

Spatial Data and Analysis

Discussion 2

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Admin stuff

1. Check the discussion syllabus in the website
2. Remember to use “[spatial]” in the subject header
3. Check discussion slides, there is useful stuff

Outline

1. Miscellaneous
2. Logic
3. Loops
4. Monte Carlo
5. School Choice
6. Additional

Miscellaneous #1: vector and locations

- ▶ For simplicity think of a vector as a $n \times 1$ matrix: $\vec{a} = \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix}$
- ▶ N locations can be described by 2 vectors of size N :

$$(\vec{x}, \vec{y}) = \begin{bmatrix} x_1 & y_1 \\ \vdots & \vdots \\ x_n & y_n \end{bmatrix}$$

where (x_i, y_i) is a single location with $i = 1, \dots, N$

- ▶ Random locations: (\vec{x}, \vec{y}) is $[\text{randn}(N, 1) \quad \text{randn}(N, 1)]$

Miscellaneous #2: matlab v/s stata

- ▶ Matlab reads the code `plot(A,B)` and plots A in the **x-axis** and B in the **y-axis** \sim `plot(x,y)`
- ▶ Stata reads the code `scatter A B` and plots A in the **y-axis** and B in the **x-axis** \sim `scatter y x`
- ▶ In a way, matlab thinks more mathematically and stata more statistically.
- ▶ When plotting locations in matlab: `plot(lon,lat)`, not the perhaps more intuitive `plot(lat,lon)`

Logical operators

- ▶ Basic logical operators we will use repeatedly:

EQUAL	:	=
GREATER	:	>
GREATER OR EQUAL	:	>=
AND	:	&
OR	:	
IS NOT	:	~=

- ▶ Useful to create indicator variables and others

Example

► Examples of logical operators:

```
1  % Parameters
2      A = [1 0 1];
3      B = [0 0 1];
4      C = [0 4 0];
5
6  % Operations
7      C = (A==1 & B==1) ;
8      D = (A==1 | B==1) ;
9      E = A (A >1)      ;
10     F = C (C >1)      ;
```

Example (cont.)

- ▶ You very rarely have to loop through a whole matrix to get only specific parts of it
- ▶ Exercise: find an approximate value for the mean of negative numbers in a normal distribution with 1,000 values

```
1      X = randn(1000,1) ;  
2      mean(X(X<0)) ;
```


if

- Sometimes `if` statements can be useful:

```
1  % Parameters
2      N = 100                      ;
3      A = ceil(randn(N,1));
4      B = NaN(N,1)                 ;
5
6  % If statement
7      if A(10, 1) == 1
8          disp('Hi everyone')
9      else disp('Its hot')
10     end
```

Comments

- Note that the command `==` is used when using a statement, and `=` when assigning values

- Short circuit behavior:

`A & B` : `A` and `B` are evaluated

`A && B` : `B` only evaluated if `A` is true

- Multiple statements need to be separated by parentheses:

```
1  % Statement 1
2      (x == 3 & x > 4) | x < 4
3  % Statement 2
4      x == 3 & (x > 4 | x < 4)
```

Loops

- ▶ Loops allow you to execute a block of code many times. For example, create a vector `x` with ten ones using a loop:

```
1  for i = 1 : 10
2      x(i, 1) = 1 ;
3  end
```

- ▶ However, this is obviously inefficient since we can create the same vector simply typing `x=ones(10,1)`.
- ▶ Knowing when to use a loop is an important part of being efficient at coding. Indentation is important for readability

Example

- ▶ Create a fake dataset with panel structure where variables are 100 ID's and 10 years:

ID	Year
1	1
⋮	⋮
1	10
⋮	⋮
100	1
⋮	⋮
100	10

Solution

```
1  % Parameters
2      ID = 100          ;
3      t  = 10           ;
4      x  = NaN(ID*t, 2) ;
5
6  % Loop
7      for i = 1:ID
8          x((i-1)*10+1:i*10, 1) = i      ;
9          x((i-1)*10+1:i*10, 2) = 1:10  ;
10     end
```

Monte Carlo

- ▶ Monte Carlo is an algorithm that uses random sampling to obtain numerical results
- ▶ This method is useful in applied statistics to obtain quick answers to a wide range of questions
- ▶ Consider $y_i = 5 + 0.5x_i + \varepsilon_i$. Then:
 1. what is the distribution of an OLS estimator using (y, x) ?
 2. what are the consequences of classical measurement error in x ?
- ▶ Assume $x \sim N(0, 1)$, $\varepsilon \sim N(0, 1)$, $x \perp \varepsilon$ and different magnitudes for the measurement error

Structure and parameters

- Think about the following structure to answer these questions:
define parameters, procedure, output

```
1  % Parameters
2      clear ; clc
3      s = 1000      ;
4      N = 100       ;
5      b = NaN(s,1);
6      g = NaN(s,1);
```

- Always useful to think about what is the output of the process you're writing

```

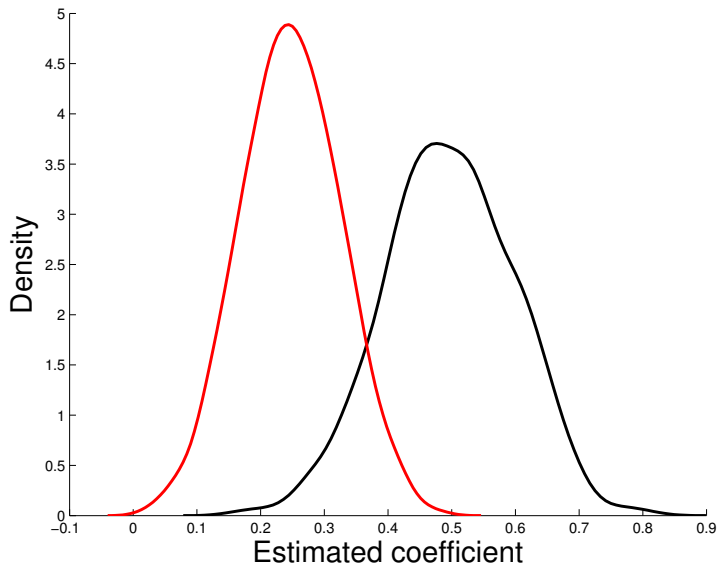
1  % Loop
2      for i = 1:s
3          % Variables
4              x = randn(N, 1) ;
5              z = x - randn(N,1) ;
6              y = 5 + .5 * x + randn(N,1);
7          % OLS coefficients
8              X = [ones(N,1) x] ;
9              beta = (X' * X) \ (X' * y) ;
10             b(i,1) = beta(2, 1);
11         % Attenuated OLS coefficients
12             Z = [ones(N,1) z] ;
13             gamma = (Z' * Z) \ (Z' * y) ;
14             g(i,1) = gamma(2, 1);
15     end

```


Output

```
1  % Figure
2      [a,b] = ksdensity(b) ;
3      [c,d] = ksdensity(g) ;
4
5      hold on
6      plot(b, a, 'k', 'LineWidth', 2)
7      plot(d, c, 'r', 'LineWidth', 2)
8          ylabel('Density'      , 'FontSize' , 20)
9          xlabel('Coefficient', 'FontSize' , 20)
10         box off
11         set(gcf, 'Color', 'w')
12     hold off
13
14     export_fig 'ols.pdf'
```

Figure: *Distribution of OLS coefficient and attenuation bias*



Application: school choice

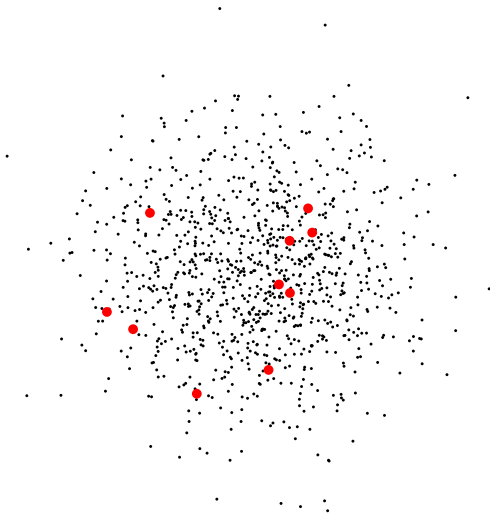
- ▶ Consider a simple model of school choice in which students enroll in the school that is closer to their home
- ▶ As researcher you are worried that students' locations could be measured with error
- ▶ This is crucial if we want to predict how many students will enroll in each school (e.g., construction of schools)
- ▶ Use point processes, distances, Monte Carlo and the model of school choice to explore the consequences of measurement error

```

1  % Parameters
2      N      = 1000      ; % students
3      S      = 10       ; % schools
4      x      = 0:.01:1   ; % size of noise
5      ERROR = NaN(N, size(x, 2)) ;
6  % Location of schools
7      lat_s = randn(S,1) ;
8      lon_s = randn(S,1) ;
9      LOC_s = [lon_s lat_s] ;
10 % Location of students
11     lat_n = randn(N,1) ;
12     lon_n = randn(N,1) ;
13     LOC_n = [lon_n lat_n] ;
14 % Distance and school choice
15     D = pdist2(LOC_n, LOC_s, 'euclidean') ;
16     [d, c] = min(D,[],2) ;

```

Figure: *Schools and students in space*



```

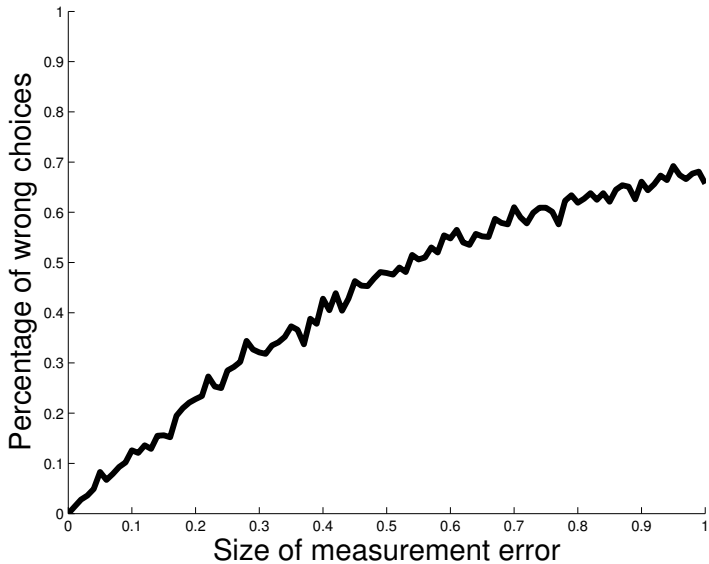
1  % Loop for different types of noise
2      for i = 1:size(x,2)
3          % Noisy location of students
4              lat_n_n = lat_n + x(1,i)*randn(N,1) ;
5              lon_n_n = lon_n + x(1,i)*randn(N,1) ;
6              LOC_n_n = [lon_n_n lat_n_n]      ;
7          % Noisy distance from students to schools
8              D_n = pdist2(LOC_n_n, LOC_s, 'euclidean') ;
9          % Noisy school choice
10             [d_n, c_n] = min(D_n,[],2) ;
11         % Actual and noisy school choice
12             ERROR(:, i) = (c ~= c_n) ;
13     end
14     % Percentage of errors in choices
15     result = sum(ERROR, 1)./N ;

```

Output

```
1 % Figure
2 plot(x, result, 'k', 'LineWidth', 4)
3     ylabel('Percentage of wrong choices', 'FontSize', 20)
4     xlabel('Size of measurement error' , 'FontSize', 20)
5     axis([0 1 0 1])
6     box off
7     set(gcf, 'Color', 'w')
8
9     export_fig 'error_1.pdf'
```

Figure: *Mistakes due to measurement error*



Incorporating uncertainty

```
1  % Loop for different location of students
2      for j = 1:M
3          % Location of students
4              lat_n = randn(N,1)      ;
5              lon_n = randn(N,1)      ;
6              LOC_n = [lon_n lat_n] ;
7          % Distance and school choice
8              D = pdist2(LOC_n, LOC_s, 'euclidean') ;
9              [d, c] = min(D,[],2) ;
10         % Loop for different types of noise
11             for i = 1 : size(x, 2)
12                 % Noisy location of students
13                     lat_n_n = lat_n + x(1,i)*randn(N,1) ;
```

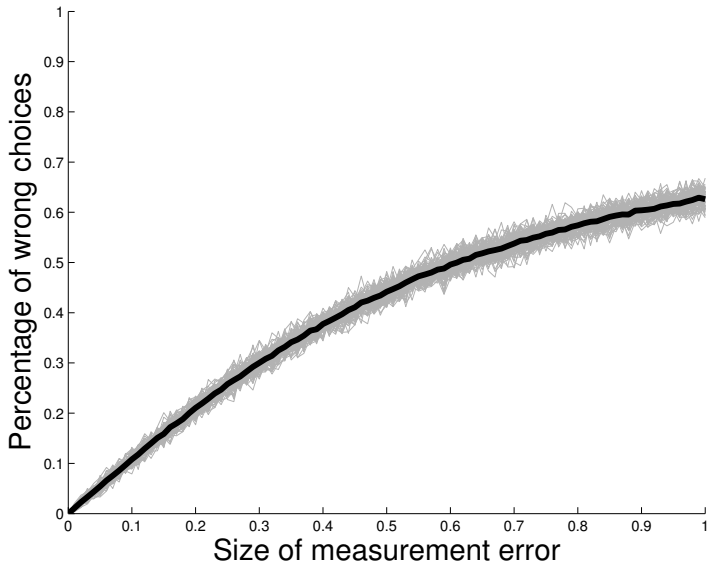
Incorporating uncertainty

```
14         lon_n_n = lon_n + x(1,i)*randn(N,1) ;
15         LOC_n_n = [lon_n_n lat_n_n] ;
16         % Noisy distance and noisy school choice
17         D_n = pdist2(LOC_n_n, LOC_s, 'euclidean') ;
18         [d_n, c_n] = min(D_n,[],2) ;
19         % Actual and noisy school choice
20         ERROR(:, i) = (c ~= c_n) ;
21     end
22     % Percentage of errors in choices
23     result(j,:) = sum(ERROR, 1)./N ;
24 end
```

Output

```
1  % Plot error
2  hold on
3  plot(x, result, 'Color', [.7 .7 .7], 'LineWidth', .5)
4  plot(x, mean(result), 'k', 'LineWidth', 4 )
5  ylabel('Percentage of wrong choices', 'FontSize' , 20)
6  xlabel('Size of measurement error' , 'FontSize' , 20)
7  axis([0 1 0 1])
8  box off
9  set(gcf, 'Color', 'w')
10 hold off
11
12 export_fig 'error_2.pdf'
```

Figure: *Incorporating uncertainty using Montecarlo*



Additional resources

1. Explore `export_fig`
2. Explore `mcode`
3. Explore options of commands we have used (e.g, `plot`, `pdist2`)