ⁵ 4	Vernalis; Ag Returns; Calaveras	



Multiple DSM2 Simulation Scenarios Were Considered

Run #	Ag Barriers	DCC	S. Delta Exports
1	Historical	Historical	Historical
2	In	Open	Historical
3	Out	Open	Historical
4	In	Closed	Historical
5	Out	Closed	Historical
6	Historical	Historical	None
7	In	Open	None
8	Out	Open	None
9	In	Closed	None
10	Out	Closed	None

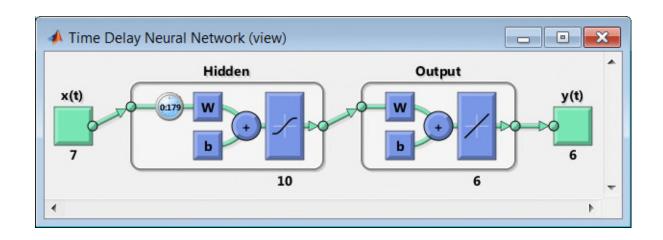


ANN Modeling Goal

- Using time history of flow, for each of the flows, estimate volumetric contribution at selected output locations
- Thus, a user would provide an estimate of the last six months of the flow inputs, and the ANN would estimate the relative contribution of each of the input volumes (% from each source by month)
- For conservative constituents, the % volumetric contribution multiplied by inflow concentration provides estimates of concentrations at output stations.
- DSM2 model time frame provides data for training: 1990-2010



ANN Model Structures



- 10 hidden neurons
- 180 days time delay
- Different input structures examined
- All possible inflows were considered for each output location, letting the ANN identify the more important flow contributions
- One ANN was developed for each station for five primary inflows (San Joaquin River at Vernalis, Mokelumne River, Calaveras River, Yolo Bypass, and Martinez); additional ANNs were developed for the Sacramento River for each month; and other ANNs for each DICU inflow.
- 187 ANNs for primary inflows (17 outputs X 11 flows)
- 153 ANNs for DICU inflows (17 outputs X 9 flows)

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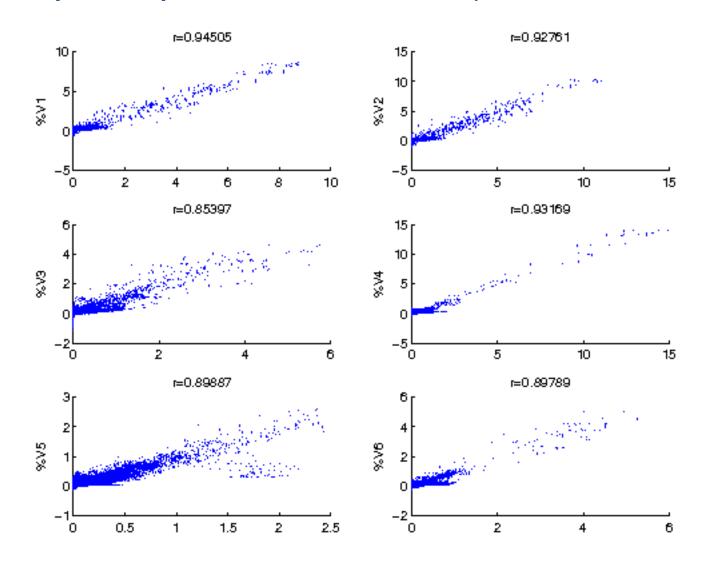


Representative Examples

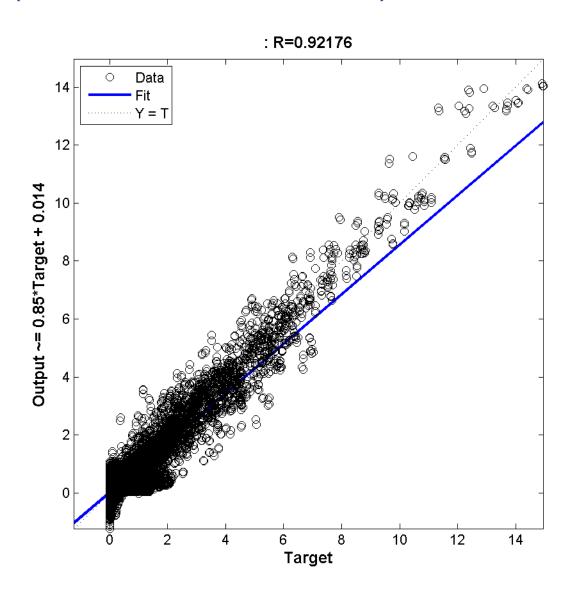
- We show the training at one station: Antioch
- Plots are shown for the major tributaries, comparing the DSM2-calculated versus the ANN-calculated flow contributions
- For each inflow, there are six possible contributions,
 from 1 to 6 months prior to the current time
- The model application is shown for individual conservative constituents (DOC, Br, EC) at different stations throughout the Delta



Results at Antioch: Contribution from Calaveras River (Each plot represents one month)

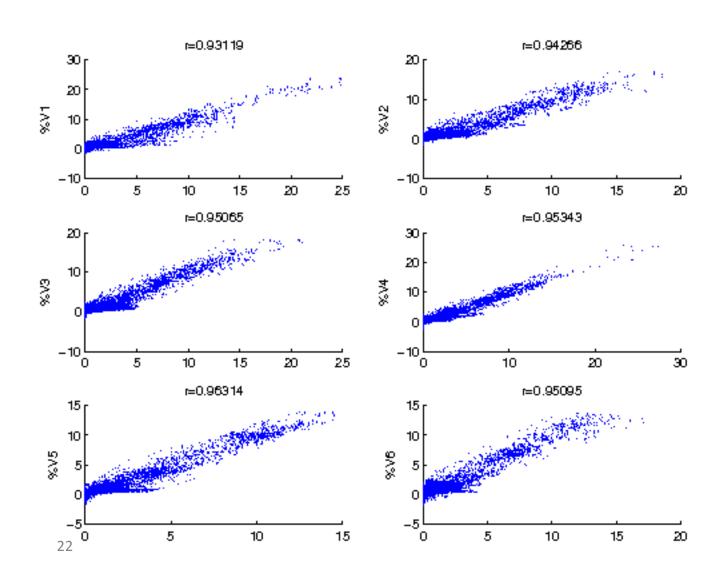


River (All months combined)





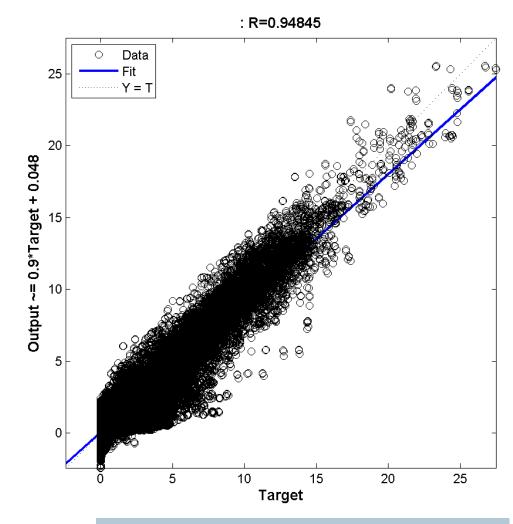
River (Each plot represents one month)





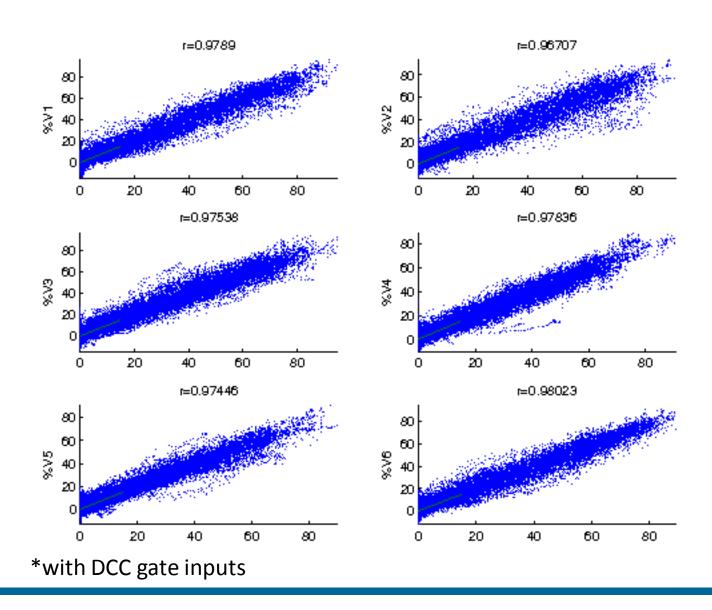
River (All months combined)







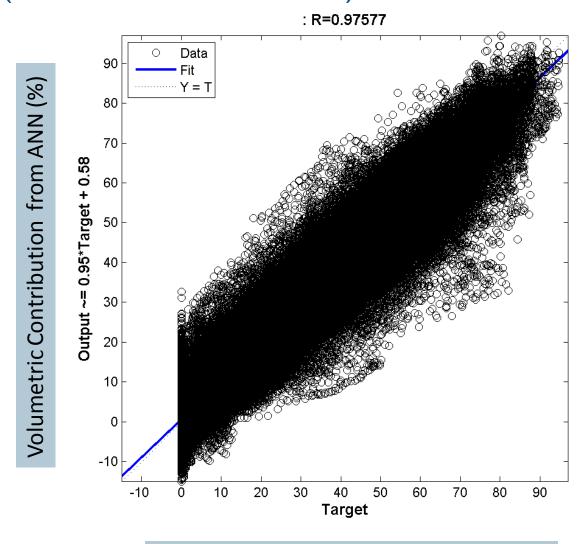
River (Each plot represents one month)*



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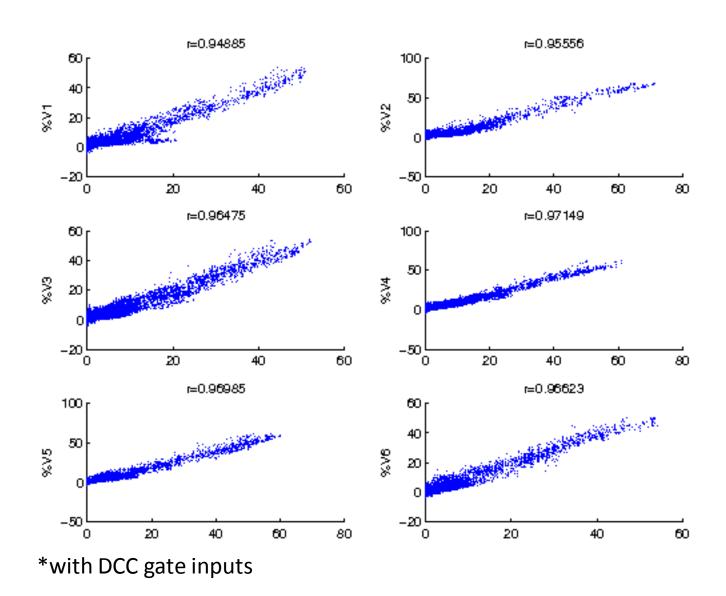


River (All months combined)





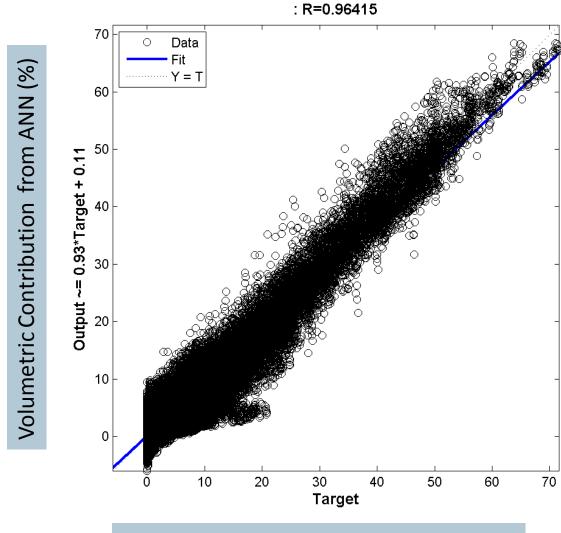
River (Each plot represents one month)*



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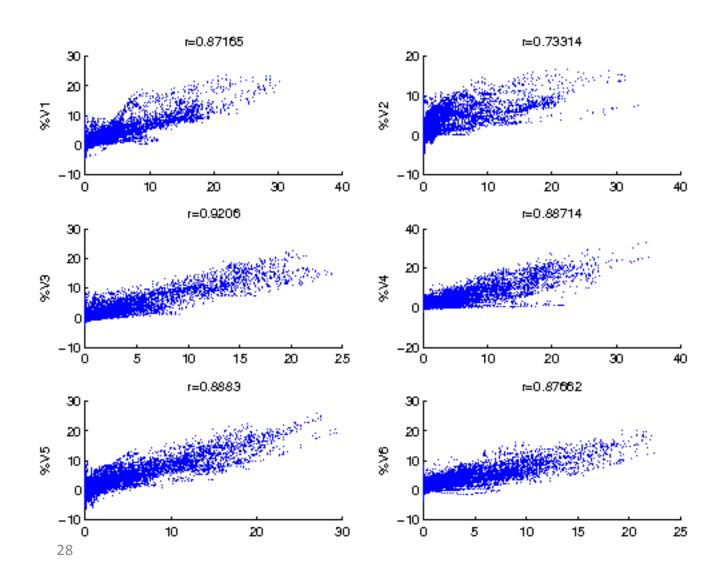


River (All months combined)





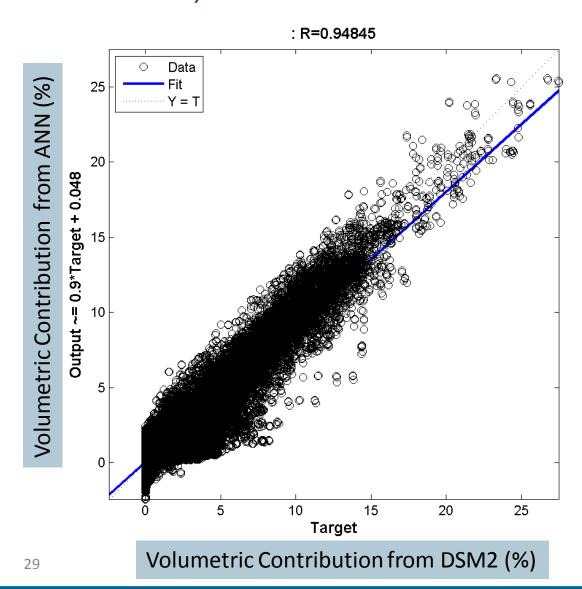
Results at Antioch: Contribution from Martinez (Each plot represents one month)



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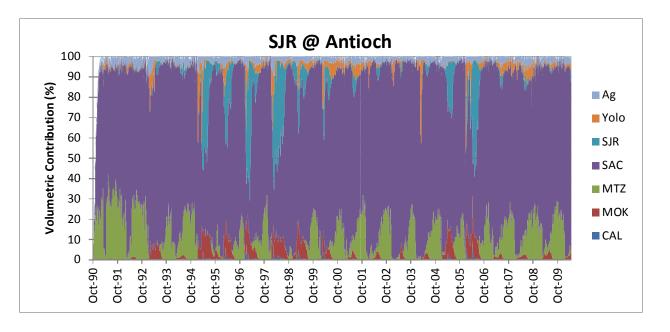
months combined)

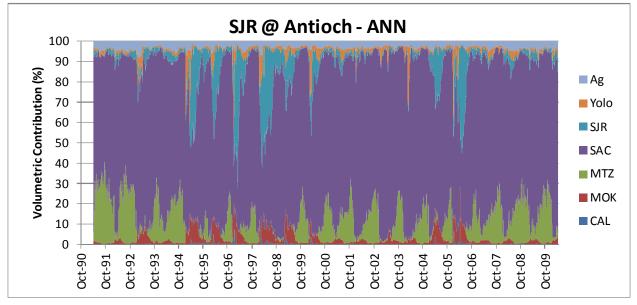




Model Comparison ANN vs. DSM2

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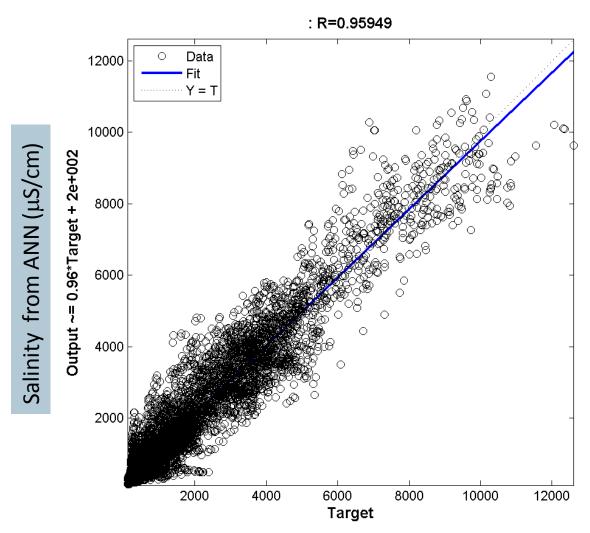


ANN Application to Compute Concentrations

- The ANN-simulated volumetric contribution from different sources, along with conservative constituent concentrations at boundaries, was used to estimate concentrations at different locations within the Delta
- Three conservative constituents were considered:
 - Electrical conductivity (EC)
 - Bromide (Br)
 - Organic carbon (OC)
- The estimated concentrations from the fingerprint ANN were compared to values estimated from the DSM2-simulated volumetric contribution
- The results suggest strong agreement between the two modeling approaches
- Three stations are shown for illustration: Antioch, San Joaquin River at Hwy 4, and Port Chicago

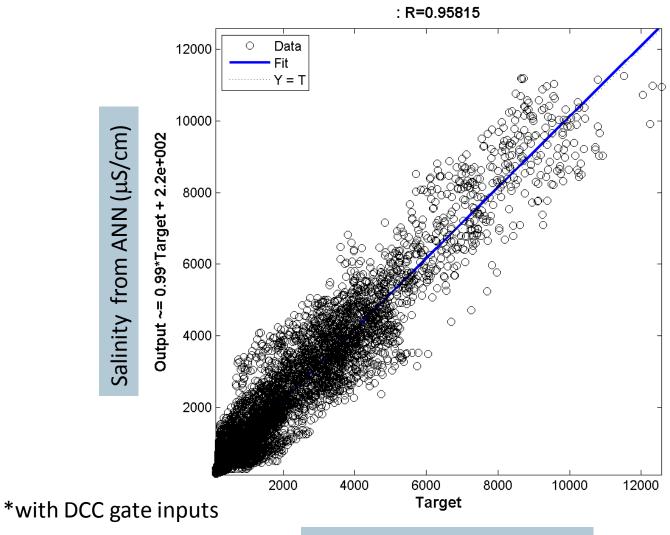


ANN Application to Predict EC at Antioch*



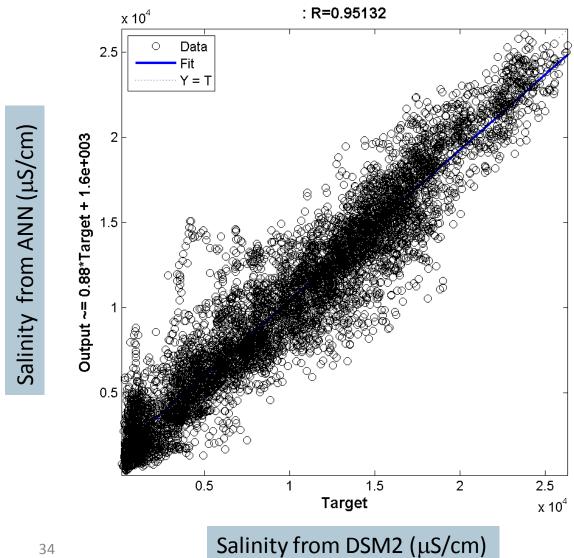


ANN Application to Predict EC at SJR @ HWY4*



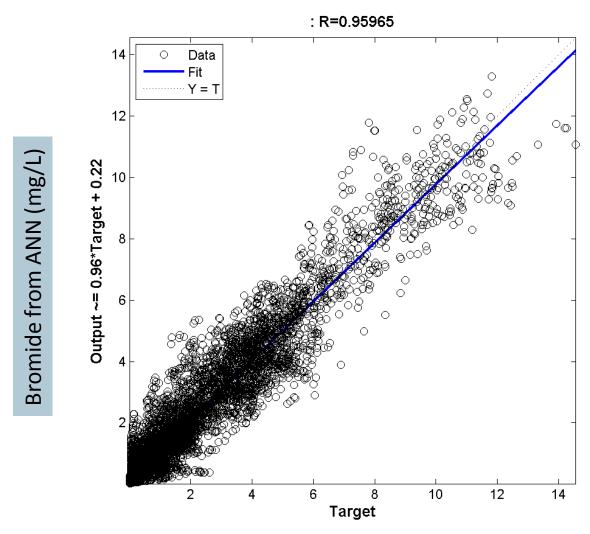


ANN Application to Predict EC at Port Chicago





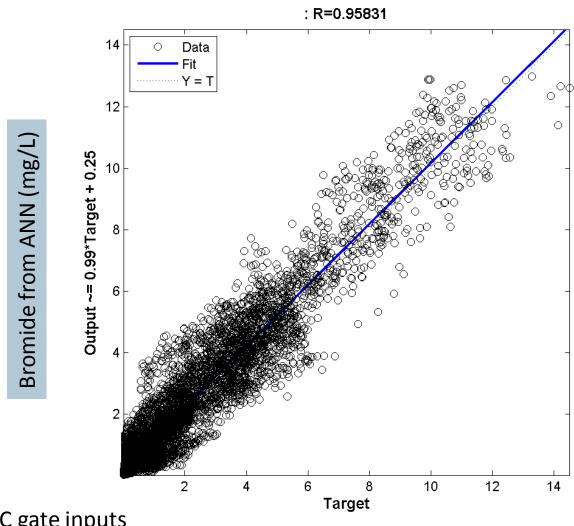
ANN Application to Predict Br at Antioch*



*with DCC gate inputs



ANN Application to Predict Br at SJR @ HWY4*



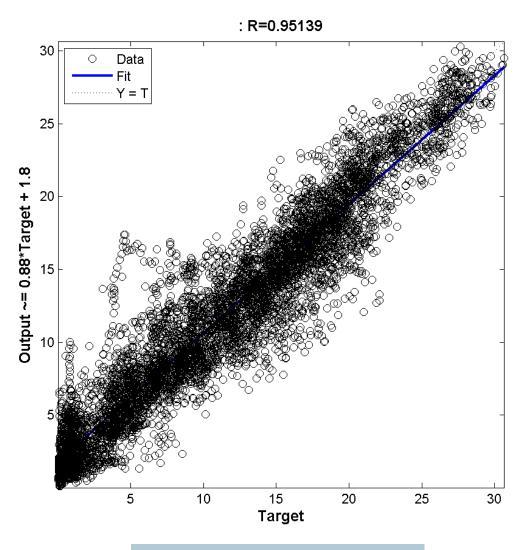
*with DCC gate inputs

Bromide from DSM2 (mg/L)



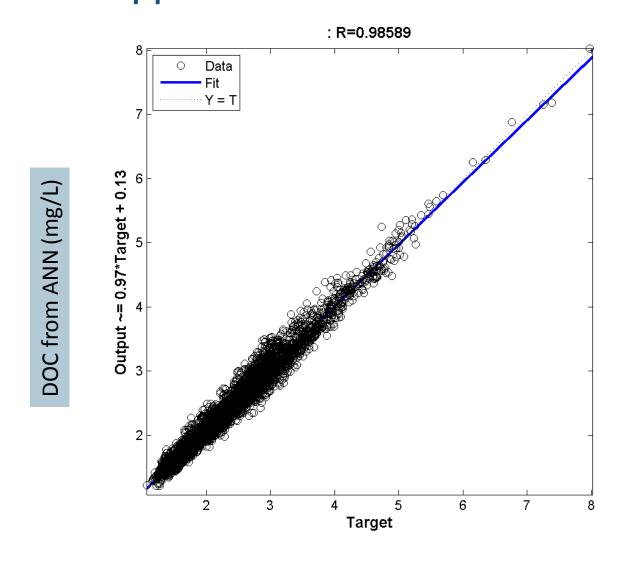
ANN Application to Predict Br at Port Chicago







ANN Application to Predict DOC at Antioch*



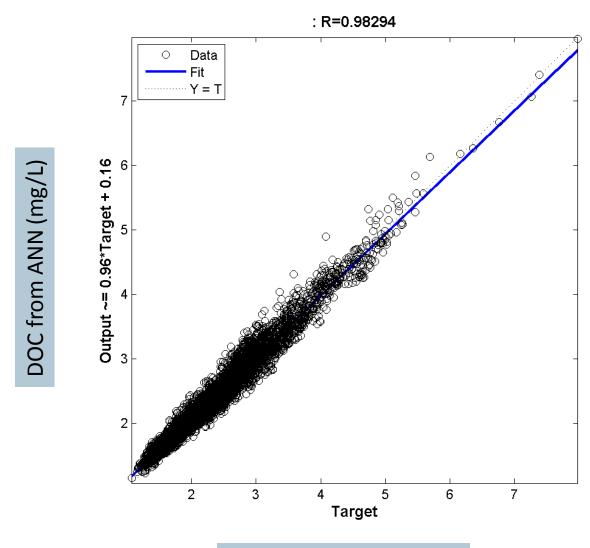
*with DC@ gate inputs

DOC from DSM2 (mg/L)

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ANN Application to Predict DOC at SJR @ HWY4*



*with DC@ gate inputs

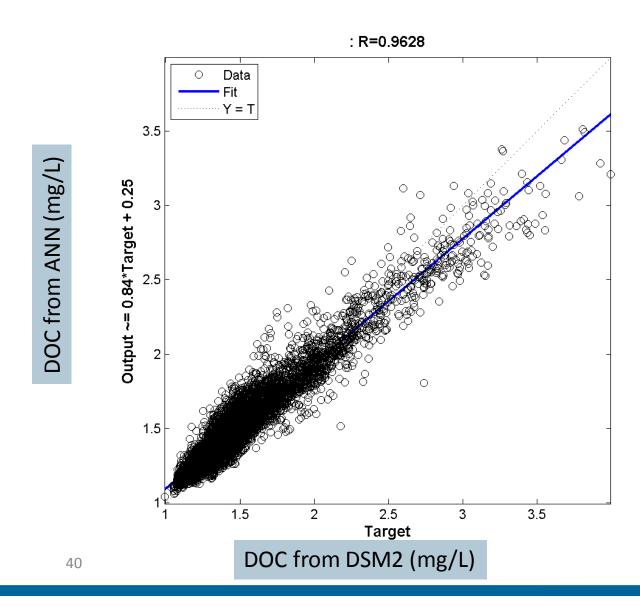
DOC from DSM2 (mg/L)

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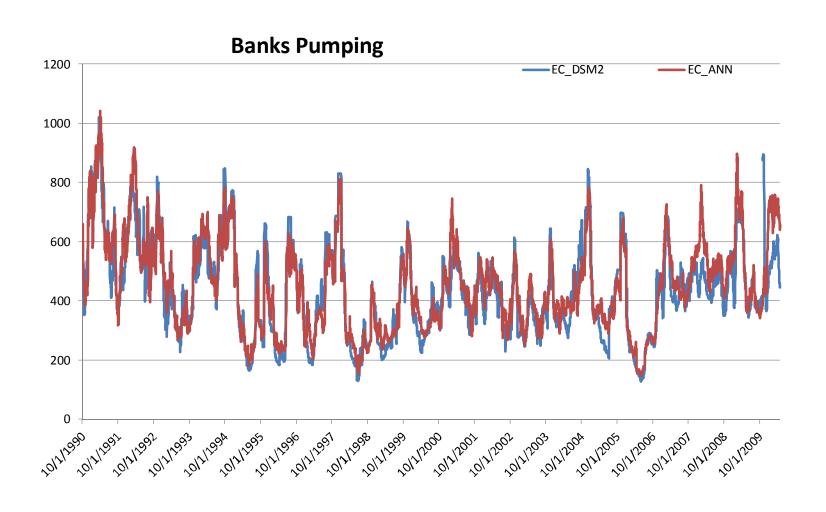
ANN Application to Predict DOC at Port Chicago

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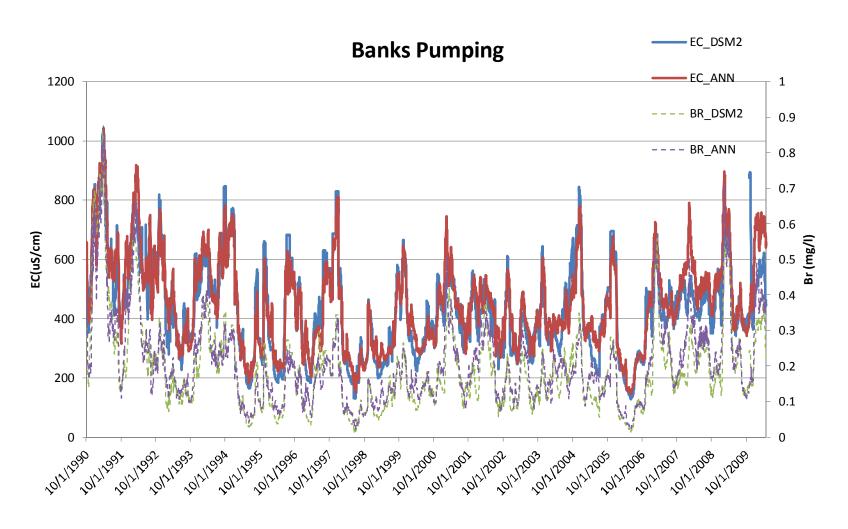


Time Series Example: Banks Pumping Plant EC



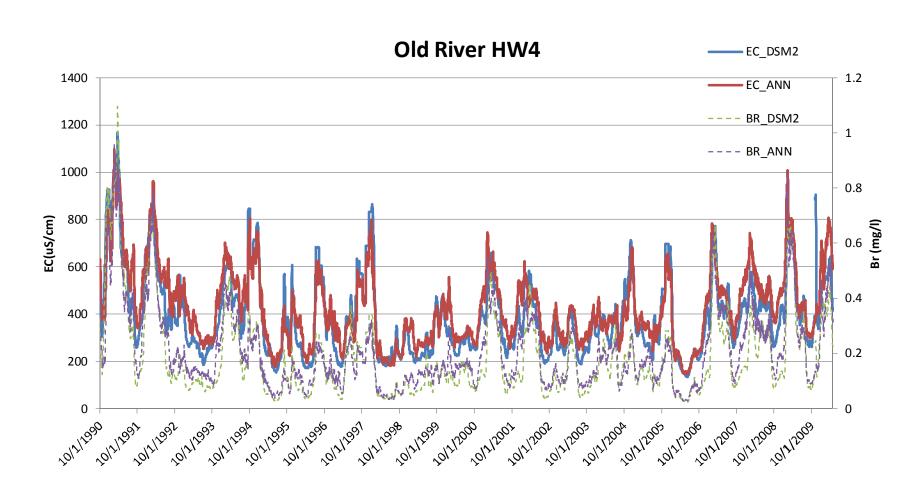


Time Series Example: Banks Pumping Plant EC and Br



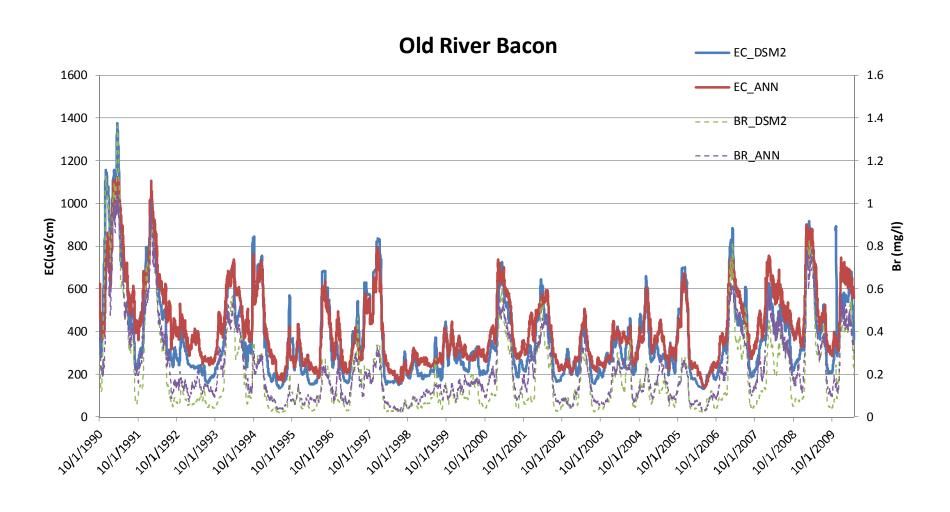


Time Series Example: Old River at Hwy 4 EC and Br





Time Series Example: Old River at Hwy 4 EC and Br





Lessons Learned

- The fingerprint process is complex, given the number of inflows and the time history for each flow
- Multiple ANNs were required to best capture the relationships; 340 ANNs being used
- Correlations for flow were reasonably good across most stations on the San Joaquin and Sacramento Rivers (>0.9), but not as good for Old River and Middle River stations (i.e., were rarely >0.95, good fits were considered to be in the >0.85 range)
- The consideration of the DCC gate status improved fits at several stations and was made part of the input structure
- The ANN application to estimate concentrations for three constituents showed very good results across 16 stations that were targeted in this work; correlations were better than for flow, and ranged from 0.90 to 0.99
- At least one station was not adequately fit using this approach (Middle River at Union Island)