



# Indicator Simulation for Permeability Modeling

- Discuss the Problem of Accounting for Secondary Data
- Review Conventional Techniques
- Sequential Indicator Simulation
- Examples



# Indicator Coding of Data

- local hard indicator data  $i(\mathbf{u}_\alpha; z)$  originating from local hard data  $z(\mathbf{u}_\alpha)$ :

$$i(\mathbf{u}_\alpha; z) = \mathbf{1} \text{ if } z(\mathbf{u}_\alpha) \leq z, = \mathbf{0} \text{ if not}$$

- local hard indicator data  $j(\mathbf{u}_\alpha; z)$  originating from ancillary information that provides hard inequality constraints on the local value  $z(\mathbf{u}_\alpha)$ . If  $z(\mathbf{u}_\alpha) \in (a_\alpha, b_\alpha]$ , then:

$$j(\mathbf{u}_\alpha; z) = \left\{ \begin{array}{l} 0 \text{ if } z \leq a_\alpha \\ \text{undefined (missing) if } z \in (a_\alpha, b_\alpha] \\ 1 \text{ if } z > b_\alpha \end{array} \right\}$$

- local soft indicator data  $y(\mathbf{u}_\alpha; z)$  originating from ancillary information providing prior (pre-posterior) probabilities about the value  $z(\mathbf{u}_\alpha)$ :

$$y(\mathbf{u}_\alpha; z) = \mathbf{Prob} \{Z(\mathbf{u}_\alpha) \leq z \mid \text{local information}\}$$

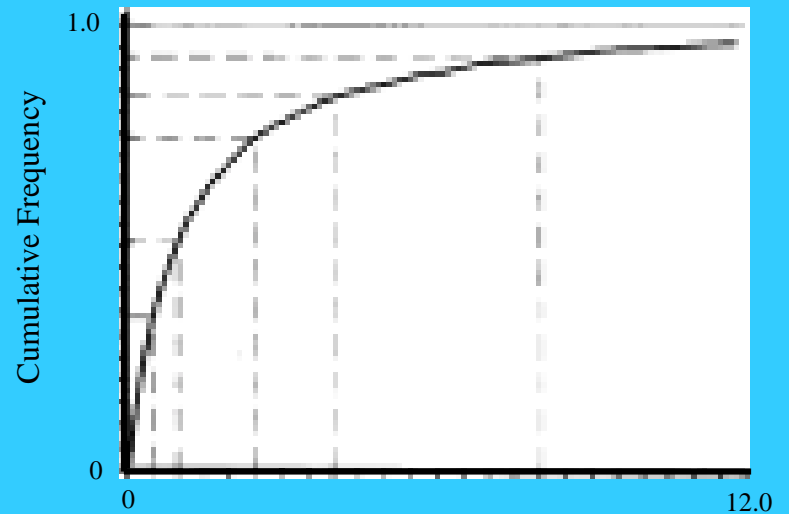
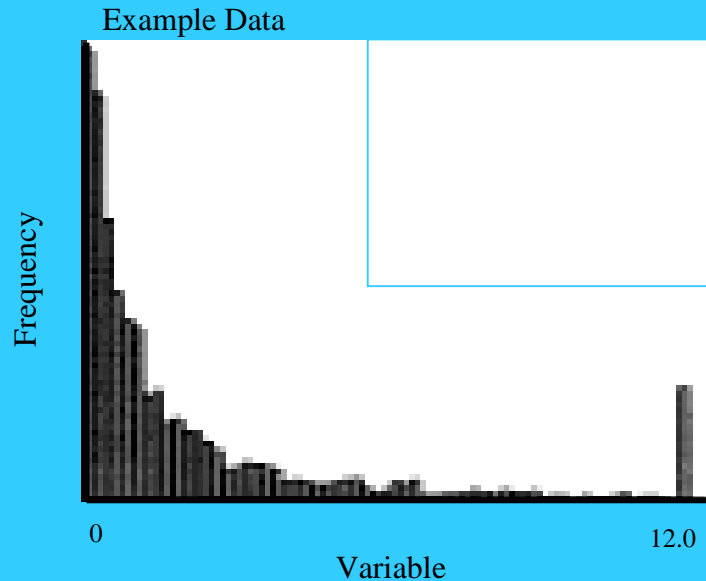
$\in [0,1]$ , and  $\neq F(z)$  : global prior as defined hereafter

- *global* prior information common to all locations  $\mathbf{u}$  within the stationary area  $A$ :

$$F(z) = \mathbf{Prob}\{Z(u) \leq z\}, \forall u \in A$$



# Indicator Kriging (IK)



- Build a local cdf conditional to surrounding data
- Can compute any of the following:
  - $E$ -type estimate (local conditional mean)
  - Probability to exceed a threshold  $z$
  - $z$ -value of any probability
  - probability intervals
  - truncated statistics



# Markov Bayes Model (1)

The IK process can be seen as a Bayesian updating of the local prior cdf into a posterior cdf using information supplied by neighboring local prior cdf's

$$[\text{Prob}\{Z(u) \leq z | (n + n')\}]_{IK}^* = \lambda_0(u)F(z) + \sum_{\alpha=1}^n \lambda_{\alpha}(u; z)i(u_{\alpha}; z) + \sum_{\alpha'=1}^{n'} v_{\alpha'}(u; z)y(u'_{\alpha'}; z)$$

The  $\lambda_{\alpha}(u; z)$ 's are the weights attached to the  $n$  neighboring hard indicator data, the  $v_{\alpha}(u; z)$ 's

are the weights attached to the  $n'$  neighboring soft indicator data, and  $\lambda_0$  is the weight attributed to the global prior cdf. To ensure unbiasedness,  $\lambda_0$  is usually set to:

$$\lambda_0(u) = 1 - \sum_{\alpha=1}^n \lambda_{\alpha}(u; z) - \sum_{\alpha'=1}^{n'} v_{\alpha'}(u; z)$$

This can be seen as an indicator cokriging that pools information of different types: the hard  $i$  and  $j$  indicator data and the soft  $y$ -prior probabilities.



# Markov Bayes Model (2)

The Markov-Bayes model is a model whereby the matrix of covariances is given by the following model.

$$\begin{aligned}C_{IY}(\mathbf{h}; \mathbf{z}) &= B(\mathbf{z})C_I(\mathbf{h}; \mathbf{z}), \quad \forall \mathbf{h} \\C_Y(\mathbf{h}; \mathbf{z}) &= B^2(\mathbf{z})C_I(\mathbf{h}; \mathbf{z}), \quad \forall \mathbf{h} > \mathbf{0} \\&= |B(\mathbf{z})|C_I(\mathbf{h}; \mathbf{z}), \quad \mathbf{h} = \mathbf{0}\end{aligned}$$

The coefficients  $B(\mathbf{z})$  are obtained from calibration of the soft  $y$ -data to the hard  $z$ -data; more precisely:

$$B(\mathbf{z}) = m^{(1)}(\mathbf{z}) - m^{(0)}(\mathbf{z}) \in [-1, +1]$$

with:

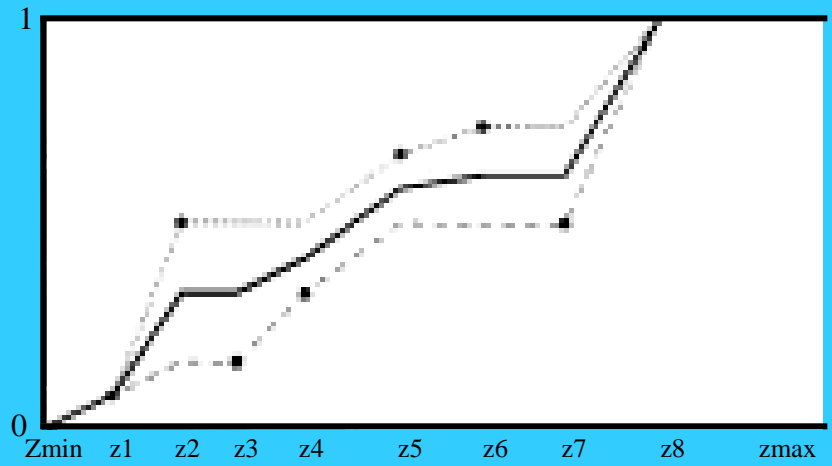
$$m^{(1)}(\mathbf{z}) = E\{y(\mathbf{u}; \mathbf{z}) \mid I(\mathbf{u}; \mathbf{z}) = \mathbf{1}\}$$

$$m^{(0)}(\mathbf{z}) = E\{y(\mathbf{u}; \mathbf{z}) \mid I(\mathbf{u}; \mathbf{z}) = \mathbf{0}\}$$

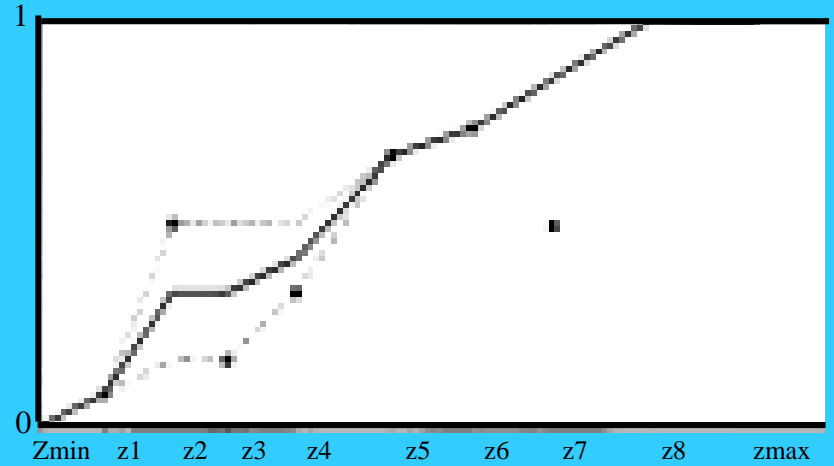


# Order Relations Corrections

Order Relations Correction

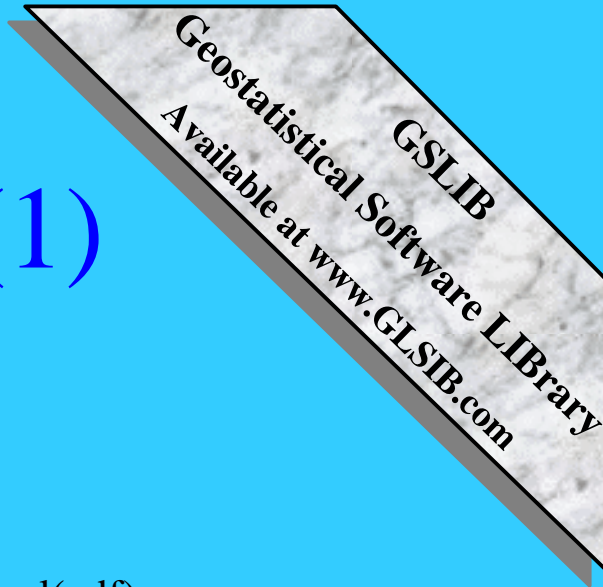


Order Relations Correction  
(no data in class  $z_6$  and  $z_7$ )





# Parameter File (1)



## Parameters for SISIM \*\*\*\*\*

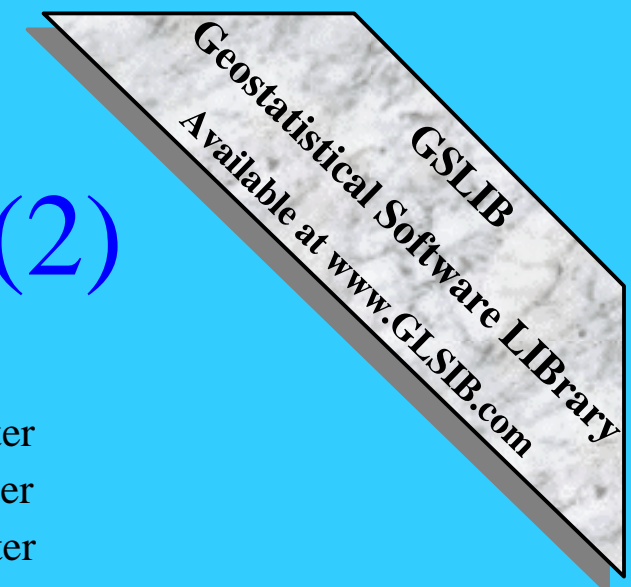
### START OF PARAMETERS:

```

1          \ 1=continuous(cdf), 0=categorical(pdf)
5          \ number thresholds/categories
0.5  1.0  2.5  5.0  10.0  \ thresholds / categories
0.12 0.29 0.50 0.74 0.88  \ global cdf / pdf
../data/cluster.dat      \ file with data
1  2  0  3              \ columns for X,Y,Z, and variable
direct.ik                \ file with soft indicator input
1  2  0  3 4 5 6 7      \ columns for X,Y,Z, and indicators
0                        \ Markov-Bayes simulation (0=no,1=yes)
0.61 0.54 0.56 0.53 0.29 \ calibration B(z) values
-1.0e21  1.0e21         \ trimming limits
0.0  30.0              \ minimum and maximum data value

```

...



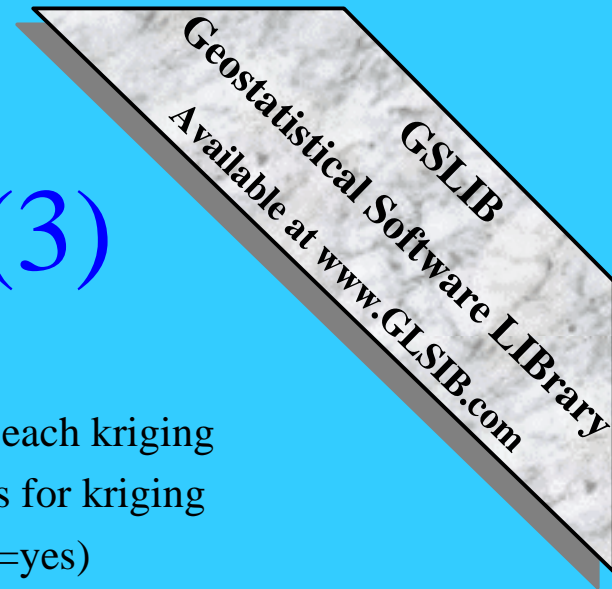
# Parameter File (2)

1	0.0	\ lower tail option and parameter
1	1.0	\ middle option and parameter
1	30.0	\ upper tail option and parameter
cluster.dat		\ file with tabulated values
3 0		\ columns for variable, weight
0		\ debugging level: 0,1,2,3
sisim.dbg		\ file for debugging output
sisim.out		\ file for simulation output
1		\ number of realizations
50 0.5 1.0		\ nx,xmn,xsiz
50 0.5 1.0		\ ny,ymn,ysiz
1 1.0 10.0		\ nz,zmn,zsiz
69069		\ random number seed
12		\ maximum original data for each kriging
12		\ maximum previous nodes for each kriging...





# Parameter File (3)



12	\ maximum previous nodes for each kriging
1	\ maximum soft indicator nodes for kriging
1	\ assign data to nodes? (0=no,1=yes)
0 3	\ multiple grid search? (0=no,1=yes),num
0	\ maximum per octant (0=not used)
20.0 20.0 20.0	\ maximum search radii
0.0 0.0 0.0	\ angles for search ellipsoid
0 2.5	\ 0=full IK, 1=median approx. (cutoff)
0	\ 0=SK, 1=OK
1 0.15	\ One nst, nugget effect
1 0.85 0.0 0.0 0.0	\ it,cc,ang1,ang2,ang3
10.0 10.0 10.0	\ a_hmax, a_hmin, a_vert
1 0.10	\ Two nst, nugget effect