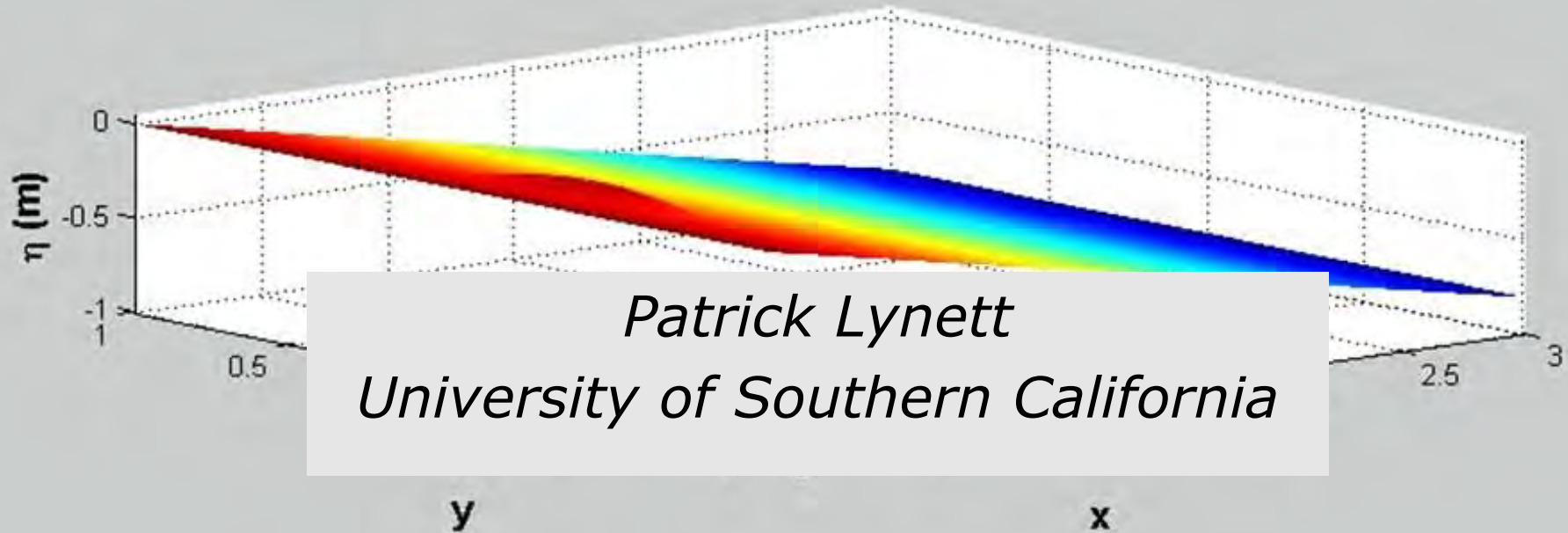
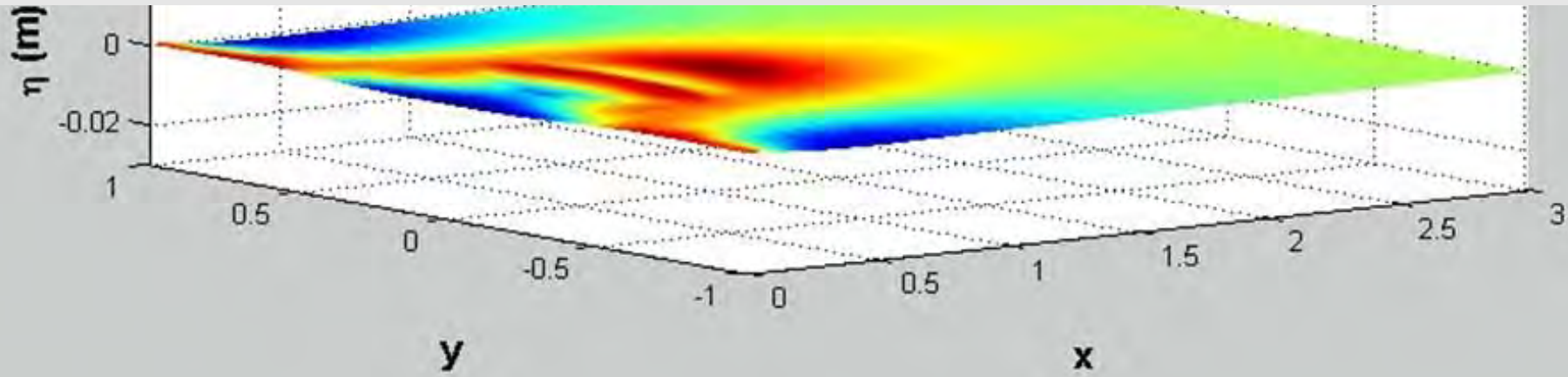
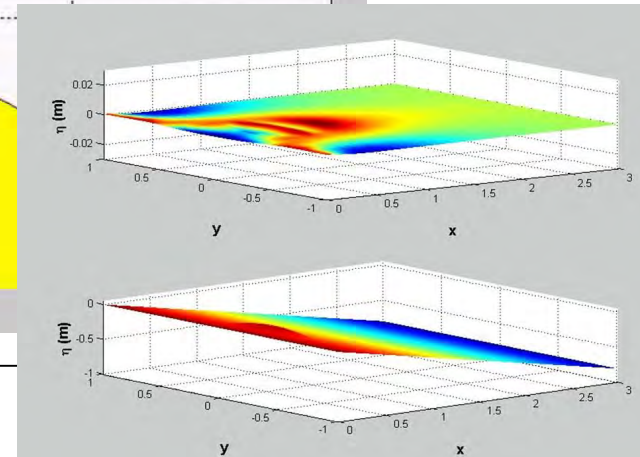
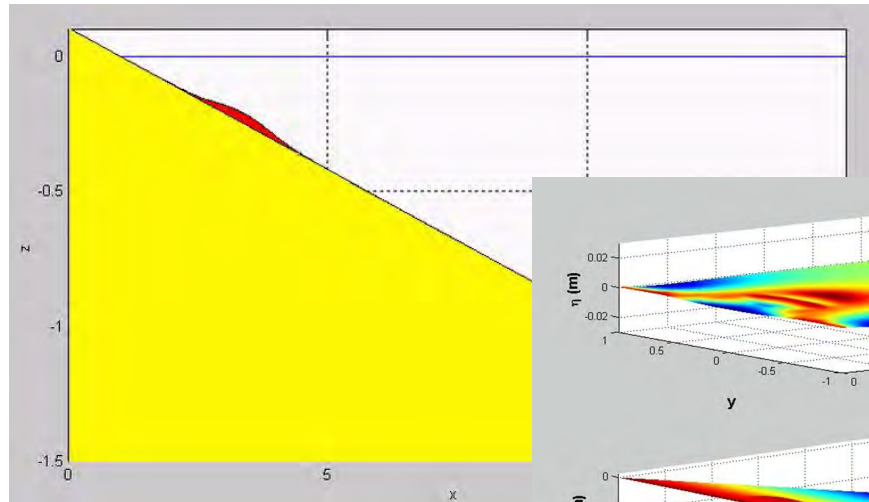
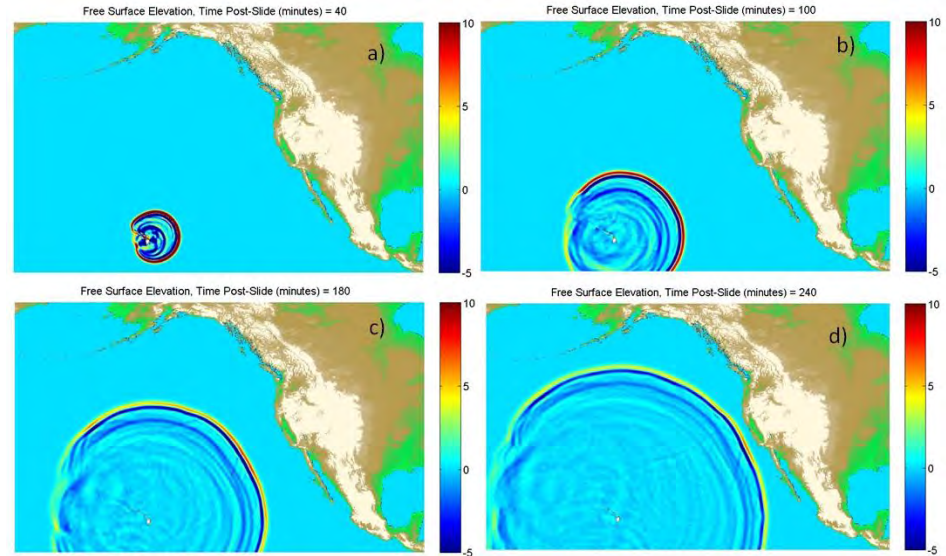


# Landslide Tsunami Modeling Galveston, TX NTHMP Meeting



# Outline & Approach

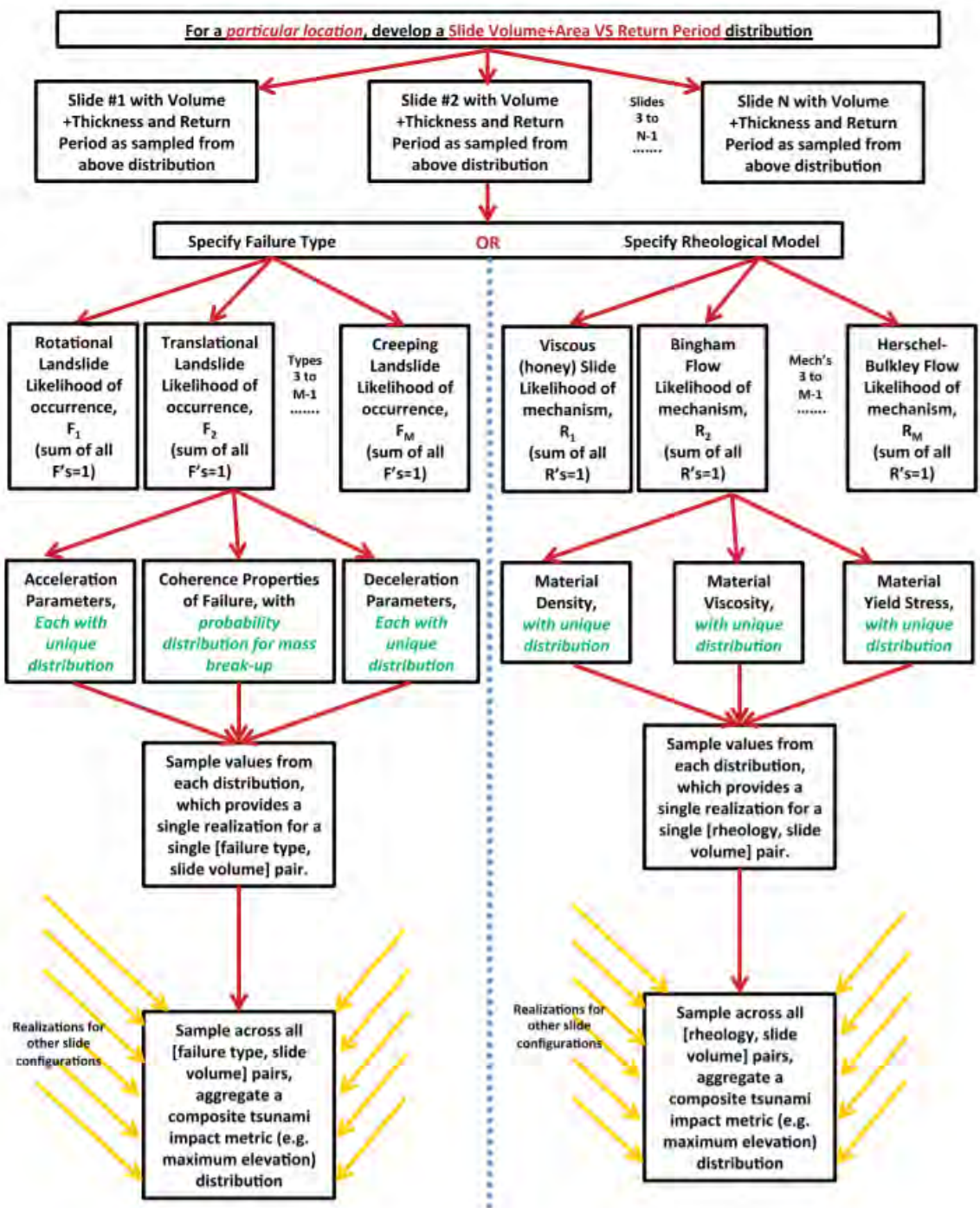
- A bit about my background and previous landslide tsunami applications
- Review of models to be used
  - Linear Mild Slope Equation model
  - Boussinesq-type equations
  - ~~OpenFOAM~~
- Benchmark #1
- Benchmark #2



# Leading thoughts

- Past few years working with USGS & NRC on NPP tsunami hazard assessment
- Would use “upper limit” conservative initial conditions for landslide sources – couldn’t justify using any particular slide motion model
- Full parameter space of potential slide motion is daunting

Geist, E. and Lynett, P. (2014) "Source Process in the Probabilistic Assessment of Tsunami Hazards." Oceanography 27(2), pp. 86-93, doi: 10.5670/oceanog.2014.43.





# Mild Slope Equation Model

(Dingemans, 1997; Bellotti *et al.*, 2008; Cecioni & Bellotti, 2010)

- Free surface evolution equations ( $z=0$ ):

$$\eta_t = G\varphi - \nabla \cdot (F\nabla\varphi) - h_t$$

$$\varphi_t = -g\eta$$

$$F = \frac{c\varphi g}{g} \quad G = \frac{w^2 - k^2 c\varphi g}{g}$$

- Mild-Slope Equation:

$$\eta_{tt} - \nabla \cdot (gF\nabla\eta) + gG\eta = h_{tt}$$

*Time-dependent*

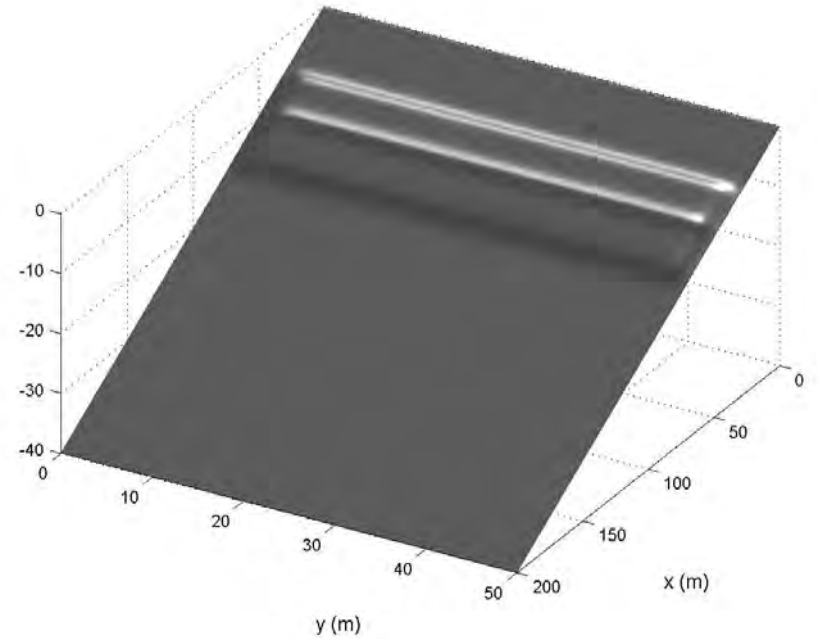
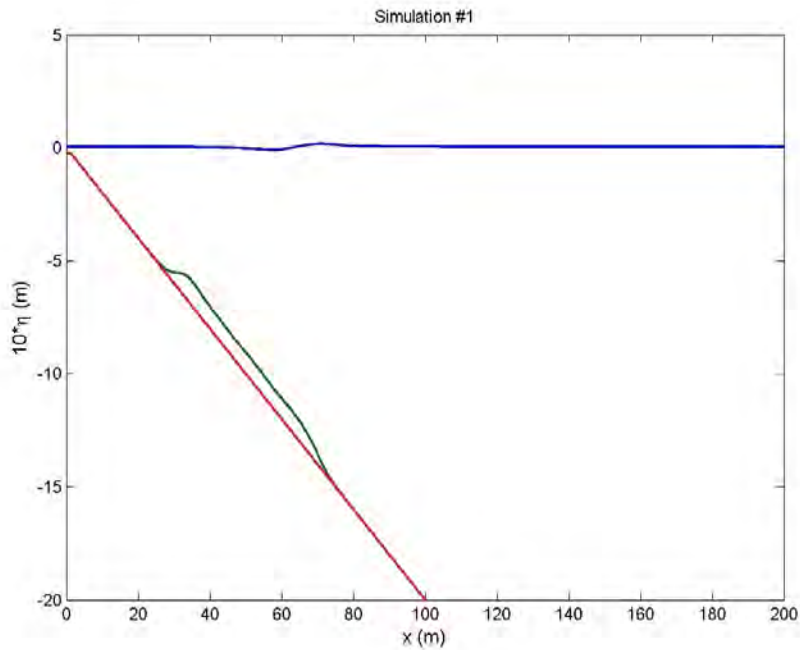


*FFT in time*

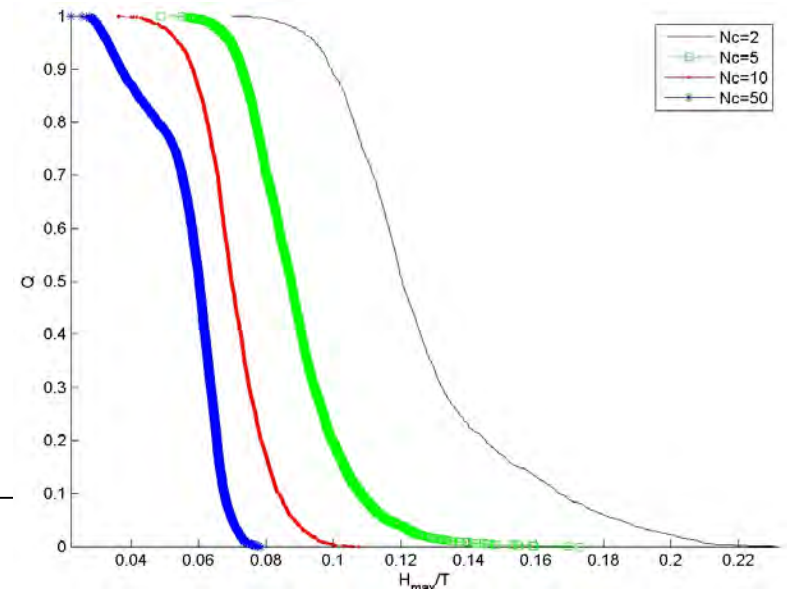
$$\nabla \cdot (c\varphi g \nabla N) + w^2 \frac{c\varphi}{c} N = \frac{1}{\cosh(kh)} H$$

*Frequency-dependent*

# Mild Slope Equation Model



*Fast & accurate for (linear)  
arbitrary slide motion  
Decent engine for MC analysis*



# Boussinesq-type Model

(Lynett & Liu, 2002)

$$\begin{aligned} & \frac{1}{\varepsilon} h_t + \zeta_t + \nabla \cdot (H \mathbf{u}_\alpha) \\ & - \mu^2 \nabla \cdot \left\{ H \left[ \left( \frac{1}{6} (\varepsilon^2 \zeta^2 - \varepsilon \zeta h + h^2) - \frac{1}{2} z_\alpha^2 \right) \nabla (\nabla \cdot \mathbf{u}_\alpha) \right. \right. \\ & \quad \left. \left. + \left( \frac{1}{2} (\varepsilon \zeta - h) - z_\alpha \right) \nabla \left( \nabla \cdot (h \mathbf{u}_\alpha) + \frac{h_t}{\varepsilon} \right) \right] \right\} = O(\mu^4) \end{aligned}$$

$$\begin{aligned} & \mathbf{u}_{\alpha t} + \varepsilon \mathbf{u}_\alpha \cdot \nabla \mathbf{u}_\alpha + \nabla \zeta \\ & + \mu^2 \frac{\partial}{\partial t} \left\{ \frac{1}{2} z_\alpha^2 \nabla (\nabla \cdot \mathbf{u}_\alpha) + z_\alpha \nabla \left[ \nabla \cdot (h \mathbf{u}_\alpha) + \frac{h_t}{\varepsilon} \right] \right\} \\ & + \varepsilon \mu^2 \left\{ \left[ \nabla \cdot (h \mathbf{u}_\alpha) + \frac{h_t}{\varepsilon} \right] \nabla \left[ \nabla \cdot (h \mathbf{u}_\alpha) + \frac{h_t}{\varepsilon} \right] \right. \\ & \quad - \nabla \left[ \zeta \left( \nabla \cdot (h \mathbf{u}_\alpha)_t + \frac{h_{tt}}{\varepsilon} \right) \right] + (\mathbf{u}_\alpha \cdot \nabla z_\alpha) \nabla \left[ \nabla \cdot (h \mathbf{u}_\alpha) + \frac{h_t}{\varepsilon} \right] \\ & \quad + z_\alpha \nabla \left[ \mathbf{u}_\alpha \cdot \nabla \left( \nabla \cdot (h \mathbf{u}_\alpha) + \frac{h_t}{\varepsilon} \right) \right] + z_\alpha (\mathbf{u}_\alpha \cdot \nabla z_\alpha) \nabla (\nabla \cdot \mathbf{u}_\alpha) \\ & \quad \left. + \frac{1}{2} z_\alpha^2 \nabla [\mathbf{u}_\alpha \cdot \nabla (\nabla \cdot \mathbf{u}_\alpha)] \right\} \\ & + \varepsilon^2 \mu^2 \nabla \left\{ -\frac{1}{2} \zeta^2 \nabla \cdot \mathbf{u}_{\alpha t} - \zeta \mathbf{u}_\alpha \cdot \nabla \left[ \nabla \cdot (h \mathbf{u}_\alpha) + \frac{h_t}{\varepsilon} \right] + \zeta \left[ \nabla \cdot (h \mathbf{u}_\alpha) + \frac{h_t}{\varepsilon} \right] \nabla \cdot \mathbf{u}_\alpha \right\} \\ & + \varepsilon^3 \mu^2 \nabla \left\{ \frac{1}{2} \zeta^2 [(\nabla \cdot \mathbf{u}_\alpha)^2 - \mathbf{u}_\alpha \cdot \nabla (\nabla \cdot \mathbf{u}_\alpha)] \right\} = O(\mu^4). \end{aligned} \tag{3.3}$$

# Boussinesq-type Model

(Lynett & Liu, 2002)

$$\frac{1}{\varepsilon} h_t + \zeta_t + \nabla \cdot (H \mathbf{u}_\alpha) - \mu^2 \nabla \cdot \left\{ H \left[ \left( \frac{1}{6} (\varepsilon^2 \zeta^2 - \varepsilon \zeta h + h^2) - \frac{1}{2} z_\alpha^2 \right) \nabla (\nabla \cdot \mathbf{u}_\alpha) + \left( \frac{1}{2} (\varepsilon \zeta - h) - z_\alpha \right) \nabla \left( \nabla \cdot (h \mathbf{u}_\alpha) + \frac{h_t}{\varepsilon} \right) \right] \right\} = O(\mu^4)$$

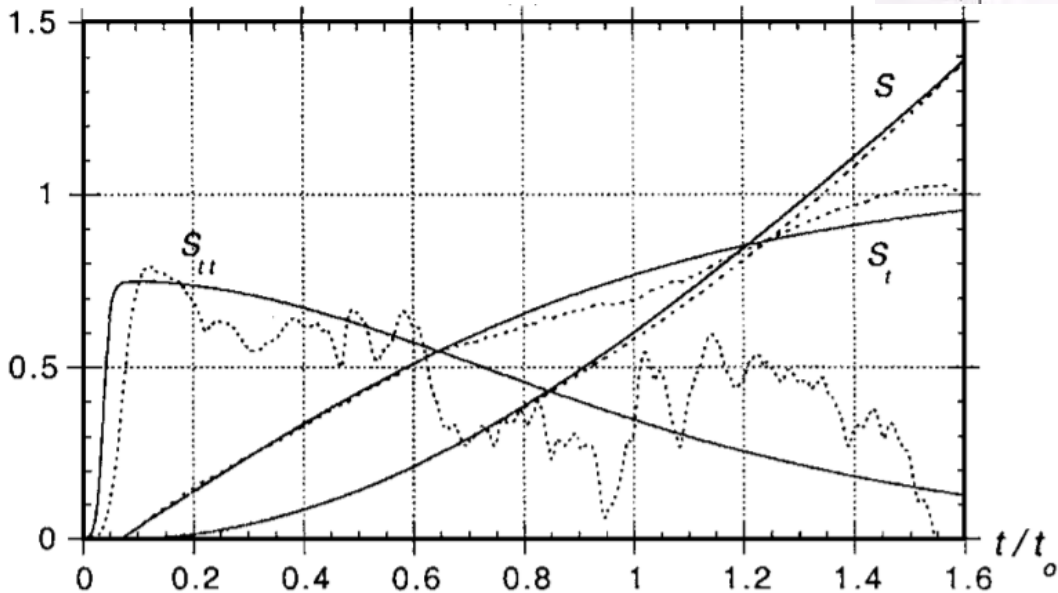
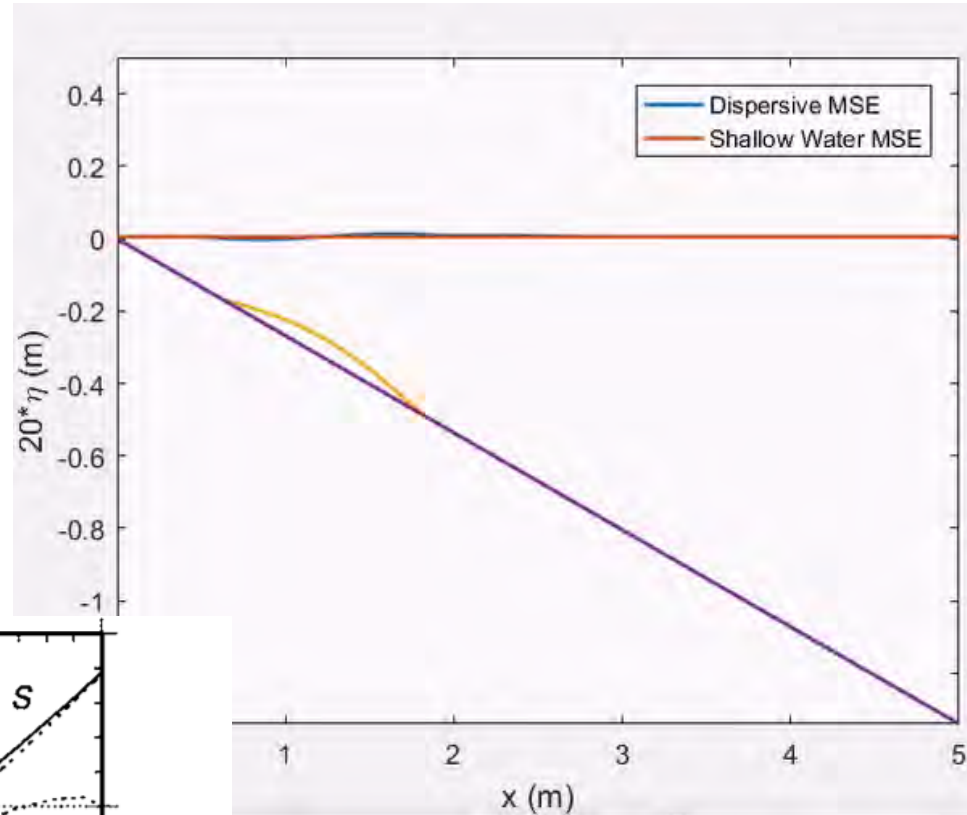
$$\begin{aligned} & \mathbf{u}_{\alpha t} + \varepsilon \mathbf{u}_\alpha \cdot \nabla \mathbf{u}_\alpha + \nabla \zeta \\ & + \mu^2 \frac{\partial}{\partial t} \left\{ \frac{1}{2} z_\alpha^2 \nabla (\nabla \cdot \mathbf{u}_\alpha) + z_\alpha \nabla \left[ \nabla \cdot (h \mathbf{u}_\alpha) + \frac{h_t}{\varepsilon} \right] \right\} \\ & + \varepsilon \mu^2 \left\{ \left[ \nabla \cdot (h \mathbf{u}_\alpha) + \frac{h_t}{\varepsilon} \right] \nabla \left[ \nabla \cdot (h \mathbf{u}_\alpha) + \frac{h_t}{\varepsilon} \right] \right. \\ & \quad - \nabla \left[ \zeta \left( \nabla \cdot (h \mathbf{u}_\alpha)_t + \frac{h_{tt}}{\varepsilon} \right) \right] + (\mathbf{u}_\alpha \cdot \nabla z_\alpha) \nabla \left[ \nabla \cdot (h \mathbf{u}_\alpha) + \frac{h_t}{\varepsilon} \right] \\ & \quad \left. + z_\alpha \nabla \left[ \mathbf{u}_\alpha \cdot \nabla \left( \nabla \cdot (h \mathbf{u}_\alpha) + \frac{h_t}{\varepsilon} \right) \right] + z_\alpha (\mathbf{u}_\alpha \cdot \nabla z_\alpha) \nabla (\nabla \cdot \mathbf{u}_\alpha) \right. \\ & \quad \left. + \frac{1}{2} z_\alpha^2 \nabla [\mathbf{u}_\alpha \cdot \nabla (\nabla \cdot \mathbf{u}_\alpha)] \right\} \\ & + \varepsilon^2 \mu^2 \nabla \left\{ -\frac{1}{2} \zeta^2 \nabla \cdot \mathbf{u}_{\alpha t} - \zeta \mathbf{u}_\alpha \cdot \nabla \left[ \nabla \cdot (h \mathbf{u}_\alpha) + \frac{h_t}{\varepsilon} \right] + \zeta \left[ \nabla \cdot (h \mathbf{u}_\alpha) + \frac{h_t}{\varepsilon} \right] \nabla \cdot \mathbf{u}_\alpha \right\} \\ & + \varepsilon^3 \mu^2 \nabla \left\{ \frac{1}{2} \zeta^2 [(\nabla \cdot \mathbf{u}_\alpha)^2 - \mathbf{u}_\alpha \cdot \nabla (\nabla \cdot \mathbf{u}_\alpha)] \right\} = O(\mu^4). \end{aligned} \tag{3.3}$$

Linear and nonlinear terms are  $h_t$ ,  $h_{xxt}$ ,  $h_{xxt}$

Depth evolution must be continuous in time and space

# Benchmark #1

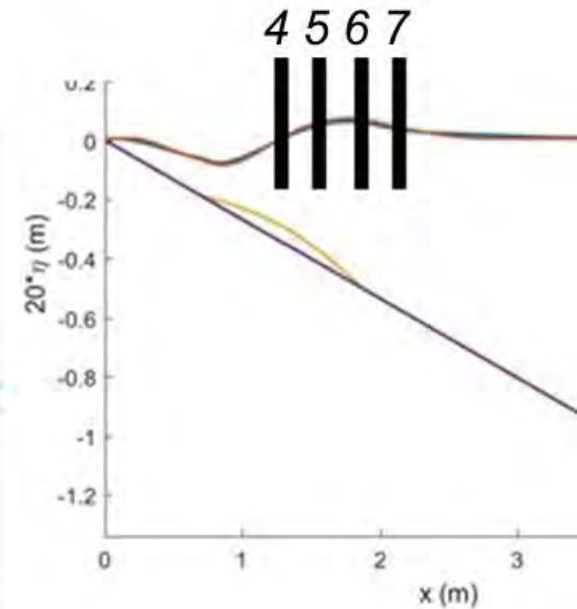
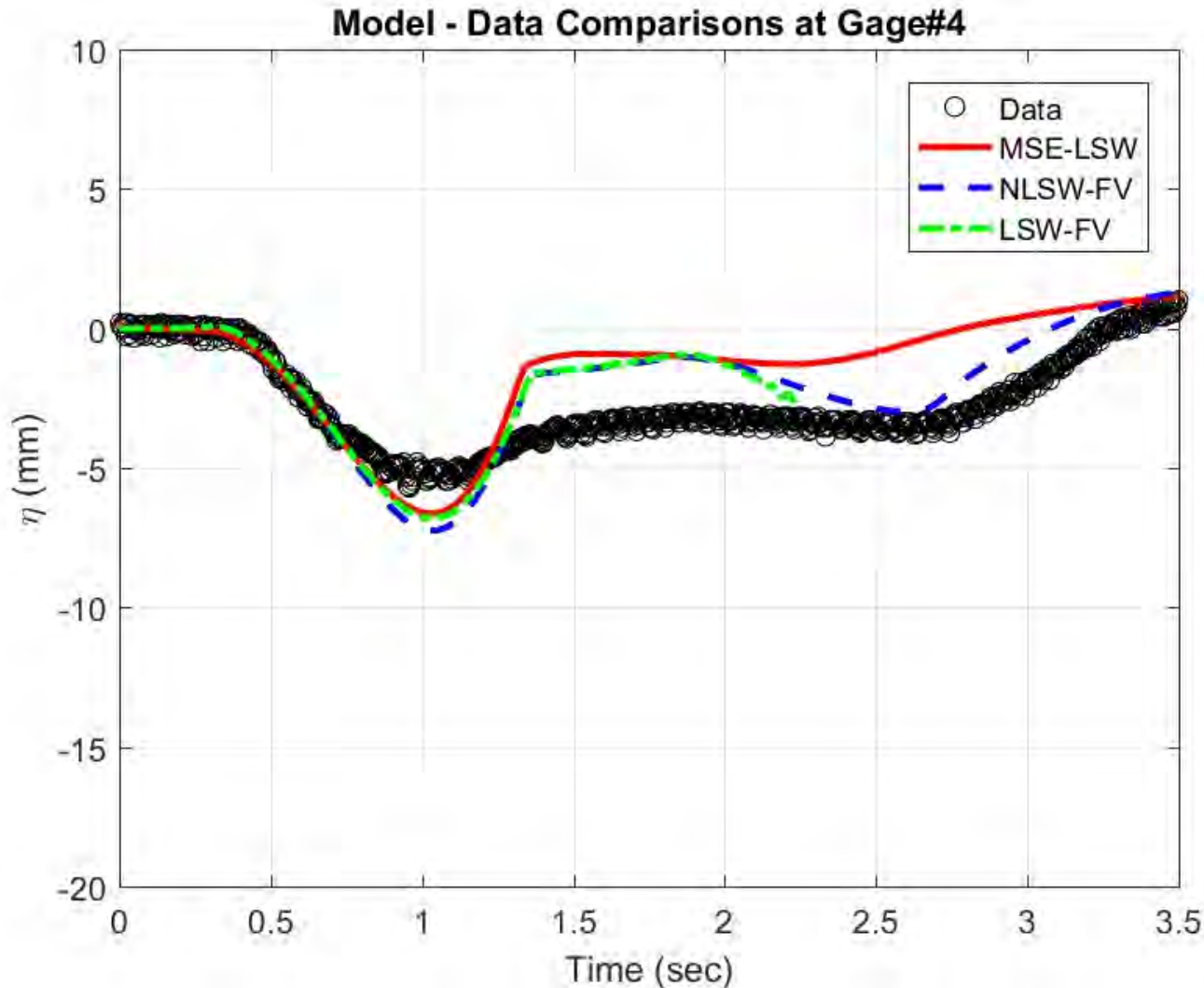
- Used the prescribed slide motion
  - Due to the max function, derivatives are not continuous in (x) – significance of slide edge effects are grid dependent
  - Did not smooth the slide shape in any way
  - Did not use the initial time acceleration correction





# Benchmark #1

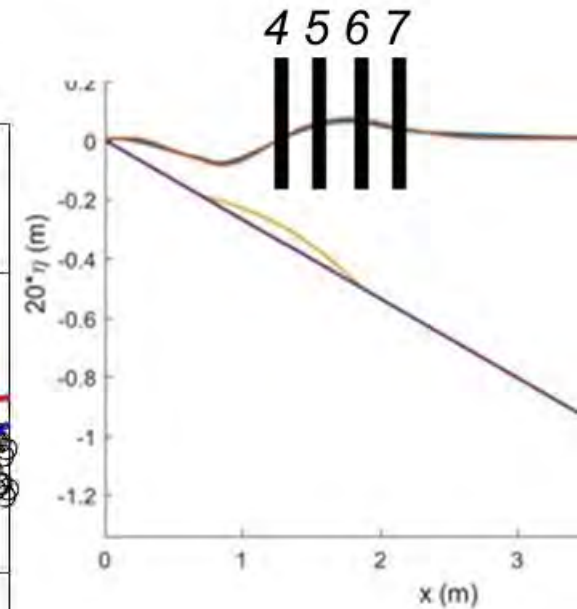
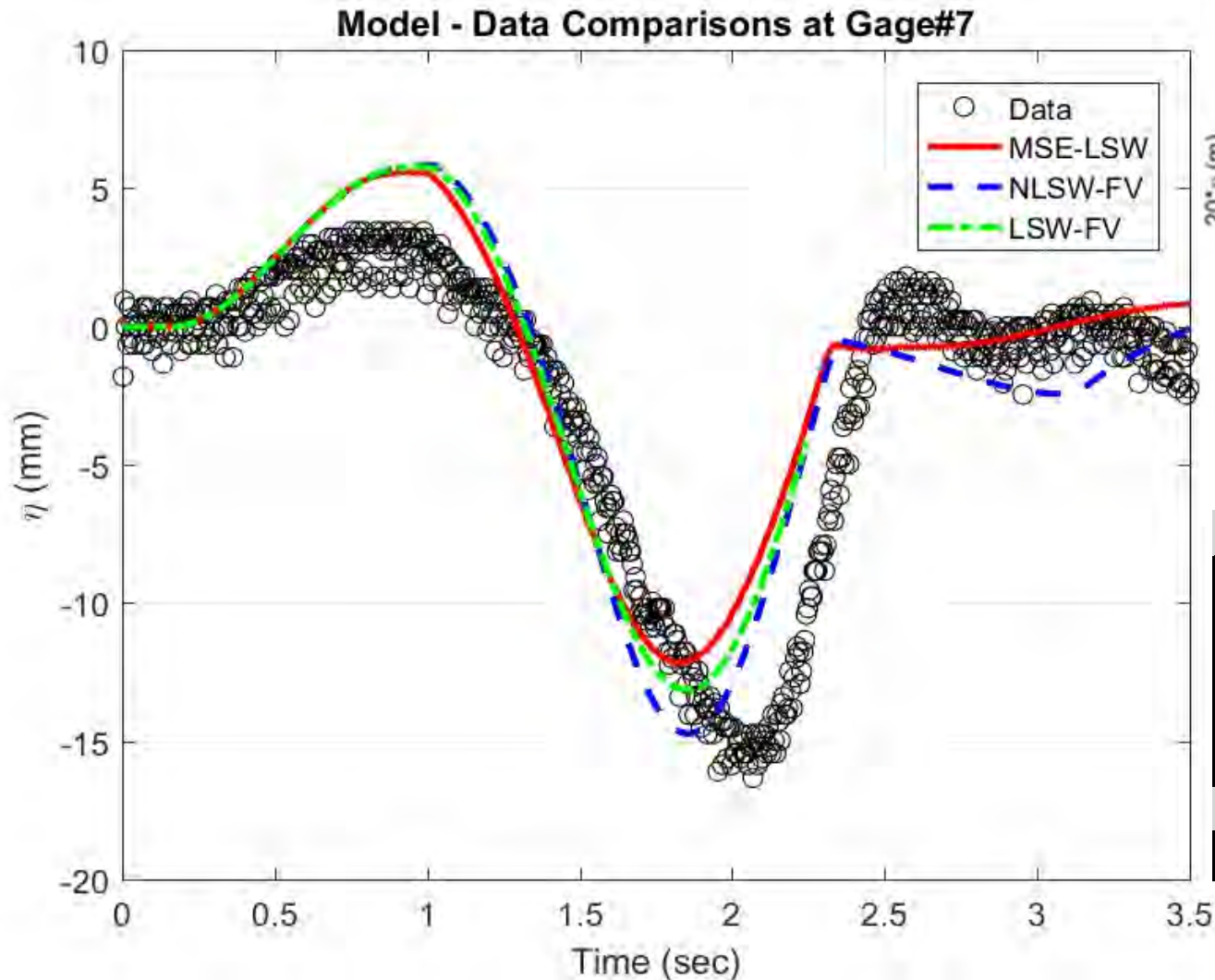
- First, do the different models produce the same results for the same setups?



depth=	0.33 m	
Period (s)	kh	c/sqrt(gh)
2	0.59	0.98
1	1.42	0.81
0.5	5.31	0.43
0.25	21.25	0.22
min resolvable kh=		103.62

# Benchmark #1

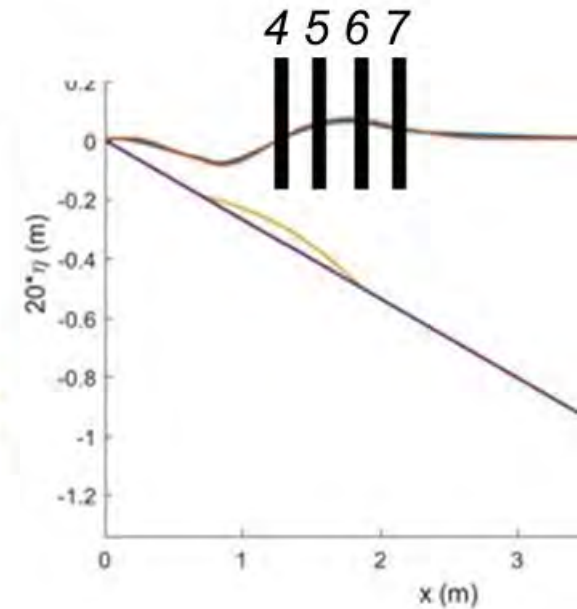
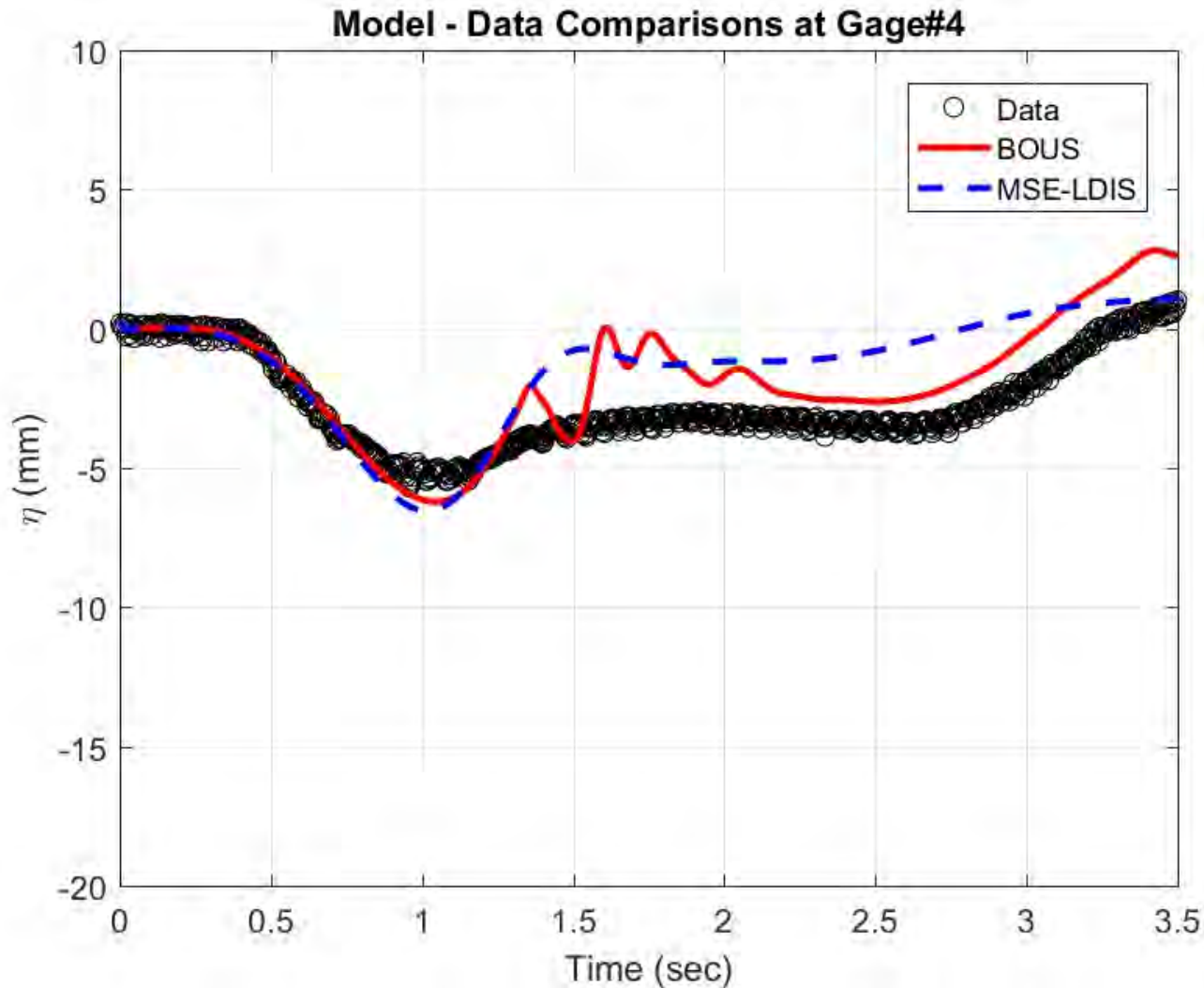
- First, do the different models produce the same results for the same setups?



depth=	0.59 m	
Period (s)	kh	c/sqrt(gh)
2	0.81	0.95
1	2.39	0.64
0.5	9.50	0.32
0.25	37.99	0.16
min resolvable kh=		185.26

# Benchmark #1

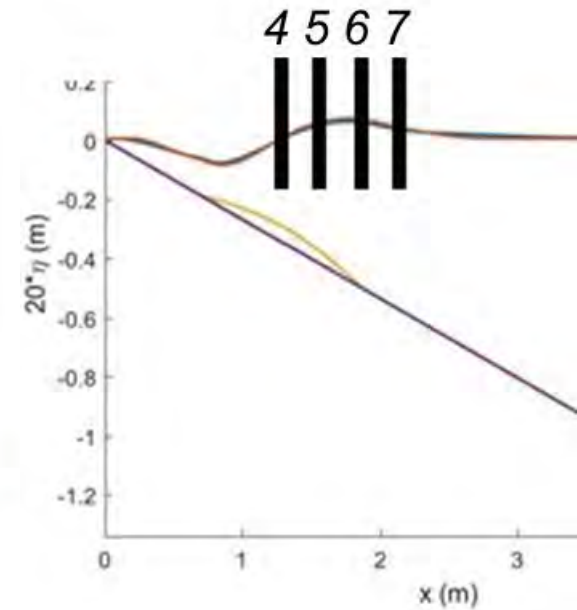
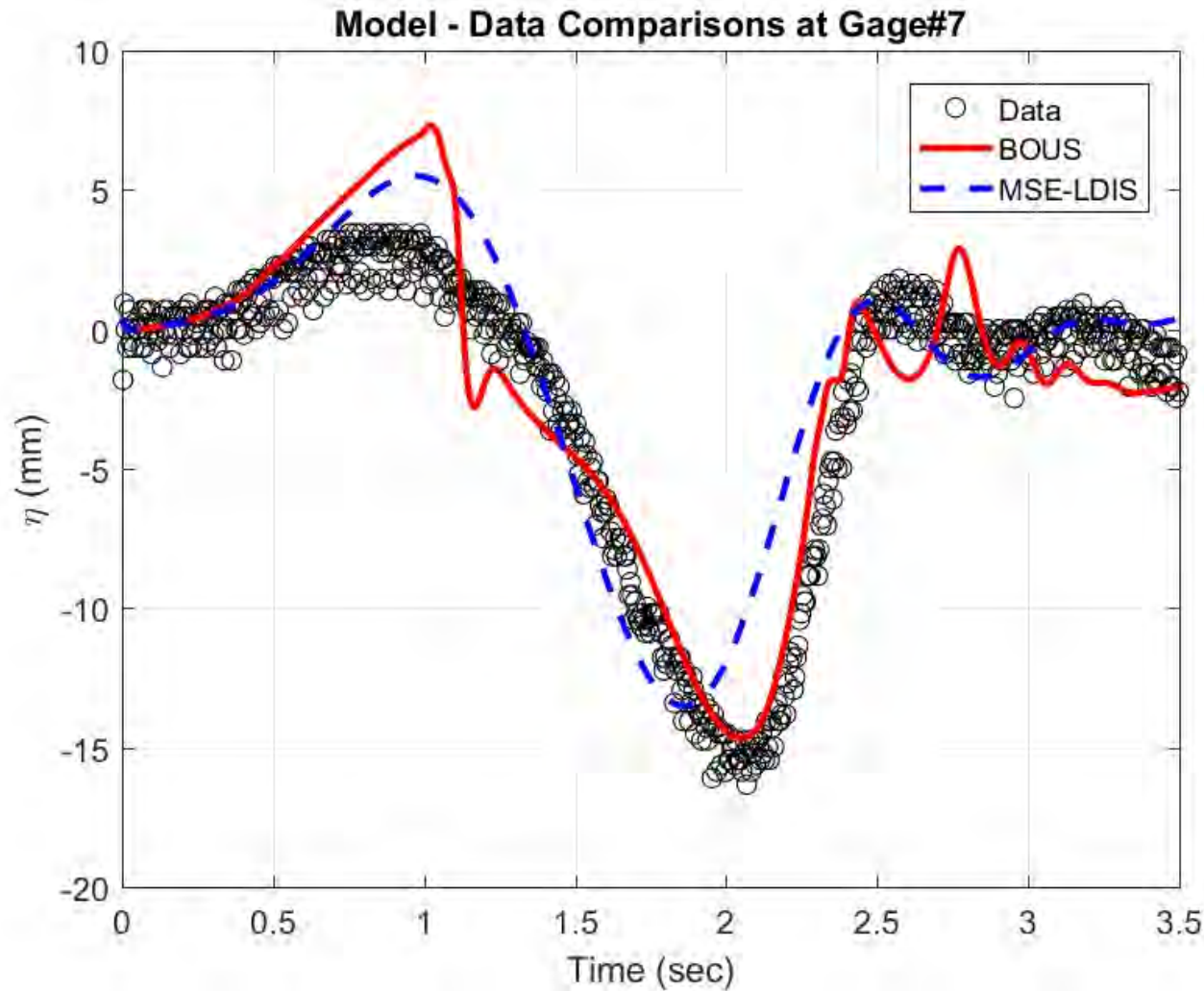
- Bous vs MSE



depth=	0.33 m	
Period (s)	kh	c/sqrt(gh)
2	0.59	0.98
1	1.42	0.81
0.5	5.31	0.43
0.25	21.25	0.22
min resolvable kh=		103.62

# Benchmark #1

- Bous vs MSE

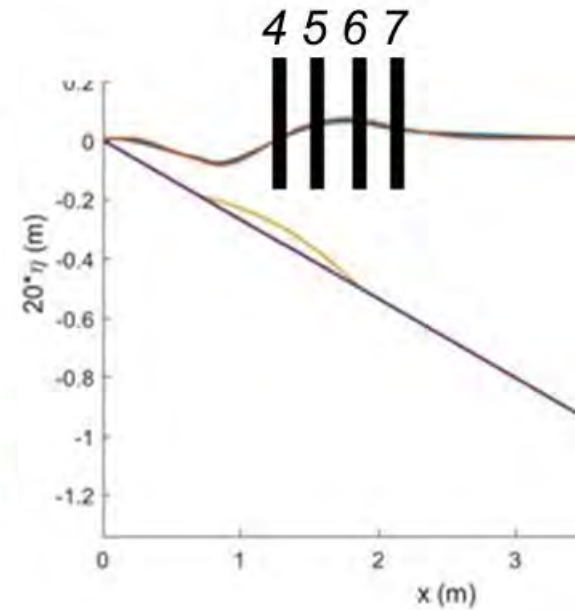
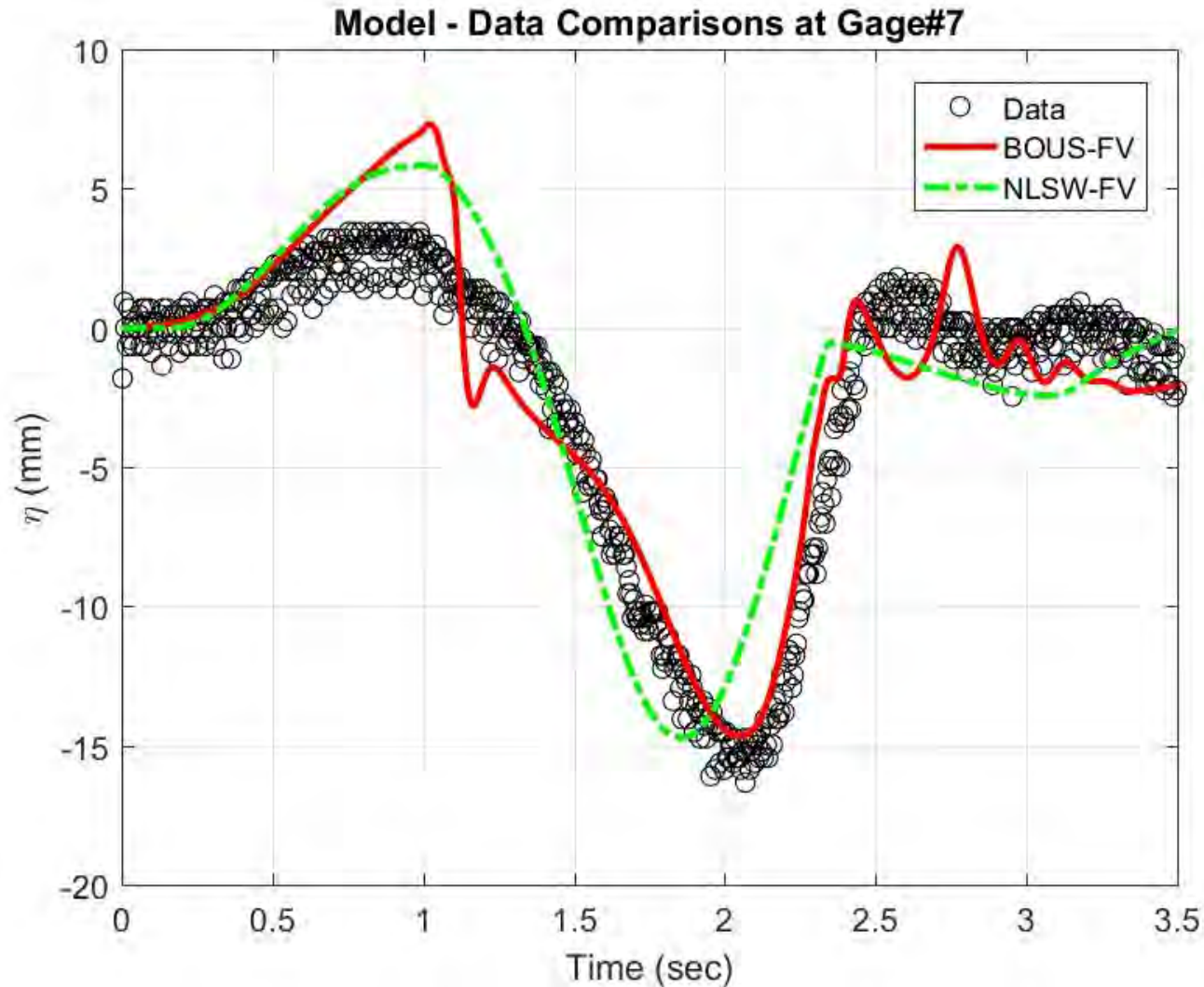


depth=	0.59 m	
Period (s)	kh	c/sqrt(gh)
2	0.81	0.95
1	2.39	0.64
0.5	9.50	0.32
0.25	37.99	0.16
min resolvable kh=		185.26



# Benchmark #1

- Bous vs NLSW – NLSW less bad?



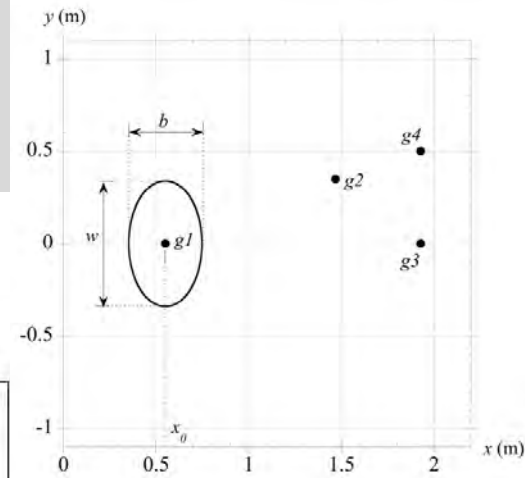
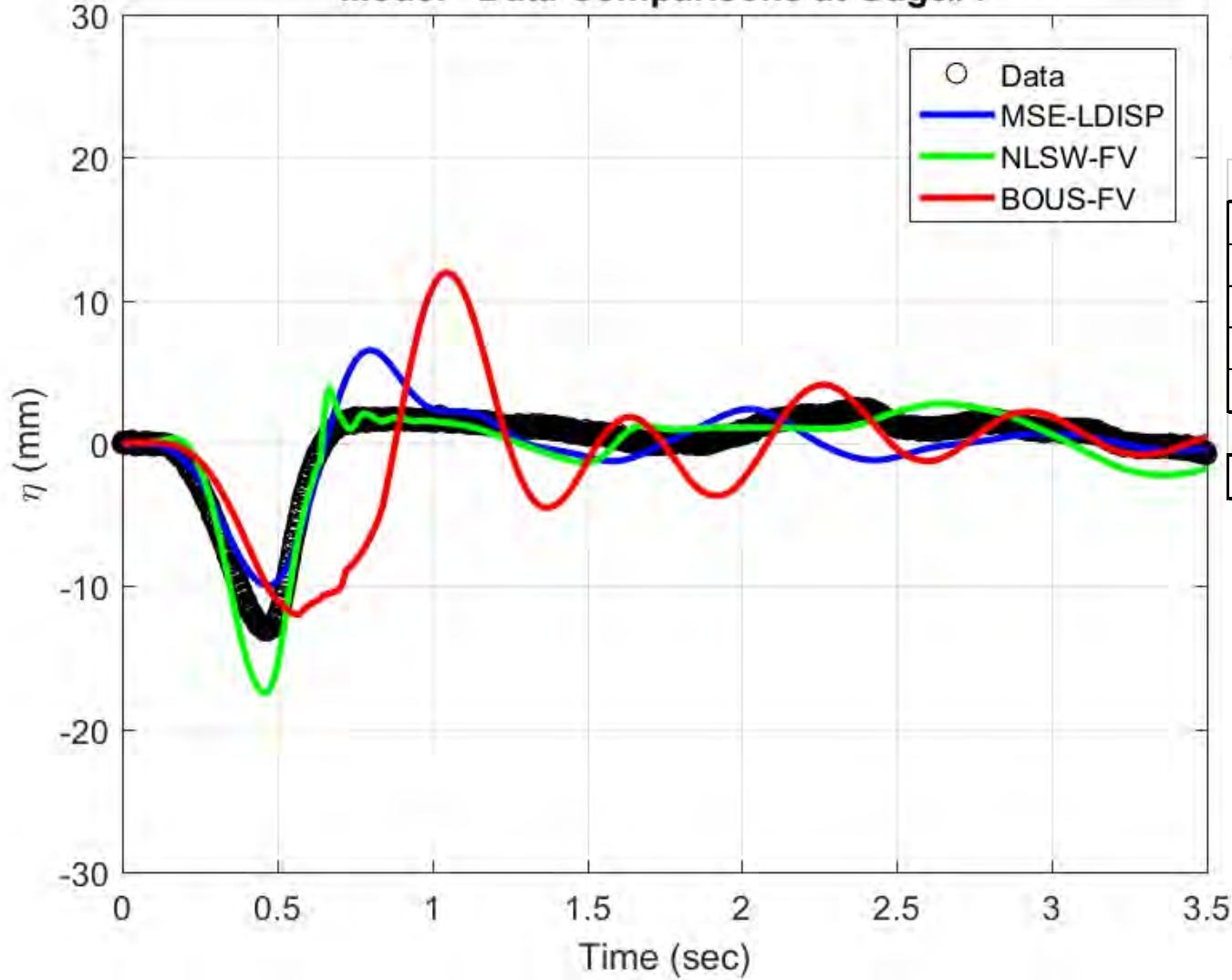
depth=	0.59 m	
Period (s)	kh	c/sqrt(gh)
2	0.81	0.95
1	2.39	0.64
0.5	9.50	0.32
0.25	37.99	0.16
min resolvable kh=		185.26



# Benchmark #2

- Initial depth = 0.061m

Model - Data Comparisons at Gage#1

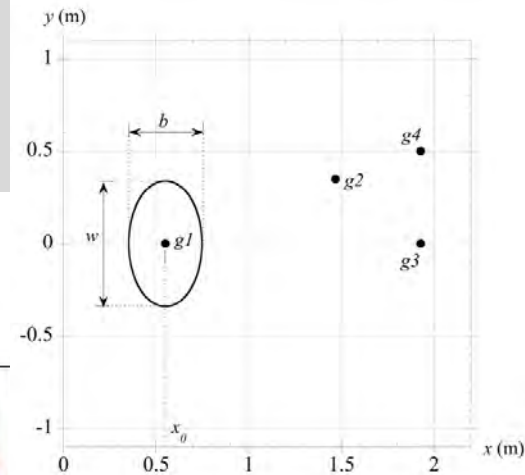
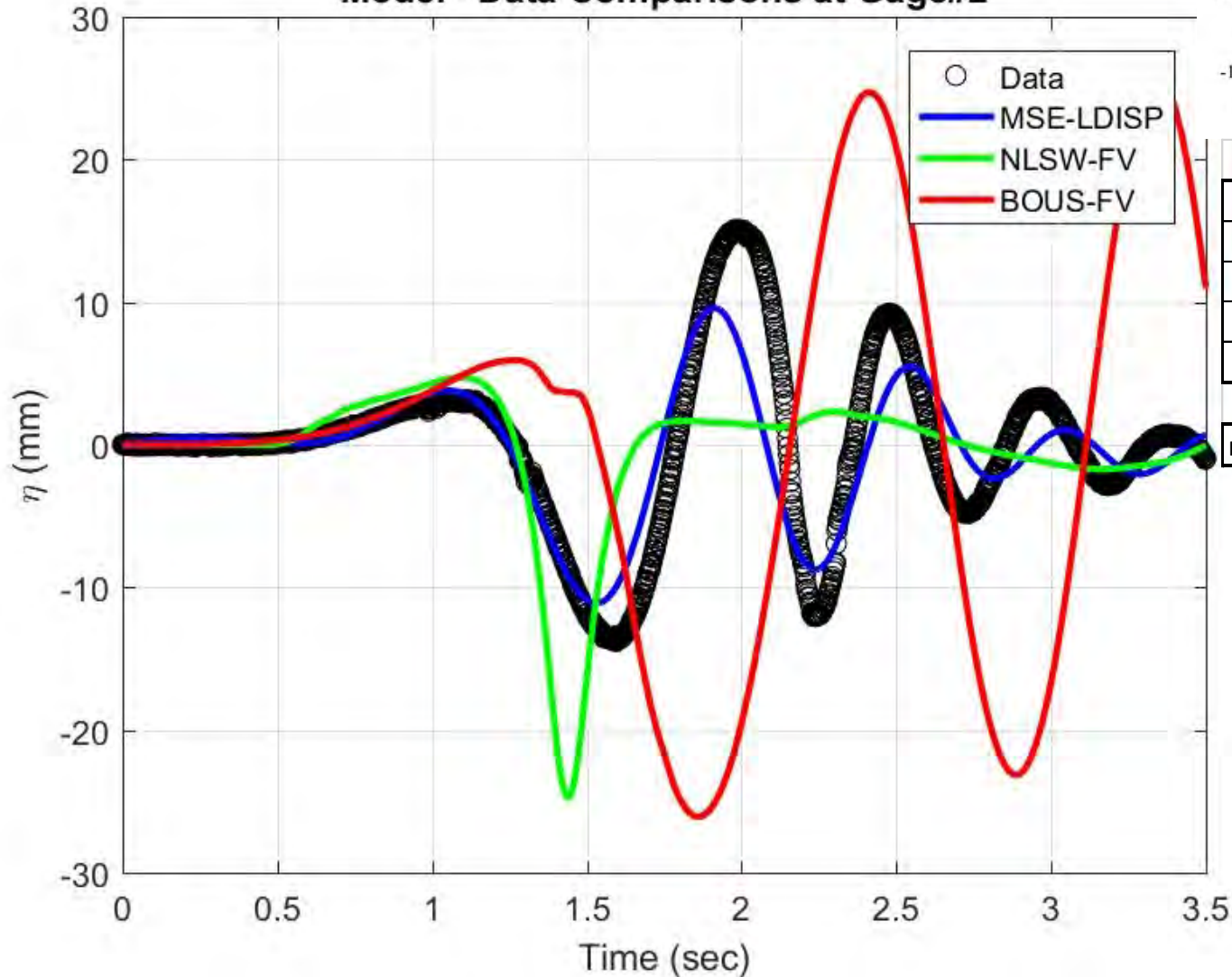


depth=	0.14 m	
Period (s)	kh	c/sqrt(gh)
2	0.38	1.00
1	0.79	0.95
0.5	2.28	0.66
0.25	9.01	0.33
min resolvable kh=		13.19

# Benchmark #2

- Initial depth = 0.061m

Model - Data Comparisons at Gage#2

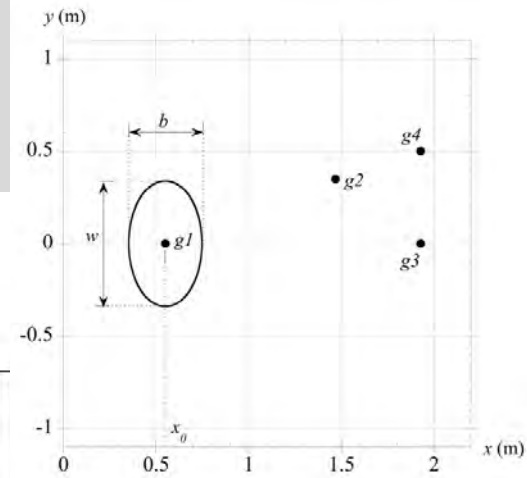
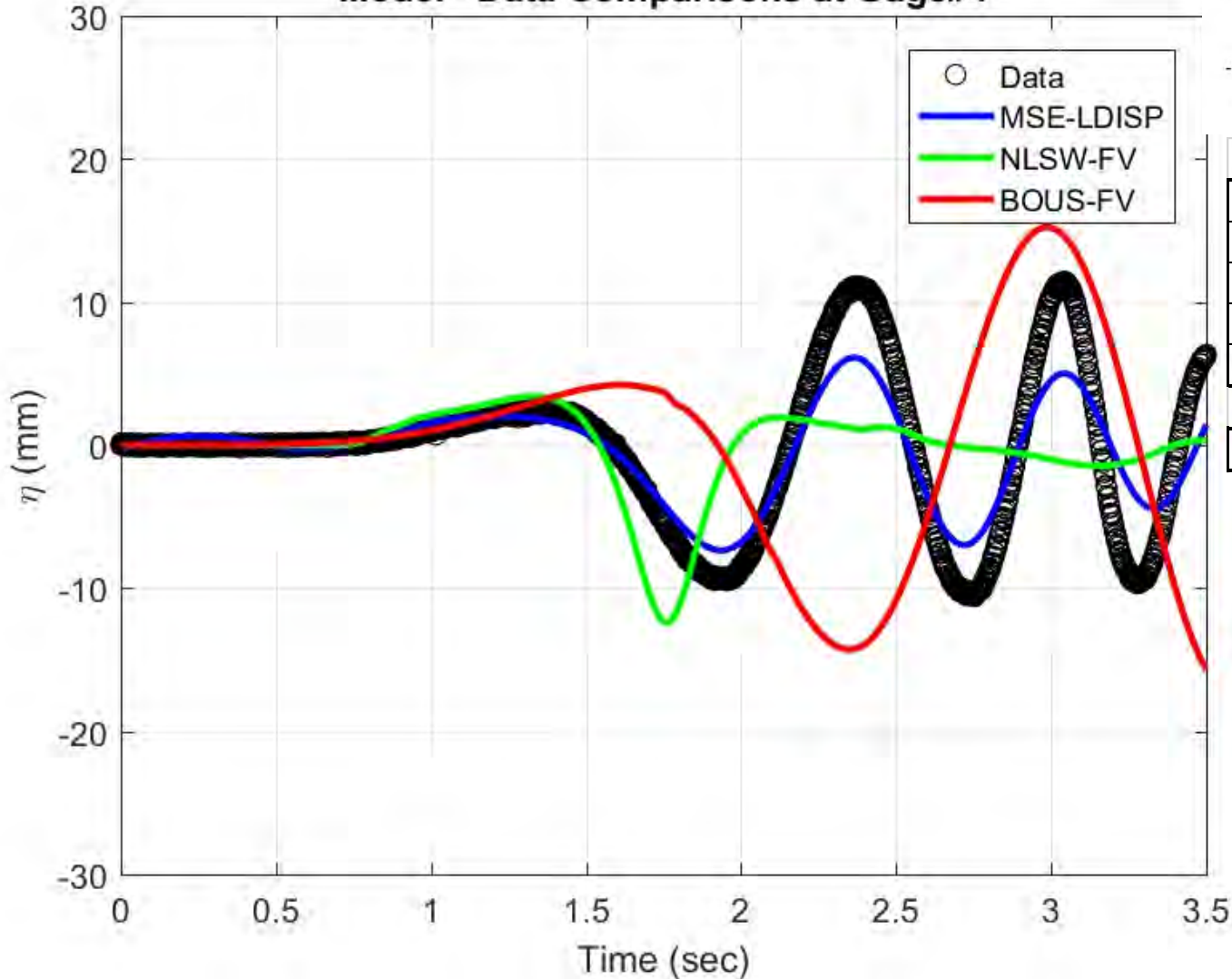


depth=	0.39 m	
Period (s)	kh	c/sqrt(gh)
2	0.64	0.98
1	1.64	0.76
0.5	6.28	0.40
0.25	25.11	0.20
min resolvable kh=		36.74

# Benchmark #2

- Initial depth = 0.061m

Model - Data Comparisons at Gage#4

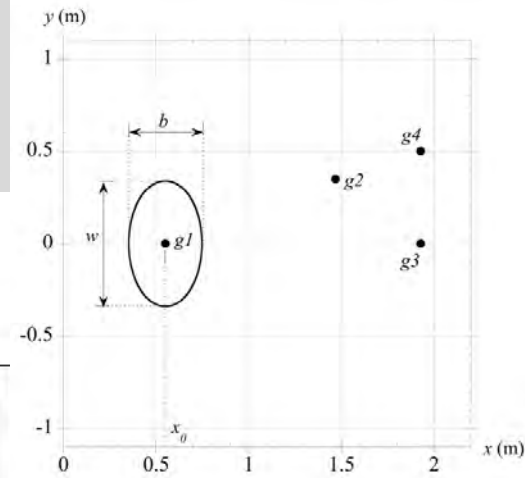
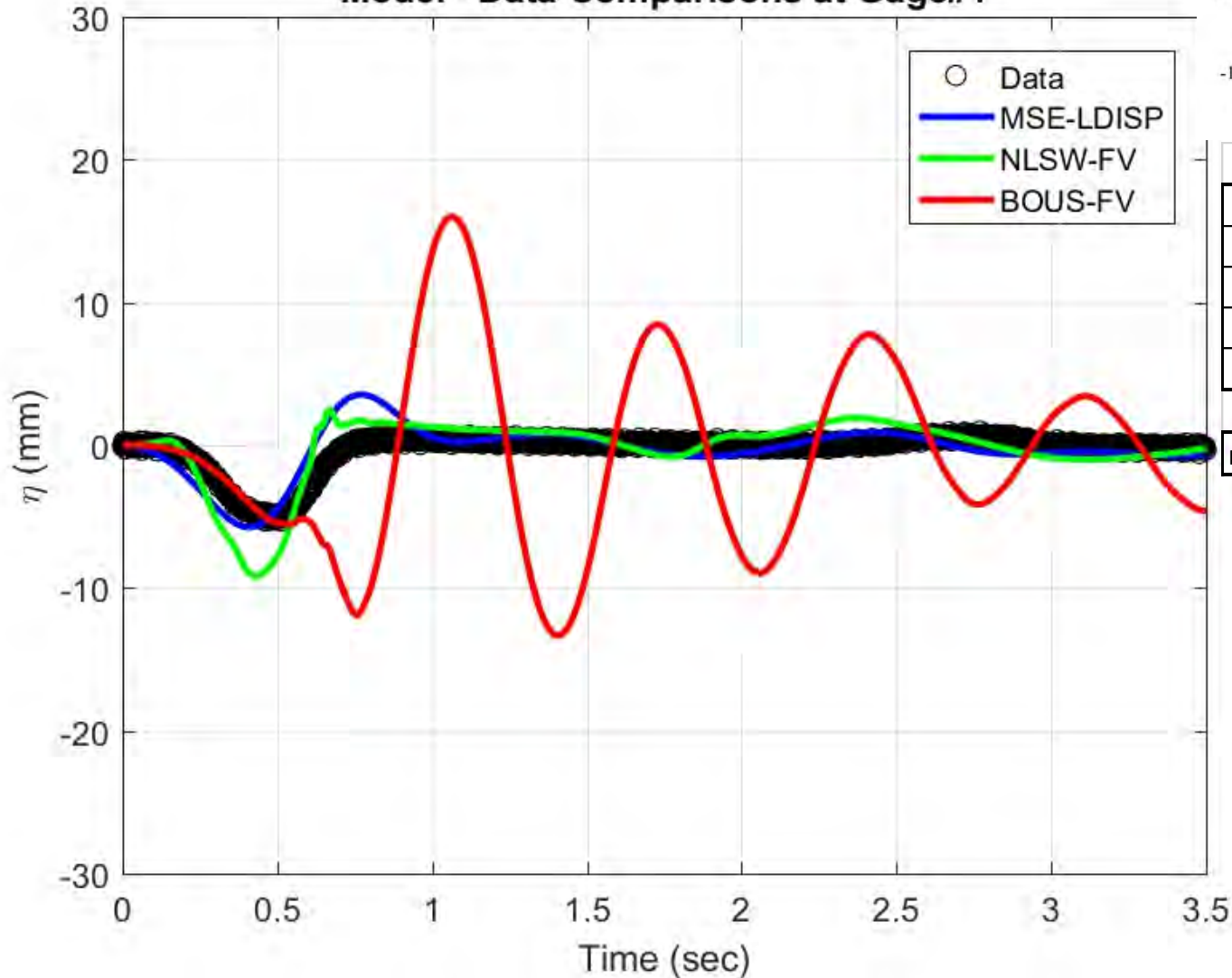


depth=	0.52 m	
Period (s)	$kh$	$c/\text{sqrt}(gh)$
2	0.75	0.96
1	2.12	0.68
0.5	8.37	0.35
0.25	33.48	0.17
min resolvable $kh$ =		48.98

# Benchmark #2

- Initial depth = 0.120m

Model - Data Comparisons at Gage#1

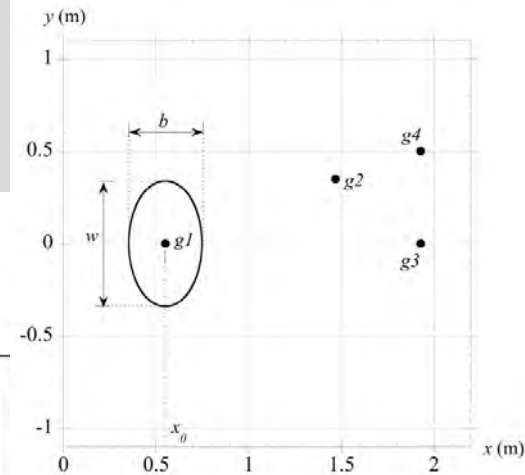
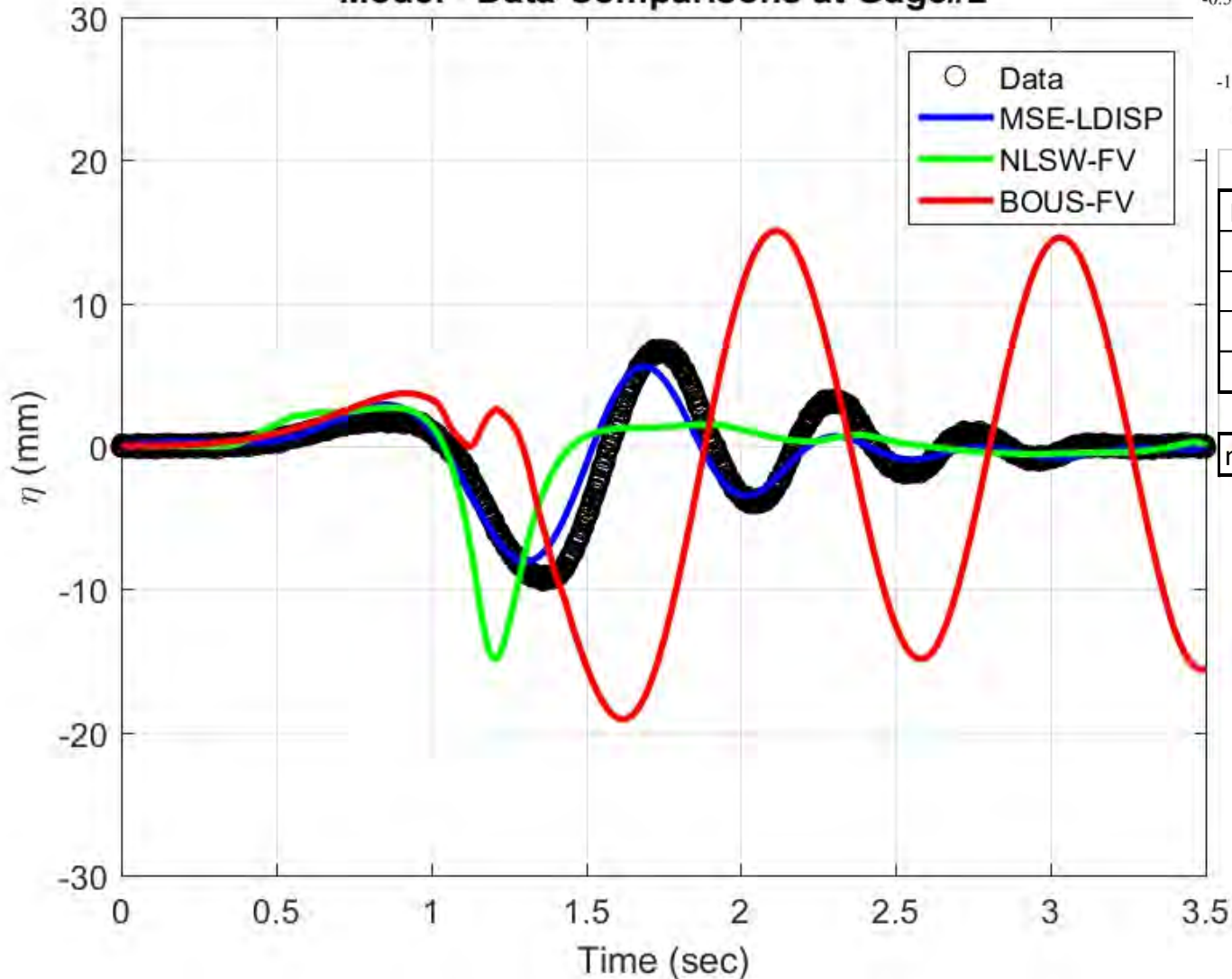


depth=	0.2 m	
Period (s)	kh	c/sqrt(gh)
2	0.45	0.99
1	0.99	0.91
0.5	3.22	0.56
0.25	12.88	0.28
min resolvable kh=		18.84

# Benchmark #2

- Initial depth = 0.120m

Model - Data Comparisons at Gage#2



depth=	0.39 m	
Period (s)	kh	c/sqrt(gh)
2	0.64	0.98
1	1.64	0.76
0.5	6.28	0.40
0.25	25.11	0.20
min resolvable kh=		36.74

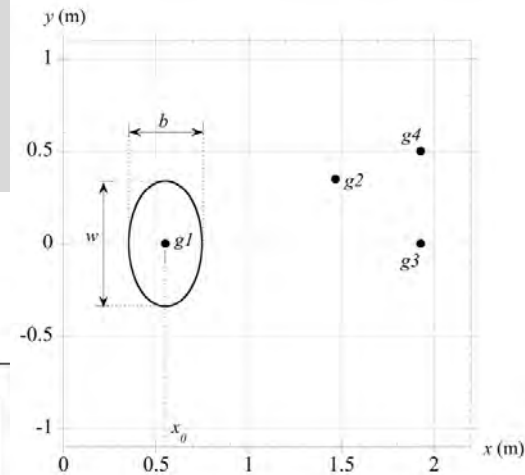
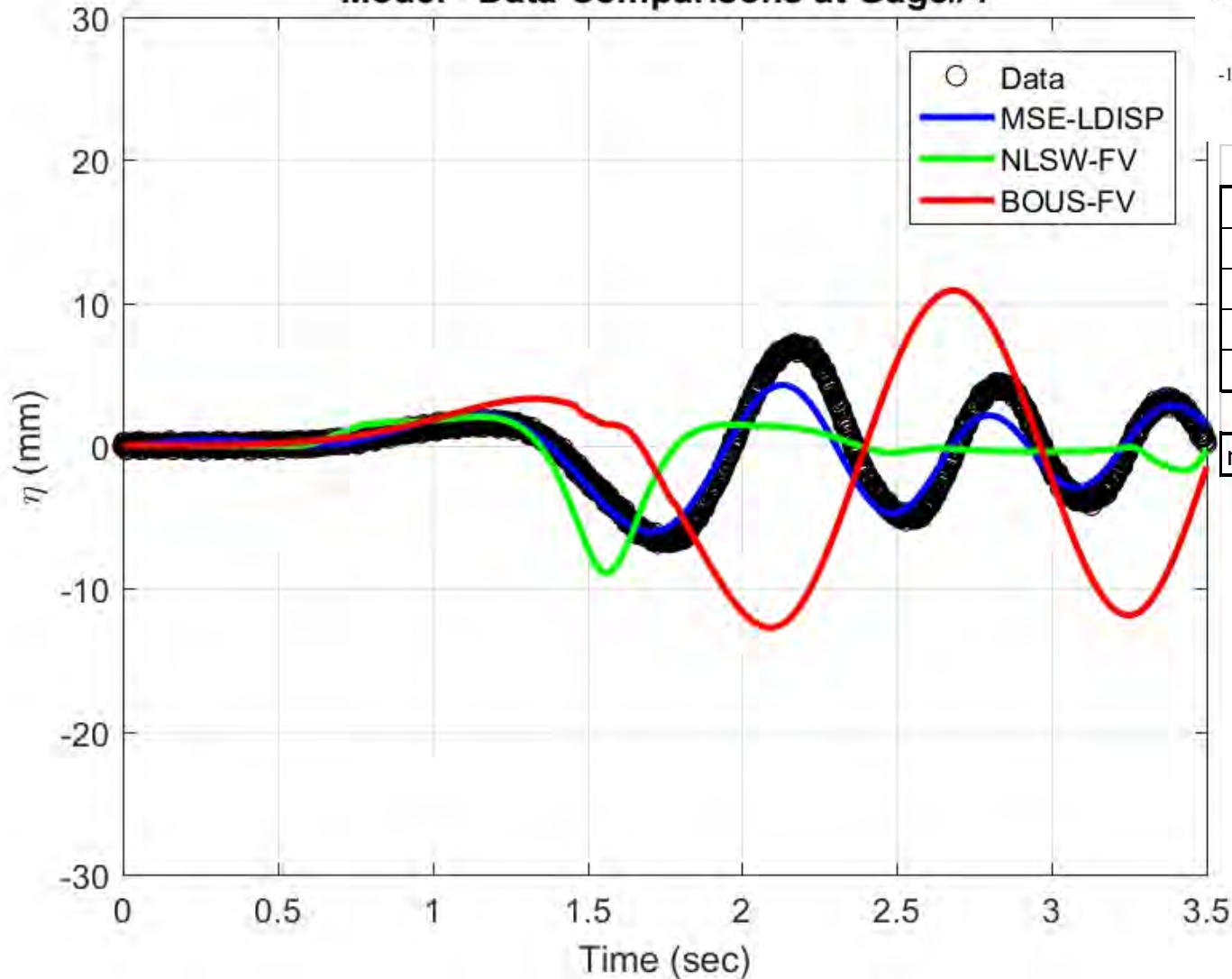
MSE better  
for this  
depth  
Nonlinearity  
less  
important



# Benchmark #2

- Initial depth = 0.120m

Model - Data Comparisons at Gage#4



depth=	0.52 m	
Period (s)	kh	c/sqrt(gh)
2	0.75	0.96
1	2.12	0.68
0.5	8.37	0.35
0.25	33.48	0.17
min resolvable kh=		48.98

NLSW better than Bous for leading wave, but  $kh \sim 3!$

# Conclusions & Thoughts

- NLSW can be “less wrong” than weakly dispersive models when generated wavenumbers exceed accuracy limitations of the weakly dispersive models
    - But hard to reconcile using NLSW for high  $kh$  forcing...
    - Finn’s filter is probably a required approach for a general application of weakly dispersive models for arbitrary bottom forcing
  - The Giorgio model (Mild Slope Equation) offers a rapid approach to estimate generated waves with arbitrary (single-valued in the horizontal) landslide shape
    - Linear
    - Needs coupling to another model for propagation away from source, viscous effects, and for runup
- 
- To what degree should we allow modelers to smooth / modify slide evolution to permit a stable / accurate result?
  - Are we benchmarking the slide evolution or the wave generation?
    - IF we are benchmarking slide motion, then we need to use slide motion benchmarks (BM6!)
    - IF we are benchmarking wave generation, we need to be more restrictive on the slide motion
  - Slides stop too! – do we need a slump-like benchmark, with a coherent de-acceleration?
  - Thinking of landslide tsunami forecast (NOT hindcast) – if you had just a landslide location, approximate mass (within 20%), approximate direction of failure (with 20%), and approximate time scale (within 50%) [this is the information we might get in near realtime from seismic inversion] – how well could we forecast the waves?
    - Where is the uncertainty, or the knowledge gaps – hydro, geo, coupling?