

The Basic Idea of Simulation

So, you estimate a model... and you want to say something *substantive* with quantities of interest:

Predicted or Expected Values of DV = $X_m \hat{\mathbf{b}}$

First Differences = $X_{+s} \hat{\mathbf{b}} - X_m \hat{\mathbf{b}}$

The problem is that our $\hat{\mathbf{b}}\mathbf{s}$ are uncertain!

The solution is we know how uncertain.

$$\hat{\mathbf{b}}_1$$

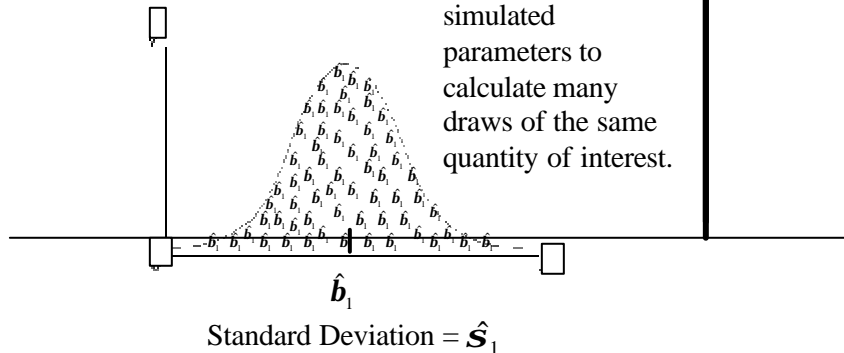
$$(\hat{\mathbf{S}}_1)$$



The Basic Idea of Simulation: Parameters

In order to capture the uncertainty, we draw simulated $\hat{\mathbf{b}}_S$ from the multivariate* normal distribution.

Then we use these simulated parameters to calculate many draws of the same quantity of interest.



The Basic Idea of Simulation: Quantities of Interest

In practice... $Y_i \sim f(\mathbf{q}_i, \mathbf{a})$, $\mathbf{q}_i = g(X_i, \mathbf{b})$
 $Y_i \sim N(\mathbf{m}_i, \mathbf{s}^2)$, $\mathbf{m}_i = g(X_i, \mathbf{b}) = \mathbf{b}_0 + X_{i1}\mathbf{b}_1 + X_{i2}\mathbf{b}_2 + \dots$

$$\hat{\mathbf{g}} = \begin{bmatrix} \hat{\mathbf{b}}_1 \\ \hat{\mathbf{b}}_2 \\ \vdots \\ \hat{\mathbf{a}}_1 \end{bmatrix} \quad \hat{\mathbf{V}}(\hat{\mathbf{g}}) = \begin{bmatrix} v_{\hat{\mathbf{b}}_1\hat{\mathbf{b}}_1} & v_{\hat{\mathbf{b}}_1\hat{\mathbf{b}}_2} & \dots & v_{\hat{\mathbf{b}}_1\hat{\mathbf{a}}_1} \\ v_{\hat{\mathbf{b}}_2\hat{\mathbf{b}}_1} & v_{\hat{\mathbf{b}}_2\hat{\mathbf{b}}_2} & \dots & v_{\hat{\mathbf{b}}_2\hat{\mathbf{a}}_1} \\ \vdots & \vdots & \ddots & \vdots \\ v_{\hat{\mathbf{a}}_1\hat{\mathbf{b}}_1} & v_{\hat{\mathbf{a}}_1\hat{\mathbf{b}}_2} & \dots & v_{\hat{\mathbf{a}}_1\hat{\mathbf{a}}_1} \end{bmatrix}$$

we simulate parameters with M draws from the multivariate normal distribution... $\tilde{\mathbf{g}} \sim N(\hat{\mathbf{g}}, \hat{\mathbf{V}})$

$$\begin{bmatrix} \tilde{\mathbf{b}}_{11} \\ \tilde{\mathbf{b}}_{21} \\ \vdots \\ \tilde{\mathbf{a}}_1 \end{bmatrix} \begin{bmatrix} \tilde{\mathbf{b}}_{12} \\ \tilde{\mathbf{b}}_{22} \\ \vdots \\ \tilde{\mathbf{a}}_2 \end{bmatrix} \dots \begin{bmatrix} \tilde{\mathbf{b}}_{1M} \\ \tilde{\mathbf{b}}_{2M} \\ \vdots \\ \tilde{\mathbf{a}}_M \end{bmatrix}$$

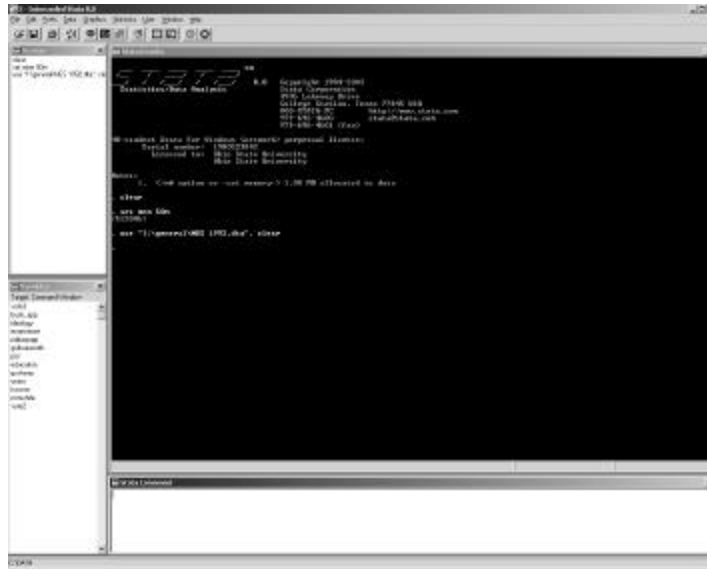
1. Choose a starting scenario, X_c .
2. Draw one value of $\tilde{\mathbf{g}}$, and compute $\tilde{\mathbf{q}}_c = g(X_c, \tilde{\mathbf{b}})$.
3. Simulate the outcome \tilde{Y}_c , by taking a random draw from $f(\tilde{\mathbf{q}}_c, \tilde{\mathbf{a}})$.
4. Repeat M times to get the distribution of Y_c .

Components of Clarify

- *estsimp* – estimates the model and simulates the parameters
 - This command **must** precede your regression command
 - e.g.: *estsimp logit y x1 x2 x3 x4*
 - This will save simulated β s to your dataset!
- *setx* – sets the values for the IVs (the Xs)
 - Used after model estimation to set values of the Xs
 - e.g.: *setx x1 mean x2 p20 x3 .4 x4[16], nocwdel*
 - functions = mean|median|min|max|p#|math|#|'macro'|varname[#]
 - reset values by re-issuing the command, e.g.: *setx x1 median*
- *simqi* – simulates the quantities of interest
 - Automates the simulation of quantities of interest for the X values you just set.
 - e.g.: *simqi, prval(1)*
 - e.g.: *simqi, fd(prval(1)) changex(x4 p25 p75)*

There are lots of options: Explore on your own!

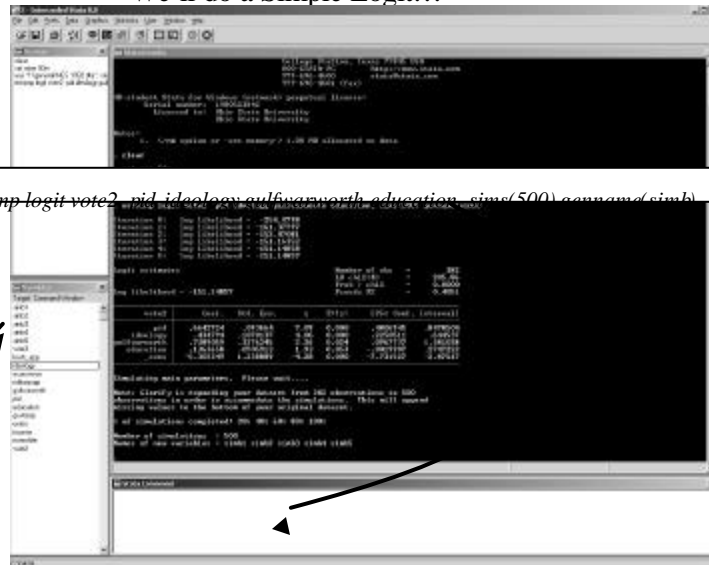
Onto the Machines...



We'll do a Simple Logit...

Type: `estsim logit vote2 id ideology gulfwar worth education sim(500) command simb`

Note that Clarify has added 5 new variables to our data set.



1. Type: `sum simb1-simb5` 2. Type: `setx mean`

Tables of First Differences

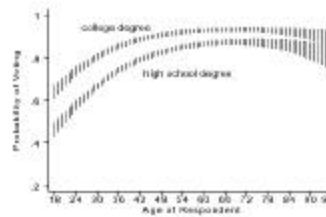
Type: `fd(prval(1)) changex(pid -3 -2)`

Probability of Bush Vote as PID Varies

Party ID	-3	-2	-1	0	1	2	3
P(Bush)	0.049	0.089	0.156	0.263	0.408	0.571	0.719
95% CI	(.025, .091)	(.056, .141)	(.111, .223)	(.204, .330)	(.332, .487)	(.475, .669)	(.613, .818)

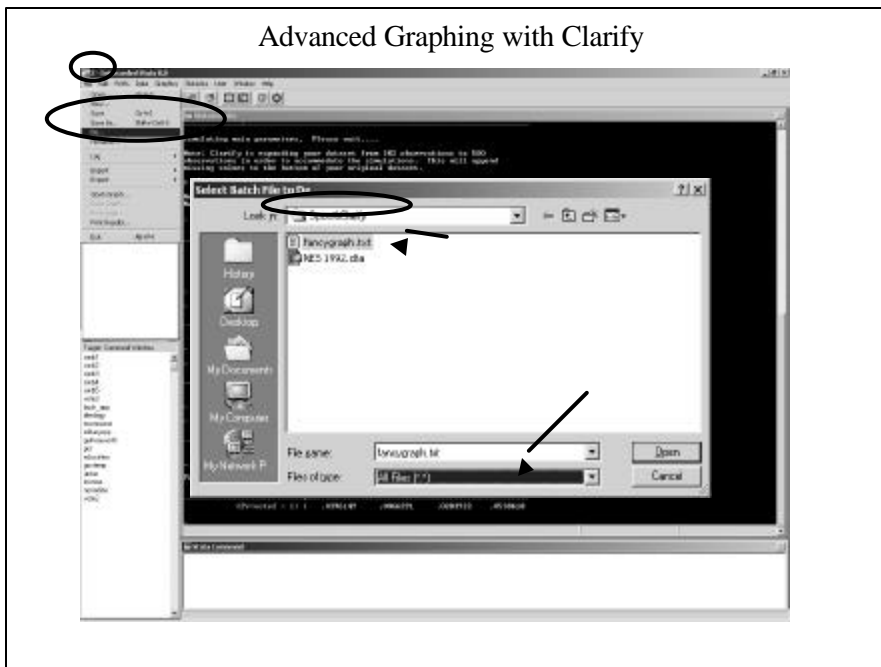
And, since we know P(Bush) is .273(.212, .339) when every variable is held at its mean, we can calculate percentage changes ourselves to increase substantive interpretability.

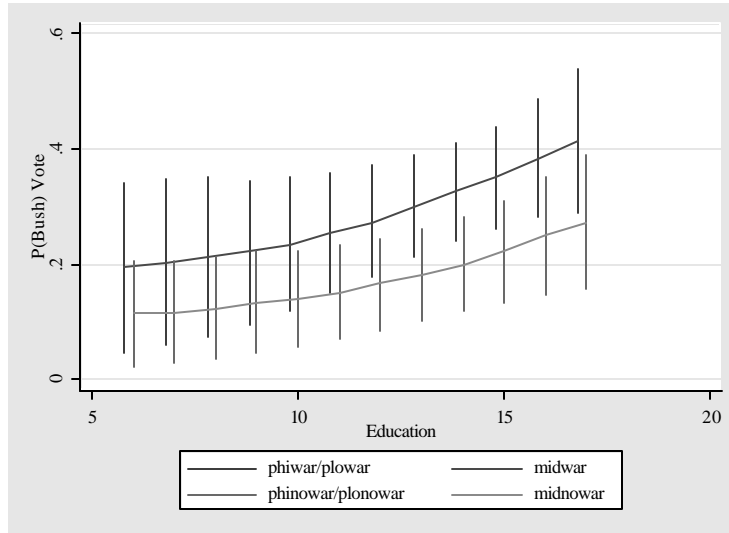
But a picture is worth a thousand words, so it would be nice to use Clarify to generate pictures like this:



From King et al. *AJPS* 2000

Advanced Graphing with Clarify





```

version 8.0
set more off
# definit;

gen plowar=.;
gen phiwar=.;
gen eduaxis = _n + 5 in 1/12;
setx gulfwarworth 1 ideology mean pid mean;
local i = 6;
while `i' <= 17 {;
    setx education `i';
    simqi, prval(1) genpr(pi);
    _pctile pi, p(2.5,97.5);
    replace plowar = r(r1) if eduaxis==`i';
    replace phiwar = r(r2) if eduaxis==`i';
    drop pi;
    local i = `i'+1;
};

gen plonowar=.;
gen phinowar=.;
setx gulfwarworth 0 ideology mean pid mean;
local i = 6;
while `i' <= 17 {;
    setx education `i';
    simqi, prval(1) genpr(pi);
    _pctile pi, p(2.5,97.5);
    replace plonowar = r(r1) if eduaxis==`i';
    replace phinowar = r(r2) if eduaxis==`i';
    drop pi;
    local i = `i'+1;
};

gen eduaxis2 = eduaxis -.2;
sort eduaxis;
gen midwar = (plowar+phiwar)/2;
gen midnowar = (plonowar+phinowar)/2;

graph twoway rspike phiwar plowar eduaxis2 || line midwar eduaxis2 || rspike
phinowar plonowar eduaxis || line midnowar eduaxis, ytitle(P(Bush) Vote)
xtitle(Education);

```


Conclusion

- Models Currently Supported by Clarify
 - regress
 - logit
 - probit
 - ologit
 - oprobit
 - mlogit
 - poisson
 - nbreg
 - sureg
 - weibull
- But, you really don't need Clarify to do this, so you can simulate quantities of interest for *any* model!
 - ✓ Easy to simulate parameters because Stata saves them after estimation!
 - ✓ Program the correct link function yourself!

