

Help File - hybreg

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Installation

1. Setup Command

This step adds `hybreg` command to Stata.

Copy the files in folder `Command_c_ado_personal` to your personal ado path, for example
`C:\ado\personal`

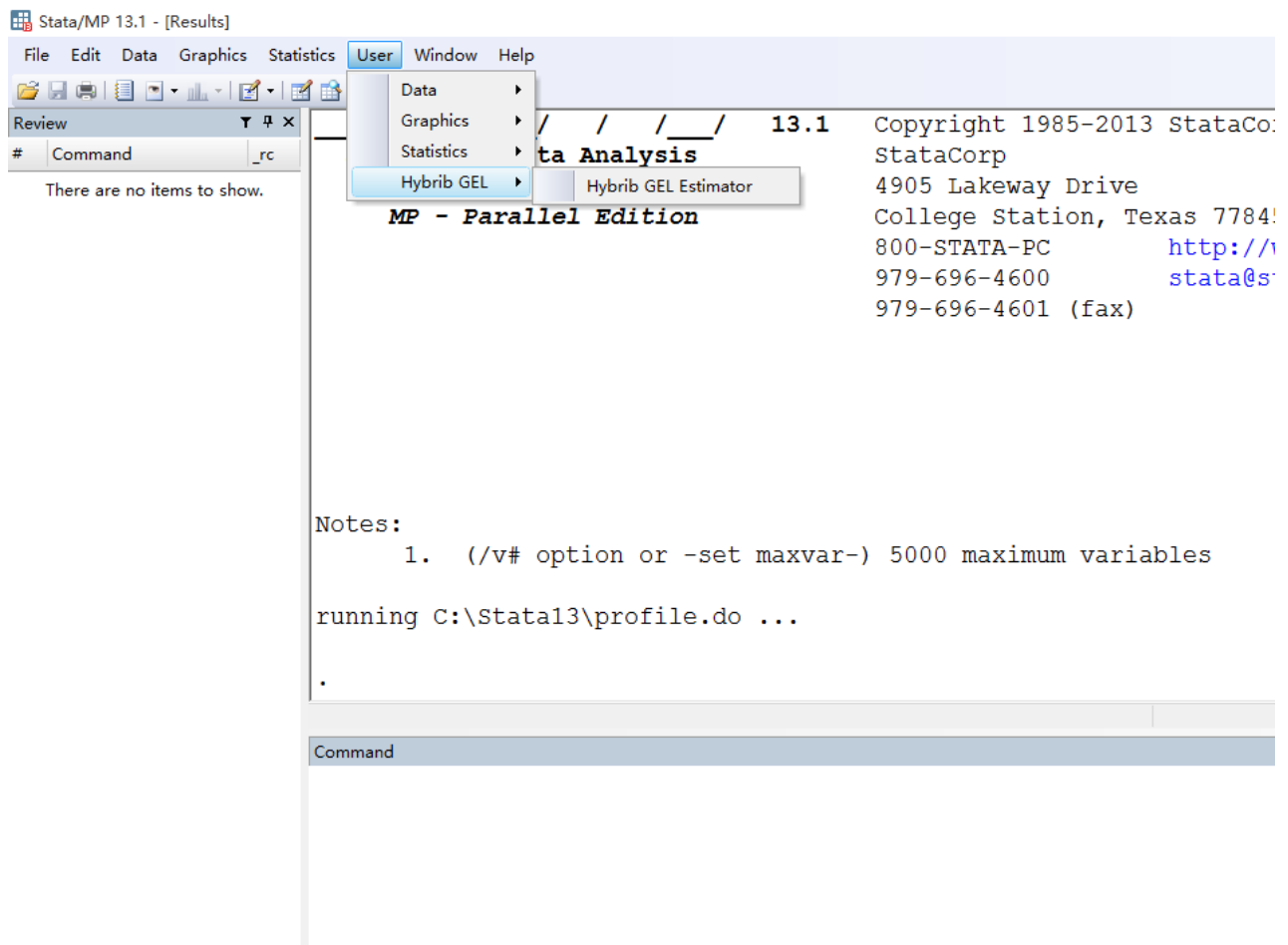
2. Setup Menu

This step adds menu items to Stata when Stata is launched.

Copy the files in folder `Menu_c_stata13` to the folder where Stata is installed, for example
`C:\Stata13`

3. Run Stata

Run Stata. You may see a line: `running C:\Stata13\profile.do ...`



The `hybreg` menu is added to your Stata. You can also run `db hybreg` to launch the GUI.

4. profile.do

If you already have `profile.ado` predefined, you can add these lines to your `profile.ado`, to enable menu.

```
window menu clear  
window menu append submenu "stUser" "Hybrib GEL"  
window menu append item "Hybrib GEL" "Hybrib GEL Estimator" "db hybreg"  
window menu refresh
```

One Endogenous Variable

First we clear the data, stars are the comments in Stata

```
clear
```

Command line is hard to use, the recommended way to use `hybreg` is the GUI graphic interface. GUI will be able to generate the command line for you.

```
sysuse hybreg_Data  
  
help hybreg  
  
db hybreg
```

hybreg - Hybrid Generalized Empirical Likelihood Estimators

Dependent Variables:
 ☐ Show trace log

Endogenous Variables:
 ☐ Add intercept

Instrumental Variables:
 ☐ Add IV intercept.

Initial Values:
 ☐ Add Lasso intercept
☐ Add Lasso normalize step

Choose a Method: Estimate Type: Optimization:

GEL	CUE	BFGS
GMM	EL	One-Dim-Optim
IV	ET	
	onestep	

OK Cancel Submit

1. IV

This is the IV estimator. You can choose `2sls`, `liml` or `gmm` for the type of estimator.

hybreg - Hybrid Generalized Empirical Likelihood Estimators

Dependent Variables:
 ☐ Show trace log

Endogenous Variables:
 ☐ Add intercept

Instrumental Variables:
 ☐ Add IV intercept.

Initial Values:
 ☐ Add Lasso intercept
☐ Add Lasso normalize step

Choose a Method: Estimate Type: Optimization:

GEL	twostep	BFGS
GMM	igmm	One-Dim-Optim
IV	2sls	
	emm	

OK Cancel Submit

```
hybreg y ( x = sz1 sz2 sz3 ), trace(0) method(IV) types(2sls) intercept(0)
tet0(tet0) lars_normalize(0) lars_intercept(0) optfct(optim)
```

Stata output :

Lasso step:

Selected Instrumental Variables: sz1

End of Lasso step.

Instrumental variables (2SLS) regression

Number of obs = 100

Wald chi2(1) = .

Prob > chi2 = .

R-squared = .

Root MSE = 4.279

y	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
-----+-----						
x	.8614375	.4464976	1.93	0.054	-.0136816	1.736557

Instrumented: x

Instruments: sz1

other IV results:

```
hybreg y ( x = sz1 sz2 sz3 ), trace(0) method(IV) types(liml) intercept(0)
tet0(tet0) lars_normalize(0) lars_intercept(0) optfct(optim)
```

Lasso step:

Selected Instrumental Variables: sz1

End of Lasso step.

Instrumental variables (LIML) regression	Number of obs =	100
	Wald chi2(1) =	.
	Prob > chi2 =	.
	R-squared =	.
	Root MSE =	4.279

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
y					
x	.8614375	.4464976	1.93	0.054	-.0136816 1.736557

Instrumented: x

Instruments: sz1

```
hybreg y ( x = sz1 sz2 sz3 ), trace(0) method(IV) types(gmm) intercept(0)
tet0(tet0) lars_normalize(0) lars_intercept(0) optfct(optim)
```

Instrumental variables (GMM) regression	Number of obs =	100
	Wald chi2(1) =	.
	Prob > chi2 =	.
	R-squared =	.
GMM weight matrix: Robust	Root MSE =	4.279

	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
y					
x	.8614375	.7173258	1.20	0.230	-.5444952 2.26737

Instrumented: x

Instruments: sz1

2. GMM

This is the GMM estimator, You can choose `onestep`, `twostep` or `igmm` for the type of estimator.

hybreg - Hybrid Generalized Empirical Likelihood Estimators

Dependent Variables:

y

Endogenous Variables:

x

Instrumental Variables:

sz1 sz2 sz3

Initial Values:

tet0

☐ Show trace log
☐ Add intercept
☐ Add IV intercept.
☐ Add Lasso intercept
☐ Add Lasso normalize step

Choose a Method:

GEL
GMM
IV

Estimate Type:

ET
onestep
twostep
igmm

Optimization:

BFGS
One-Dim-Optim

OK

Cancel

Submit

GMM: onestep

```
hybreg y ( x = sz1 sz2 sz3 ), trace(0) method(GMM) types(onestep) intercept(0)
tet0(tet0) lars_normalize(0) lars_intercept(0) optfct(optimize)
```

Lasso step:

Selected Instrumental Variables: sz1

End of Lasso step.

Step 1

Iteration 0: GMM criterion $Q(b)$ = .6815524

Iteration 1: GMM criterion $Q(b)$ = 1.071e-26

Iteration 2: GMM criterion $Q(b)$ = 5.835e-33

GMM estimation

Number of parameters = 1

Number of moments = 1

Initial weight matrix: Unadjusted

Number of obs = 100

		Robust				
		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
-----+-----						
/b_x		.8614375	.7173258	1.20	0.230	-.5444952 2.26737

Instruments for equation 1: sz1

GMM: twostep

```
hybreg y ( x = sz1 sz2 sz3 ), trace(0) method(GMM) types(twostep) intercept(0)
tet0(tet0) lars_normalize(0) lars_intercept(0) optfct(optimize)
```

Lasso step:

Selected Instrumental Variables: sz1

End of Lasso step.

Step 1

Iteration 0: GMM criterion $Q(b)$ = .6815524

Iteration 1: GMM criterion $Q(b)$ = 1.071e-26

Iteration 2: GMM criterion $Q(b)$ = 5.835e-33

Step 2

Iteration 0: GMM criterion $Q(b)$ = 1.235e-34

Iteration 1: GMM criterion $Q(b)$ = 1.235e-34

GMM estimation

Number of parameters = 1

Number of moments = 1

Initial weight matrix: Unadjusted

Number of obs = 100

GMM weight matrix: Robust

```
-----
          |               Robust
          |               Coef.   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
      /b_x |   .8614375   .7173258    1.20   0.230   - .5444952    2.26737
-----
```

Instruments for equation 1: sz1

GMM: igmm


```
hybreg y ( x = sz1 sz2 sz3 ), trace(0) method(GMM) types(igmm) intercept(0)
tet0(tet0) lars_normalize(0) lars_intercept(0) optfct(optimize)
```

Lasso step:

Selected Instrumental Variables: sz1

End of Lasso step.

Step 1

Iteration 0: GMM criterion $Q(b) = .6815524$

Iteration 1: GMM criterion $Q(b) = 1.071e-26$

Iteration 2: GMM criterion $Q(b) = 5.835e-33$

Step 2

Iteration 0: GMM criterion $Q(b) = 1.235e-34$

Iteration 1: GMM criterion $Q(b) = 1.235e-34$

iterative GMM weight matrix converged

iterative GMM parameter vector converged

GMM estimation

Number of parameters = 1

Number of moments = 1

Initial weight matrix: Unadjusted

Number of obs = 100

GMM weight matrix: Robust

		Robust				
		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
-----+-----						
/b_x		.8614375	.7173258	1.20	0.230	-.5444952 2.26737

Instruments for equation 1: sz1

3. GEL

This is the GEL estimator, You can choose CUE , EL or ET for the type of estimator.

hybreg - Hybrid Generalized Empirical Likelihood Estimators

Dependent Variables:
y

Endogenous Variables:
x

Instrumental Variables:
sz1 sz2 sz3

Initial Values:
tet0

☐ Show trace log
☐ Add intercept
☐ Add IV intercept.
☐ Add Lasso intercept
☐ Add Lasso normalize step

Choose a Method: Estimate Type: Optimization:

GEL	CUE	BFGS
GMM	EL	One-Dim-Optim
IV	ET	
	onestep	

OK Cancel Submit

GEL: CUE

```
hybreg y ( x = sz1 sz2 sz3 ), trace(0) method(GEL) types(CUE) intercept(0)
tet0(tet0) lars_normalize(0) lars_intercept(0) optfct(optimize)
```

Type of GEL: CUE

Coefficients:

Variable	Estimate	t value	Std. Error	Pr(> t)
x	.8613381	.7173808	1.200671	.229879

Lambda:

Variable	Estimate	t value	Std. Error	Pr(> t)
h1	-2.92e-06	0	.	.

Over-identifying restrictions tests: degrees of freedom is 0

Variable	statistics	p-value
LR test	1.59e-08	.
LM test	3.87e-08	.
J test	1.92e-08	.

Convergence code for the coefficients: There is no convergence code for optimize

GEL: ET

```
hybreg y ( x = sz1 sz2 sz3 ), trace(0) method(GEL) types(ET) intercept(0)
tet0(tet0) lars_normalize(0) lars_intercept(0) optfct(optimize)
```

Lasso step:

Selected Instrumental Variables: sz1

End of Lasso step.

Type of GEL: ET

Coefficients:

Variable	Estimate	t value	Std. Error	Pr(> t)
x	.859578	.718029	1.197136	.2312537

Lambda:

Variable	Estimate	t value	Std. Error	Pr(> t)
h1	2.65e-14	0	.	.

Over-identifying restrictions tests: degrees of freedom is 0

Variable	statistics	p-value
LR test	-7.55e-15	.
LM test	3.19e-24	.
J test	6.71e-06	.

Convergence code for the coefficients: There is no convergence code for optimize

GEL: EL

```
hybreg y ( x = sz1 sz2 sz3 ), trace(0) method(GEL) types(EL) intercept(0)
tet0(tet0) lars_normalize(0) lars_intercept(0) optfct(optimize)
```

Lasso step:

Selected Instrumental Variables: sz1

End of Lasso step.

Type of GEL: EL

Coefficients:

Variable	Estimate	t value	Std. Error	Pr(> t)
x	.8613522	.7173686	1.200711	.2298634

Lambda:

Variable	Estimate	t value	Std. Error	Pr(> t)
h1	-1.78e-06	1.56e-10	-11385.17	0

Over-identifying restrictions tests: degrees of freedom is 0

Variable	statistics	p-value
LR test	1.41e-08	.
LM test	1.44e-08	.
J test	1.41e-08	.

Convergence code for the coefficients: There is no convergence code for optimize

Error Cases

Case 1

```
hybreg y ( x = sz1 sz2 sz3 ), trace(0) method(GEL) types(CUE) intercept(0)
tet0(tet0) lars_normalize(0) lars_intercept(0) optfct(optim)
```

Lasso step:

Selected Instrumental Variables: sz1

End of Lasso step.

The length of initial value does not match the number of regressors

The length of initial value is 3

The the number of regressors is 1.

You may want to choose one dimensional optimization.

Only when the dimension of regressor is 1, you can choose between the algorithm BFGS
> or one dimensional optimization. In that case, the former is unreliable. If one d
> imensional optimization is chosen, you should provide the upper and lower bound of
> the interval in which the algorithm seeks the solution.r(232);

Only when the dimension of endogenous variable is 1, you can choose between the algorithm `optim` (BFGS algorithm) or `optimize` (One Dimensional Optimization). In that case, the former is unreliable. If `optimize` is chosen, you should provide the upper and lower bound of the interval in which the algorithm seeks the solution.

Case 2

```
hybreg y ( x = sz3 ), trace(0) method(GEL) types(CUE) intercept(0) tet0(tet0)
lars_normalize(0) lars_intercept(0) optfct(optimize)
```

Lasso step:

Selected Instrumental Variables:

End of Lasso step.

Equation not identified, must have at least as many instruments not in the regression.

There are 1 endogenous variable, 0 instrumented variables
r(481);

You should provide more instrumental variables.

Multiple Endogenous Variables

There are two endogenous variables. z_1 z_2 are correlated with x_1 , and z_3 is correlated with x_2 .

```
sysuse hybreg_multi
```

```
generate var7 = 0.5 in 1
```

```
replace var7 = 0.5 in 2
```

```
rename var7 tet0
```

```
db hybreg
```

1. IV

hybreg - Hybrid Generalized Empirical Likelihood Estimators

Dependent Variables:
y

Endogenous Variables:
x1 x2

Instrumental Variables:
z1 z2 z3

Initial Values:
tet0

☐ Show trace log

☐ Add intercept

☐ Add IV intercept.

☐ Add Lasso intercept

☐ Add Lasso normalize step

Choose a Method: Estimate Type: Optimization:

GEL
GMM
IV

twostep
igmm
2sls
zmm

BFGS
One-Dim-Optim

OK Cancel Submit

```
hybreg y ( x1 x2 = z1 z2 z3 ), trace(0) method(IV) types(2sls) intercept(0) tet0(t
> et0) lars_normalize(0) lars_intercept(0) optfct(optim)
```

Lasso step:

Selected Instrumental Variables: z1 z2 z3

End of Lasso step.

note: z3 dropped due to collinearity

Instrumental variables (2SLS) regression

Number of obs = 100
Wald chi2(2) = .
Prob > chi2 = .
R-squared = .
Root MSE = 10.63

y	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
-----+-----						
x1	-2.474199	15.54964	-0.16	0.874	-32.95093	28.00253
x2	7.23444	26.98233	0.27	0.789	-45.64996	60.11884

Instrumented: x1 x2

Instruments: z1 z2

2. GMM



GMM: onestep


```
hybreg y ( x1 x2 = z1 z2 z3 ), trace(0) method(GMM) types(onestep) intercept(0) te
> t0(tet0) lars_normalize(0) lars_intercept(0) optfct(optim)
```

Lasso step:

Selected Instrumental Variables: z1 z2 z3

End of Lasso step.

Step 1

Iteration 0: GMM criterion $Q(b)$ = 21.065446

Iteration 1: GMM criterion $Q(b)$ = .05915936

Iteration 2: GMM criterion $Q(b)$ = .05915936

GMM estimation

Number of parameters = 2

Number of moments = 3

Initial weight matrix: Unadjusted

Number of obs = 100

		Robust					
		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
-----+-----							
/b_x1		1.077991	.2798355	3.85	0.000	.5295239	1.626459
/b_x2		1.066366	.3252707	3.28	0.001	.428847	1.703885

Instruments for equation 1: z1 z2 z3

GMM: twostep

```
hybreg y ( x1 x2 = z1 z2 z3 ), trace(0) method(GMM) types(twostep) intercept(0) te
> t0(tet0) lars_normalize(0) lars_intercept(0) optfct(optim)
```

Lasso step:

Selected Instrumental Variables: z1 z2 z3

End of Lasso step.

Step 1

Iteration 0: GMM criterion Q(b) = 6.0499342

Iteration 1: GMM criterion Q(b) = .05915936

Iteration 2: GMM criterion Q(b) = .05915936

Step 2

Iteration 0: GMM criterion Q(b) = .00775748

Iteration 1: GMM criterion Q(b) = .00774086

Iteration 2: GMM criterion Q(b) = .00774086

GMM estimation

Number of parameters = 2

Number of moments = 3

Initial weight matrix: Unadjusted

Number of obs = 100

GMM weight matrix: Robust

		Robust				
		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]

	+	-----				
/b_x1		1.066874	.2792201	3.82	0.000	.5196123 1.614135
/b_x2		1.07838	.3246282	3.32	0.001	.4421205 1.71464

Instruments for equation 1: z1 z2 z3

.

GMM: igmm

```
hybreg y ( x1 x2 = z1 z2 z3 ), trace(0) method(GMM) types(igmm) intercept(0) tet0(
> tet0) lars_normalize(0) lars_intercept(0) optfct(optim)
```

Lasso step:

Selected Instrumental Variables: z1 z2 z3

End of Lasso step.

Step 1

Iteration 0: GMM criterion $Q(b)$ = 21.065446

Iteration 1: GMM criterion $Q(b)$ = .05915936

Iteration 2: GMM criterion $Q(b)$ = .05915936

Step 2

Iteration 0: GMM criterion $Q(b)$ = .00775748

Iteration 1: GMM criterion $Q(b)$ = .00774086

Iteration 2: GMM criterion $Q(b)$ = .00774086

Step 3

Iteration 0: GMM criterion $Q(b)$ = .00774782

Iteration 1: GMM criterion $Q(b)$ = .0077478

Step 4

Iteration 0: GMM criterion $Q(b)$ = .00774844

Iteration 1: GMM criterion $Q(b)$ = .00774844

Step 5

Iteration 0: GMM criterion $Q(b)$ = .00774847

Iteration 1: GMM criterion $Q(b)$ = .00774847

iterative GMM weight matrix converged

iterative GMM parameter vector converged

GMM estimation

Number of parameters = 2

Number of moments = 3

Initial weight matrix: Unadjusted

Number of obs = 100

GMM weight matrix: Robust

		Robust					
		Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
-----+-----							
/b_x1		1.066662	.2791977	3.82	0.000	.5194451	1.61388
/b_x2		1.078813	.3246118	3.32	0.001	.4425856	1.715041

Instruments for equation 1: z1 z2 z3

3. GEL

hybreg - Hybrid Generalized Empirical Likelihood Estimators

Dependent Variables:
y

Endogenous Variables:
x1 x2

Instrumental Variables:
z1 z2 z3

Initial Values:
tet0

☐ Show trace log

☐ Add intercept

☐ Add IV intercept.

☐ Add Lasso intercept

☐ Add Lasso normalize step

Choose a Method: Estimate Type: Optimization:

GEL CUE BFGS

GMM EL One-Dim-Optim

IV ET

onestep

OK Cancel Submit

GEL: CUE

Stata result:

```
hybreg y ( x1 x2 = z1 z2 z3 ), trace(0) method(GEL) types(CUE) intercept(0) tet0(t  
> et0) lars_normalize(0) lars_intercept(0) optfct(optim)
```

Dependent Variables: y

Endogenous Variables: x1 x2

Instrumental Variables: z1 z2 z3

Lasso step:

Selected Instrumental Variables: z1 z2 z3

End of Lasso step.

Type of GEL: CUE

Coefficients:

Variable	Estimate	t value	Std. Error	Pr(> t)
x1	1.067199	.2788049	3.827761	.0001293
x2	1.072571	.3136692	3.419432	.0006275

Lambda:

Variable	Estimate	t value	Std. Error	Pr(> t)
h1	.0397516	.0504093	.7885762	.4303597
h2	-.0404333	.0520053	-.7774844	.436873
h3	.0008132	.0024646	.329964	.7414272

Over-identifying restrictions tests: degrees of freedom is 1

Variable	statistics	p-value
LR test	.7740856	.3789561
LM test	.608162	.4354807
J test	.985876	.3207524

Convergence code for the coefficients: Yes

GEL: ET

Stata result:

```
. hybreg y ( x1 x2 = z1 z2 z3 ), trace(0) method(GEL) types(ET) intercept(0) tet0(te
> t0) lars_normalize(0) lars_intercept(0) optfct(optim)
```

Dependent Variables: y

Endogenous Variables: x1 x2

Instrumental Variables: z1 z2 z3

Lasso step:

Selected Instrumental Variables: z1 z2 z3

End of Lasso step.

Type of GEL: ET

Coefficients:

Variable	Estimate	t value	Std. Error	Pr(> t)
-----+-----				
x1	1.065136	.2807192	3.794313	.0001481
x2	1.07529	.3149465	3.4142	.0006397

Lambda:

Variable	Estimate	t value	Std. Error	Pr(> t)
-----+-----				
h1	.0441171	.0493637	.893715	.3714744
h2	-.0449138	.0509266	-.8819327	.3778132
h3	.0008183	.0024135	.3390621	.7345629

Over-identifying restrictions tests: degrees of freedom is 1

Variable	statistics	p-value
-----+-----		
LR test	.8306702	.3620788
LM test	.7817546	.3766051
J test	.9443374	.3311649

Convergence code for the coefficients: Yes

GEL: EL

This runs a little bit slower.

Stata result:

```
. hybreg y ( x1 x2 = z1 z2 z3 ), trace(0) method(GEL) types(EL) intercept(0) tet0(te
> t0) lars_normalize(0) lars_intercept(0) optfct(optim)
Dependent Variables: y
Endogenous Variables: x1 x2
Instrumental Variables: z1 z2 z3
```

Lasso step:

Selected Instrumental Variables: z1 z2 z3

End of Lasso step.

numerical derivatives are approximate

nearby values are missing

.....

numerical derivatives are approximate

nearby values are missing

Type of GEL: EL

Coefficients:

Variable	Estimate	t value	Std. Error	Pr(> t)
x1	1.064391	.2828743	3.76277	.000168
x2	1.075661	.3163445	3.400284	.0006732

Lambda:

Variable	Estimate	t value	Std. Error	Pr(> t)
h1	.0472081	.0484491	.9743853	.3298653
h2	-.0480656	.049983	-.9616395	.3362307
h3	.0007425	.0023687	.3134372	.7539485

Over-identifying restrictions tests: degrees of freedom is 1

Variable	statistics	p-value
LR test	.8707327	.3507525
LM test	.9292014	.3350705
J test	.9094023	.3402731

Convergence code for the coefficients: Yes

Error Cases

```
hybreg y ( x1 x2 = z1 ), trace(0) method(GEL) types(ET) intercept(0) tet0(tet0) la
> rs_normalize(0) lars_intercept(0) optfct(optim)
```

Dependent Variables: y

Endogenous Variables: x1 x2

Instrumental Variables: z1

Lasso step:

equation not identified, must have at least as many instruments not in the regressio
> n as there are instrumented variables.

Ramdon Data Generation

hybreg_data.dta

There is only one endogenous variable, and z1 is correlated with x.

```
x = z1 + nu
eps = sqrt(z1^2+z2^2+z3^2)*eps
y = x + eps
```

hybreg_multi.dta

There are two endogenous variables. z1 z2 are correlated with x1, and z3 is correlated with x2.

```
z = c(z1,z2,z3)
beta1 = c(1,1,0)
beta2 = c(0,0,1)
x1 = z %%% beta1 + nu
x2 = z %%% beta2 + nu
eps = sqrt(z1^2+z2^2+z3^2)*eps
y = x1 + x2 + eps
```