

# Spatial Sampling Design and Strategies

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## Outline

- 1 Overview
- 2 Regular Grid Designs
- 3 Cyclic Sampling Designs
- 4 Design of Experiment

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## Spatial sampling design

- Example: a study of old-growth northern hardwood forests (Miller et al., 2002).
  - Consideration of biodiversity in natural resource management.
  - Spatial patterns of forest understorey vegetation (herbs, shrubs, tree seedlings, saplings).
  - Different species exhibit different spatial patterns within a given environment?
  - Biotic and abiotic factors in the environment are related to a species' spatial pattern?
- An important question: *where* should data be collected?
- The purpose is to design a sampling scheme that ensures scientific objectivity.

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## Spatial sampling design

- Suppose the study area of interest is  $D$ .
- Suppose measurements of  $Z$  will be taken at locations  $\mathbf{s}_1, \mathbf{s}_2, \dots, \mathbf{s}_n$  in  $D$ , where  $\mathbf{s} = (x, y)$  and  $n$  is the sample size. Where should they be?
  - It depends!
  - Possible objectives
    - Estimation of mean (e.g. average soil P in a field)
    - Estimation of variogram (e.g. map of soil P in the field)
    - Comparison of treatments (e.g. effect of a new fertilizer)
  - Possible prior information
    - Accessible study area and sampling locations
    - Affordable sample size
    - Condition of a study area

## Related subjects

- Survey sampling: design-based sampling versus model-based sampling (Grujiter and Braak, 1990; Särndal et al., 1992)
- Design of experiment and optimal design (Mead et al., 1993)
- Spatial sampling design and optimal sampling (Webster and Oliver, 2001)
- An excellent review article: Stein and Christien (2003)

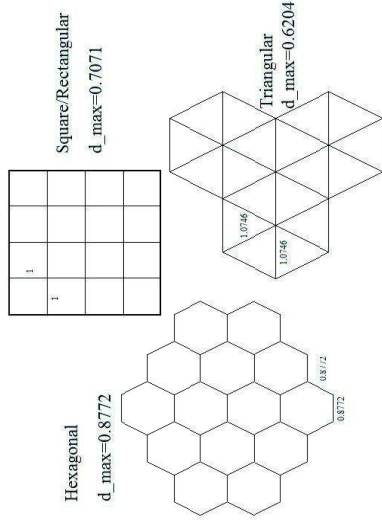
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## Regular grids

- Triangular or isometric grid: tiling plane regularly with equilateral triangles.
- Rectangular grid: tiling plane regularly with squares.
- Hexagonal grid: tiling plane regularly with hexagons.

## Regular grids



$d_{\max}$  = maximum distance from any point in  $D$  to the nearest grid point.

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## And the winner is...

- A triangular grid is the most efficient design with the smallest  $d_{\max}$ .
- That is, for the same sampling intensity, it places the sampling locations as far apart as possible while minimizing the area that is under-represented.
- A triangular grid is most efficient for most bounded variograms that have finite ranges.

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## A plausible scenario

- The goal is to estimate the overall mean  
$$\mu = E(Z).$$
- Assume a regular spatial sampling grid with a fixed sampling density.
- Assume an exponential semivariogram for the spatial correlation structure.

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## Remarks

- Under some other assumptions, a hexagonal grid is the most efficient design.
- For convenience, a rectangular grid is often the preferred design in field work.
- Major drawbacks of a regular grid include poor variogram estimates at short distances and the potential problems of systematic design (as versus randomized design).

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## Main idea

- To compensate for poor variogram estimates using regular grid designs, an improved method was proposed by Clayton and Hudelson (1995).
- The main idea is to use a regular grid system, but sample at unequal spacings.
- In one dimension (1D), the design allows the estimation of variogram at all multiples of the smallest lag with a minimum number of sampling locations.

## 1D transect

- Let  $\times$  = sample;  $\circ$  = skip (sampling).
- A 3/7 cyclic sampling design with 2 repeats looks like:  
with lag distances  

$\times$	$\times$	$\circ$	$\times$	$\circ$	$\circ$	$\times$	$\times$	$\circ$	$\times$	$\circ$	$\circ$	$\circ$
$\times$	1	-	3	-	-	-	7	-	-	-	-	-
-	$\times$	-	2	-	-	-	6	7	-	-	-	-
-	-	-	$\times$	-	-	-	4	5	-	7	-	-

## 1D transect

- Choice of specific sampling pattern is important.
- Why not  
with lag distances  

$\times$	$\times$	$\times$	$\circ$	$\circ$	$\circ$	$\circ$	$\times$	$\times$	$\circ$	$\circ$	$\circ$	$\circ$
$\times$	1	2	-	-	-	-	7	-	-	-	-	-
-	$\times$	1	-	-	-	-	6	7	-	-	-	-
-	-	-	$\times$	-	-	-	5	6	7	-	-	-
- Lag distances 3 and 4 are missed.

## Remarks

- For each lag distance, the proposed 3/7 design gives enough data for making the confidence intervals of the variogram small.
- There are more 7-lag distances than others in a 3/7 design, which cannot be avoided.
- Other possible cyclic sampling designs are: 4/13, 5/21, 6/31, etc. (Clinger and Ness, 1976).

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## 2D region

- Extension to a 2D region is straightforward, but only approximately optimal.
- A 3/7 design for both the x-axis and y-axis:  

X	X	O	X	O	O	X	X	O	X	O	O	O
X	X	O	X	O	O	X	X	O	X	O	O	O
O	O	O	O	O	O	O	O	O	O	O	O	O
X	X	O	X	O	O	X	X	O	X	O	O	O
O	O	O	O	O	O	O	O	O	O	O	O	O
O	O	O	O	O	O	O	O	O	O	O	O	O
O	O	O	O	O	O	O	O	O	O	O	O	O
- One can have different cyclic sampling designs for rows and columns.
- See Miller et al. (2002) for more details of the understory vegetation example.

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## Sampling design in practice

In practice, how to choose a particular cyclic sampling design and hence the sample size?

- 1 Conduct a pilot study to obtain a rough estimate of the range, sill, and nugget.
- 2 Simulate data on a grid with the finest grain scale possible for sampling, based on the estimated range, sill, and nugget.
- 3 Sample from the simulated data according to different sampling designs.
- 4 For each sample, compute the fitted range, sill, and nugget, and the confidence intervals of the variogram.
- 5 Evaluate the effect of different designs on the confidence interval width.
- 6 Consult a statistician!

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## Example: Nitrogen cycling

- Assume exponential variogram model with
  - $r = 2$  (i.e. 95% effective  $r = 6$ ).
  - $r = 1$  (i.e. 95% effective  $r = 3$ ).
- Assume a  $25 \times 25$  grid structure at a 2-m increment.
- Compare the use of 2D 3/7 cyclic sampling design with 1, 2, or 3 repeats.

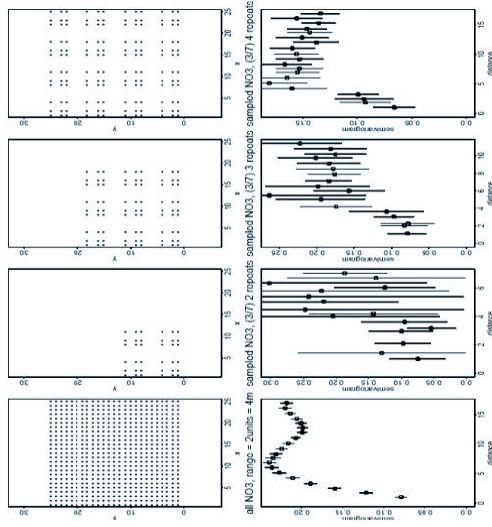
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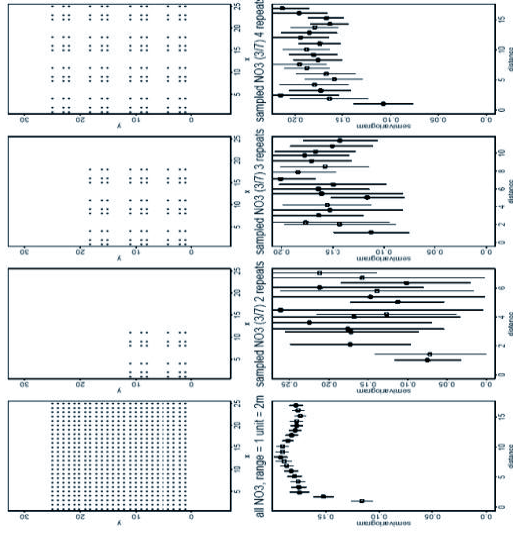
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$r = 2$



$r = 1$



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## Main idea

- In many field experiments, blocking is used to account for experimental unit (EU) heterogeneity, assuming that EUs within block homogeneous.
- Often there is spatial correlation within a block.
- If the goal is to have equal precision for the tests of treatment differences, it would make sense to design the experiment so that all treatments are equally near each other.

## Example

Block	Treatment			
1	C	A	B	D
2	B	D	C	A
3	A	B	D	C
4	D	B	C	A

Distance between plots

Contrast	Block 1	Block 2	Block 3	Block 4	Average
A vs B	1	3	1	2	1.75
A vs C	1	1	3	1	1.50
A vs D	2	2	2	3	2.25
B vs C	2	2	2	1	1.75
B vs D	1	1	1	1	1.00
C vs D	3	1	1	2	1.75
Average					1.67

## Nearest neighbor approach

- Instead of distance, look at neighbors of each treatment:  
A vs B: 2 A vs C: 3 A vs D: 0  
B vs C: 1 B vs D: 4 C vs D: 2
- Similar problem as before. While switching block 4 would help, we can do better.
- There are 12 neighbor pairs and 6 trt pairs:
 

Block	Treatment			
1	C	A	B	D
2	B	D	A	C
3	A	D	C	B
4	D	C	B	A
- Arrangement above is balanced for nearest neighbors and distance.
- Often correlation in both directions (2D). Similar approaches apply.

## Average distance balanced design

- Not a balanced design since some treatments are on average closer than others.
- Simple switch in block 4 to DACB would result in much closer average distances.
- A strategy may be to strive for an average distance balanced design.

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