# Value and Consumption Low vs Higher Interest Rates Results Comparison

This is the example vignette for function: <a href="main\_bisec\_vec">snw\_vfi\_main\_bisec\_vec</a> from the <a href="PrjOptiSNW Package">PrjOptiSNW Package</a>. This function solves for the V(states) for individuals at lower and higher savings interest rate. Note that welfare improves for all when interest rate goes up for savings in a model where borrowing is not allowed. However, an increase change in the interest rate generates both an income effect (higher resources) and also changes relative price of consumption today vs tomorrow. The change in income increases incentive to consume, the change in relative price depresses incentives to consume today. The combined effect of rising interest rate on savings on consumption/savings differs by the state-space, households might overall consume more or less depending on their state-space.

## Solve Model at 4 Percent Interest Rate

Solve the benchmark model at 4 percent savings interest rate.

```
% mp_params = snw_mp_param('default_dense');
mp_params = snw_mp_param('default_docdense');
mp_params('beta') = 0.95;
fl_higher_r = 0.04;
fl_lower_r = 0.02;
mp_params('r') = fl_higher_r;
mp_controls = snw_mp_control('default_test');
mp_controls('bl_print_vfi') = false;
mp_controls('bl_print_vfi_verbose') = false;
mp_controls('bl_timer') = true;
[V_ss_r04,~,cons_ss_r04,~] = snw_vfi_main_bisec_vec(mp_params, mp_controls);
```

Completed SNW VFI MAIN BISEC VEC; SNW MP PARAM=default docdense; SNW MP CONTROL=default test; time=496.1817

## Solve Model at 2 Percent Interest Rate

Solve the benchmark model at 2 percent savings interest rate.

```
mp_params('r') = fl_lower_r;
mp_controls = snw_mp_control('default_test');
mp_controls('bl_print_vfi') = false;
mp_controls('bl_print_vfi_verbose') = false;
mp_controls('bl_timer') = true;
[V_ss_r02,~,cons_ss_r02,~] = snw_vfi_main_bisec_vec(mp_params, mp_controls);
```

 ${\tt Completed SNW\_VFI\_MAIN\_BISEC\_VEC; SNW\_MP\_PARAM= default\_docdense; SNW\_MP\_CONTROL= default\_test; time=495.0637}$ 

# **Generate Interest Rate Comparison Matrixes**

Take the difference between 4 percent and 2 percent savings interest rate results. When interest rates are higher, greater incentive to save, but leads to heterogeneous responses by income and other characteristics, note that this changes both relative prices as well as total resource/budget, so there is both income and price

effects that differ in magnitudes depending on the individual's statespace. Welfare does improve for all higher higher r. Welfare is converted to units in fixed life-time consumption.

```
gamma = mp_params('gamma');
mn_V_gain_r = snw_hh_welfare(V_ss_r04, gamma) - snw_hh_welfare(V_ss_r02, gamma);
mn_C_gain_r = cons_ss_r04 - cons_ss_r02;
fl_r_gap = fl_higher_r - fl_lower_r;
```

## **Dense Param Results Define Frames**

Define the matrix dimensions names and dimension vector values. Policy and Value Functions share the same ND dimensional structure.

```
% Grids:
age_grid = 18:100;
agrid = mp_params('agrid')';
eta_H_grid = mp_params('eta_H_grid')';
eta_S_grid = mp_params('eta_S_grid')';
ar_st_eta_HS_grid = string(cellstr([num2str(eta_H_grid', 'hz=%3.2f;'), num2str(eta_S_grid', 'wz
edu_grid = [0,1];
marry_grid = [0,1];
kids_grid = (1:1:mp_params('n_kidsgrid'))';
% NaN(n_jgrid,n_agrid,n_etagrid,n_educgrid,n_marriedgrid,n_kidsgrid);
cl_mp_datasetdesc = {};
cl_mp_datasetdesc{1} = containers.Map({'name', 'labval'}, {'age', age_grid});
cl_mp_datasetdesc{2} = containers.Map({'name', 'labval'}, {'savings', agrid});
cl_mp_datasetdesc{3} = containers.Map({'name', 'labval'}, {'eta', 1:length(eta_H_grid)});
cl_mp_datasetdesc{4} = containers.Map({'name', 'labval'}, {'edu', edu_grid});
cl_mp_datasetdesc{5} = containers.Map({'name', 'labval'}, {'marry', marry_grid});
cl_mp_datasetdesc{6} = containers.Map({'name', 'labval'}, {'kids', kids_grid});
```

# Analyze Difference in V and C with Higher and Lower Savings Interest Rate

The difference between V and C with higher and lower r.

```
% Generate some Data
mp_support_graph = containers.Map('KeyType', 'char', 'ValueType', 'any');
mp_support_graph('cl_st_xtitle') = {'Savings States, a'};
mp_support_graph('st_legend_loc') = 'eastoutside';
mp_support_graph('bl_graph_logy') = true; % do not log
mp_support_graph('it_legend_select') = 21; % how many shock legends to show
mp_support_graph('cl_colors') = 'jet';
```

 $MEAN(MN_V_GAIN(A,Z))$ 

Tabulate value and policies along savings and shocks:

```
% Set
ar_permute = [1,4,5,6,3,2];
% Value Function
st_title = ['MEAN(MN_V_Gain(A,Z)), r_gap=' num2str(fl_r_gap) ];
tb_az_v = ff_summ_nd_array(st_title, mn_V_gain_r, true, ["mean"], 4, 1, cl_mp_datasetdesc, ar_p
```

```
savings
   group
                      mean_eta_1
                                  mean_eta_2
                                              mean_eta_3
                                                          mean_eta_4
                                                                      mean_eta_5
                                                                                  mean_eta_6
                                                                                              mean_
     1
                      4.6652e-05
                                   4.802e-05
                                              4.9487e-05
                                                          5.1045e-05
                                                                      5.2688e-05
                                                                                  5.4417e-05
                                                                                              5.623
     2
           0.00051498
                       4.725e-05
                                  4.8601e-05
                                              5.0052e-05
                                                          5.1595e-05
                                                                      5.3225e-05
                                                                                  5.4941e-05
                                                                                              5.674
     3
           0.0041199
                      5.1273e-05
                                  5.2516e-05
                                              5.3868e-05
                                                           5.532e-05
                                                                      5.6863e-05
                                                                                  5.8496e-05
                                                                                              6.022
                                                                      6.5996e-05
     4
            0.013905
                       6.118e-05
                                  6.2221e-05
                                              6.3373e-05
                                                          6.4635e-05
                                                                                  6.7452e-05
                                                                                              6.900
     5
            0.032959
                      7.7395e-05
                                  7.8196e-05
                                              7.9136e-05
                                                          8.0192e-05
                                                                      8.1347e-05
                                                                                  8.2598e-05
                                                                                              8.394
     6
            0.064373
                      9.9291e-05
                                  9.9974e-05
                                              0.00010079
                                                          0.00010171
                                                                      0.00010271
                                                                                  0.00010381
                                                                                              0.000
% Consumption
st_title = ['MEAN(MN C Gain(A,Z)), r_gap=' num2str(fl_r_gap) ];
tb_az_c = ff_summ_nd_array(st_title, mn_C_gain_r, true, ["mean"], 4, 1, cl_mp_datasetdesc, ar_r
savings
                      mean eta 1
                                   mean eta 2
   group
                                                mean eta 3
                                                             mean eta 4
                                                                         mean eta 5
                                                                                      mean eta 6
     1
                               0
                                                                     0
```

9.0392e-06

5.5968e-05

0.00017196

0.0001617

5.4936e-05

9.0202e-06

5.437e-05

0.00016519

0.00014482

2.3556e-05

9.0007e-06

5.3887e-05

0.00015748

0.00014062

1.5431e-05

8.9809e-06

5.3137e-05

0.0001458

0.00014373

2.7756e-05

#### Graph Mean Values:

2

3

4

5

0.00051498

0.0041199

0.013905

0.032959

0.064373

9.0758e-06

6.9258e-05

0.0001131

0.00026095

0.00022015

```
st_title = ['MEAN(MN\_V\_GAIN(A,Z)), r\_gap=' num2str(fl_r_gap) ''];
mp_support_graph('cl_st_graph_title') = {st_title};
mp_support_graph('cl_st_ytitle') = {'MEAN(MN\_V\_GAIN(a,z))'};
ff_graph_grid((tb_az_v{1:end, 3:end})', ar_st_eta_HS_grid, agrid, mp_support_graph);
```

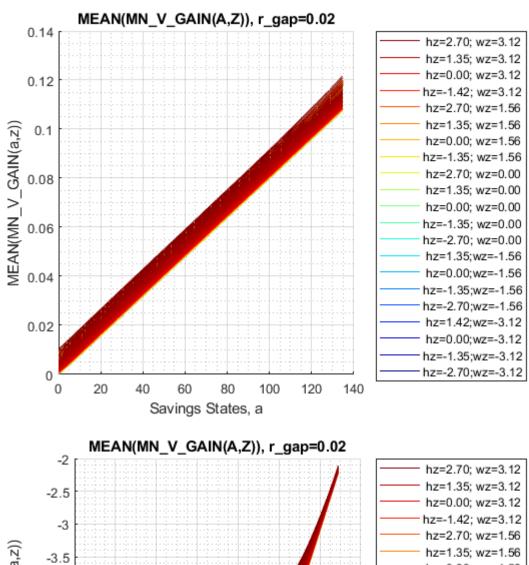
9.0577e-06

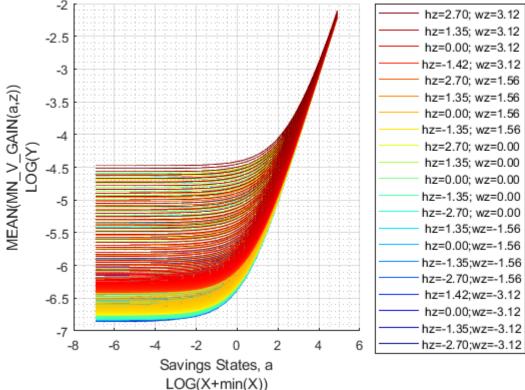
5.9037e-05

0.00016511

0.00020294

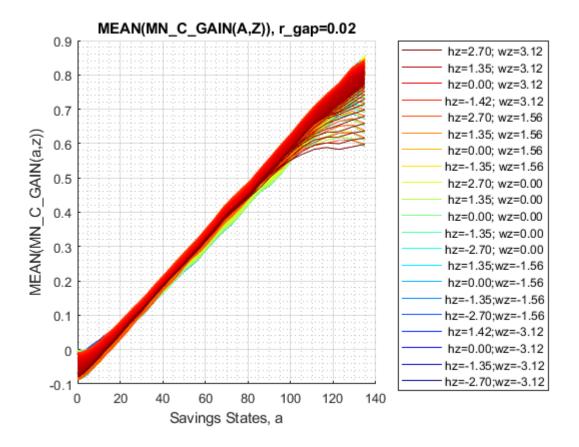
0.00011898

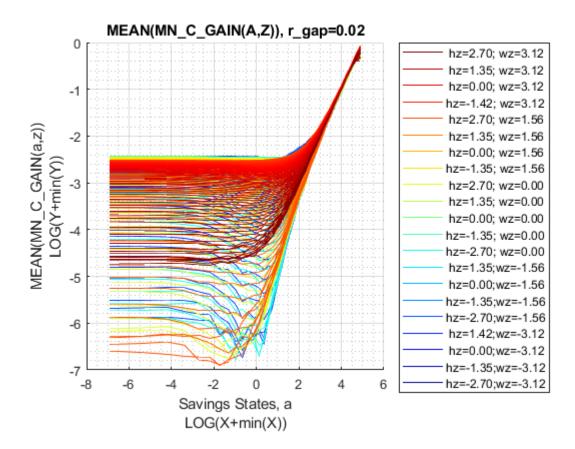




Graph Mean Consumption:

```
st_title = ['MEAN(MN\_C\_GAIN(A,Z)), r\_gap=' num2str(fl_r_gap) ''];
mp_support_graph('cl_st_graph_title') = {st_title};
mp_support_graph('cl_st_ytitle') = {'MEAN(MN\_C\_GAIN(a,z))'};
ff_graph_grid((tb_az_c{1:end, 3:end})', ar_st_eta_HS_grid, agrid, mp_support_graph);
```





# **Analyze Kids and Marriage and Age**

Aggregating over education, savings, and shocks, what are the differential effects of Marriage and Age.

```
% Generate some Data
mp_support_graph = containers.Map('KeyType', 'char', 'ValueType', 'any');
ar_row_grid = [...
    "k0M0", "K1M0", "K2M0", "K3M0", "K4M0", ...
    "k0M1", "K1M1", "K2M1", "K3M1", "K4M1"];
mp_support_graph('cl_st_xtitle') = {'Age'};
mp_support_graph('st_legend_loc') = 'best';
mp_support_graph('bl_graph_logy') = true; % do not log
mp_support_graph('st_rounding') = '6.2f'; % format shock legend
mp_support_graph('cl_scatter_shapes') = {...
    'o', 'd', 's', 'x', '*'};
mp_support_graph('cl_colors') = {...
    'red', 'red', 'red', 'red'...
    'blue', 'blue', 'blue', 'blue'};
```

MEAN(VAL(KM,J)), MEAN(AP(KM,J)), MEAN(C(KM,J))

Tabulate value and policies:

```
% Set
% NaN(n_jgrid,n_agrid,n_etagrid,n_educgrid,n_marriedgrid,n_kidsgrid);
ar_permute = [2,3,4,1,6,5];
% Value Function
```

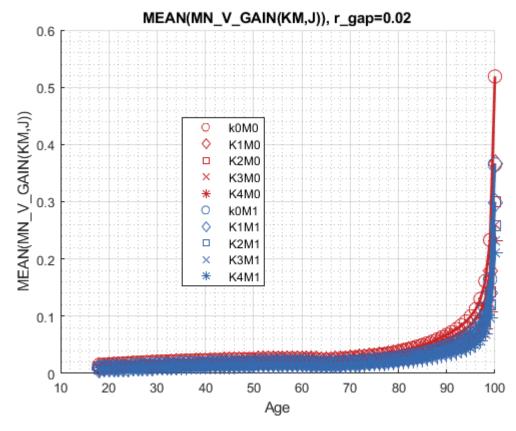
```
tb_az_v = ff_summ_nd_array(st_title, mn_v_gain_r, true, ["mean"], 3, 1, cl_mp_datasetdesc, ar_r
group
           kids
                   marry
                           mean_age_18
                                        mean_age_19
                                                      mean_age_20
                                                                    mean_age_21
                                                                                  mean_age_22
                                                                                               mean_age_23
     1
            1
                             0.015087
                                          0.015546
                                                        0.016026
                                                                       0.01653
                                                                                    0.017025
                                                                                                 0.017508
     2
            2
                    0
                             0.011707
                                          0.012061
                                                        0.012444
                                                                      0.012858
                                                                                    0.013278
                                                                                                 0.013703
     3
            3
                    0
                             0.010499
                                          0.010777
                                                        0.011081
                                                                      0.011412
                                                                                    0.011749
                                                                                                 0.012091
     4
            4
                    0
                            0.0094894
                                         0.0097107
                                                       0.0099547
                                                                      0.010224
                                                                                     0.0105
                                                                                                 0.010781
     5
            5
                    0
                              0.00886
                                         0.0090319
                                                       0.0092234
                                                                     0.0094371
                                                                                   0.0096558
                                                                                                0.0098787
            1
     6
                    1
                             0.010229
                                          0.010593
                                                        0.010984
                                                                      0.011405
                                                                                    0.011831
                                                                                                 0.012259
     7
            2
                    1
                             0.008528
                                         0.0087978
                                                       0.0090932
                                                                     0.0094182
                                                                                   0.0097531
                                                                                                 0.010096
     8
            3
                    1
                            0.0079012
                                         0.0081252
                                                       0.0083716
                                                                     0.0086439
                                                                                   0.0089244
                                                                                                0.0092117
     9
            4
                    1
                            0.0073246
                                         0.0075062
                                                       0.0077075
                                                                     0.0079319
                                                                                   0.0081637
                                                                                                0.0084017
    10
                                                                                                0.0078372
                              0.00696
                                         0.0071075
                                                       0.0072717
                                                                     0.0074553
                                                                                    0.007644
% Consumption Function
st_title = ['MEAN(MN_C_Gain(KM,J)), r_gap=' num2str(fl_r_gap) ];
tb_az_c = ff_summ_nd_array(st_title, mn_C_gain_r, true, ["mean"], 3, 1, cl_mp_datasetdesc, ar_r
```

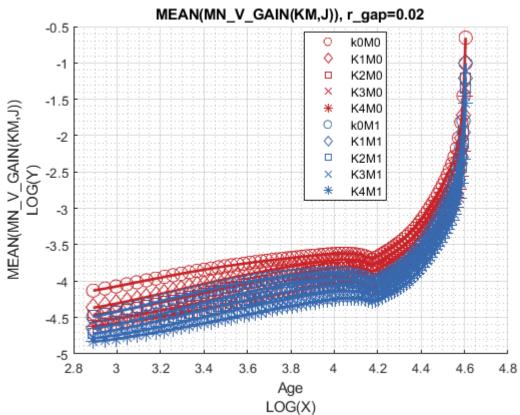
st title = ['MEAN(MN V Gain(KM,J)), r gap=' num2str(fl r gap) ];

xxx MEAN(I	MN_C_Gair	n(KM,J)),	r_gap=0.02 xxx	XXXXXXXXXXXXXXXX	XXXXXXXX			
group	kids	marry	mean_age_18	mean_age_19	mean_age_20	mean_age_21	mean_age_22	mean_age_2
								<del></del>
1	1	0	0.055699	0.056856	0.058874	0.061289	0.064119	0.067235
2	2	0	0.051038	0.052145	0.054007	0.055436	0.057546	0.060312
3	3	0	0.050801	0.05338	0.054644	0.055147	0.05643	0.058389
4	4	0	0.050711	0.052919	0.054103	0.053992	0.05454	0.055872
5	5	0	0.052196	0.053297	0.054608	0.053912	0.053811	0.054449
6	1	1	-0.022203	-0.022187	-0.021707	-0.021466	-0.020507	-0.018751
7	2	1	-0.019706	-0.020509	-0.020942	-0.02173	-0.021723	-0.020839
8	3	1	-0.016291	-0.017215	-0.017899	-0.019714	-0.02062	-0.020552
9	4	1	-0.013267	-0.013589	-0.013895	-0.016006	-0.017459	-0.018261
10	5	1	-0.0131	-0.012701	-0.011796	-0.013633	-0.014917	-0.015545

#### Graph Mean Values:

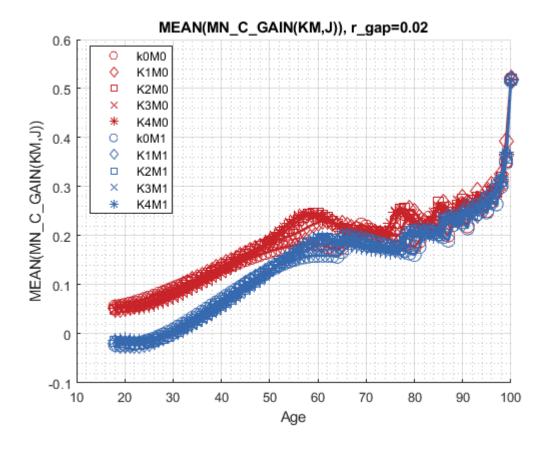
```
st_title = ['MEAN(MN\_V\_GAIN(KM,J)), r\_gap=' num2str(fl_r_gap) ''];
mp_support_graph('cl_st_graph_title') = {st_title};
mp_support_graph('cl_st_ytitle') = {'MEAN(MN\_V\_GAIN(KM,J))'};
ff_graph_grid((tb_az_v{1:end, 4:end}), ar_row_grid, age_grid, mp_support_graph);
```

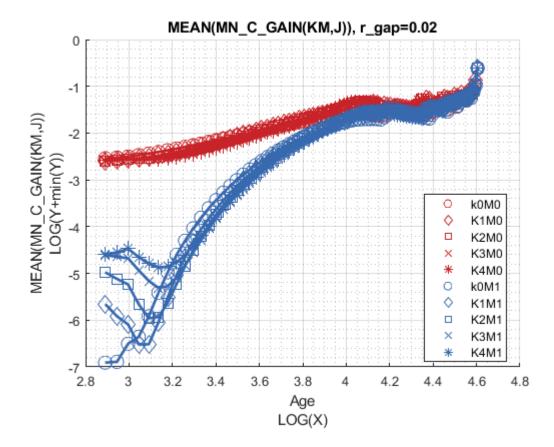




**Graph Mean Consumption:** 

```
st_title = ['MEAN(MN\_C\_GAIN(KM,J)), r\_gap=' num2str(fl_r_gap) ''];
mp_support_graph('cl_st_graph_title') = {st_title};
mp_support_graph('cl_st_ytitle') = {'MEAN(MN\_C\_GAIN(KM,J))'};
ff_graph_grid((tb_az_c{1:end, 4:end}), ar_row_grid, age_grid, mp_support_graph);
```





# **Analyze Education and Marriage**

Aggregating over education, savings, and shocks, what are the differential effects of Marriage and Age.

```
% Generate some Data
mp_support_graph = containers.Map('KeyType', 'char', 'ValueType', 'any');
ar_row_grid = ["E0M0", "E1M0", "E0M1", "E1M1"];
mp_support_graph('cl_st_xtitle') = {'Age'};
mp_support_graph('st_legend_loc') = 'best';
mp_support_graph('bl_graph_logy') = true; % do not log
mp_support_graph('st_rounding') = '6.2f'; % format shock legend
mp_support_graph('cl_scatter_shapes') = {'*', 'p', '*', 'p' };
mp_support_graph('cl_colors') = {'red', 'red', 'blue', 'blue'};
```

MEAN(VAL(EM,J)), MEAN(AP(EM,J)), MEAN(C(EM,J))

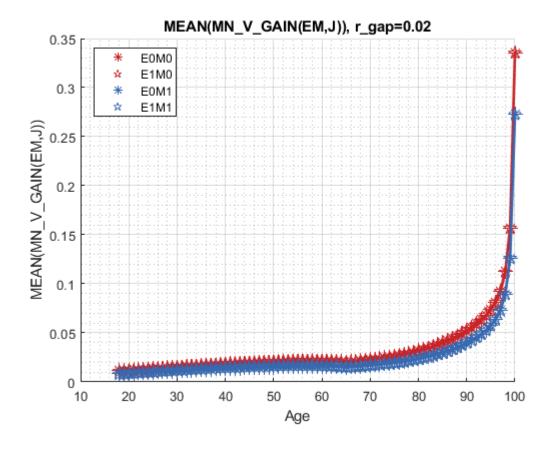
Tabulate value and policies:

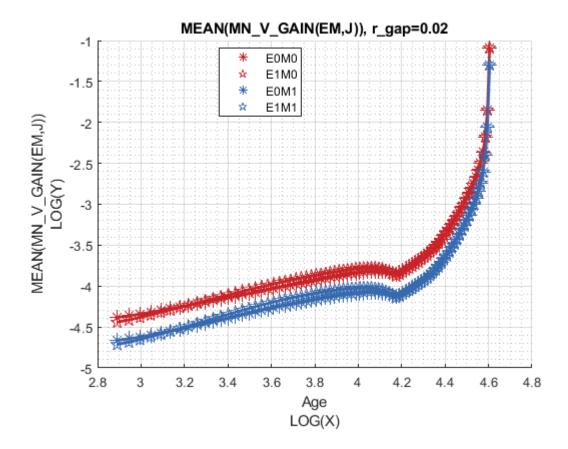
1	0	0	0.011474	0.011736	0.012012	0.012305	0.012599	0.012892	
2	1	0	0.010783	0.011115	0.011479	0.011879	0.012285	0.012692	
3	0	1	0.0084099	0.0086195	0.0088444	0.0090861	0.0093328	0.0095836	
4	1	1	0.0079671	0.0082324	0.0085268	0.0088556	0.0091936	0.0095387	
<pre>% Consumption st_title = ['MEAN(MN_C_Gain(EM,J)), r_gap=' num2str(fl_r_gap) ]; tb_az_c = ff_summ_nd_array(st_title, mn_C_gain_r, true, ["mean"], 3, 1, cl_mp_datasetdesc, ar_</pre>									

xxx MEAN(MN_C_Gain(EM,J)), r_gap=0.02 xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx								
grou	p edu	marry	mean_age_18	mean_age_19	mean_age_20	mean_age_21	mean_age_22	mean_age_23
1	0	0	0.06916	0.070507	0.072258	0.07446	0.076888	0.07952
2	1	0	0.035018	0.036932	0.038237	0.03745	0.037691	0.038982
3	0	1	-0.0014877	-0.001014	-0.00023405	0.00057466	0.0018134	0.0035145
4	1	1	-0.032339	-0.033466	-0.034261	-0.037595	-0.039904	-0.041093

## Graph Mean Values:

```
st_title = ['MEAN(MN\_V\_GAIN(EM,J)), r\_gap=' num2str(fl_r_gap) ''];
mp_support_graph('cl_st_graph_title') = {st_title};
mp_support_graph('cl_st_ytitle') = {'MEAN(MN\_V\_GAIN(EM,J))'};
ff_graph_grid((tb_az_v{1:end, 4:end}), ar_row_grid, age_grid, mp_support_graph);
```





## Graph Mean Consumption:

```
st_title = ['MEAN(MN\_C\_GAIN(EM,J)), r\_gap=' num2str(fl_r_gap) ''];
mp_support_graph('cl_st_graph_title') = {st_title};
mp_support_graph('cl_st_ytitle') = {'MEAN(MN\_C\_GAIN(EM,J))'};
ff_graph_grid((tb_az_c{1:end, 4:end}), ar_row_grid, age_grid, mp_support_graph);
```

