

# Spaghetti Plot Demo

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## Get Data

This example uses data from <https://stats.idre.ucla.edu/stata/examples/mlm-imm/introduction-to-multilevel-modeling-by-kreft-and-de-leeuwchapter-4-analyses/>

```
. use https://stats.idre.ucla.edu/stat/examples/imm/imm23, clear  
. label variable ses "Socioeconomic Status" // correct spelling of variable label
```

## Basic Spaghetti Plot

```
. spagplot math ses, id(schid)  
. graph export graph1.png, width(500) replace  
file graph1.png saved as PNG format
```

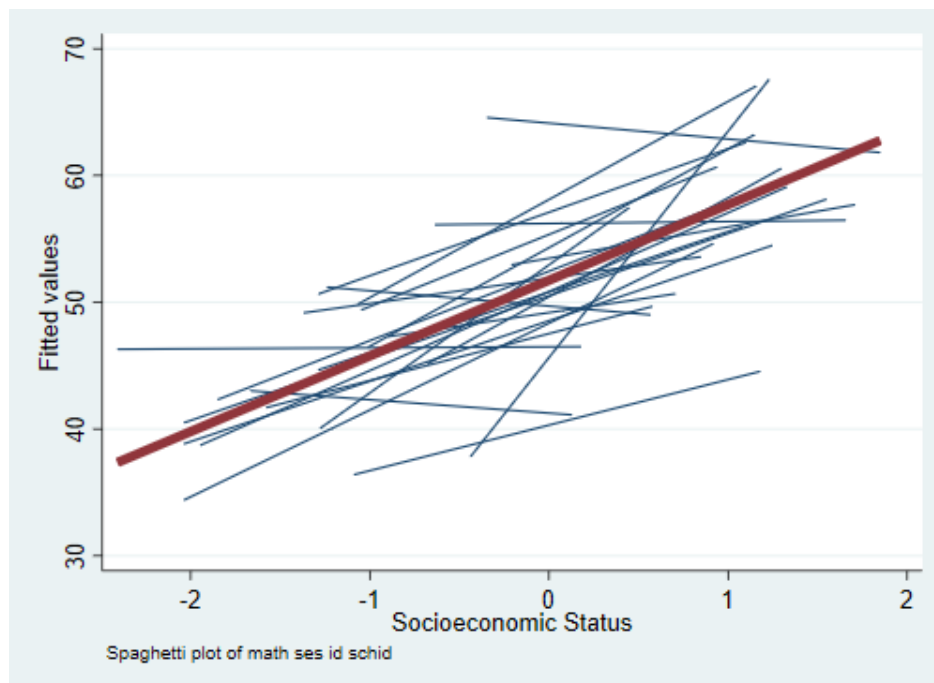


Figure 1: Basic Spaghetti Plot

## Add Better Scheme

Schemes are very helpful in making better looking Stata graphs. A useful Stata scheme is `s1color`. Useful user written schemes are `lean2`, `plottig` (type `findit lean2` or `findit plottig` to install these), and my own [Michigan Stata graph scheme](#).

```
. spagplot math ses, id(schid) ///  
> scheme(michigan) ///  
> title("Spaghetti Plot of Math Score By SES") ///  
> note(" ") // blank "note" since title explains this graph  
  
. graph export graph2.png, width(500) replace  
file graph2.png saved as PNG format
```

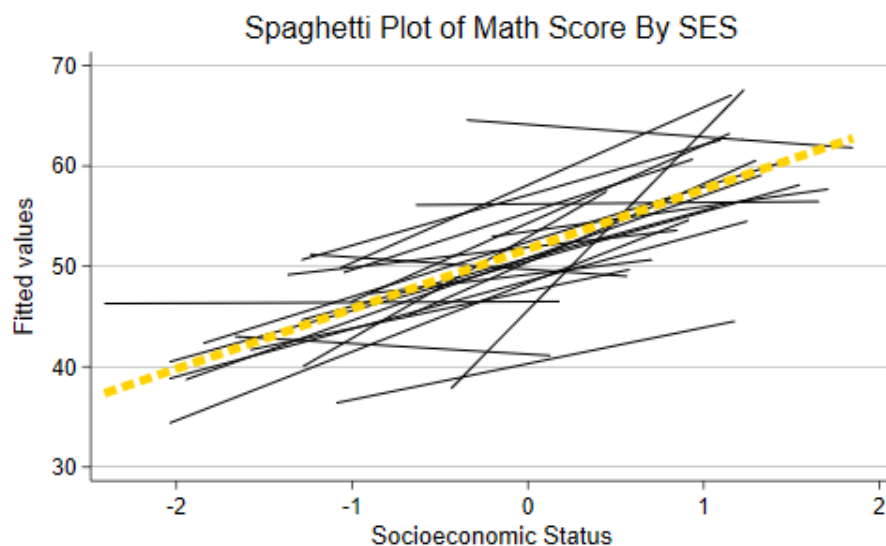


Figure 2: Spaghetti Plot With Better Options

## Now Try twoway Syntax

```
. twoway lfit math ses, scheme(michigan) title("Math Score By SES")  
  
. graph export graph3.png, width(500) replace  
file graph3.png saved as PNG format
```

## Separate Panels For Schools

This ONLY works well with a limited number of schools.

```
. twoway lfit math ses, scheme(michigan) by(schid, title("Math Score By SES"))  
  
. graph export graph4.png, width(1000) replace  
file graph4.png saved as PNG format
```

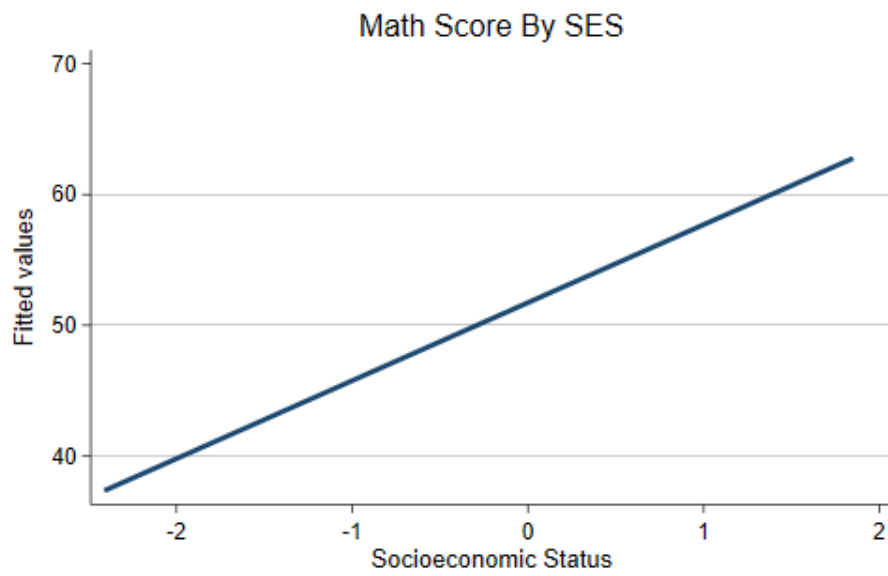


Figure 3: Initial `twoway` Graph

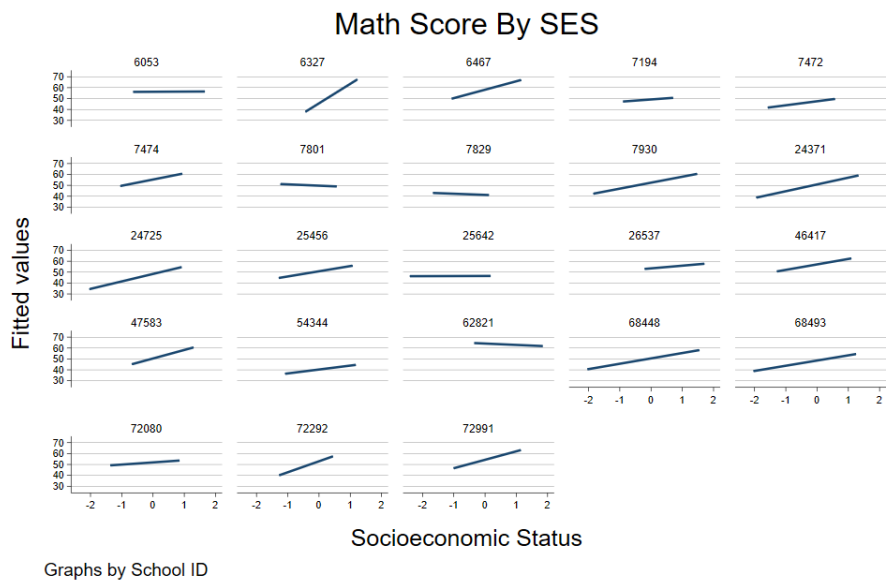
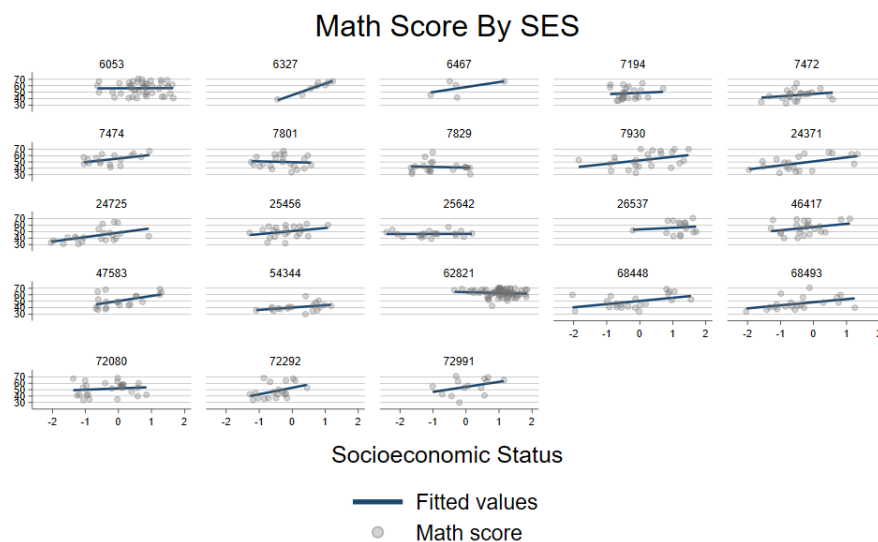


Figure 4: Separate Panels For Schools

## Separate Panels For Schools With Scatterplots

```
. twoway (lfit math ses) ///
> (scatter math ses, mcolor(gs7%30)), /// color gs7 @ 30% transparency
> scheme(michigan) by(schid, title("Math Score By SES"))

. graph export graph5.png, width(1000) replace
file graph5.png saved as PNG format
```



Graphs by School ID

Figure 5: Separate Panels For Schools With Scatterplots

## “Model Based” Spaghetti Plot

A sometimes unacknowledged point is that spaghetti plots—unless we take steps to correct this—reflect *unadjusted*, or *bivariate* associations.

We may sometimes wish to develop a spaghetti plot that reflects the *adjusted* estimates from our models.

To do this we first estimate a multilevel model.

```
. mixed math ses meanses || schid: // multilevel model; random intercept; no random effects
Performing EM optimization:
Performing gradient-based optimization:
Iteration 0: log likelihood = -1871.9169
Iteration 1: log likelihood = -1871.9169
Computing standard errors:
Mixed-effects ML regression
Group variable: schid
Number of obs = 519
Number of groups = 23
Obs per group:
    min = 5
    avg = 22.6
    max = 67
Wald chi2(2) = 69.58
Prob > chi2 = 0.0000
Log likelihood = -1871.9169
```

math	Coefficient	Std. err.	z	P> z	[95% conf. interval]
------	-------------	-----------	---	------	----------------------

ses	3.88476	.6096853	6.37	0.000	2.689799	5.079722
meanses	3.281962	1.464135	2.24	0.025	.4123106	6.151614
_cons	51.48904	.7582764	67.90	0.000	50.00284	52.97523

Random-effects parameters	Estimate	Std. err.	[95% conf. interval]	
schid: Identity				
var(_cons)	8.931927	3.813085	3.868681	20.62184
var(Residual)	75.21885	4.778177	66.41333	85.19187

LR test vs. linear model:  $\chi^2(01) = 25.58$  Prob  $\geq \chi^2 = 0.0000$

NB that this is a model with only a random intercept,  $u_0$  and no random slopes e.g.  $u_1$ , etc. . . .

## “Simple” Predicted Values

```
. predict yhat
(option xb assumed)

. spagplot yhat ses, id(schid) scheme(michigan)

. graph export graph6A.png, width(500) replace
file graph6A.png saved as PNG format
```

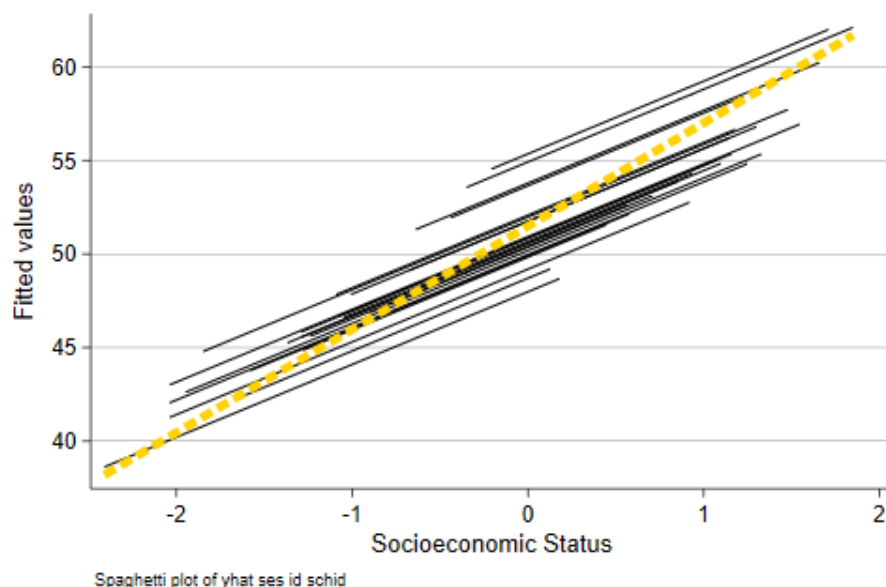


Figure 6: Spaghetti Plot With Predicted Values

## “Model Based” Predicted Values

The spaghetti plots so far give an indication of different slopes per school. Below we outline a procedure for (a) developing a spaghetti plot of adjusted estimates; and (b) ensuring that the plot reflects the *exact* structure of the model e.g. random intercept only, or random intercept + random slope(s).

To carry out this procedure we employ the `_b` notation in Stata. For example, `_b[_cons]` indicates the intercept of the model while `_b[ses]` indicates the slope attached to `ses`.

We need to carry out a few preliminary calculations.

1. Estimate (`predict`) the random effect(s).
2. Estimate the mean values (`summarize`) of variables that we are going to hold constant.
3. Generate predicted values ( $\hat{y}$ ) using the `_b` notation (`generate yhat = ...`).
4. Graph the spaghetti plot (`twoway connect`).

## Estimate The Random Effects

```
. mixed math ses meanses || schid: // multilevel model; random intercept; no random effects
Performing EM optimization:
Performing gradient-based optimization:
Iteration 0:   log likelihood = -1871.9169
Iteration 1:   log likelihood = -1871.9169
Computing standard errors:
Mixed-effects ML regression              Number of obs   =       519
Group variable: schid                   Number of groups =        23
Obs per group:
    min =          5
    avg =       22.6
    max =         67
Wald chi2(2) =       69.58
Prob > chi2   =       0.0000
Log likelihood = -1871.9169
```

math	Coefficient	Std. err.	z	P> z	[95% conf. interval]	
ses	3.88476	.6096853	6.37	0.000	2.689799	5.079722
meanses	3.281962	1.464135	2.24	0.025	.4123106	6.151614
_cons	51.48904	.7582764	67.90	0.000	50.00284	52.97523

Random-effects parameters	Estimate	Std. err.	[95% conf. interval]	
schid: Identity				
var(_cons)	8.931927	3.813085	3.868681	20.62184
var(Residual)	75.21885	4.778177	66.41333	85.19187

```
LR test vs. linear model: chibar2(01) = 25.58           Prob >= chibar2 = 0.0000

. predict u0, reffects
```

## Estimate the Mean Values of Relevant Variables

```
. summarize meanses
```

Variable	Obs	Mean	Std. dev.	Min	Max
meanses	519	-.0012717	.6206429	-1.0685	1.17625

The mean of `meanses` is -0.00127.

## Estimate Predicted Values

We are using  $\beta_0$ , the random intercept  $u_0$ ,  $\beta_{ses}$  multiplied by the actual value of `ses`, and  $\beta_{meanses}$  multiplied by the mean of `meanses`.

```
. generate yhat2 = _b[_cons] + u0 + _b[ses] * ses + _b[meanses] * -.0012717
```

## Graph The Spaghetti Plot

```
. twoway scatter yhat2 ses, scheme(michigan)

. graph export graph6B.png, width(500) replace
file graph6B.png saved as PNG format
```

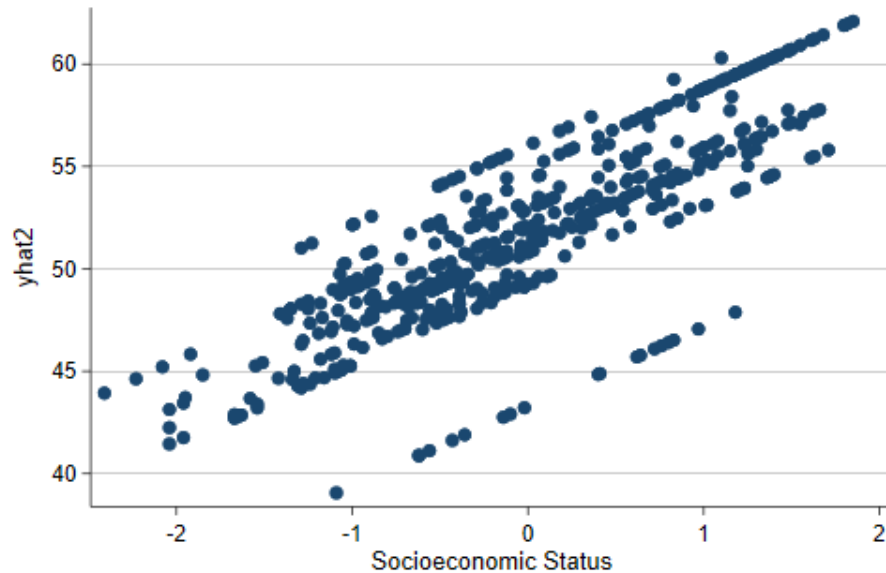


Figure 7: Initial “Model Based” Spaghetti Plot

We still have a small amount of work to do to make this look more “spaghetti plot like”.

We are going to use `twoway connect` to create connected line plots. We employ option `c(L)` to ensure that only ascending values are connected: i.e. each Level 2 unit has their own regression line. For `c(L)` to work properly we are going to need to sort the data by *school* and *ses*. Lastly, we’re going to change the `msymbol` so that we do not see dots, but only lines.

```
. sort schid ses // sort on Level 2 units and x values

. twoway connect yhat2 ses, ///
> lcolor("0 39 76") /// Michigan blue for connecting lines
> title("Model Based Spaghetti Plot") /// title
> xtitle("Socioeconomic Status") /// title for x axis
> ytitle("Model Predicted Values") /// title for y axis
> c(L) /// connect only ascending values
> msymbol(none) /// no marker symbol; only lines
> scheme(michigan) // michigan scheme

. graph export graph7.png, width(500) replace
file graph7.png saved as PNG format
```

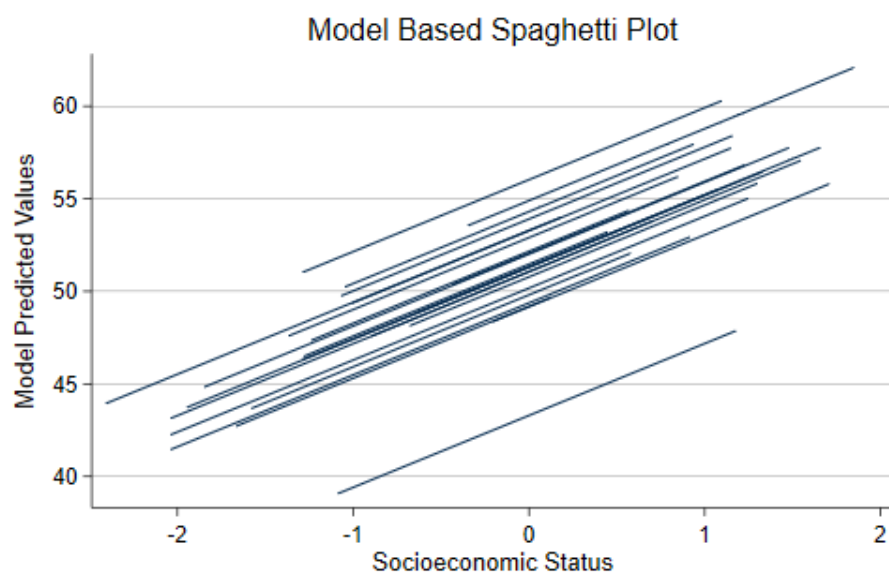


Figure 8: Finalized “Model Based” Spaghetti Plot