

GMM using Moment Equations

In the study of interest rate structure, the continuous time model can be specified as:

$$r_{t+\Delta t} - r_t = (\alpha + \beta r_t) \Delta t + \varepsilon_{t+\Delta t} \quad (1)$$

where: $E[\varepsilon_{t+\Delta t}] = 0$ and $E[\varepsilon_{t+\Delta t}^2] = \sigma^2 r_t^{2\gamma} \Delta t$

Then, the model can be transformed to be discrete time model by setting $\Delta t = 1$. The discrete time model can be stated as:

$$r_{t+1} - r_t = \Delta r_t = \alpha + \beta r_t + \varepsilon_{t+1} \quad (2)$$

where: $E[\varepsilon_{t+1}] = 0$ and $E[\varepsilon_{t+1}^2] = \sigma^2 r_t^{2\gamma}$

The above model consists of four parameters including $\alpha, \beta, \sigma^2, \gamma$. The model can be estimate using GMM. Four moment condition equations can be stated as:

$$\begin{aligned} \text{Zero mean condition:} & \quad E(\varepsilon_{t+1}) = 0 \\ \text{Orthogonality condition:} & \quad E(\varepsilon_{t+1} r_t) = 0 \\ \text{Variance condition:} & \quad E(\varepsilon_{t+1}^2 - \sigma^2 r_t^{2\gamma}) = 0 \\ \text{Zero covariance condition:} & \quad E((\varepsilon_{t+1}^2 - \sigma^2 r_t^{2\gamma}) r_t) = 0 \end{aligned}$$

The above model can be claimed as unrestricted model for other eight interest rate structure models which can be indicated as follows:

Model	α	β	σ^2	γ
Unrestricted				
Merton		0		0
Vasicek				0
CIR SR				0.5
Dothan	0	0		1
GBM	0			1
Brennan & Schwartz				1
CIR VR	0	0		1.5
CEV	0			

Unrestricted model has four unknown parameters to be estimated and four moment condition equations, thus, method of estimate will be method of moment since it is exact identify equations model. Other eight restricted models will be estimated using GMM.

GMM can be performed by using the following command:

Unrestricted model:

```
. gmm (dr-{alpha}-{beta}*r) ((dr-{alpha}-{beta}*r)*r) ((dr-{alpha}-{beta}*r)^2-
{sigma2}*r^(2*{gamma})) (((dr-{alpha}-{beta}*r)^2-{sigma2}*r^(2*{gamma}))*r)
winitia(identity)
```

```
warning: 1 missing value returned for equation 1 at initial values
warning: 1 missing value returned for equation 2 at initial values
warning: 1 missing value returned for equation 3 at initial values
warning: 1 missing value returned for equation 4 at initial values
```

Step 1

```
numerical derivatives are approximate
flat or discontinuous region encountered
Iteration 0: GMM criterion Q(b) = .00005673
Iteration 1: GMM criterion Q(b) = .00005192 (backed up)
Iteration 2: GMM criterion Q(b) = .00004097
Iteration 3: GMM criterion Q(b) = 3.562e-06
Iteration 4: GMM criterion Q(b) = 2.174e-07
Iteration 5: GMM criterion Q(b) = 1.644e-07
```

Step 2

```
Iteration 0: GMM criterion Q(b) = .01229538
Iteration 1: GMM criterion Q(b) = .0055188
Iteration 2: GMM criterion Q(b) = .00045234
Iteration 3: GMM criterion Q(b) = 8.141e-07
Iteration 4: GMM criterion Q(b) = 2.461e-12
Iteration 5: GMM criterion Q(b) = 1.451e-23
```

GMM estimation

```
Number of parameters = 4
Number of moments = 4
Initial weight matrix: Identity
GMM weight matrix: Robust
Number of obs = 1429
```

	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
/alpha	-.0009679	.0012837	-0.75	0.451	-.0034839	.0015481
/beta	.0014035	.0005181	2.71	0.007	.0003882	.0024189
/sigma2	.0006419	.0001156	5.55	0.000	.0004153	.0008686
/gamma	.1273995	.0720773	1.77	0.077	-.0138695	.2686685

```
Instruments for equation 1: _cons
Instruments for equation 2: _cons
Instruments for equation 3: _cons
Instruments for equation 4: _cons
```

```
. estat overid
```

Test of overidentifying restriction:

Hansen's J chi2(0) = 2.1e-20 (p = .)

```
. est store Unrestricted
```

Merton model:

```
. gmm (dr-{alpha}) ((dr-{alpha})*r) ((dr-{alpha})^2-{sigma2}) (((dr-{alpha})^2-
{sigma2}*r) winitia(identity)
```

```
warning: 1 missing value returned for equation 1 at initial values
...
warning: 1 missing value returned for equation 4 at initial values
```

Step 1
 Iteration 0: GMM criterion $Q(b) = .00005673$
 Iteration 1: GMM criterion $Q(b) = 6.386e-07$
 Iteration 2: GMM criterion $Q(b) = 6.382e-07$

Step 2
 Iteration 0: GMM criterion $Q(b) = .00760718$
 Iteration 1: GMM criterion $Q(b) = .00652536$
 Iteration 2: GMM criterion $Q(b) = .00652532$

GMM estimation

Number of parameters = 2
 Number of moments = 4
 Initial weight matrix: Identity
 GMM weight matrix: Robust
 Number of obs = 1429

	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
/alpha	.0021651	.0007043	3.07	0.002	.0007847	.0035456
/sigma2	.0008218	.0000853	9.63	0.000	.0006546	.000989

Instruments for equation 1: _cons
 Instruments for equation 2: _cons
 Instruments for equation 3: _cons
 Instruments for equation 4: _cons

. estat overid

Test of overidentifying restriction:

Hansen's J $\chi^2(2) = 9.32468$ (p = 0.0094)

. est store Merton

Vasicek model:

. gmm (dr-{alpha}-{beta}*r) ((dr-{alpha}-{beta}*r)*r) ((dr-{alpha}-{beta}*r)^2-
 {sigma2}) (((dr-{alpha}-{beta}*r)^2-{sigma2})*r) winitial(identity)

warning: 1 missing value returned for equation 1 at initial values

...

warning: 1 missing value returned for equation 4 at initial values

Step 1
 Iteration 0: GMM criterion $Q(b) = .00005673$
 Iteration 1: GMM criterion $Q(b) = 3.000e-09$
 Iteration 2: GMM criterion $Q(b) = 2.107e-09$

Step 2
 Iteration 0: GMM criterion $Q(b) = .00263124$
 Iteration 1: GMM criterion $Q(b) = .00226079$
 Iteration 2: GMM criterion $Q(b) = .00226074$

GMM estimation

Number of parameters = 3
 Number of moments = 4
 Initial weight matrix: Identity
 GMM weight matrix: Robust
 Number of obs = 1429

	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
/alpha	-.0003515	.0012376	-0.28	0.776	-.002777	.0020741
/beta	.0012479	.0005106	2.44	0.015	.0002471	.0022487
/sigma2	.0007951	.0000862	9.23	0.000	.0006262	.0009641

Instruments for equation 1: _cons
 Instruments for equation 2: _cons
 Instruments for equation 3: _cons
 Instruments for equation 4: _cons

```
. estat overid
    Test of overidentifying restriction:
    Hansen's J chi2(1) = 3.23059 (p = 0.0723)
. est store Vasicek
```

CIR SR model:

```
. gmm (dr-{alpha}-{beta}*r) ((dr-{alpha}-{beta}*r)*r) ((dr-{alpha}-{beta}*r)^2-
{sigma2}*r) (((dr-{alpha}-{beta}*r)^2-{sigma2}*r)*r) winitial(identity) nolog
```

warning: 1 missing value returned for equation 1 at initial values

...
warning: 1 missing value returned for equation 4 at initial values

Final GMM criterion Q(b) = .0117512

GMM estimation

```
Number of parameters = 3
Number of moments = 4
Initial weight matrix: Identity
GMM weight matrix: Robust
Number of obs = 1429
```

	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
/alpha	-.0031028	.0011644	-2.66	0.008	-.005385	-.0008207
/beta	.0018442	.0005066	3.64	0.000	.0008514	.0028371
/sigma2	.0002351	.0000278	8.45	0.000	.0001805	.0002896

```
Instruments for equation 1: _cons
Instruments for equation 2: _cons
Instruments for equation 3: _cons
Instruments for equation 4: _cons
```

```
. estat overid
    Test of overidentifying restriction:
    Hansen's J chi2(1) = 16.7924 (p = 0.0000)
```

```
. est store CIR_SR
. est table Unrestricted Merton Vasicek CIR_SR, star(0.1 0.05 0.01)
```

Variable	Unrestricted	Merton	Vasicek	CIR_SR
alpha				
_cons	-.00096793	.00216513***	-.00035145	-.00310284***
beta				
_cons	.00140354***		.00124786**	.00184424***
sigma2				
_cons	.00064191***	.00082182***	.00079513***	.00023507***
gamma				
_cons	.12739949*			

legend: * p<.1; ** p<.05; *** p<.01

```
. *Test Unrestricted vs Merton
. test (_b[/beta]=0) (_b[/gamma]=0)
```

```
( 1) [beta]_cons = 0
( 2) [gamma]_cons = 0
```

```
chi2( 2) = 8.96
Prob > chi2 = 0.0113
```

GMM Examples Simulated Data

Example 1

```
set obs 500
capture drop y x z u
mat mCorr = (1, .7, .5 \ .7, 1, 0 \ .5, 0, 1)
corr2data x z u, mean(10 10 0) sd(3 4 20) corr(mCorr) n(500) clear
corr
g y=1+2*x+u
reg y x
ivregress gmm y (x=z)
```

```
. set obs 500
number of observations (_N) was 500, now 500

. capture drop y x z u

. mat mCorr = (1, .7, .5 \ .7, 1, 0 \ .5, 0, 1)

. corr2data x z u, mean(10 10 0) sd(3 4 20) corr(mCorr) n(500) clear
(obs 500)

. corr
(obs=500)
```

	x	z	u
x	1.0000		
z	0.7000	1.0000	
u	0.5000	0.0000	1.0000

```
. g y=1+2*x+u
. reg y x
```

Source	SS	df	MS	Number of obs	=	500
Model	127744.002	1	127744.002	F(1, 498)	=	424.96
Residual	149700.001	498	300.602411	Prob > F	=	0.0000
Total	277444.003	499	556.000006	R-squared	=	0.4604
				Adj R-squared	=	0.4593
				Root MSE	=	17.338

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]
x	5.333333	.2587168	20.61	0.000	4.825022 5.841644
_cons	-32.33333	2.70086	-11.97	0.000	-37.63982 -27.02685

```
. ivregress gmm y (x=z)
```

```
Instrumental variables (GMM) regression
Number of obs = 500
Wald chi2(1) = 19.64
Prob > chi2 = 0.0000
R-squared = 0.2806
Root MSE = 19.98

GMM weight matrix: Robust
```

y	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]
x	2	.4513481	4.43	0.000	1.115374 2.884626
_cons	.9999991	4.62114	0.22	0.829	-8.057269 10.05727

```
Instrumented: x
Instruments: z
```

Example 2

```

set obs 500
capture drop y x z* u
mat mCorr = (1, .7, .7, .5 \ .7, 1, .4, 0 \ .7, .4, 1, 0 \ .5, 0, 0, 1)
corr2data x z1 z2 u, mean(10 20 15 0) sd(3 4 5 20) corr(mCorr) n(500) clear
corr
g y=1+2*x+u
reg y x
ivregress gmm y (x=z1)
ivregress gmm y (x=z1 z2)
ivregress 2sls y (x=z1 z2)

```

```

. set obs 500
number of observations (_N) was 0, now 500

. capture drop y x z* u

. mat mCorr = (1, .7, .7, .5 \ .7, 1, .4, 0 \ .7, .4, 1, 0 \ .5, 0, 0, 1)

. corr2data x z1 z2 u, mean(10 20 15 0) sd(3 4 5 20) corr(mCorr) n(500) clear
(obs 500)

. corr
(obs=500)

```

	x	z1	z2	u
x	1.0000			
z1	0.7000	1.0000		
z2	0.7000	0.4000	1.0000	
u	0.5000	-0.0000	-0.0000	1.0000

```

. g y=1+2*x+u
. reg y x

```

Source	SS	df	MS	Number of obs	=	500
Model	127743.999	1	127743.999	F(1, 498)	=	424.96
Residual	149700	498	300.602411	Prob > F	=	0.0000
				R-squared	=	0.4604
				Adj R-squared	=	0.4593
Total	277443.999	499	555.999998	Root MSE	=	17.338

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
x	5.333333	.2587168	20.61	0.000	4.825022	5.841644
_cons	-32.33333	2.70086	-11.97	0.000	-37.63982	-27.02685

```

. ivregress gmm y (x=z1)

```

```

Instrumental variables (GMM) regression
Number of obs      =      500
Wald chi2(1)       =      24.63
Prob > chi2         =      0.0000
R-squared           =      0.2806
Root MSE           =      19.98

GMM weight matrix: Robust

```

y	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
x	2	.4030215	4.96	0.000	1.210092	2.789908
_cons	.9999999	4.004532	0.25	0.803	-6.848738	8.848738

```

Instrumented:  x
Instruments:   z1

```

```
. ivregress gmm y (x=z1 z2)
```

```
Instrumental variables (GMM) regression      Number of obs   =      500
                                             wald chi2(1)    =      35.10
                                             Prob > chi2      =      0.0000
                                             R-squared       =      0.2806
                                             Root MSE       =      19.98

GMM weight matrix: Robust
```

	y	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
	x	2	.3375669	5.92	0.000	1.338381	2.661619
	_cons	1	3.373732	0.30	0.767	-5.612393	7.612393

```
Instrumented:  x
Instruments:   z1 z2
```

```
. estat overid
```

Test of overidentifying restriction:

Hansen's J chi2(1) = 1.1e-14 (p = 1.0000)

```
. ivregress 2sls y (x=z1 z2)
```

```
Instrumental variables (2SLS) regression      Number of obs   =      500
                                             wald chi2(1)    =      31.50
                                             Prob > chi2      =      0.0000
                                             R-squared       =      0.2806
                                             Root MSE       =      19.98
```

	y	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
	x	2	.3563483	5.61	0.000	1.30157	2.69843
	_cons	1	3.673801	0.27	0.785	-6.200517	8.200518

```
Instrumented:  x
Instruments:   z1 z2
```

Example 3

```
set obs 500
capture drop y x z* u
mat mCorr = (1, .7, .3, .5 \ .7, 1, .4, .2 \ .3, .4, 1, .3 \ .5, .2, .3, 1)
corr2data x z1 z2 u, mean(10 20 15 0) sd(3 4 5 20) corr(mCorr) n(5000) clear
corr
g y=1+2*x+u
reg y x
ivregress gmm y (x=z1)
ivregress gmm y (x=z1 z2)
estat overid
ivregress 2sls y (x=z1 z2)
```

```
. set obs 500
number of observations (_N) was 0, now 500

. capture drop y x z* u

. mat mCorr = (1, .7, .3, .5 \ .7, 1, .4, .2 \ .3, .4, 1, .3 \ .5, .2, .3, 1)

. corr2data x z1 z2 u, mean(10 20 15 0) sd(3 4 5 20) corr(mCorr) n(5000) clear
(obs 5,000)

. corr
(obs=5,000)
```

		x	z1	z2	u
	x	1.0000			
	z1	0.7000	1.0000		
	z2	0.3000	0.4000	1.0000	
	u	0.5000	0.2000	0.3000	1.0000

```
. g y=1+2*x+u
```

```
. reg y x
```

Source	SS	df	MS	Number of obs	=	5,000
Model	1279743.99	1	1279743.99	F(1, 4998)	=	4264.96
Residual	1499700	4,998	300.060025	Prob > F	=	0.0000
				R-squared	=	0.4604
				Adj R-squared	=	0.4603
Total	2779444	4,999	556	Root MSE	=	17.322

y	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
x	5.333333	.081666	65.31	0.000	5.173232	5.493435
_cons	-32.33333	.852611	-37.92	0.000	-34.00483	-30.66184

```
. ivregress gmm y (x=z1)
```

```
Instrumental variables (GMM) regression
```

Number of obs	=	5,000
wald chi2(1)	=	1044.34
Prob > chi2	=	0.0000
R-squared	=	0.4274
Root MSE	=	17.841

GMM weight matrix: Robust

y	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
x	3.904762	.1208297	32.32	0.000	3.66794	4.141584
_cons	-18.04762	1.229494	-14.68	0.000	-20.45738	-15.63785

```
Instrumented: x
Instruments: z1
```

```
. ivregress gmm y (x=z1 z2)
```

```
Instrumental variables (GMM) regression
```

Number of obs	=	5,000
wald chi2(1)	=	1103.18
Prob > chi2	=	0.0000
R-squared	=	0.4315
Root MSE	=	17.776

GMM weight matrix: Robust

y	Coef.	Robust Std. Err.	z	P> z	[95% Conf. Interval]	
x	3.997248	.1203476	33.21	0.000	3.761371	4.233125
_cons	-19.00297	1.224456	-15.52	0.000	-21.40286	-16.60308

```
Instrumented: x
Instruments: z1 z2
```

```
. estat overid
```

Test of overidentifying restriction:

Hansen's J chi2(1) = 293.132 (p = 0.0000)

```
. ivregress 2sls y (x=z1 z2)
```

```
Instrumental variables (2SLS) regression
```

Number of obs	=	5,000
wald chi2(1)	=	1100.94
Prob > chi2	=	0.0000
R-squared	=	0.4305
Root MSE	=	17.792

y	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
x	3.97411	.119773	33.18	0.000	3.739359	4.208861
_cons	-18.7411	1.223875	-15.31	0.000	-21.13985	-16.34235

```
Instrumented: x
Instruments: z1 z2
```