



Reconstruction of Soil Stress-Strain Response Using Optimisation

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- Started as a PDRA as part of the IROE group ~2 months ago
- Almost finished PhD at University of Sheffield (viva delayed due to Covid) (Supervisors: Colin C Smith; Jonathan A Black)

- Current research is on recovery of geotechnichal design parameters from undersea soils via geophysics/seismic data
- Aiming to introduce data science techniques to geotechnics

• This presentation is a summary of my PhD project





- It is necessary to find soil properties for engineering design
 - Either for direct use or as parameters for further numerical modelling
- Many such methods are used, either in the lab or in the field
 - Triaxial
 - Shear vane
 - Cone penetrometer
 - Shear box
 - Scale model testing
 - + more
- However all methods have pros and cons





Potential Issues

- Sample disturbance
- Sample representativeness
- Result interpretation
- Inaccuracy in measured soil response could cause e.g. mismatch between physical and numerical modelling

- Image processing now allows the strain field to be reconstructed in plane strain model tests (and in 3D with transparent soils).
- Hence the physical model test itself (e.g. bearing capacity, retaining wall) can be the 'element' test.



• Conservation of Energy:

• Work done by applied loading e.g. a footing displacing due to load must equal work done by deformation.

• External work = Internal work



- For homogenous isotropic undrained soil (Clay) a number of assumptions can be made:
 - Associative flow, i.e. same principal angles for stress and strain
 - No volumetric strain, i.e. total volume of soil stays the same, it just moves around
 - Simple isotropic relationship between shear stress and strain i.e. $t=f(\varepsilon_s)$
- Resulting the simplest form of the equilibrium equation:

$$\int_{u^{j-1}}^{u^j} P \, du = \left(2 \cdot \int_{\epsilon_s^{j-1}}^{\epsilon_s^j} t \, d\epsilon_s \cdot V \right)$$



"Segment" based method

• Define a piecewise representation of the shear-stress shear strain curve in *m* parts

Methodology:

Southam

- This means there are a set of m unknown values of τ to be identified
- Optimization is used to find the curve that produces the lowest difference between internal and external energy







• Capture the physical model test in n images

- Log *n* sets of load-displacement data corresponding to each image
- Discretise each image into p patches each of which provide shear strain increment $\Delta \gamma$
- Strain can be calculated from displacement field using Constant Strain Triangle elements







- External work $W_{\rm ext}$ for each image can be computed from the load-displacement data

- Shear strain at each (discretised) point in the image can be determined from the displacement field
- Internal work W_{int} can be computed by integrating the shear strains with the shear stresses found via the (unknown) soil response
- Find the response that minimises rms of $(W_{ext} W_{int})$ across *all* image data (typical solve time ~ 1-5 mins).

Validation: Rotating wall FEA, "Segment" approach



Input Curve

0.1

Output Curve

0.12

0.14



Southampton





- A suite of physical model tests carried out in order to obtain high quality image and loading data
- 1g footing tests with 20mm and 40mm footings
- GeoPIV-RG used to obtain imaging data
- 1g strain based actuator used at a rate ensuring undrained conditions
- Supplementary testing (triaxial, shear vane) taken to provide comparison data
- Methodology works well for artificial datasets goal is to find robustness when dealing with real data





Southampton Physical Model Testing: Results



- Overall findings are that the methodology works but is not robust to flaws in the datasets.
- It is possible to use the methodology to recover stress-strain responses but care must be taken to ensure datasets are of sufficient quality.
- Flaws that cause some internal work to be "missing" such as poor PIV texture or movement outside area of interest cause stress-strain response to be higher and steeper.
- Some flaws such as random noise result in "extra" internal work, causing recovered response to be lower/shallower
- A subset of collected data will be presented to illustrate these points.



Dataset 4





Dataset 4





Dataset 4







- Methodology shows promise, working well for "perfect" artificial datasets and some "real" datasets but is not robust to flaws or omission in the available data.
- It is possible that this could be solved through software changes but experimental design is likely the best means to ensure successful recovery of stress-strain response.
- Detailed recommendations for designing experiments to utilise the methodology can be found in my thesis.
- Key point is to ensure **All** internal work is counted.





End of Presentation

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