



## ESG4 Conference @ UCSB

4th IASPEI / IAEE International Symposium:  
Effects of Surface Geology on Seismic Motion

### ENHANCING SITE RESPONSE MODELING THROUGH DOWNHOLE ARRAY RECORDINGS

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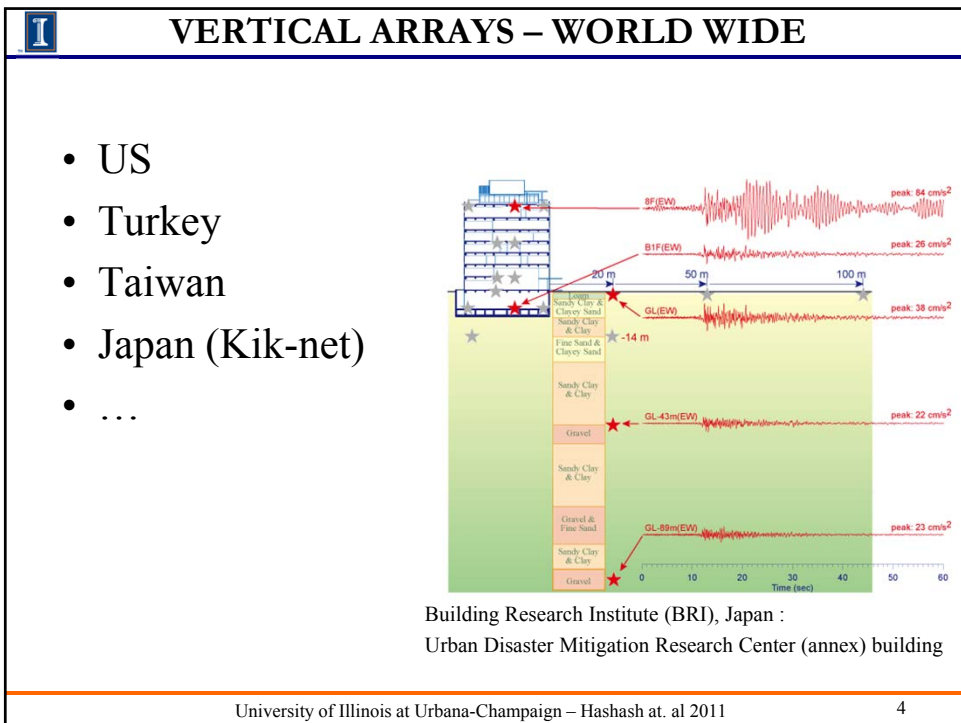
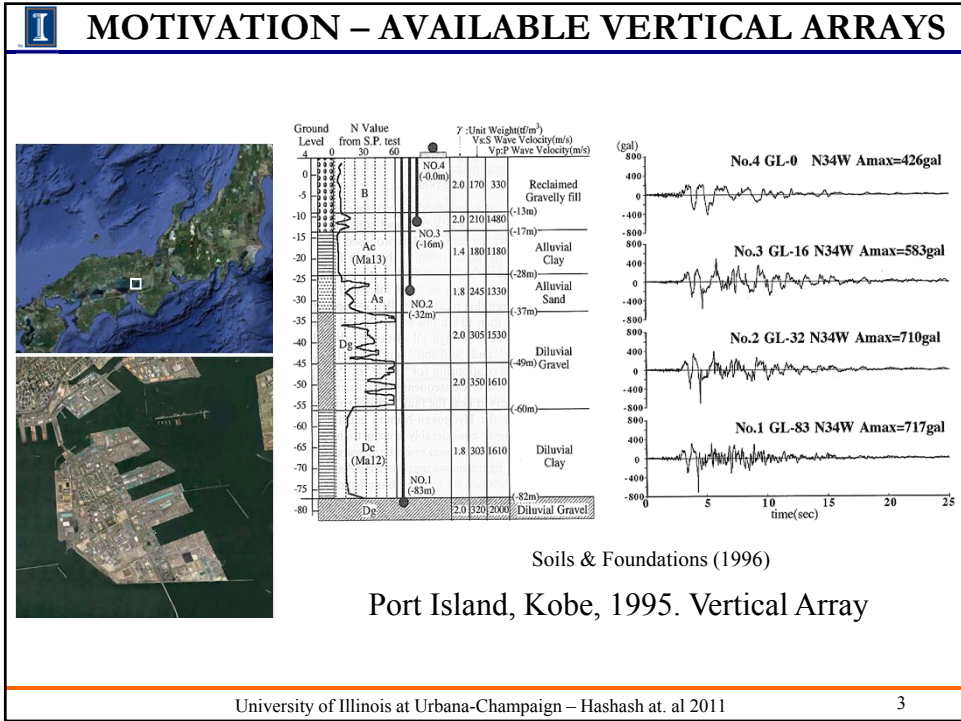


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## OUTLINE

- Motivation and background
- SelfSim inverse analysis framework development with porewater pressure development
- Learning soil behavior from synthetically generated downhole arrays
- Learning soil behavior from field array measurements – Wildlife array, CA
- Simulation of downhole arrays from 2011 Tohoku, Japan, earthquake.
- Concluding remarks



## I MOTIVATION

...it is an inverse problem

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## I BACKGROUND – INVERSE PROBLEMS

- Output is known (surface)
- Input is unknown
- System is unknown

- Output is known (surface)
- Input is known
- System is unknown

-Deconvolution  
-Challenging  
-Uniqueness? Need many records

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**I** **INVERSE ANALYSIS - OPTIMIZATION**

The diagram illustrates the process of inverse analysis through optimization. It features a central blue box labeled 'CONSTITUTIVE MODEL'. To its left is a green box labeled 'STRAIN', and to its right is another green box labeled 'STRESS'. Arrows point from 'STRAIN' to the 'CONSTITUTIVE MODEL' and from the 'CONSTITUTIVE MODEL' to 'STRESS'. Below the 'CONSTITUTIVE MODEL' box, three arrows point upwards towards it, labeled 'Nonlinearity', 'Hardening & Softening', and 'Anisotropy'.

- Requires pre-defined constitutive model
- Difficult to develop & to refine for new loading conditions

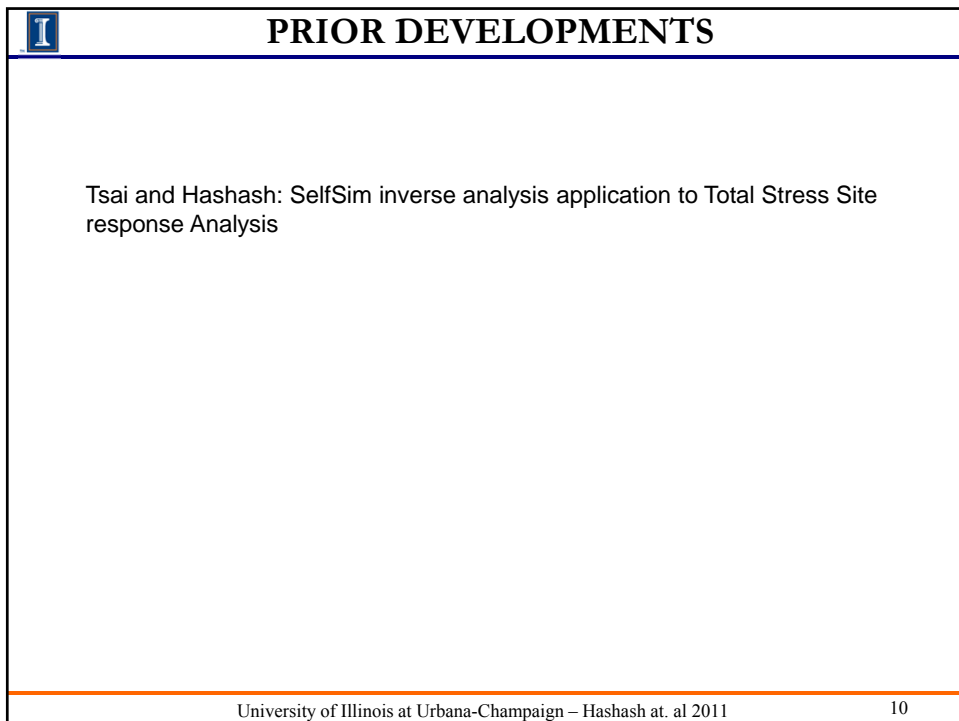
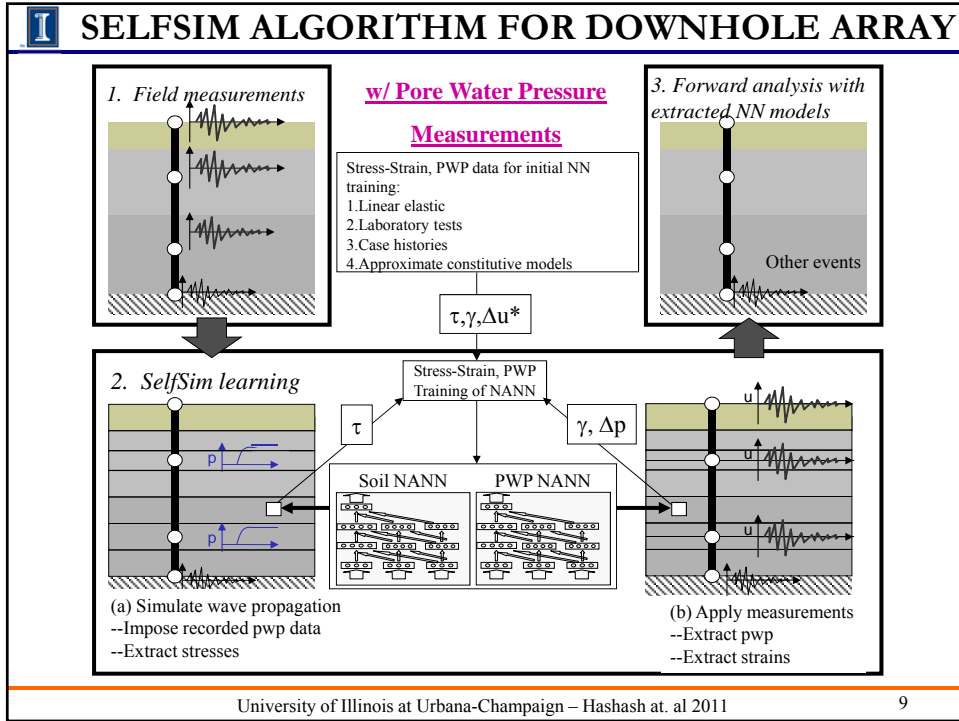
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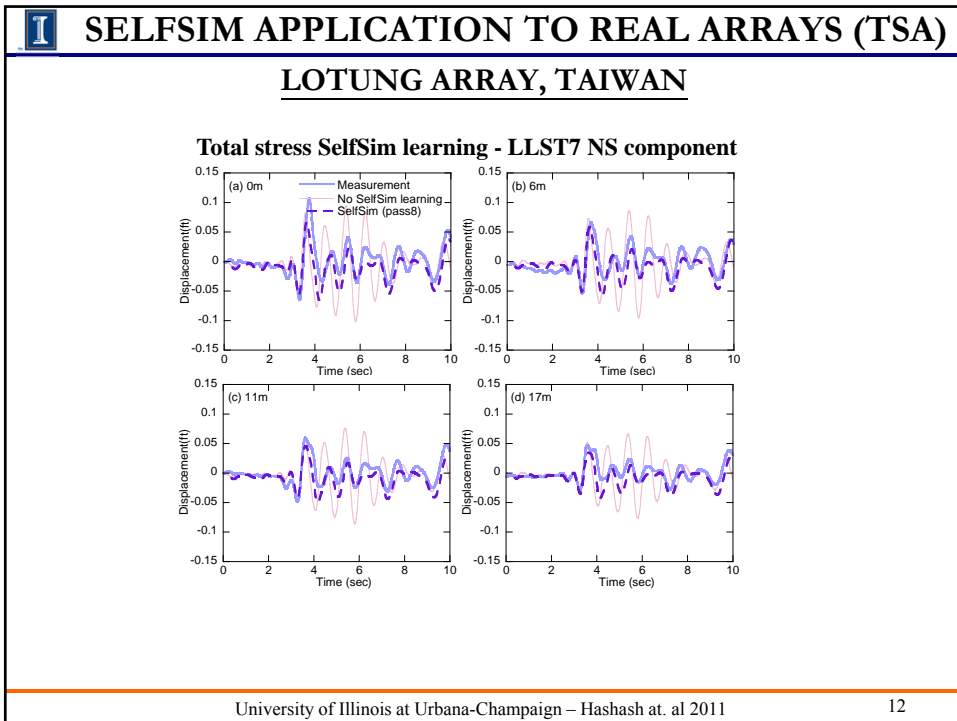
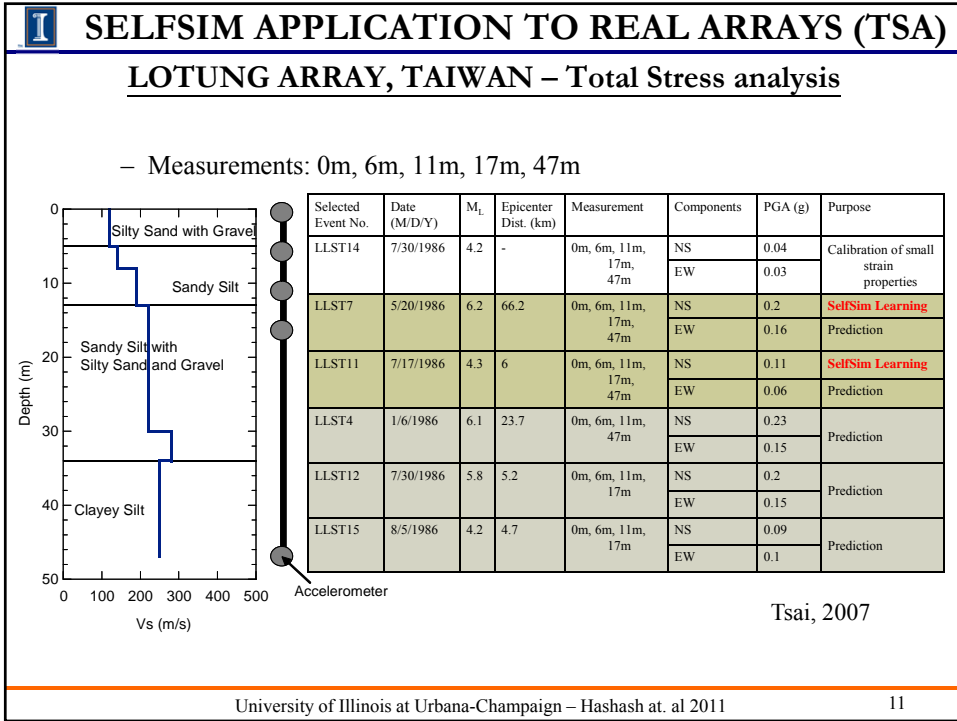
**I** **INVERSE ANALYSIS – SELFSIM LEARNING**

The diagram illustrates self-sim learning for inverse analysis. It features a central blue box labeled 'Self Learning Simulations' which contains a neural network diagram with three layers of nodes. To its left is a green box labeled 'STRAIN', and to its right is another green box labeled 'STRESS'. Arrows point from 'STRAIN' to the 'Self Learning Simulations' box and from the 'Self Learning Simulations' box to 'STRESS'.

- Continuously evolving soil model (Artificial Neural Network, NN, Material Model)
- Material behavior in the form of a stress-strain-porewater pressure database
- Learn new behavior unconstrained by prior assumptions of soil behavior

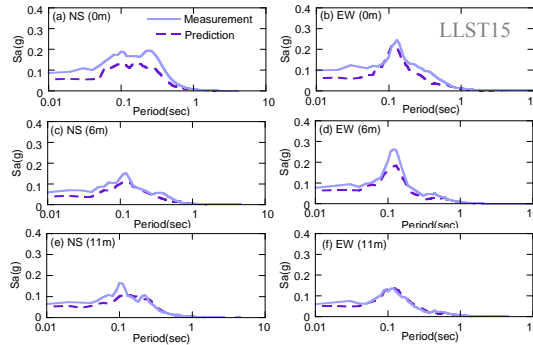
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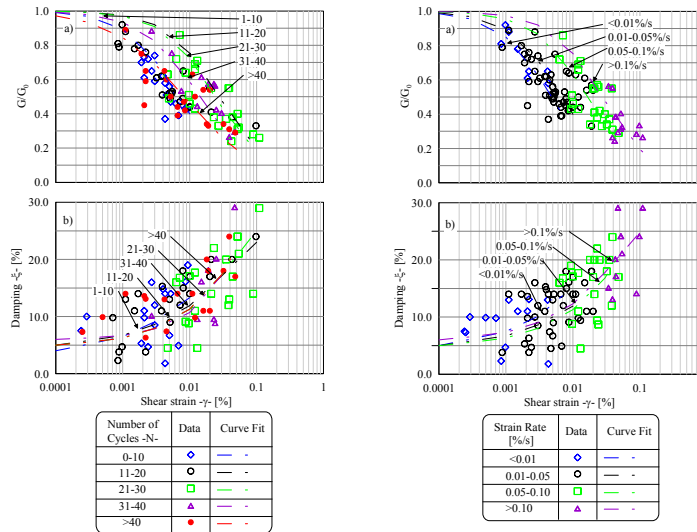
# I PREDICTION FROM SELFSIM MODEL (TSA)

## LOTUNG ARRAY, TAIWAN



- Tsai (2007) found for total stress analyses...
  - SelfSim is capable of reproducing downhole array measurements
  - Learning from multiple records is required to capture nonlinear behavior over a wide range of strain

# I LEARNED DYNAMIC SOIL BEHAVIOR



**I** ...EXTENSION TO POREWATER PRESSURE GENERATION

- Develop NN soil model with PWP generation capability
- Soil column discretization

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**I** NEURAL NETWORK MATERIAL MODEL

(a) 3 history points

stress

strain

$(\gamma_{i-2}, \tau_{i-2}, p_{i-2})$

$(\gamma_{i-1}, \tau_{i-1}, p_{i-1})$

$(\gamma_i, \tau_i, p_i)$

$(\gamma_{i+1}, p_{i+1}, ?)$

(b) **PWP NN**

Output layer (1 node)

Hidden Layers (19 nodes ea.)

Input Layer (4 nodes)

$(|\gamma_i|, |\gamma_{rev}|, \gamma_i - \gamma_{rev}, |\gamma_{i+1}|)$

$\Delta p$

$p_{i+1} = p_i + \Delta p$

(c) **SOIL NN**

Output layer (1 node)

Hidden Layers (19 nodes ea.)

Input Layer (11 nodes)

$(\gamma_i, \tau_i, p_i), (\gamma_{i-1}, \tau_{i-1}, p_{i-1}), (\gamma_{i-2}, \tau_{i-2}, p_{i-2}), (\gamma_{i+1}, p_{i+1})$

$\tau_{i+1}$

- 3-point model demonstrated for stress-strain-pore pressure coupled behavior (a).
- Proposed NN architectures for (b) pore pressure response model, and (c) stress-strain model illustrating how the pore pressure output is given as input to the soil model.

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## I SELFSIM MESH GENERATION

○ Acceleration Recording
△ Pore Pressure Recording
□ / ■ Integration Point (Data Extracted / Not Extracted)

Downhole Array Profile

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## I SYNTHETIC VERTICAL ARRAY

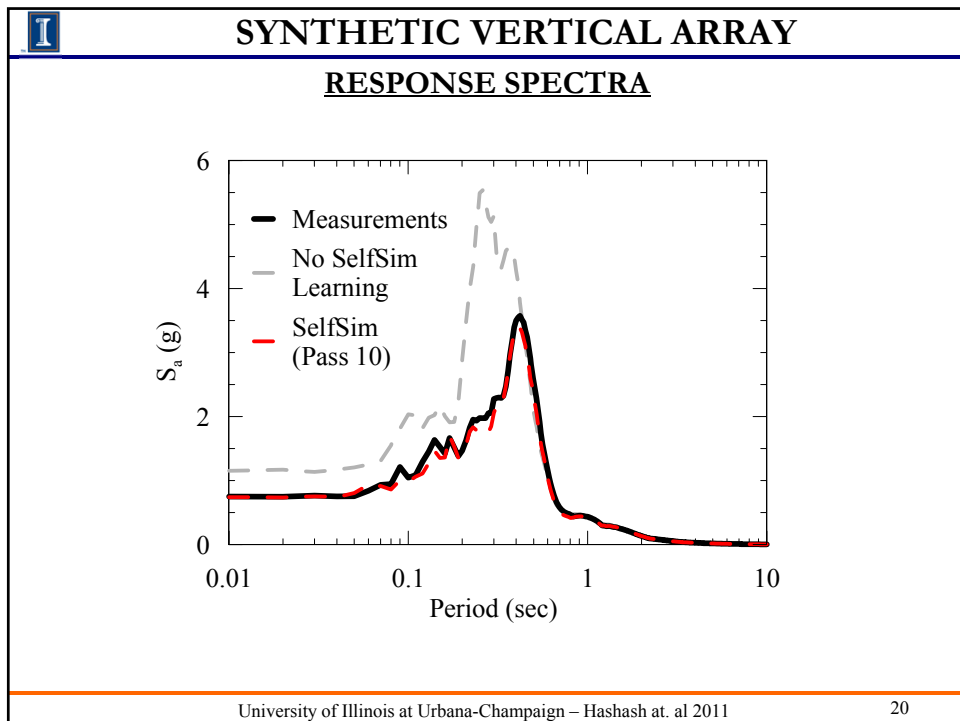
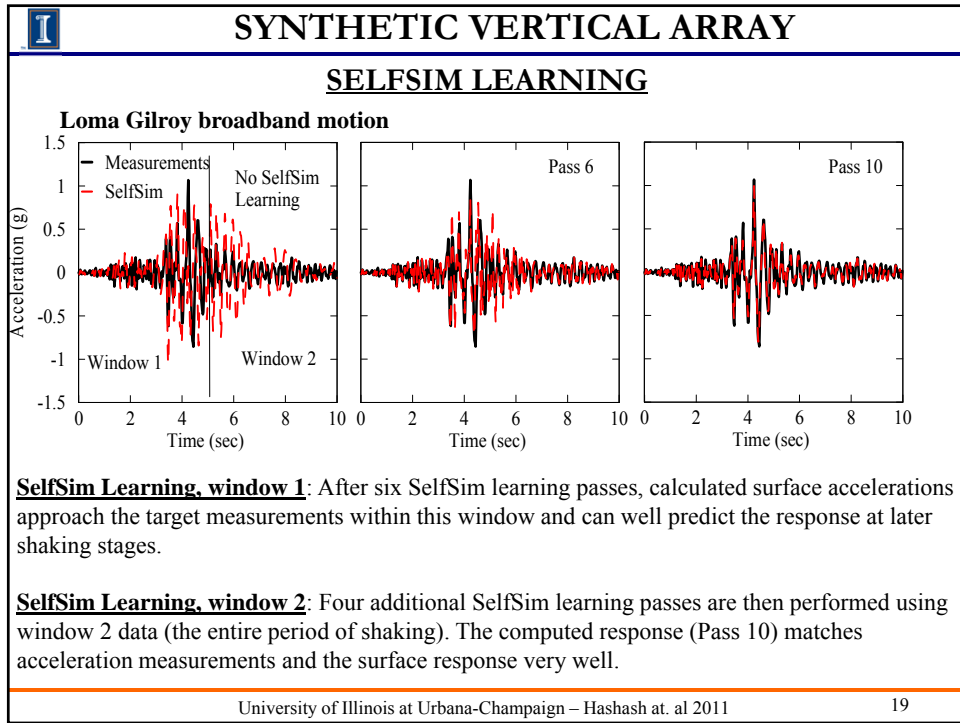
### PROFILE

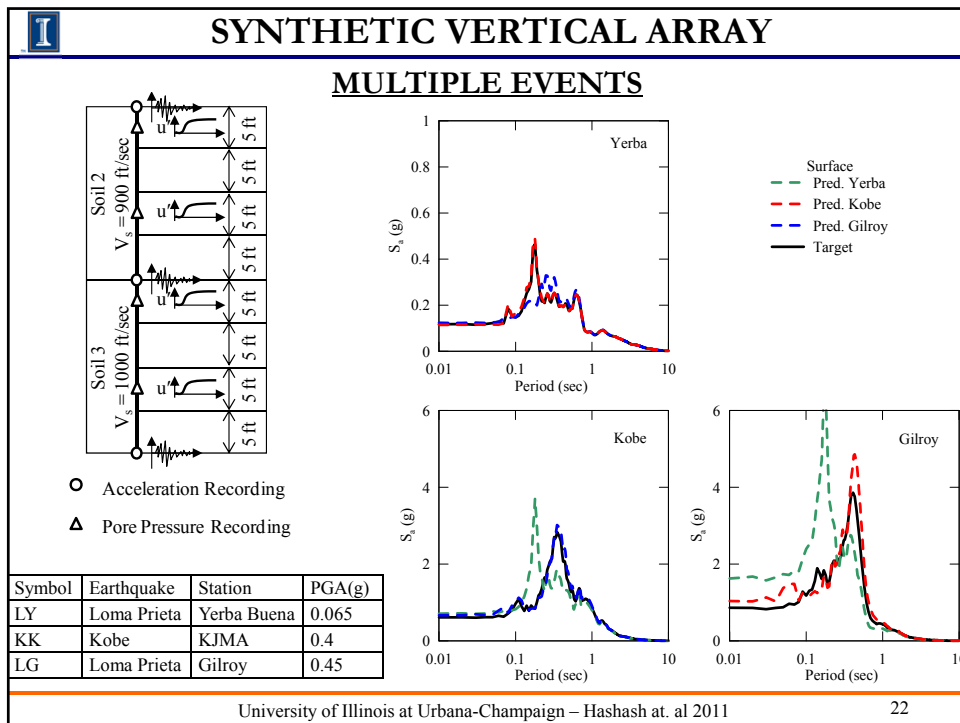
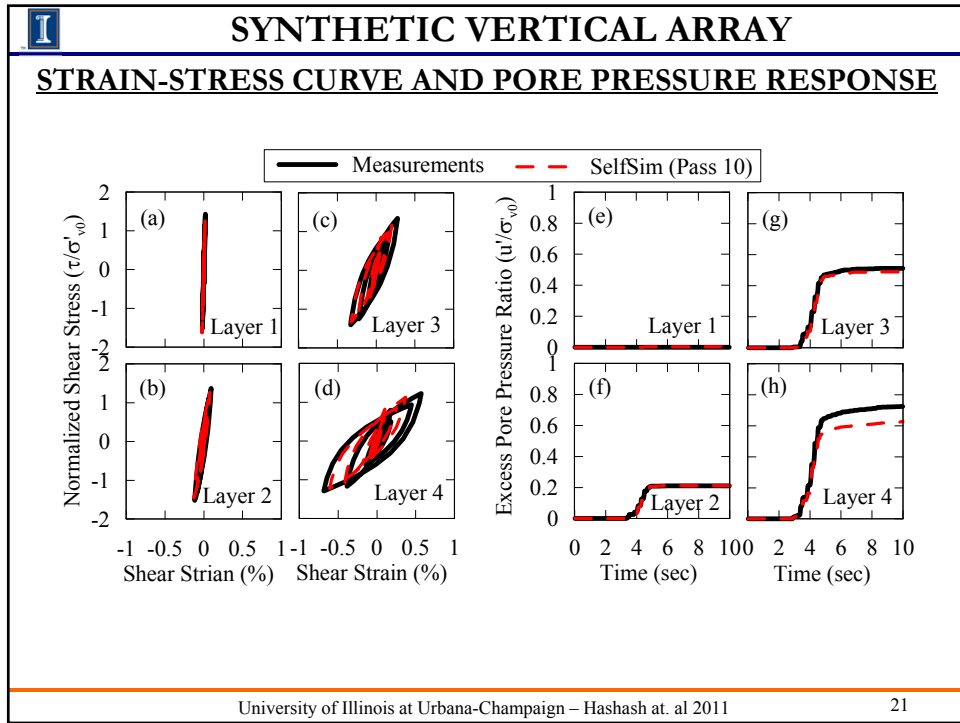
Development of synthetic vertical array data follows three steps:

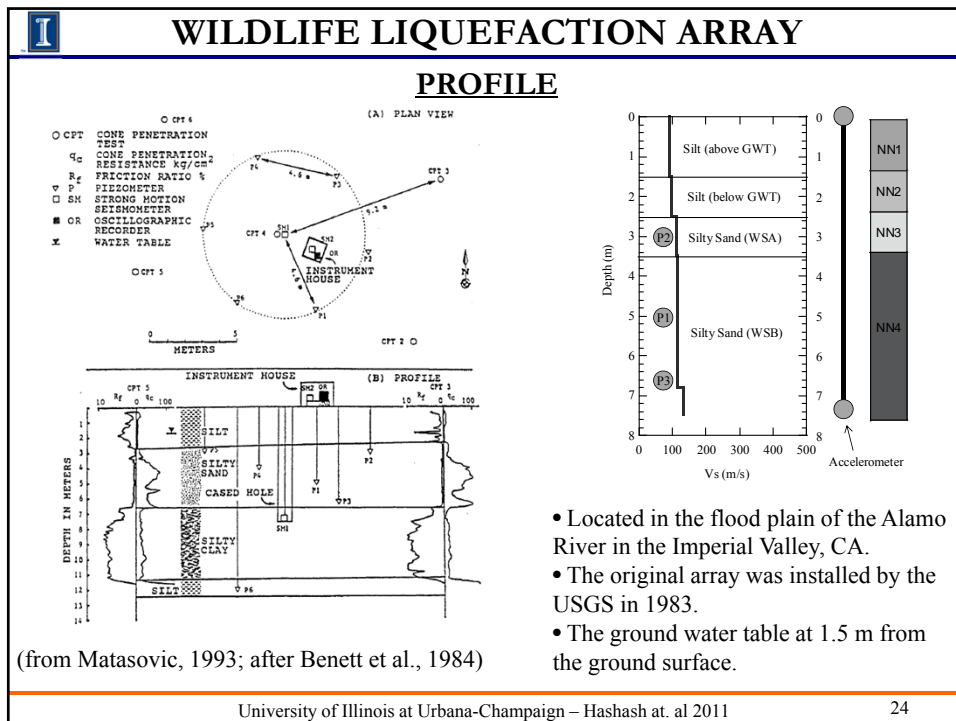
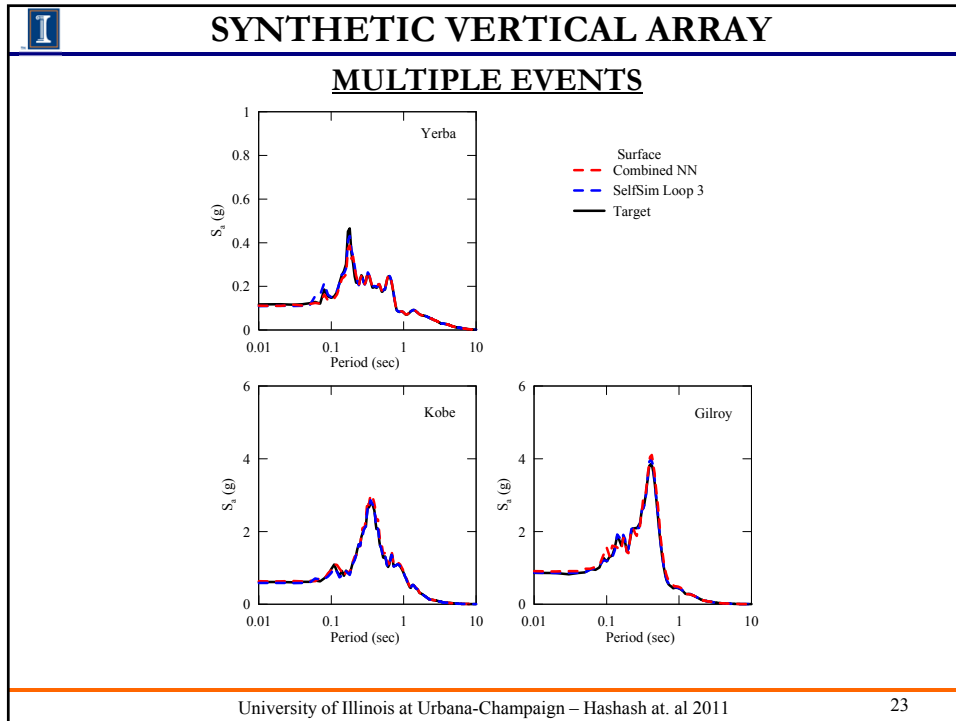
1. Select a soil profile with known nonlinear soil behavior at the site.
2. Use DEEPSOIL to propagate a motion at bedrock through soil columns with known soil behavior using the hyperbolic soil model.
3. The output displacement, velocity, acceleration, and pore water pressure data of certain layers are used as synthetic array data.

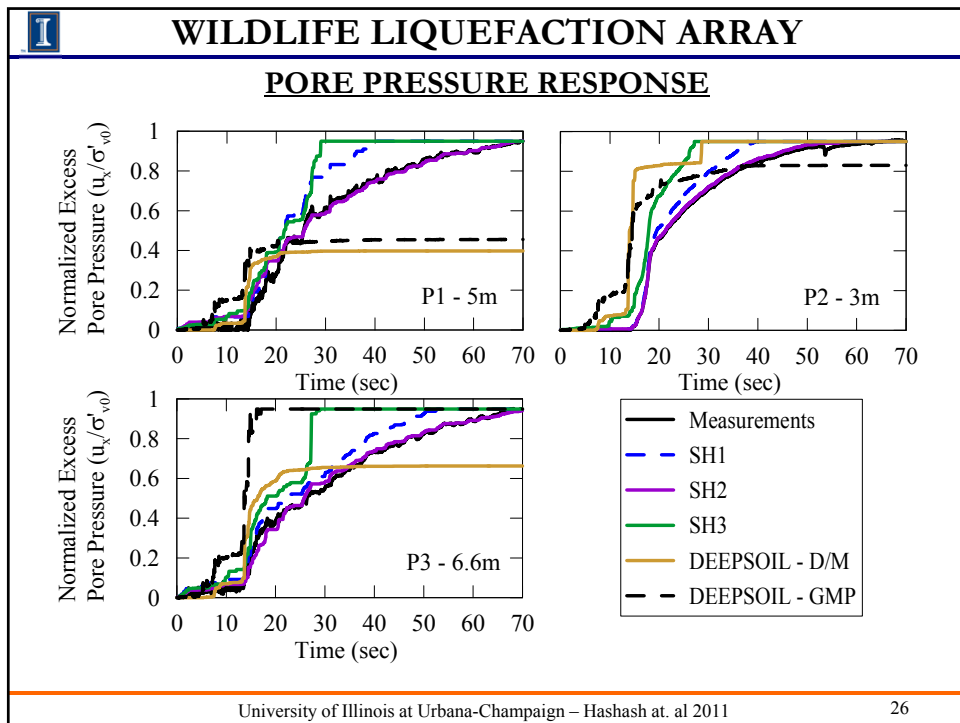
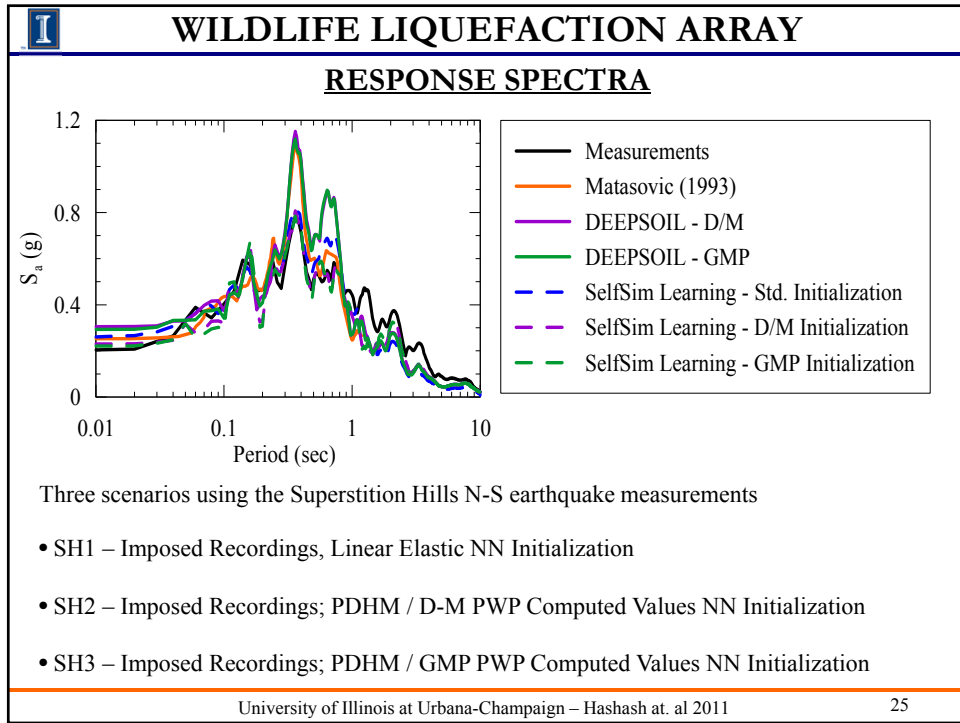
○ Acceleration Recording  
△ Pore Pressure Recording

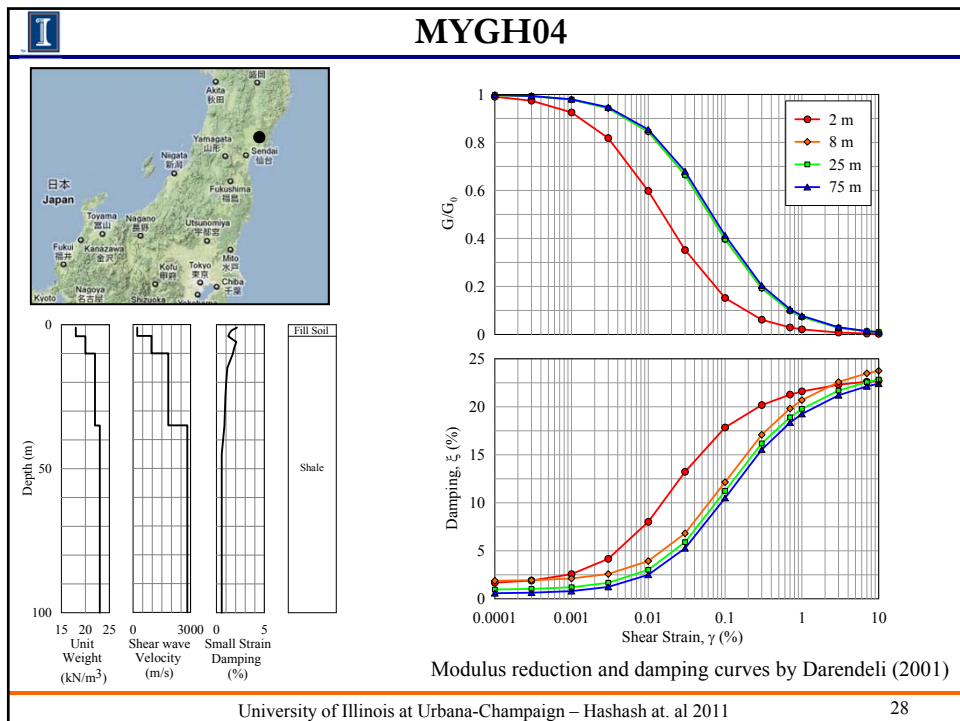
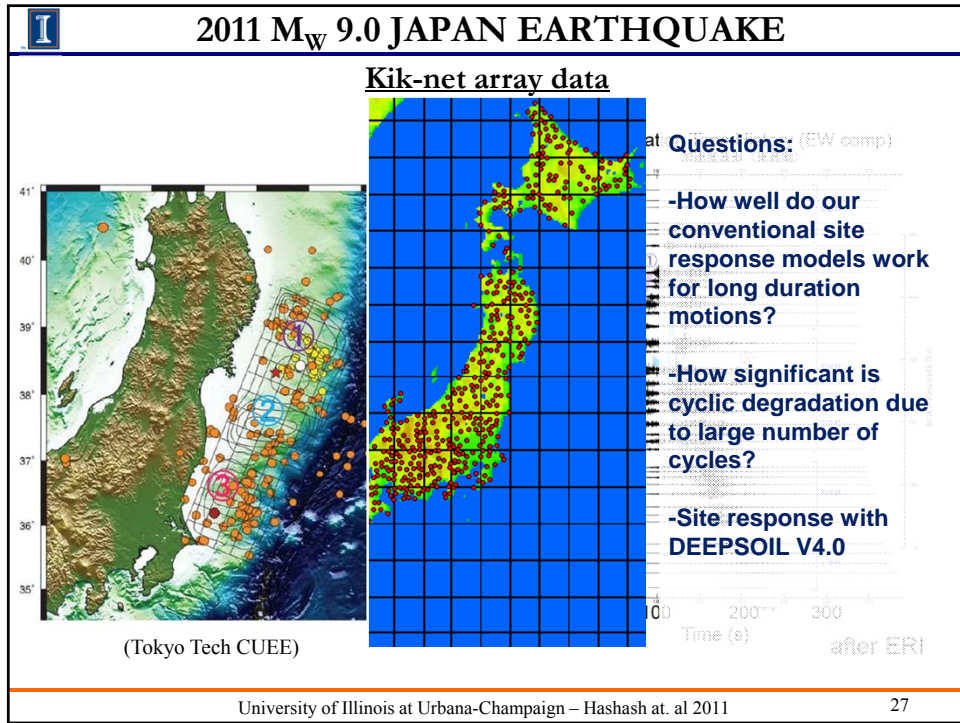
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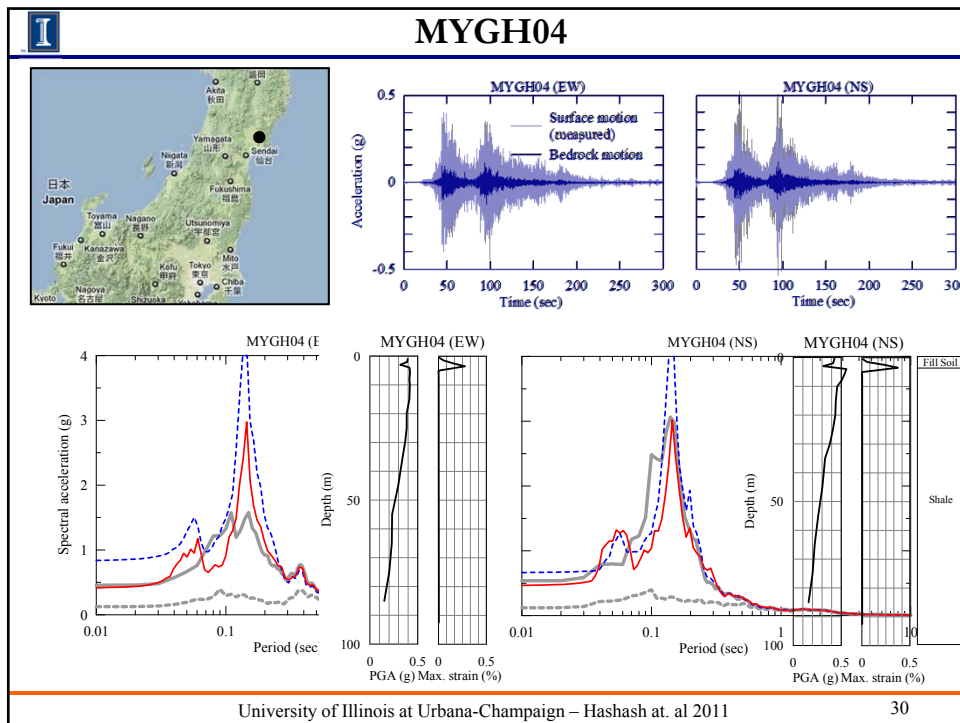
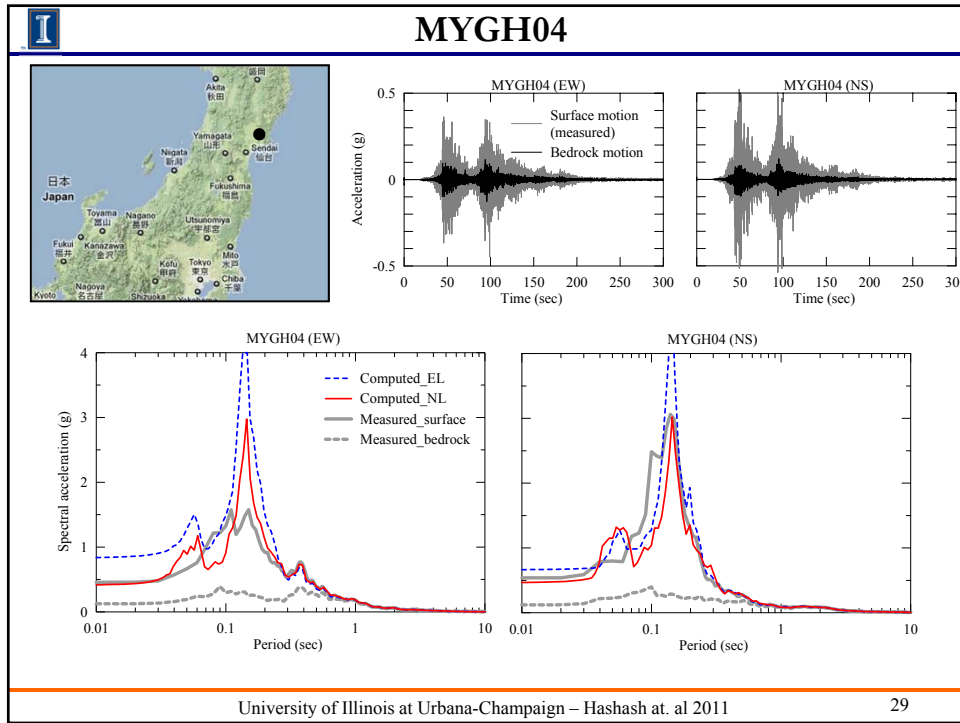












**I****CONCLUDING REMARKS**

- Vertical Arrays provide valuable information for improving our site response models
- SelfSim inverse analysis framework has been extended to learn from vertical arrays with pore water pressure (PWP) measurements.
- Need more array data with PWP to evaluate and improve the reliability of our models.
- Preliminary one-dimensional site response analyses were conducted using the program DeepSoil v4.0 to reproduce the soil response from the downhole arrays for the March 11 2011, Tohoku, Japan, earthquake.
  - Current site response analysis are able to capture key features of the response at some of the stations.
  - Need for more detailed characterization of the sites
  - Influence if cyclic degradation?
  - There is much to learn from the data.