

**NC STATE UNIVERSITY**

# **Bayesian Estimation of Soil-Water Characteristic Curves**

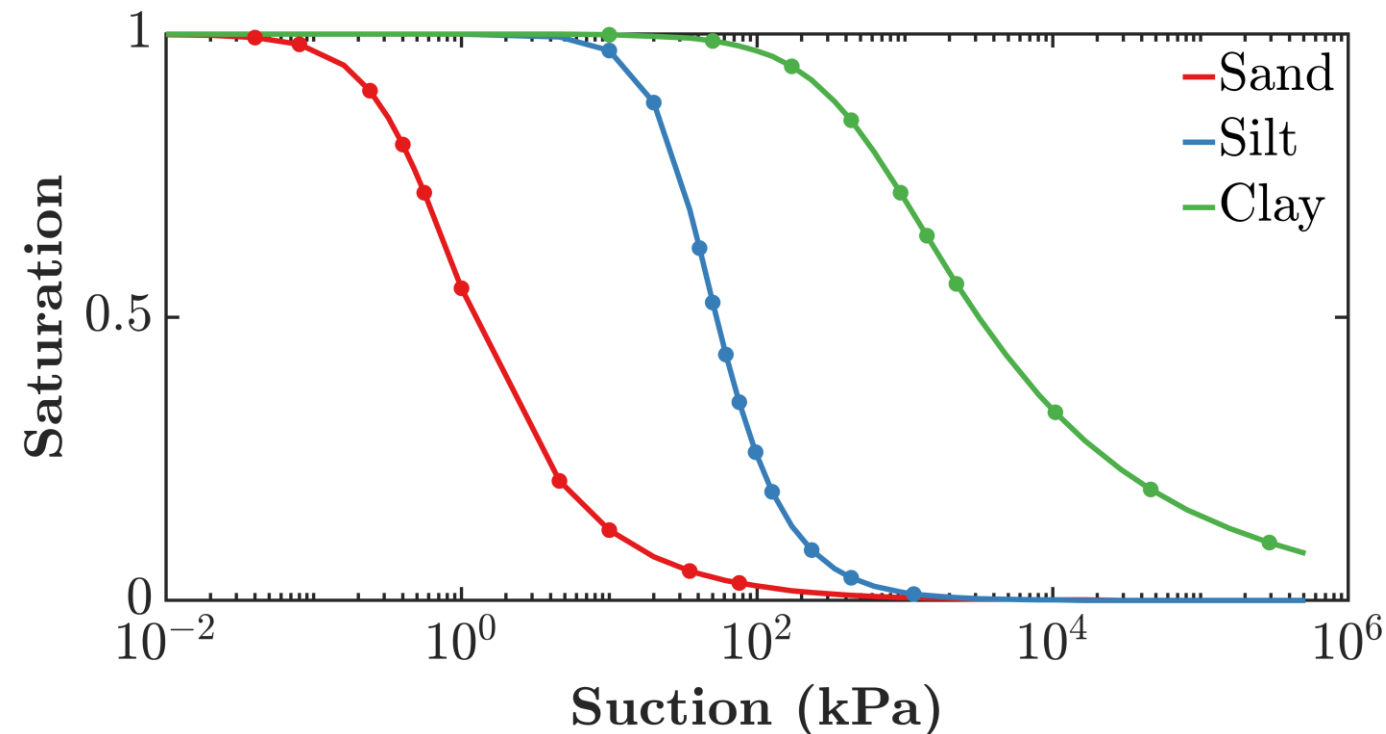
**ST540: Final Project  
Presentation: 5/1/2024**

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# What is SWCC?

- The Soil Water Characteristic Curve (SWCC) is an important prerequisite for studying the mechanical properties of unsaturated soil.



# Why is Bayesian Analysis used for SWCC?

- ❑ The issues are:
  - The experimental measurement for the SWCC is time-consuming
  - Empirical methods suggested to estimate the SWCC have substantial uncertainties
  
- ❑ Suggest a Bayesian approach to estimate the SWCC from a limited number of tests:
  - take advantage of prior information: database of SWCC
  - define prior distribution of SWCC parameters based on soil properties from quick and low-cost lab tests (e.g. soil texture, void ratio, and effective grain size)



# Data and Data Analysis

❑ Data Source: SoilVision Database

❑ SWCC model: van Genuchten (1980)

$$g_i = g(\psi_i | \alpha, n) = \frac{1}{[1 + (\alpha\psi)^n]^{1-\frac{1}{n}}}$$

❑ Parameter for the model:

- $\alpha$ : the fitting parameter related to the air-entry value of soil (kPa)
- $n$ : the fitting parameter related to the slope of the SWCC

❑ Covariates:

- Soil texture: percentage of clay ( $C_i$ ) and coarse grain ( $C_g$ )
- Void ratio ( $e$ )
- Effective grain size ( $D_{10}$ )



# How to Establish our Model for SWCC?

$S_i \sim N(g_i, \sigma^2)$  for homoskedastic  
 $S_i \sim N(g_i, \sigma^2 g_i(1 - g_i))$  for heteroskedastic

$$g_i = g(\psi_i | \alpha, n) = \frac{1}{[1 + (\alpha\psi)^n]^{1-\frac{1}{n}}}$$

(Zhang et al, 2021)

## Multi Linear Regression (Database)

$$(\ln \alpha, \ln n)^T \sim N(\boldsymbol{\mu}, \boldsymbol{\Sigma})$$

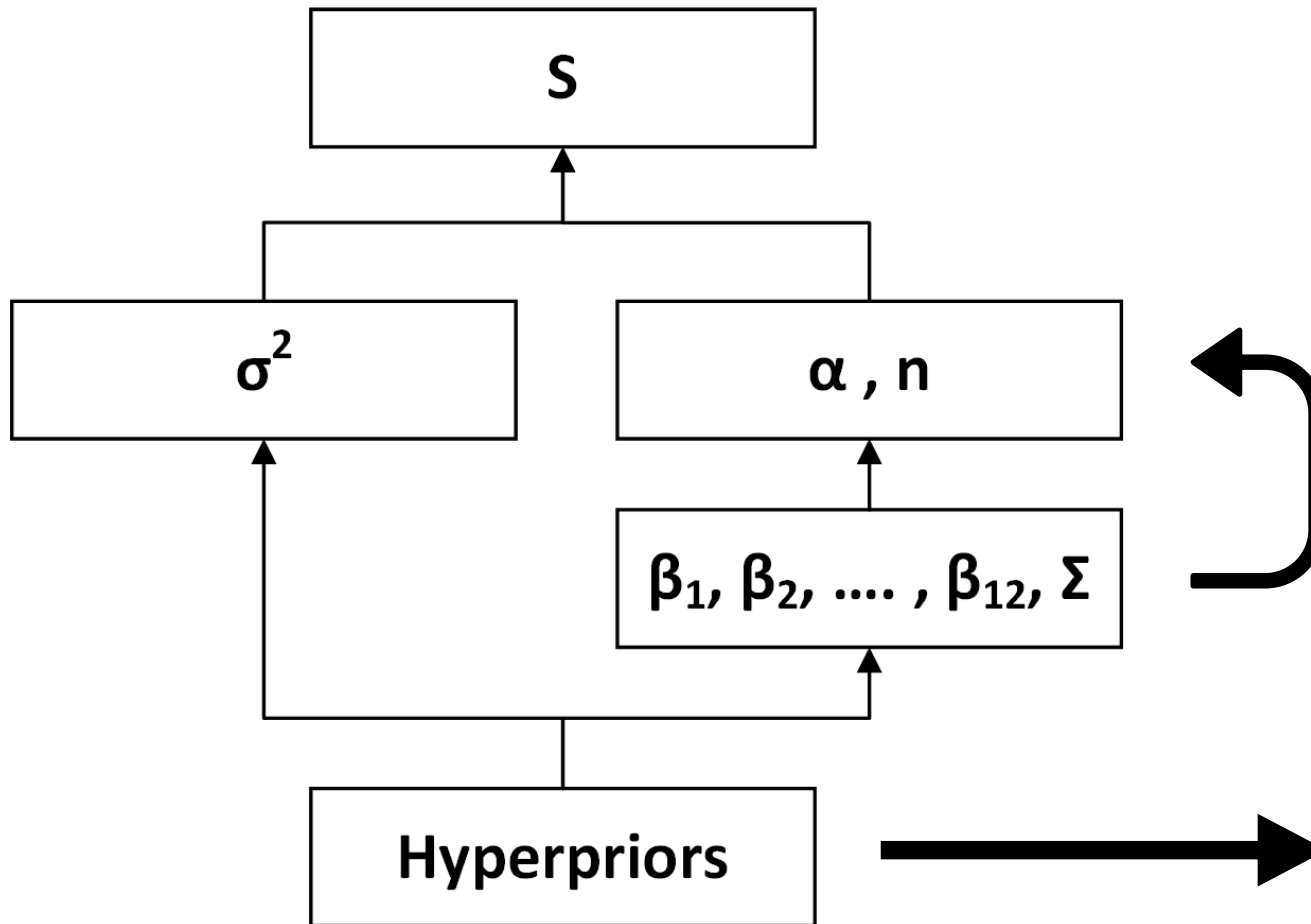
$$\mu_{\ln \alpha} = \beta_1 + \beta_2 X_1 + \beta_3 X_2 + \beta_4 X_3 + \beta_5 X_4$$

$$\mu_{\ln n} = \beta_6 + \beta_7 X_5 + \beta_8 X_6 + \beta_9 X_7 + \beta_{10} X_8$$

$$\beta_{ij} \sim N(0, 0.001)$$

$$\boldsymbol{\Sigma} \sim \text{InvWishart}(2.1, \mathbf{I}_2/2.1)$$

$$\sigma^2 \sim \text{InvGamma}(2.366, 0.012) \text{ informative}$$



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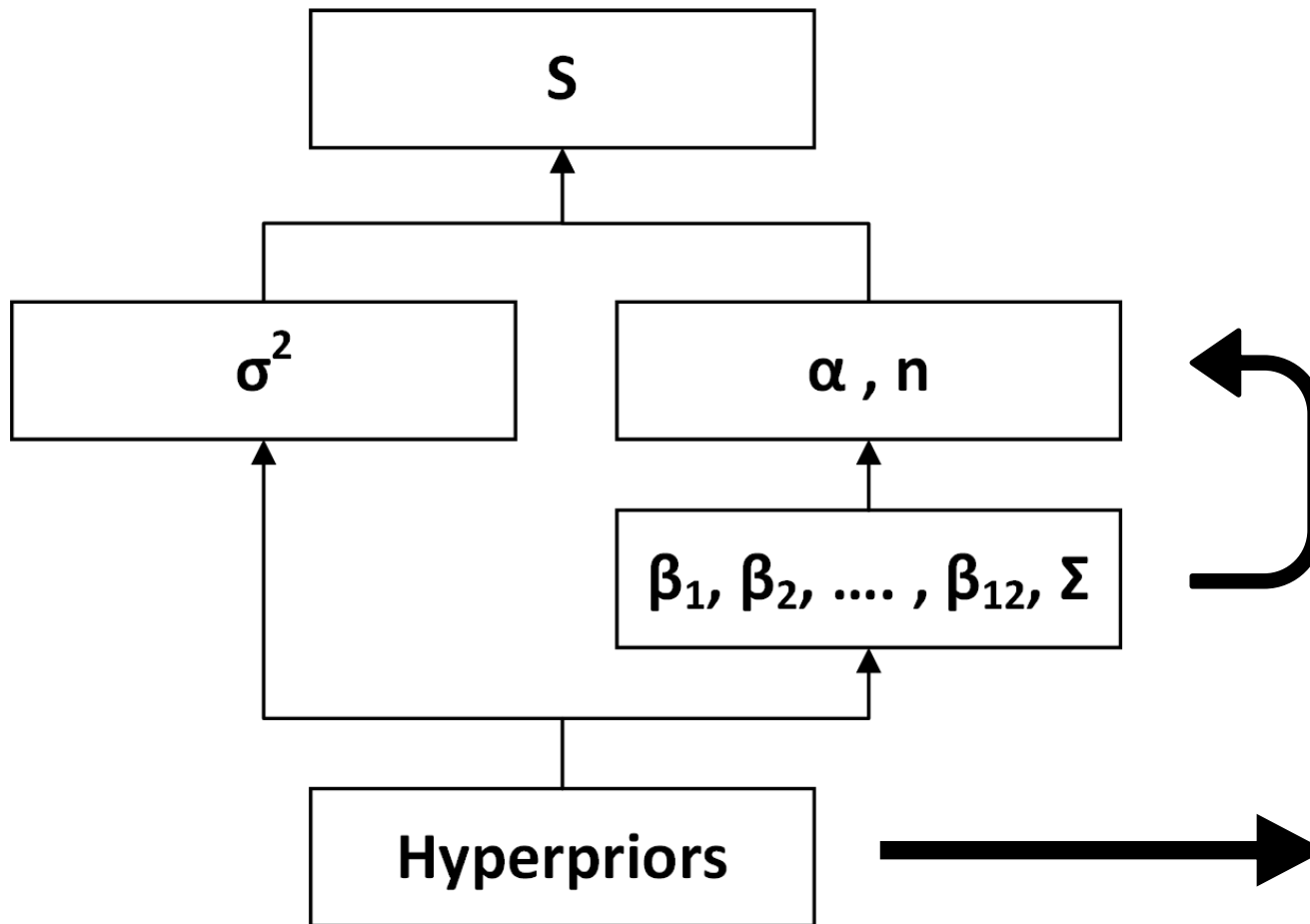
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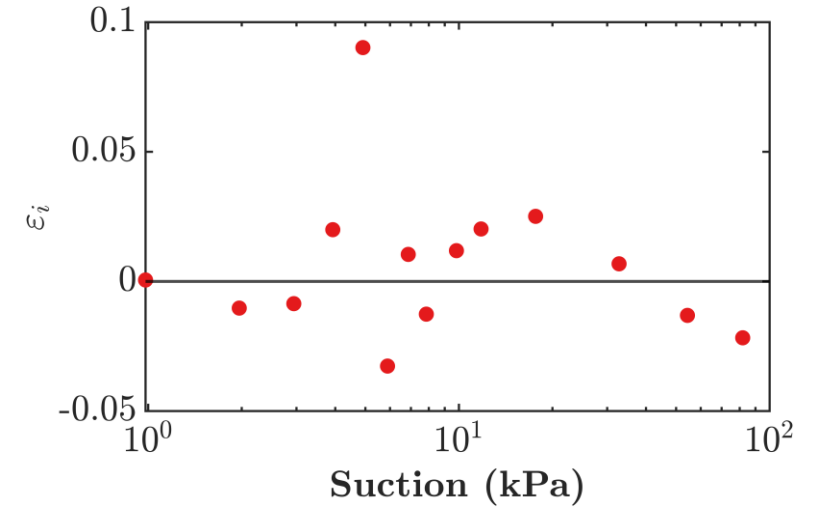
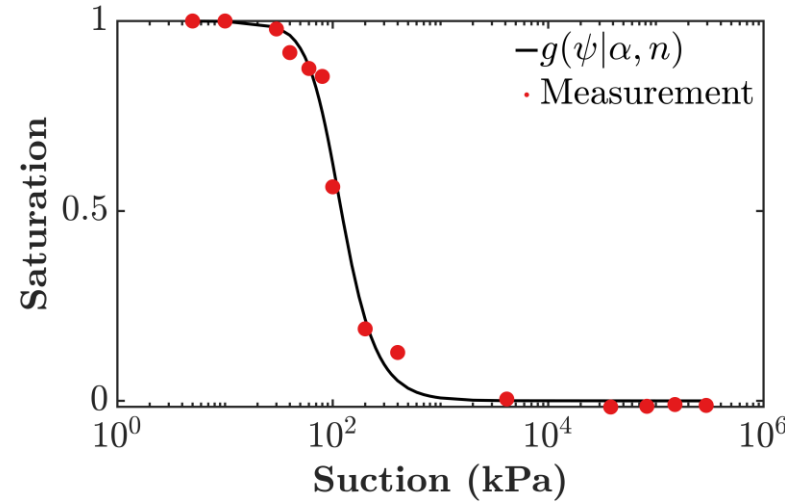
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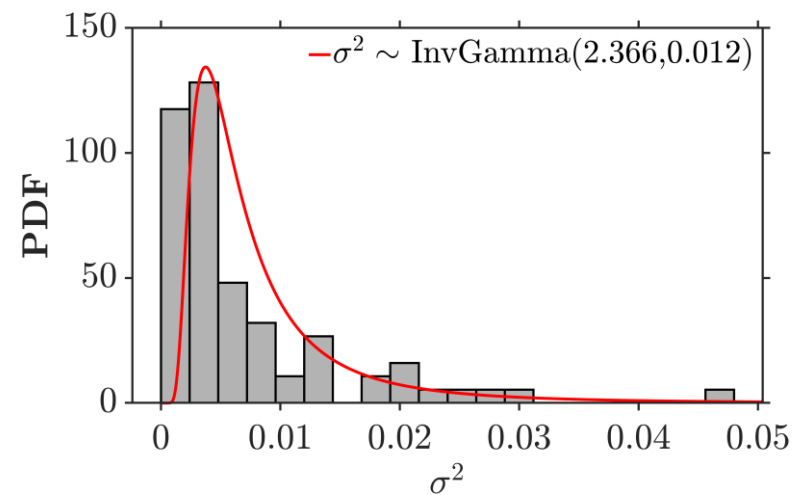


# How to Establish our Model for SWCC?

❑ Heteroskedastic?



❑ Informative prior for  $\sigma^2$ ?



# Model Selection and Fitting

❑ Model selection:

Multi Linear Regression

Model	$\alpha$	n	Min (ESS)	Dw	Penalty	DIC
A	$C_i, C_g, e, D_{10}$	$C_i, C_g, e, D_{10}$	26413.9	439.1	13.3	452.7
B	$\ln C_i, \ln C_g, \ln e, \ln D_{10}$	$\ln C_i, \ln C_g, \ln e, \ln D_{10}$	27290.1	436.7	13.34	450.1
C	$C_i, C_g, e, D_{10}$	$\ln C_i, \ln C_g, \ln e, \ln D_{10}$	27948.7	438.9	13.71	452.6

❑ Model fit:

Model	Min ( $\alpha$ )	Min (n)	Max ( $\alpha$ )	Max (n)	Range ( $\alpha$ )	Range (n)
A	0.377	0	0.574	0.484	0.654	0.981
B	0.090	0	0.456	0.238	0.861	0.955
C	0.397	0	0.589	0.236	0.651	0.937

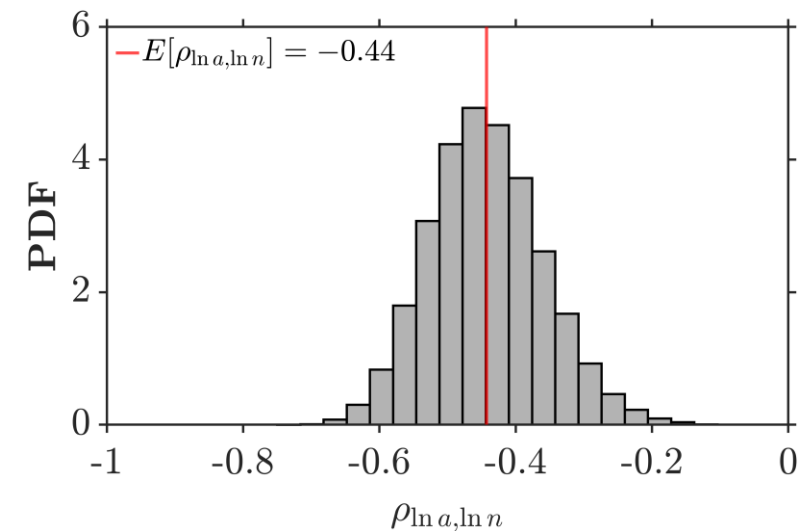
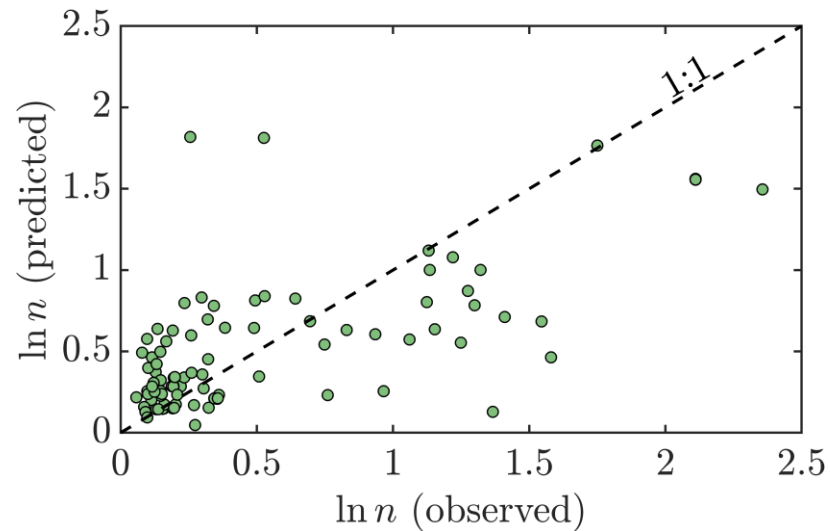
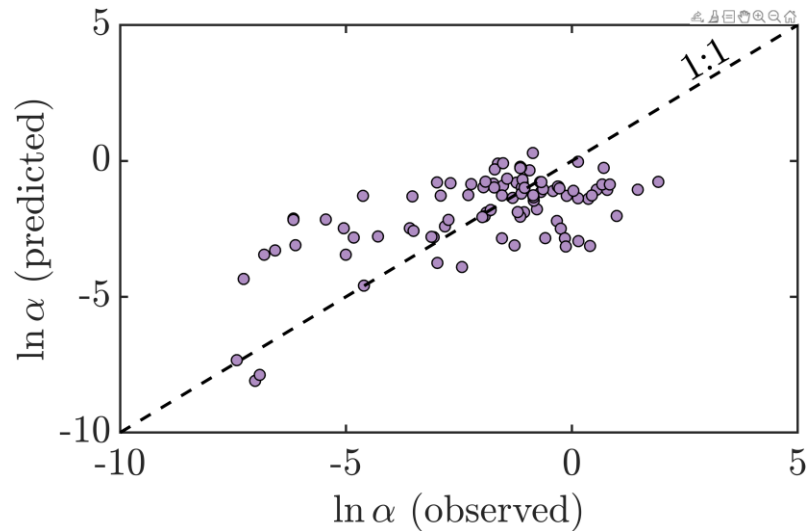




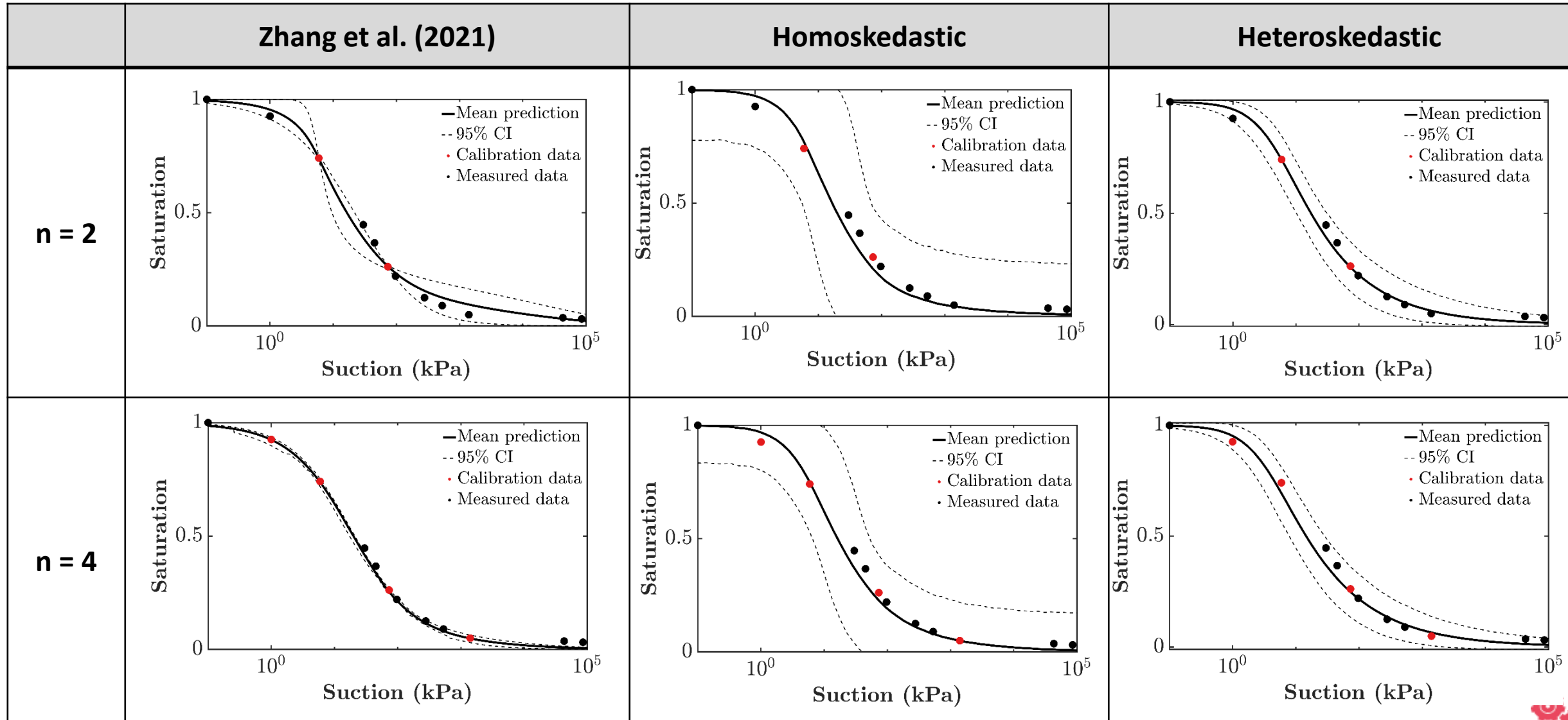
# Cross Validation

- Five equally-sized groups / Predictions setting aside one group at a time and fitting the model with the observations of the remaining groups

Model	Parameter	Bias	MSE	MAD	COV	WIDTH
B	$\alpha$	0.021	2.891	1.334	0.447	6.786
	n	-0.001	0.167	0.282	0.937	1.603



# Results of Informative Model



# Conclusion

- ❑ Limitations and improvement of our model
  - The value of  $\min = 0$ : the model did not fit the data sufficiently
    - add more test data (95 samples; more than 200 samples)
  - Using linear regression
    - use higher-order polynomial or a second-order polynomial regression
  - Limited covariates
    - use more covariates (currently use 4 covariates)



# Reference

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- Reich, B. J., Ghosh, S. K. (2019). Bayesian Statistical Methods.
- van Genuchten, M.T. (1980). Closed-form equation for predicting the hydraulic conductivity of unsaturated soils. *Soil science society of America Journal*. **44**: 892 – 898.
- Zhang, J., Yang, S., Zhang, L. L., Zhou, M. L. (2021). Bayesian estimation of soil-water characteristic curve. *Canadian Geotechnical Journal*. **59**: 569-582.



# Thank you



**ST540: Geotechnical Engineers**