

### Models can be repositories of lots of different types of information.



## The hydrologic system of can be broadly classified into four subsystems.







## Generally only one subsystem is simulated, with various methods for accounting for fluxes from other subsystems.





"The stage of the groundwater reservoir is assumed to be independent of the interflow from the river..."



"RIV does not simulate surfacewater flow in the river – only the river/aquifer seepage."



4



Hydrologic subsystems can have have length and time scales that vary over orders of magnitude.



## Integrated hydrologic models simulate two or more linked subsystems.







Weakly coupled models pass water from one subsystem to another, but only in one direction.



Iteratively coupled models solve each subsystem independently, and water is passed back and forth every timestep.







Fully coupled models solve all subsystems simultaneously using a single system of equations.

Detailed representation of physical processes! But...

...computationally very expensive

...high degree of parameterization

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9

# Model development and evaluation should be centered around the question(s) being asked.









# Late-summer streamflow is the primary concern in the Scott Valley.

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The Scott Valley Integrated Hydrologic Model (SVIHM) is the combination of three models and two types of coupling.



Iteratively
Coupled



# Building the model is often the easier part of model development.



### <sup>14</sup> Sensitivity analysis and calibration are where modelers often spend most of their time.



Simple "black box" model





### Parameters (model input) that change model outputs when perturbed are considered sensitive.



**Less Sensitive** 

15



**Very Sensitive** 

### **Not Sensitive**



Relatively simple integrated models can have a large number of parameters that need to be evaluated.



# Integrated groundwater-surface-water models tend to have a high degree of nonlinearity.





Streamflow and depth depend on groundwater levels

Groundwater levels depend on streamflow and depth



17





# Soil-Water Budget Model (SWBM) parameters are most sensitive in SVIHM.



Graduate Grou

# Low flow (<100 cfs) observations contribute the most information about SVIHM parameters.



### **5 Minute Question Break**

Coming Up

Model Calibration Uncertainty Analysis Integrated Hydrologic Model Applications





Values of model parameters are often poorly known early in model development, and are estimated using available data.



### "Inverse" modeling

**Observed Value = 75** Simulated Value = 72







"SVIHM parameters"





Model results that do not agree with observations can identify problems with your conceptual model.

	<u>(WY 1991-2011)</u>					
Landuse	рЕТ	aET	Applied Water	SW Irrigation	GW Pumping	GW Recharge
			In,	In/yr		
Alfalfa	42.0	40.1	33.1	4.1	29.0	14.6
Grain	16.2	16.1	14.1	2.1	11.9	18.4
Pasture	40.0	33.9	29.7	20.8	9.0	17.2
ET/Nolrr	11.2	10.8	0.0	0.0	0.0	10.8
noET/noIrr	0.0	0.0	0.0	0.0	0.0	21.5

Initial Results

Foglia et al. (2013a)

Alfalfa ~ 14,000 acres (~50% of irrigated land in Scott Valley)

#### Scott Valley Three Year Alfalfa Study (2012-2014)

2012	Field 1	Field 2	Field 3	
aET	30.8	28.9	28.7	
Rain + Irrig.	18.9	21.7	22.3	
2013	Field 1	Field 2	Field 4	
aET	32.8	30.4	35.4	
Rain + Irrig.	32.3	27.0	28.5	
2014	Field 1	Field 2	Field 4	
aET	32.9	28.0	29.1	
Rain + Irrig.	29.5	26.1	22.5	
Avera	Average aET ~ 31 in/yr			

Average Rain + GW Irrigation ~ 26 in/yr

Where is the additional ~5 in/yr of water coming from?





Soil moisture profiles of alfalfa fields continuously dry out during the growing season which limits groundwater recharge.



24

### Incorporating deficit irrigation on alfalfa fields improves agreement between simulated values and observations.

Foglia et al. (2018)

Landuse	рЕТ	aET	Applied Water	SW Irrigation	GW Pumping	GW Recharge
Landuse			ln/	/yr		
Alfalfa	42.0	40.1	33.1	4.1	29.0	14.6
Grain	16.2	16.1	14.1	2.1	11.9	18.4
Pasture	40.0	33.9	29.7	20.8	9.0	17.2
ET/NoIrr	11.2	10.8	0.0	0.0	0.0	10.8
noET/noIrr	0.0	0.0	0.0	0.0	0.0	21.5
Foglia et al. (2013a)						
		<u>Ca</u>	librated Re	esults		
		<u> </u>	WY 1991-2	011)		
	рЕТ	aET	Applied	SW	GW	GW
Landuse			Water	Irrigation	Pumping	Recharge
			Ir	n/yr		
Alfalfa	39.2	36.8	21.5	2.8	18.7	6.3
Grain	16.1	16.1	10.3	1.6	8.7	10.6
Pasture	38.2	34.8	26.0	20.5	5.5	11.6
ET/Nolrr	14.0	11.0	0.0	0.0	0.0	10.8
noET/noIrr	0.0	0.0	0.0	0.0	0.0	21.6

#### Scott Valley Three Year Alfalfa Study (2012-2014)

2012	Field 1	Field 2	Field 3			
aET	30.8	28.9	28.7			
Rain + Irrig.	18.9	21.7	22.3			
2013	Field 1	Field 2	Field 4			
aET	32.8	30.4	35.4			
Rain + Irrig.	32.3	27.0	28.5			
Average GW Irrigation ~ 21 in/yr						
ali	54.5	20.0	2J.1			
Rain + Irrig.	29.5	26.1	22.5			

Orloff et al. (2017)

### 35% reduction in GW irrigation 57% reduction in GW recharge



Numerical models will never match the real-world perfectly, so we look for models that are "good enough" to answer our key questions.



Real World





Great Model











26

Simulated streamflow and groundwater elevations in SVIHM show strong agreement with observations.





27

Simulated streamflow and groundwater elevations in SVIHM show strong agreement with observations.



Well-calibrated models can provide a lot of information about system behavior.

Oro Fino Creek Confluence Dry sections are consistent spatially but vary in Island Road Bridge length/duration 58 Eller Lane Bridge Etna Model appears to capture a Creek Confluence large degree of hyporheic 73 exchange 76. French Creek Confluence 79. 82. Legacy mine tailings section Tailings goes dry every year regardless Section of water year type





Log Flux (m<sup>3</sup>/day)

0

-1

-2

-3

Stream-Aquifer Flux

202-02

37

Shackleford Creek Confluence

# Uncertainty analysis is an important but often overlooked step during model evaluation.



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#### **Observation error**

30



#### Structural error



### There are many ways to evaluate model parameter uncertainty.

Linear 95% confidence intervals calculated by UCODE



Uncertainty within a single calibration

# Uncertainty across calibrations

UCDAVIS Hydrologic Scier Graduate Group

31



Tolley et al. (2019)

### There are many ways to evaluate model parameter uncertainty. 32



# Distributions of possible model parameter values





# Model(er)s can provide a range of predicted values that incorporate uncertainty about parameters, model structure.



### Weakly coupled integrated models (e.g., SVIHM) can properly and efficiently reproduce observed groundwater-surface-water interactions.

34



### **Project Contributors**

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- Professor Laura Foglia
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- Steve Orloff
- Ryan Hines
- Jakob Neumann
- Siskiyou County Resources Conservation District
- Scott Valley Groundwater Advisory Committee
- Scott Valley Stakeholders





### **Questions?**

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