

Methods for Estimating Major Angles of Anisotropy in the Subsurface

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Overview

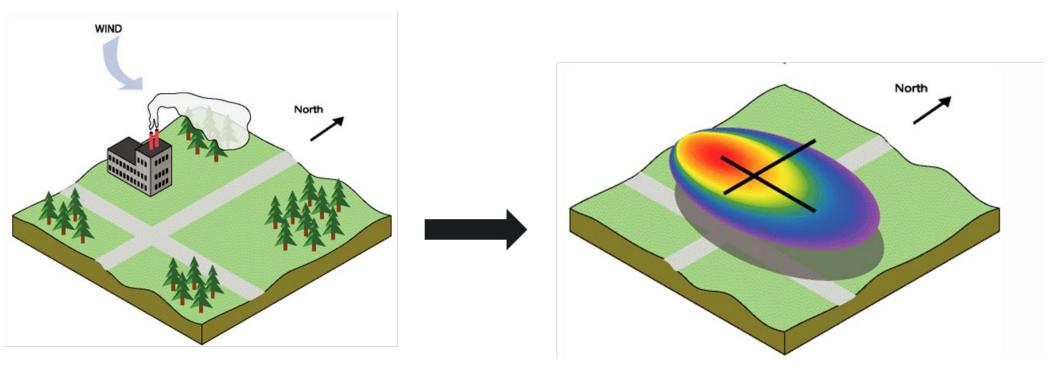
- Background
 - Anisotropy
 - Variogram analysis & automation
- Methodology
- Results/Hanford case study
- Implementation in Visual Sample Plan



Anisotropy

Occurs when spatial dependence is stronger in some directions than others

- Ex: due to wind directions surface
- Ex: due to geomorphology subsurface
- Some geospatial models incorporate anisotropy

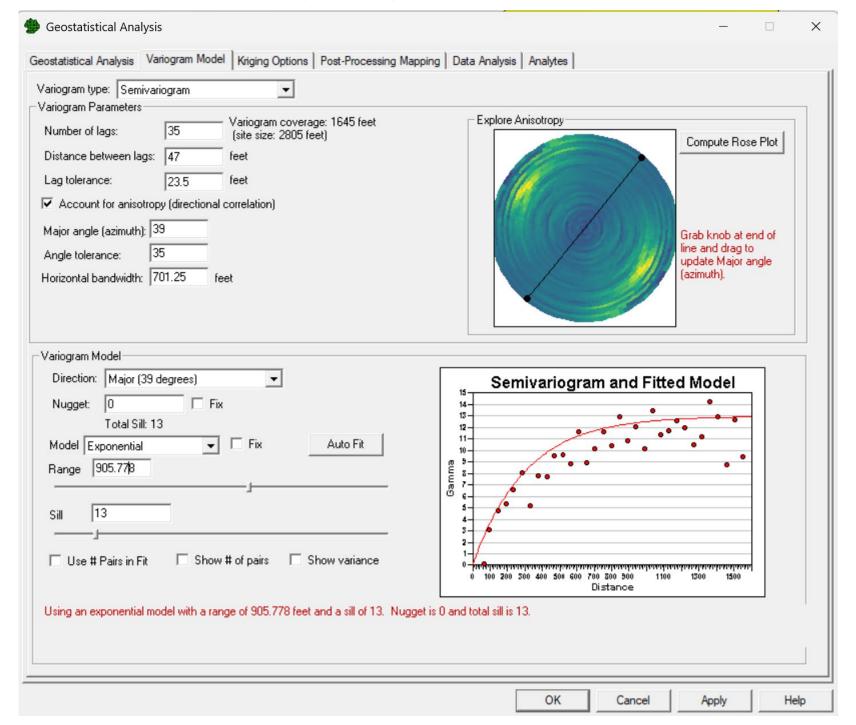




Variogram Analysis

- Statistical representation of correlation as a function of distance.
 - Isotropy: variograms are the same in all directions.
 - Anisotropy: variograms differ depending on direction.
- Added Visual Sample
 Plan functionality that
 automates the variogram
 analysis for easier
 implementation.

Visual Sample Plan (VSP) www.pnnl.gov/projects/visual-sample-plan





Approach

- Literature review identified few approaches to determining directions of anisotropy. PNNL identified MOI and CTI as most promising.
- Mass moment of inertia (MOI) (Hassanpour and Deutsch 2008)
 - Employs principles of MOI tensor, commonly used in classical mechanics to describe the rotation of a body, analyzing eigenvectors to estimate anisotropy directions (Goldstein et al. 2001).
 - Mathematical robustness and versatility in handling data from different dimensions and scales offer an approach to understanding spatial orientation.
- Covariance tensor identity (CTI) (Chorti and Hristopulos 2008)
 - Utilizes mathematics of covariance functions to determine anisotropy directions through solutions of nonlinear equations.
- Both methods previously developed in 2D frameworks. PNNL developed in 3D.
 - MOI: PNNL adapted the 2D MOI method to 3D by introducing a third dimension along the z-axis, resulting in a 3D tensor matrix.
 - CTI: PNNL developed a 3D approximation, where we applied the 2D CTI method to 2D slices of the 3D volume and calculated principal angle of anisotropy for each and then averaged the slice estimates.



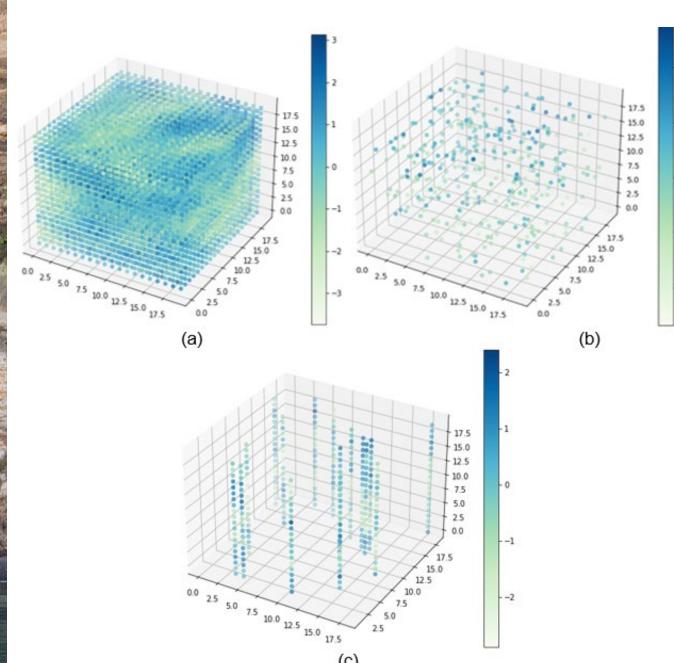
Approach (cont.)

- PNNL adapted 2D MOI and CTI frameworks for 3D datasets for application to subsurface soil samples and surveys.
 - Demonstrated ability to systematically identify principal directions of anisotropy and evaluated performance for multiple types of subsurface datasets (randomly samples points, densely sampled boreholes/wells, sparsely sampled boreholes).
 - Initially developed and tested on simulated 2D and 3D datasets to provide a proof of concept.
 - Applied MOI and CTI to case study with data from Hanford Site.

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Method Evaluation



Simulated Data

- (a) structured
- (b) unstructured
- (c) borehole data

Generated from 3D variogram models:

- Angles of anisotropy equal to $(\alpha, \beta, \gamma) = (45^{\circ}, 25^{\circ}, 10^{\circ})$.
- Axes are in meters.

Dimension	Data Type and Number of Observations					
			Borehole			
	Structured	Unstructured	16	100		
			Boreholes	Boreholes		
2D	2D 400 400		n/a	n/a		
3D	8000	400	320	2000		



Results

	Sample	True Angle of	Estimated Angle of Anisotropy				
Data Type	Size		MOI -	– 2D	CTI – 2D		
	(# Points)	Anisotropy (°)	Mean (°)	MAD (°)	Mean (°)	MAD (°)	
Structured	400	30	33	15	33	6	
Structured	400	45	44	8	45	5	
Structured	400	60	60	17	58	6	
Unstructured	400	30	33	11	43	22	
Unstructured	400	45	33	13	11	36	
Unstructured	400	60	57	10	57	11	

		True Angles of	Estimated Angles of Anisotropy					
Data Type	Sample Size		MOI	– 3D	CTI – 3D			
		Anisotropy (°)	Mean (°)	MAD (°)	Mean (°)	MAD (°)		
Structured	8000 points	45, 25, 10	40, 30, 19	7, 7, 11	62, 31, 15	24, 6, 7		
Structured		60, 30, 10	52, 20, 17	22, 11, 9	66, 34, 9	7, 6, 4		
Unstructured	400 points	45, 25, 10	49, 17, 20	18, 11, 13	n/a	n/a		
		60, 30, 10	68, 19, 18	27, 13, 12	n/a	n/a		
Borehole	16 boreholes	45,25,10	59, 23, 32	20, 12, 25	n/a	n/a		
Dorenole	320 points	60,30,10	57, 15, 55	8, 17, 46	n/a	n/a		
Borehole	100 boreholes	45,25,10	46, 17, 18	15, 10, 11	55, -2, 5	17, 27, 7		
	2000 points	60,30,10	58, 16, 19	18, 15, 12	64, -1, 5	4, 31, 5		

MAD = mean absolute deviation (uncertainty)

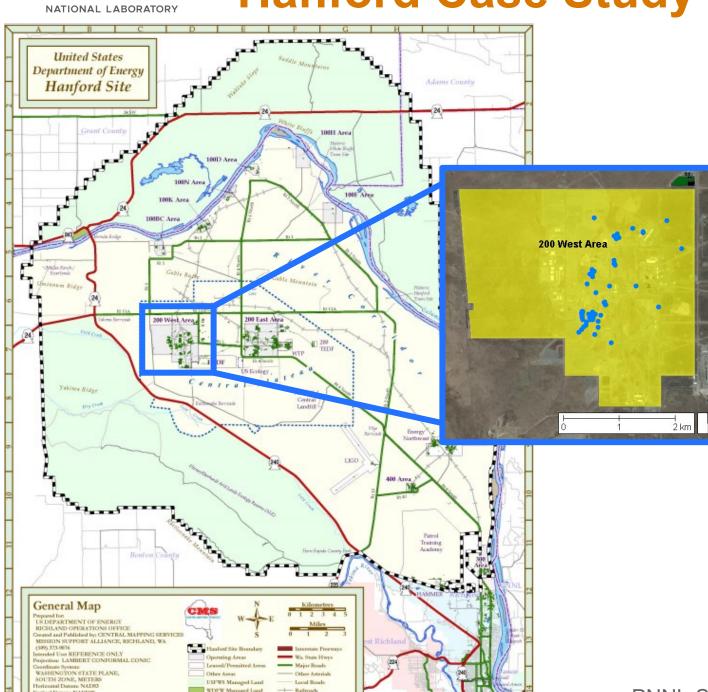
Assess Performance

Compare estimated to true angles of anisotropy.

- MOI estimates within 5-10° of the true angles of anisotropy with MAD ±20°
- CTI estimates deviated from true angles of anisotropy more than MOI estimates, with similar MAD



Hanford Case Study – Dataset



Geographic extent of 200 West Area is on the western side of the Hanford Site.

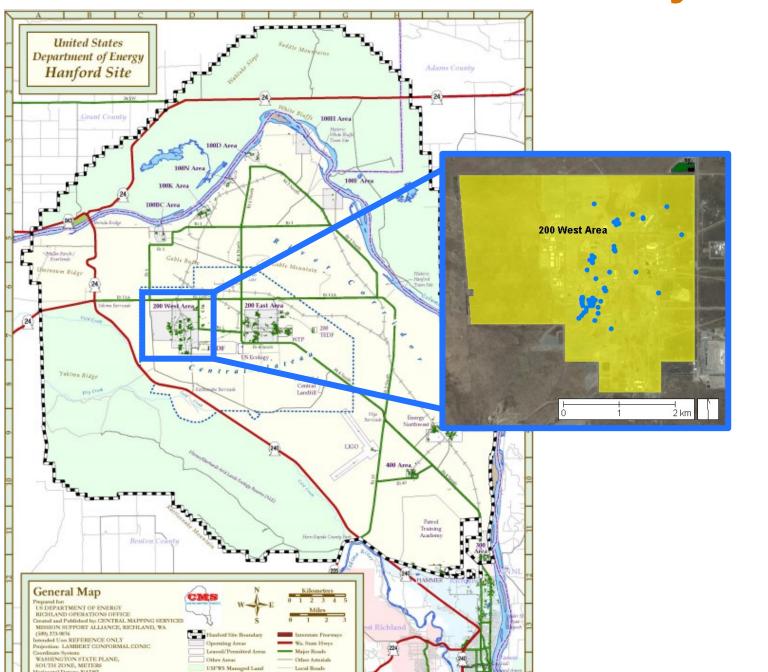
Curated borehole dataset, with input from Central Plateau Cleanup Company (CPCCo)—a prime contractor to the Department of Energy (DOE) environmental cleanup of legacy nuclear waste at the Hanford Site (CPCCo 2024) with expertise in Hanford Site data collection.

Dataset:

- Subset of all boreholes in the 200 West Area at locations shown with blue points
- Active in varying and timespans between 2002 and 2008
- Depths ranging 12–160 meters (40–525 ft)
- 8,800 datapoints total



Hanford Case Study – Geologic Properties



Columbia River drainage paths in this area have changed from the inferred drainage path during the middle-to-late Ringold time (Reidel 2013) to current drainage path.

- Current Columbia River drainage path runs east and then north of where the 200 West Area is located, represented in blue and purple.
- Inferred Columbia River drainage path during the Ringold Formation runs south and west of where the 200 West Area is located, along where WA-243 is currently located (red line running from the bottom right at roughly a 135° angle (from the southeast corner of the map) up to where it intersects WA-24 to the northwest of the 200 West Area).



Generalized Stratigraphy of 200 West Area

	Sedimentary Unit	Depositional	Lithology
		Environment	
Hf1	Hanford formation – unit 1	Fluvial, mega floods	Gravel dominated quartz and basalt
Hf2	Hanford formation – unit 2	Fluvial, mega floods	Sand dominated quartz and basalt
HF3	Hanford formation – unit 3	Fluvial, mega floods	Gravel dominated quartz and basalt
O	Cold Creek unit	Lacustrine(?)	Silt and fine grained sand
CCUC	Cold Creek unit – caliche	Lacustrine(?)	Silt and sand with a caliche overprint
г	Ringold Formation	Fluvial	International Control
	member of Taylor Flat	riaviai	Interstratified deposits of fine-grained fluvial sand and silt.
vie)	l -	Fluvial	
Ringold (Rwie)	member of Taylor Flat Ringold Formation member of Wooded		sand and silt. Well-rounded fluvial gravel of mixed lithologies, in a sand and silt matrix;

younger

older

Applied MOI and CTI methods to the analysis dataset

- 3D volume: single anisotropy estimate throughout
- 2D layered approach: different anisotropy estimate by layer



Analysis Dataset

- Data
 - Spectral gamma gross counts
 - Potassium-40 (naturally occurring radionuclide present across the Hanford Site)
 - (x, y, z) coordinates
 - Sample dates and times
 - Geologic formation within each borehole and (x, y, z) coordinate
- 130 observations per borehole, on average
 - As few as 13 observations
 - Up to 486 observations

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Borehole Identifier	X	у	Number of Observations	
299-E25-184	136006.490	575323.615	62	
299-E28-8	137074.300	573698.100	347	
299-W10-21	137154.721	566583.991	259	
299-W11-14	136913.962	567457.917	302	
299-W11-43	136971.040	567269.740	486	
299-W11-68	136782.453	566970.884	117	
299-W11-69	136834.817	566929.293	121	
299-W11-80	136873.284	566996.371	55	
299-W11-81	136801.847	567024.200	56	
299-W11-82	136407.518	566933.853	78	
299-W11-86	136610.040	568143.530	179	
299-W11-88	137113.090	567874.670	217	
299-W14-1	136311.178	566953.318	253	
299-W14-2	136340.409	566932.320	248	
299-W14-3	136342.965	566940.409	250	
299-W14-52	135936.513	566855.414	29	
299-W14-71	135567.810	567733.430	321	
299-W14-72	135941.280	567328.440	307	
299-W15-10	135981.140	566448.629	254	
299-W15-204	135827.610	566610.306	27	
299-W15-209	136140.101	566545.409	55	



Performance – 3D Volume Application

- Unlike simulated data, ground truth is unknown with real data
 - Use bootstrap re-sample to approximate sampling from boreholes at the site
 - Approximate the 'truth' using the whole dataset
 - Note that the Reidel 2013 hypothesis leads to 135° angle

Radiological	Samula Sina	Angles of Anisotropy	Estimated Angles of Anisotropy (°) – Sampled Dataset				
Measurement	Sample Size	(°)	MOI -	3D	CTI – 3D		
		Ground Truth	Mean (°)	MAD (°)	Mean (°)	MAD (°)	
Potassium-40	70 boreholes 500 points	137, 83, 99	131, 83, 92	7, 1, 8	n/a	n/a	
Gross Gamma	71 boreholes 500 points	125, 77, 96	125, 80, 89	2, 1, 7	n/a	n/a	
Potassium-40	50 boreholes 500 points	137, 83, 99	141, 83, 96	14, 4, 16	n/a	n/a	
Gross Gamma	50 boreholes 500 points	125, 77, 96	129, 75, 90	6, 4, 8	n/a	n/a	

MOI accurately approximates ground truth

- Potassium-40: ground truth 137°, 83°, 99° ↔ MOI estimates 131°, 83°, 92° (MAD 7°, 1°, 8°)



Performance – 2D Layered Application

Badiala di al			Ground Truth		Estimated Angle of Anisotropy			
Radiological Measurement	Layer	Sample Size	Angle of A	Anisotropy	MOI – 2D		CΠ – 2D	
Measurement			MOI (°)	СП (°)	Mean (°)	MAD (°)	Mean (°)	MAD (°)
	Hf1	66 boreholes 500 points	59	82	62	5	68	14
	Hf2	45 boreholes 500 points	40	76	40	1	66	11
Potassium-40	Hf3	22 boreholes 300 points	32	47	34	3	48	3
	CCU	32 boreholes 200 points	41	16	41	1	40	27
	Rwie	16 boreholes 300 points	113	74	121	9	40	34
	Hf1	66 boreholes 500 points	58	73	58	1	67	7
	Hf2	45 boreholes 500 points	45	79	45	1	74	5
Gross Gamma	Hf3	22 boreholes 300 points	38	49	39	1	49	3
	CCU	32 boreholes 200 points	44	71	44	1	71	1
	Rwie	16 boreholes 300 points	136	83	137	8	63	20

- Geologic processes reflected in different layers: angles of anisotropy differ by geologic layer \rightarrow consider both 2D layers and 3D layers in future studies.
- Method results
 - MOI estimates closer to ground truth than CTI
 - MOI MAD values less than or equal to than CTI MAD values in all layers



Ringold Formation

- MOI estimated angles of anisotropy within geologic layer associated with Ringold Formation are consistent with inferred Columbia River drainage path (Reidel 2013)
 - Roughly 135° from the northwest corner of that map (north is 0°, east 90°, and northeast to southwest is 135°)
 - Validating accuracy of MOI estimate in that geologic layer
- Summary
 - 2D layered approach: MOI method consistently resulted in estimates similar to ground truth, whereas CTI method, while accurate in some layers, was inconsistent.
 - 3D volume approach: CTI method could not be applied due to limitations.



Findings & Conclusions

- Automating the selection of principal directions of anisotropy is particularly important for practitioners considering surveys for subsurface decommissioning and new borehole and well placement.
 - Both 2D layered approach and 3D volume approach should be examined.
 - Especially when geologic/other potential subsurface conditions could affect layers.
- PNNL assessed MOI and CTI as potential methods that could be implemented in VSP to assist with determining the principal directions of anisotropy.
 - Both methods performed well in structured datasets and produced less accurate estimates in unstructured datasets (simulations).
 - MOI produced more accurate and precise results than the CTI, when compared to ground truth, in simulated and real borehole data.
 - MOI method estimated primary and secondary angles of anisotropy within default angular tolerance (22.5°) applied in practice by most geostatistical software (Müller et al. 2022).
- MOI method is more reliable and versatile than CTI.

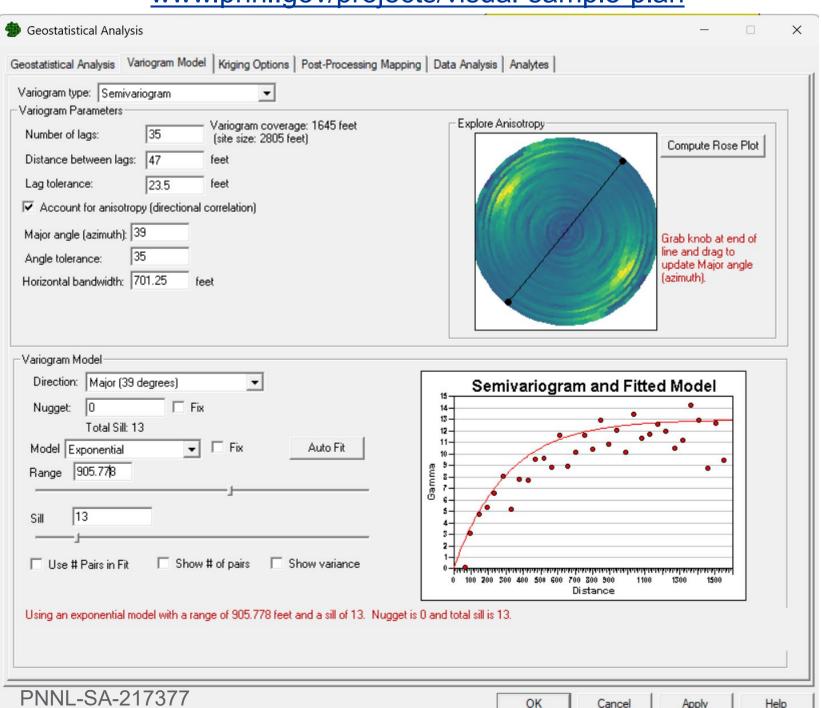


PNNL implementing MOI in VSP.

- Surface (2D) functionality is available in current VSP (Version 7.24, released 7/28/2025).
- Subsurface (3D) functionality is under development.

Visual Sample Plan (VSP)

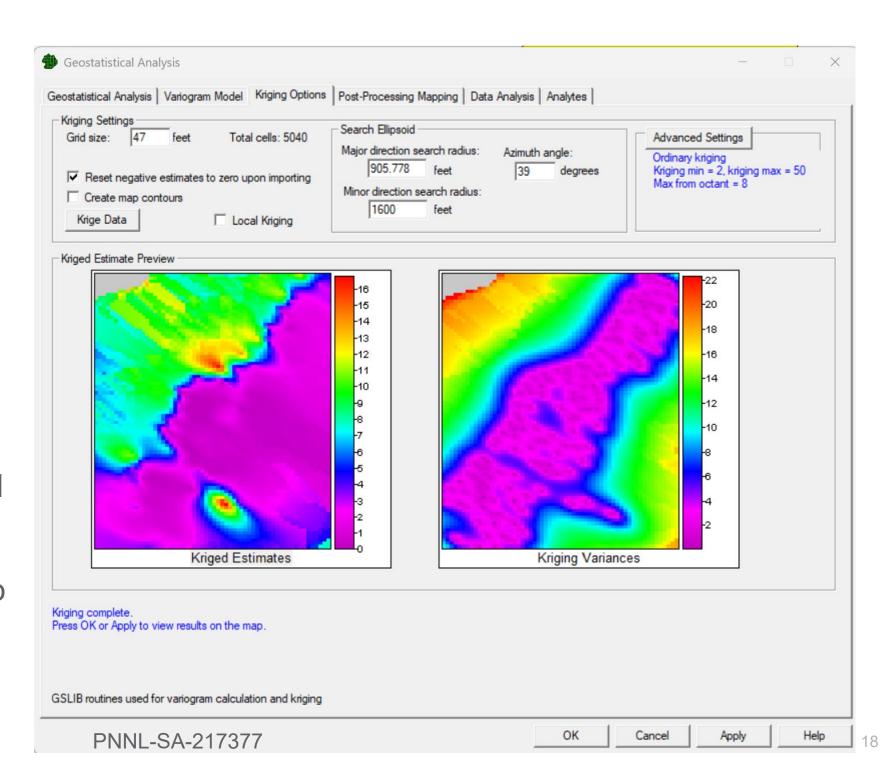
www.pnnl.gov/projects/visual-sample-plan





Functionality

- VSP calculates variograms in all directions to produce a rose plot diagram for visualization of anisotropy
- 2D MOI method automatically determines the principal angle of anisotropy, with option for the user to adjust based on visual inspection of the rose plot diagram
- VSP automatically calculates and fits variogram models for the major and minor directions of anisotropy, and allows the user to adjust the fit parameters for a better fit







Thank you

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