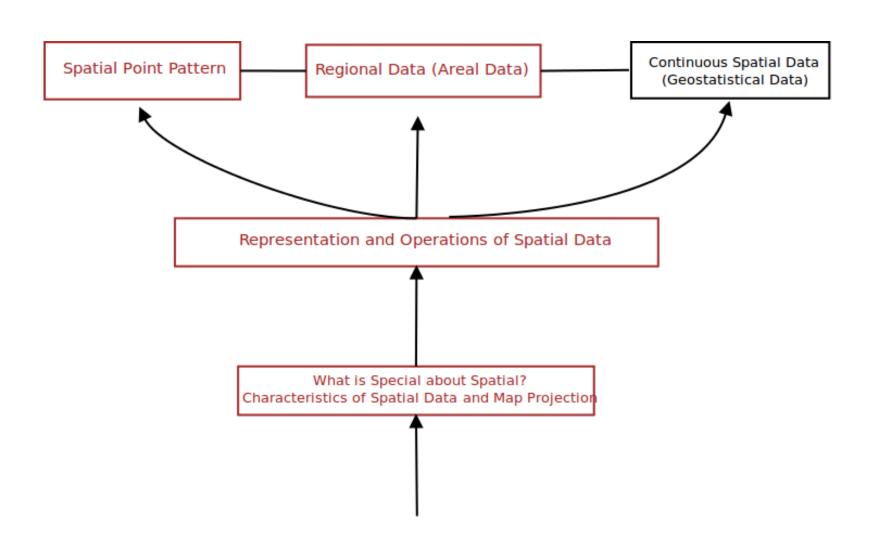
# Spatial Analysis and Modeling (GIST 4302/5302)

Guofeng Cao
Department of Geosciences
Texas Tech University

#### Outline of This Week

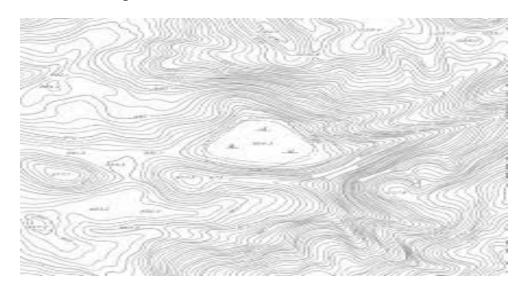
- Last topic, we learned:
  - Spatial autocorrelation of areal data
  - Spatial regression
- This topic, we will learn:
  - Spatial fields
  - Interpolation
    - Deterministic interpolation
    - -Geostatistics (Kriging family of methods)



## Spatial Fields

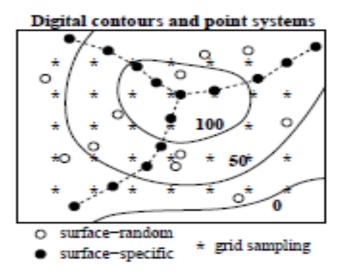
- Scalar versus vector fields:
  - scalar: quantity characterized only by its magnitude
    - scalar fields have a single value associated with each location
    - examples: temperature, elevation, precipitation
  - vector: quantity characterized by its magnitude and orientation (e.g., wind speed and direction)
    - vector fields have multiple values associated with each location
    - examples: <a href="http://hint.fm/wind/">http://hint.fm/wind/</a>

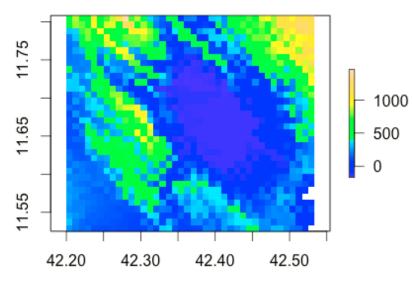
#### Surface Representation: Contours



- accuracy of digital sample depends on scale and accuracy of source analog map
- details falling between contour lines are lost
- oversampling of steep slopes (many contours) relative to gentle ones (few contours)
- many surface processing operations (e.g., slope calculation or point value determination) are extremely difficult to automate

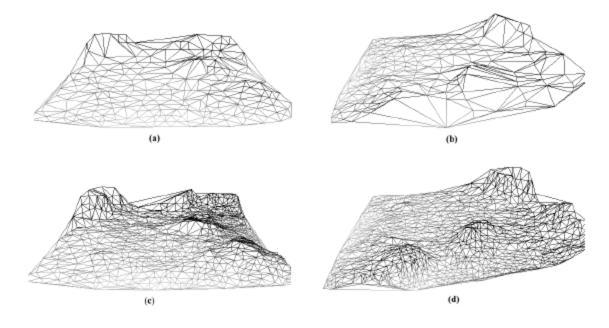
#### Surface Representation: Point Systems





- uniform data density enables display and surface processing
- no need to store spatial coordinates, just a single point and the grid spacing and orientation
- much larger sample size is required to enhance details (spatial resolution)
- Details/accuracy is controlled by the cell size/resolution
- Value of each cell is homogeneous represented by one single value

# Surface Representation: Triangulated Irregular Network



- extremely compact way of storing fields, and their properties (e.g., slope, aspect)
- can capture important surface characteristics
- accuracy depends on accuracy of underlying field (assumed known)

## Sampling Spatial Fields

#### Sampling schemes:

- collection of measurements at a set of locations(e.g., precipitation at rain gauges, elevation spot heights)
- regular grids obtained from aerial and/or satellite remote sensing (such measurements are area integrals)
- digitized contour maps = points from digitized
   contours derived from analog topographic maps

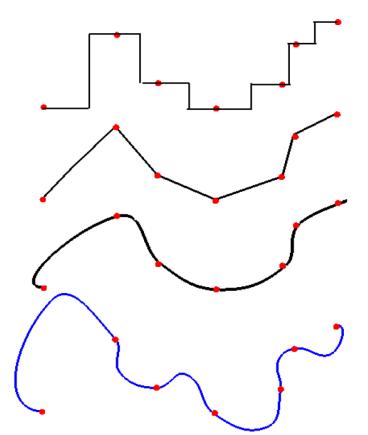
### Sampling Spatial Fields

#### Issues to consider:

- data constitute a sample of the underlying continuous field (exhaustive sampling is almost always impossible)
- measurements might have both spatial and temporal components
- often, data are not collected at random ⇒ biased and non-random sampling
- sometime, contours maps are derived from spot heights ⇒ digitized contour maps should be treated with caution
- All measurements are subjective to <u>uncertainty</u> (spatial uncertainty, more in the next lecture)

#### Spatial Interpolation

- Why spatial interpolation:
  - Observations/samples are sparse
- Interpolation: discrete->continuous
- Underline Rationale
  - Again, TFL
- It is difficult



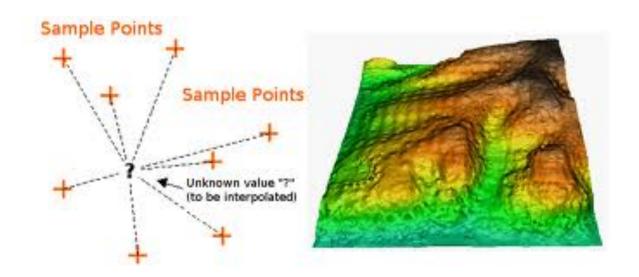
### **Spatial Interpolation**

#### General formulation of spatial interpolation

unknown value  $z(s_i)$  at any non-sampled location  $s_i$  expressed as weighted average of n sample data  $\{z(s_\alpha), \ \alpha = 1,...n\}$ :

$$z(s_i) = \sum_{\alpha=1}^n w_{i\alpha} z(s_{\alpha})$$

 $w_{i\alpha}$  denotes weight given to datum  $z(s_{\alpha})$  for prediction at location  $s_i$ 

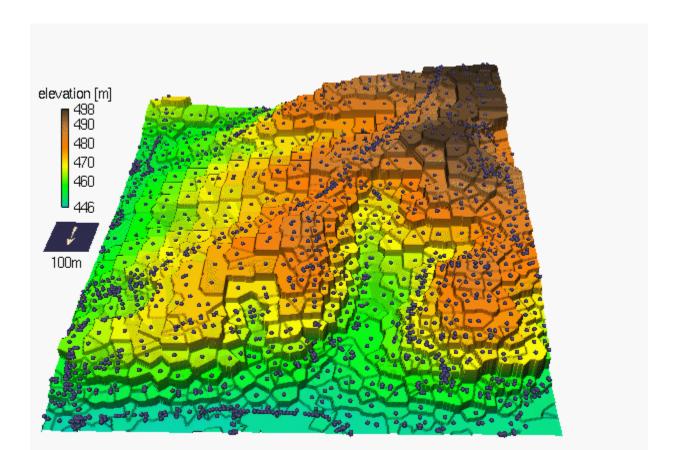


## Spatial Interpolation Methods

- Deterministic Interpolators
  - Nearest Neighbor/Natural neighbor
  - Trend Surface
  - Inverse distance weighted method
  - Spatial spline
  - Triangulation
- Stochastic Interpolators
  - Kriging
  - Outcome the credibility information compared to the deterministic interpolators

# Spatial Interpolation: Nearest Neighbor

- Assign value of nearest sample point
- Thiessen Polygons/Voronoi diagram



# Spatial Interpolation: Nearest Neighbor

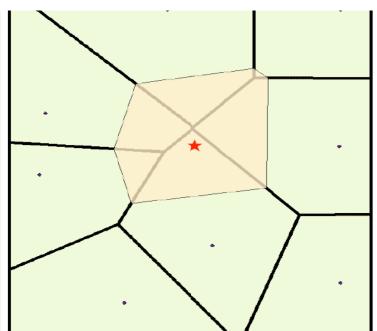
 Datum closest to the prediction location receives all weights

$$z(s_i) = \sum_{\alpha=1}^n w_{i\alpha} z(s_{\alpha}) = z(s_{\alpha}) + \sum_{\alpha=1}^{n-1} 0 * z(s_{\alpha})$$

- Unbiased estimation  $\sum_{\alpha=1}^{n} w_{i\alpha} = 1$
- set of predicted values form discontinuous (patchy) surface

### Natural Neighbor Interpolation

- Finds the closest subset of input samples to a query point and applies weights to them based on proportionate areas in order to interpolate a value
- "Area-stealing"
- Local interpolation: using only a subset of samples that surround a query point

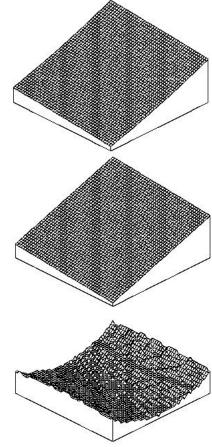


## Spatial Interpolation: Trend Surface

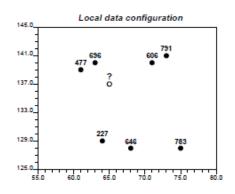
 explicit mathematical function(s) of coordinates that interpolates or approximates (smooths) the surface. For example:

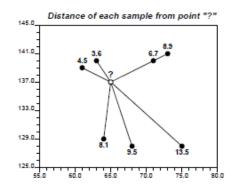
$$z(s_i) = a_0 + a_1 * x + a_2 * y$$
or
$$z(s_i) = a_0 + a_1 * x^2 + a_2 * y^2 + a_3 xy$$

- surface operations (e.g., curvature) and values can be analytically computed
- Fit polynomial equation to sample points
- Goal is to minimize deviations between sample points and surface
- arbitrary choice of number and type of functions
- local versus global fitting



#### Spatial Interpolation: Inverse Distance





#### Procedure:

• predict unknown value  $z(\mathbf{s}_i)$  at any non-sampled location  $\mathbf{s}_i$  as weighted linear combination of  $n(\mathbf{s}_i)$  nearby data  $z(\mathbf{s}_{\alpha})$ :

$$\hat{z}(\mathbf{s}_i) = \sum_{\alpha=1}^{n(\mathbf{s}_i)} w_{i\alpha} z(\mathbf{s}_\alpha)$$

where  $w_{i\alpha}$  denotes weight received by sample  $z(\mathbf{s}_{\alpha})$  for prediction at location  $\mathbf{s}_i$ 

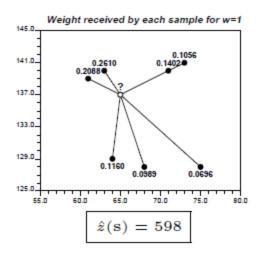
• make weight  $w_{i\alpha}$  inversely proportional to power k of distance  $h_{i\alpha} = ||\mathbf{s}_i - \mathbf{s}_{\alpha}||$ :

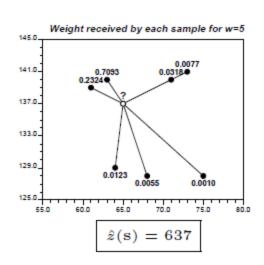
$$w_{i\alpha} = \frac{h_{i\alpha}^{-k}}{\sum_{\alpha=1}^{n(\mathbf{s}_i)} h_{i\alpha}^{-k}} = \frac{1/h_{i\alpha}^k}{\sum_{\alpha=1}^{n(\mathbf{s}_i)} 1/h_{i\alpha}^k}$$

#### Spatial Interpolation: Inverse Distance

#### **Characteristics:**

- unbiased interpolation procedure, since  $\sum_{\alpha=1}^{n(\mathbf{s}_i)} w_{i\alpha} = 1$
- "exact" interpolator:  $\hat{z}(\mathbf{s}_{\alpha}) = z(\mathbf{s}_{\alpha}), \ \forall \alpha$
- exponent k controls importance of data closer to  $s_i$ ; e.g., k=2: inverse distance squared interpolation



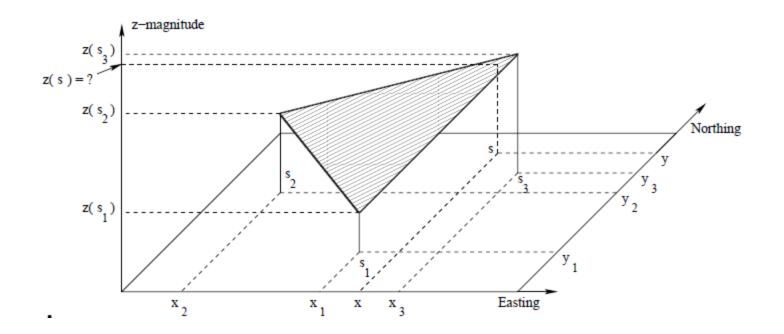


#### Spatial Spline

- Estimates values using a mathematical function that minimizes overall surface curvature
  - smooth surface
  - passes exactly through the input points

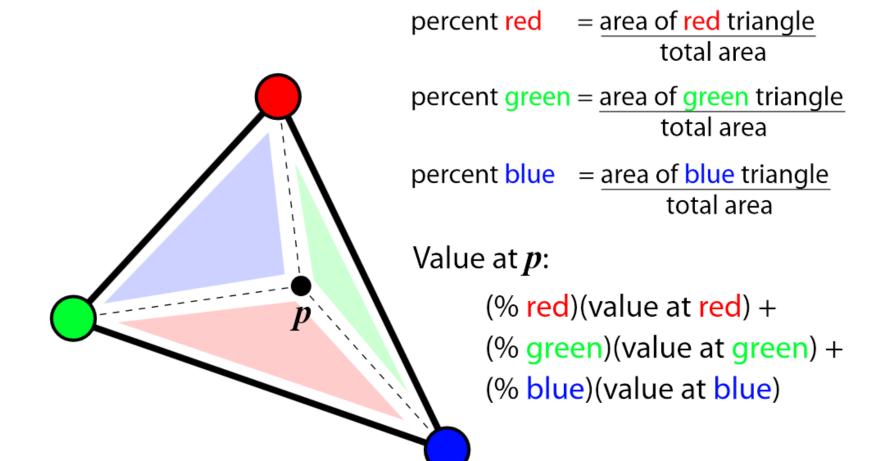
## Spatial Interpolation: Triangulation

Barycentric Interpolation



## Spatial Interpolation: Triangulation

Barycentric Interpolation



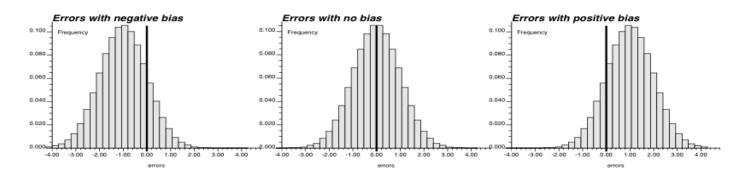
### **Evaluating Prediction Performance**

#### Cross-validation:

- Loop over sample locations:
- hide a sample datum
- predict it from the remaining data using one of the spatial interpolation method
- repeat until all sample locations are visited and crossvalidation predictions are computed

## **Evaluating Prediction Performance**

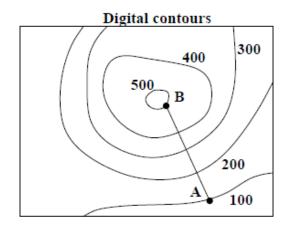
- Compare distribution of predicted values to that of true values for:
  - reproduction of mean (for possible bias), median,
     variance, and other summary statistics
  - reproduction of entire distribution of true values (QQ plot)

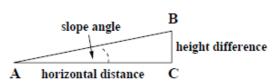


#### Surface Derivatives: Slope and Gradient

#### Gradient:

- vector quantity specified by (i) magnitude, and (ii) direction
- gradient magnitude = maximum rate of change of elevation at a point (slope)
- gradient direction = direction of steepest slope trough that point (aspect)





calculating the tangent of the slope angle:

$$\tan(\theta) = \frac{\text{height difference}}{\text{horizontal distance}} = \frac{BC}{AC} \Rightarrow \theta = \arctan(\frac{BC}{AC})$$

in Matlab: theta=rad2deg(atan(BC/AC))

#### Surface Derivatives: Slope and Gradient

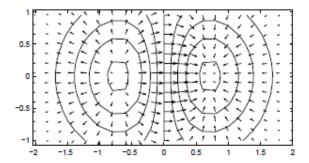
#### Gradient calculations:

- in TIN surface representation, gradient at s = gradient of containing Delaunay triangle
- in raster surface representation, gradient at s calculated using a square window (typically  $9 \times 9$ ) centered at s. Slope  $\theta$  and aspect  $\alpha$  are calculated as:

$$\theta = \sqrt{\theta_x^2 + \theta_y^2}$$
 and  $a = \arctan(\frac{\theta_x^2}{\theta_y^2})$ 

where  $\theta_x$  and  $\theta_y$  denote directional derivatives along x and y aspect  $\alpha$  measured from vertical to direction of steepest slope;  $\alpha = \alpha + 180$  if  $\theta_y > 0$ , and  $\alpha = \alpha + 360$  if  $\theta_x > 0$  and  $\theta_y < 0$ 

 alternatively, a local mathematical surface is fitted within each window, and its derivative is analytically calculated



• End of this topic