

## 2020 V and C without Unemployment

This is the example vignette for function: [snw\\_a4chk\\_wrk\\_bisec\\_vec](#) from the [PrjOptiSNW Package](#). This function solves for the V(states, check) for individuals working. Dense solution. Bisection, most time for the test here taken to generate the income matrixes. But these can be generated out of the check loops.

### Test SNW\_A4CHK\_WRK\_BISEC\_VEC Defaults Dense

Call the function with default parameters. Solve first for non-covid value and policy. Then depending on 2020 taxes, solve for 2020 policy and value.

```
mp_params = snw_mp_param('default_docdense');
% mp_params = snw_mp_param('default_dense');
mp_params('beta') = 0.95;
mp_controls = snw_mp_control('default_test');
mp_controls('bl_print_vfi') = false;
mp_controls('bl_timer') = true;
[V_ss,~,cons_ss,~] = 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=501.6001

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CONTAINER NAME: mp\_outcomes ND Array (Matrix etc)

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	i	idx	ndim	numel	rowN	colN	sum	mean	std	coefvari
	—	—	—	—	—	—	—	—	—	—
V_VFI	1	1	6	4.37e+07	83	5.265e+05	-6.6619e+08	-15.245	21.865	-1.4343
ap_VFI	2	2	6	4.37e+07	83	5.265e+05	1.3967e+09	31.962	36.426	1.1397
cons_VFI	3	3	6	4.37e+07	83	5.265e+05	2.3276e+08	5.3263	8.4413	1.5848

xxx TABLE:V\_VFI XXXXXXXXXXXXXXXXXXXXXXX

	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499
	—	—	—	—	—	—	—	—	—
r1	-293.96	-293.57	-291.09	-285.44	-276.41	-4.3584	-4.2643	-4.1713	-4.0795
r2	-284.42	-284.03	-281.55	-275.97	-267.24	-4.2519	-4.1612	-4.0717	-3.9832
r3	-274.87	-274.48	-272.03	-266.62	-258.33	-4.1429	-4.0559	-3.9698	-3.8847
r4	-265.22	-264.86	-262.58	-257.53	-249.74	-4.0309	-3.9475	-3.8649	-3.7833
r5	-256.51	-256.17	-254.04	-249.3	-241.96	-3.9252	-3.8452	-3.7659	-3.6873
r79	-13.642	-13.628	-13.535	-13.298	-12.896	-0.22092	-0.21058	-0.20086	-0.19173
r80	-12.283	-12.269	-12.176	-11.939	-11.537	-0.16979	-0.16182	-0.1543	-0.14722
r81	-10.605	-10.591	-10.498	-10.261	-9.8589	-0.11712	-0.11163	-0.10646	-0.10157
r82	-8.3494	-8.3358	-8.2424	-8.0055	-7.6035	-0.065333	-0.062242	-0.05936	-0.056635
r83	-5.0665	-5.0529	-4.9595	-4.7226	-4.3206	-0.020968	-0.019972	-0.019038	-0.018161

xxx TABLE:ap\_VFI XXXXXXXXXXXXXXXXXXXXXXX

	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	c526500
	—	—	—	—	—	—	—	—	—	—
r1	0	0	0.00051498	0.0066578	0.021589	112.13	117.67	123.4	129.31	135.72
r2	0	0	0.00051498	0.0057684	0.020245	112.17	117.71	123.43	129.34	135.76
r3	0	0	0.00020768	0.0041456	0.018539	112.2	117.73	123.45	129.37	135.78
r4	0	0	0.00010346	0.0041199	0.018307	112.86	118.39	124.11	130.03	136.44
r5	0	0	5.2907e-06	0.0041199	0.018091	113.53	119.07	124.79	130.71	137.12
r79	0	0	0	0	0	81.091	85.364	89.335	93.258	97.348
r80	0	0	0	0	0	76.124	79.747	83.431	86.986	90.578
r81	0	0	0	0	0	67.945	70.639	73.673	76.991	81.091
r82	0	0	0	0	0	50.126	53.467	56.302	57.884	60.587

r83	0	0	0	0	0	0	0	0	0	0
xxx TABLE:cons_VFI xxxxxxxxxxxxxxxxxxxx										
	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	
r1	0.036717	0.037251	0.040477	0.044486	0.049324	12.265	12.55	12.844	13.145	1
r2	0.036717	0.037251	0.040477	0.045375	0.050668	12.501	12.787	13.082	13.383	1
r3	0.036717	0.037251	0.040784	0.046998	0.052374	12.755	13.042	13.337	13.638	1
r4	0.038144	0.038678	0.042314	0.048449	0.054031	13	13.289	13.584	13.883	1
r5	0.039534	0.040068	0.043802	0.049839	0.055635	13.236	13.525	13.821	14.116	1
r79	0.19737	0.19791	0.20163	0.21175	0.23145	35.811	37.362	39.409	41.7	4
r80	0.19737	0.19791	0.20163	0.21175	0.23145	40.752	42.953	45.286	47.946	5
r81	0.19737	0.19791	0.20163	0.21175	0.23145	48.909	52.039	55.022	57.919	6
r82	0.19737	0.19791	0.20163	0.21175	0.23145	66.71	69.193	72.375	77.007	
r83	0.19737	0.19791	0.20163	0.21175	0.23145	116.82	122.65	128.66	134.88	1

```

welf_checks = 2; % 2 checks is $200 dollar of welfare checks
xi=1; % xi=0 full income loss from covid shock, xi=1, no covid income losses
b=1; % when xi=1, b does not matter, no income losses
TR = 100/58056;
mp_params('TR') = TR;
mp_params('xi') = xi;
mp_params('b') = b;
% if = mp_params('a2_covidyr_manna_heaven'), V_emp_2020 same as V_ss if b=1
% or xi=1.
% if = mp_params('a2_covidyr_tax_fully_pay'), V_emp_2020 differ due to 2020
% tax differences
mp_params('a2_covidyr') = mp_params('a2_covidyr_manna_heaven');
% mp_params('a2_covidyr') = mp_params('a2_covidyr_tax_fully_pay');
[V_emp_2020,~,cons_emp_2020,~] = snw_vfi_main_bisec_vec(mp_params, mp_controls, V_ss);

```

Completed SNW\_VFI\_MAIN\_BISEC\_VEC 1 Period Unemp Shock;SNW\_MP\_PARAM=default\_docdense;SNW\_MP\_CONTROL=default\_test;time

xx										
CONTAINER NAME: mp_outcomes ND Array (Matrix etc)										
xx										
	i	idx	ndim	numel	rowN	colN	sum	mean	std	coefvari
V_VFI	1	1	6	4.37e+07	83	5.265e+05	-6.6619e+08	-15.245	21.865	-1.4343
ap_VFI	2	2	6	4.37e+07	83	5.265e+05	1.3967e+09	31.962	36.426	1.1397
cons_VFI	3	3	6	4.37e+07	83	5.265e+05	2.3276e+08	5.3263	8.4413	1.5848
xxx TABLE:V_VFI xxxxxxxxxxxxxxxxxxxx										
	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	
r1	-293.96	-293.57	-291.09	-285.44	-276.41	-4.3584	-4.2643	-4.1713	-4.0795	
r2	-284.42	-284.03	-281.55	-275.97	-267.24	-4.2519	-4.1612	-4.0717	-3.9832	
r3	-274.87	-274.48	-272.03	-266.62	-258.33	-4.1429	-4.0559	-3.9698	-3.8847	
r4	-265.22	-264.86	-262.58	-257.53	-249.74	-4.0309	-3.9475	-3.8649	-3.7833	
r5	-256.51	-256.17	-254.04	-249.3	-241.96	-3.9252	-3.8452	-3.7659	-3.6873	
r79	-13.642	-13.628	-13.535	-13.298	-12.896	-0.22092	-0.21058	-0.20086	-0.19173	
r80	-12.283	-12.269	-12.176	-11.939	-11.537	-0.16979	-0.16182	-0.1543	-0.14722	
r81	-10.605	-10.591	-10.498	-10.261	-9.8589	-0.11712	-0.11163	-0.10646	-0.10157	
r82	-8.3494	-8.3358	-8.2424	-8.0055	-7.6035	-0.065333	-0.062242	-0.05936	-0.056635	
r83	-5.0665	-5.0529	-4.9595	-4.7226	-4.3206	-0.020968	-0.019972	-0.019038	-0.018161	

xxx TABLE:ap_VFI xxxxxxxxxxxxxxxxxxxx										
	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	c526500

[illegible]

	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	c526500
r1	0.036717	0.037251	0.040477	0.044486	0.049324	12.265	12.55	12.844	13.145	13.446
r2	0.036717	0.037251	0.040477	0.045375	0.050668	12.501	12.787	13.082	13.383	13.684
r3	0.036717	0.037251	0.040784	0.046998	0.052374	12.755	13.042	13.337	13.638	13.939
r4	0.038144	0.038678	0.042314	0.048449	0.054031	13	13.289	13.584	13.883	14.182
r5	0.039534	0.040068	0.043802	0.049839	0.055635	13.236	13.525	13.821	14.116	14.411
r79	0.19737	0.19791	0.20163	0.21175	0.23145	35.811	37.362	39.409	41.7	44.146
r80	0.19737	0.19791	0.20163	0.21175	0.23145	40.752	42.953	45.286	47.946	50.606
r81	0.19737	0.19791	0.20163	0.21175	0.23145	48.909	52.039	55.022	57.919	60.716
r82	0.19737	0.19791	0.20163	0.21175	0.23145	66.71	69.193	72.375	77.007	80.389
r83	0.19737	0.19791	0.20163	0.21175	0.23145	116.82	122.65	128.66	134.88	141.3

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CONTAINER NAME: mp\_container\_map ND Array (Matrix etc)

XX

	i	idx	ndim	numel	rowN	colN	sum	mean	std
	—	—	—	—	—	—	—	—	—
C_W	1	1	6	4.37e+07	83	5.265e+05	2.3278e+08	5.3269	8.4414
C_W_minus_C_ss	2	2	6	4.37e+07	83	5.265e+05	25096	0.00057428	0.00077346
V_W	3	3	6	4.37e+07	83	5.265e+05	-6.6561e+08	-15.231	21.817
V_W_minus_V_ss	4	4	6	4.37e+07	83	5.265e+05	5.8108e+05	0.013297	0.069536
mn_MPC	5	5	6	4.37e+07	83	5.265e+05	7.2848e+06	0.1667	0.22452

## 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=%3.2f;')], 'wz=%3.2f;'));
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 Check

The difference between V and V with Check, marginal utility gain given the check.

```
% 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\_CHECK(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_W_GAIN_CHECK(A,Z)), welf_checks=' num2str(welf_checks) ', TR=' num2str(m
tb_az_v = ff_summ_nd_array(st_title, mn_V_W_gain_check, true, ["mean"], 4, 1, cl_mp_datasetdesc
```

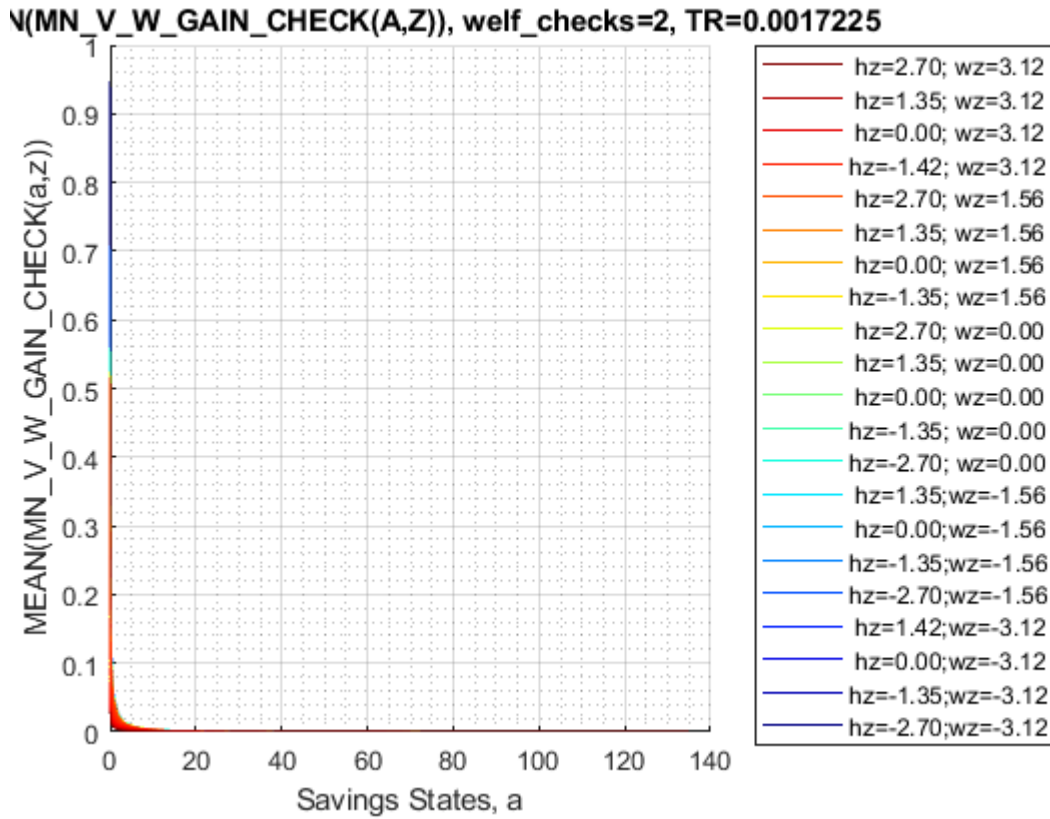
xxx	MEAN(MN_V_W_GAIN_CHECK(A,Z)), welf_checks=2, TR=0.0017225	xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx							
group	savings	mean_eta_1	mean_eta_2	mean_eta_3	mean_eta_4	mean_eta_5	mean_eta_6	mean_eta_7	mean_eta_8
1	0	0.94877	0.84952	0.761	0.68204	0.61169	0.54906	0.4906	0.4312
2	0.00051498	0.93806	0.84058	0.75354	0.6758	0.60645	0.54467	0.4862	0.4278
3	0.0041199	0.78107	0.71106	0.6451	0.58402	0.52829	0.47789	0.42789	0.37789
4	0.013905	0.59646	0.55133	0.50699	0.46454	0.42482	0.38817	0.34817	0.30817
5	0.032959	0.43075	0.404	0.37635	0.3489	0.3225	0.29766	0.27166	0.24566
6	0.064373	0.309	0.29293	0.27555	0.25773	0.24021	0.22343	0.20643	0.18943

```
% Consumption
st_title = ['MEAN(MN_MPC_W_GAIN_CHECK(A,Z)), welf_checks=' num2str(welf_checks) ', TR=' num2str(m
tb_az_c = ff_summ_nd_array(st_title, mn_MPC_W_gain_share_check, true, ["mean"], 4, 1, cl_mp_dat
```

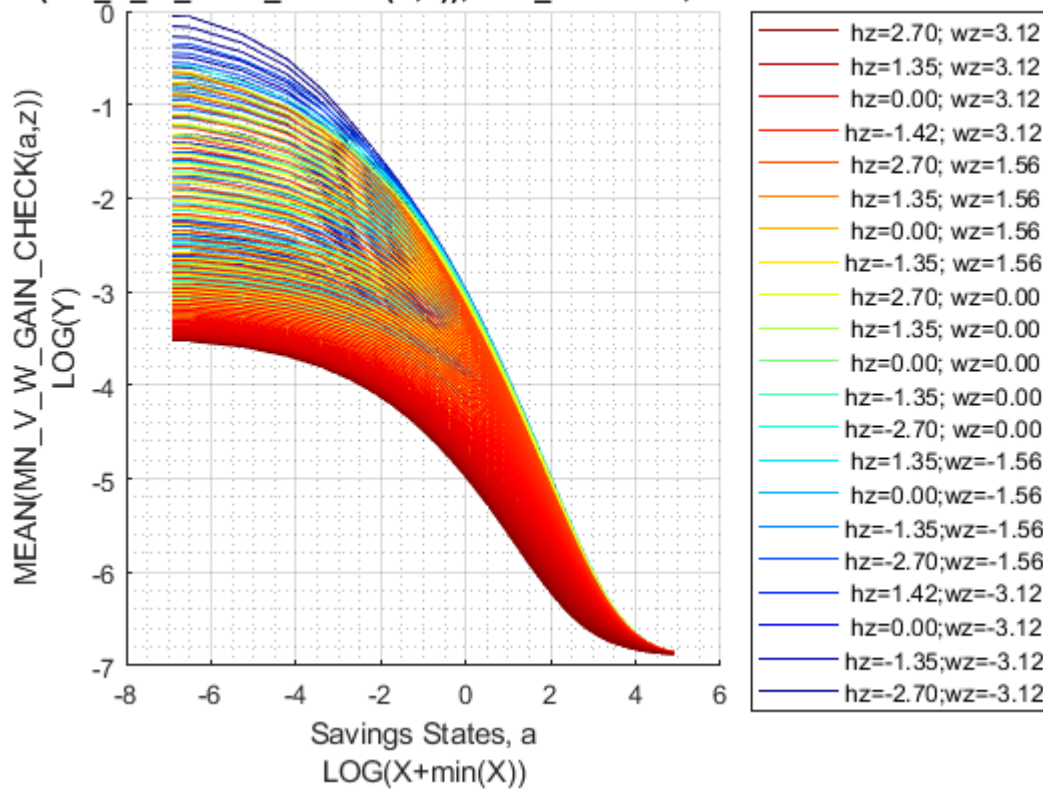
xxx	MEAN(MN_MPC_W_GAIN_CHECK(A,Z)), welf_checks=2, TR=0.0017225	xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx							
group	savings	mean_eta_1	mean_eta_2	mean_eta_3	mean_eta_4	mean_eta_5	mean_eta_6	mean_eta_7	mean_eta_8
1	0	0.99793	0.99517	0.99272	0.99139	0.99133	0.99164	0.99164	0.99164
2	0.00051498	0.99755	0.99428	0.99139	0.98982	0.98974	0.99011	0.99011	0.99011
3	0.0041199	0.88288	0.87696	0.87559	0.87507	0.87497	0.87532	0.87532	0.87532
4	0.013905	0.79641	0.79399	0.79087	0.78961	0.78966	0.79044	0.79044	0.79044
5	0.032959	0.71442	0.70961	0.70732	0.70675	0.70741	0.70894	0.70894	0.70894
6	0.064373	0.64147	0.64074	0.64117	0.64223	0.64376	0.6456	0.6456	0.6456

Graph Mean Values:

```
st_title = ['MEAN(MN\_V\_W\_GAIN\_CHECK(A,Z)), welf\_checks=' num2str(welf_checks) ', TR=' num2str(TR)];
mp_support_graph('cl_st_graph_title') = {st_title};
mp_support_graph('cl_st_ytitle') = {'MEAN(MN\_V\_W\_GAIN\_CHECK(a,z))'};
ff_graph_grid((tb_az_v{1:end, 3:end})), ar_st_eta_HS_grid, agrid, mp_support_graph);
```



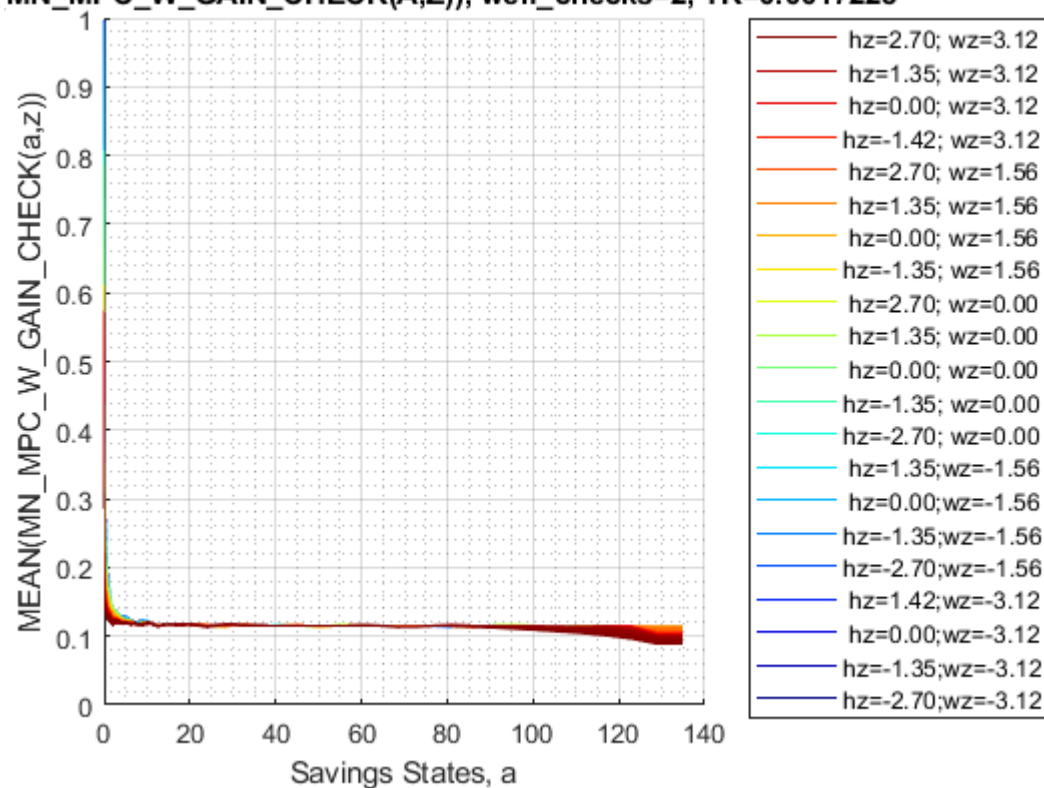
AN(MN\_V\_W\_GAIN\_CHECK(A,Z)), welf\_checks=2, TR=0.0017225



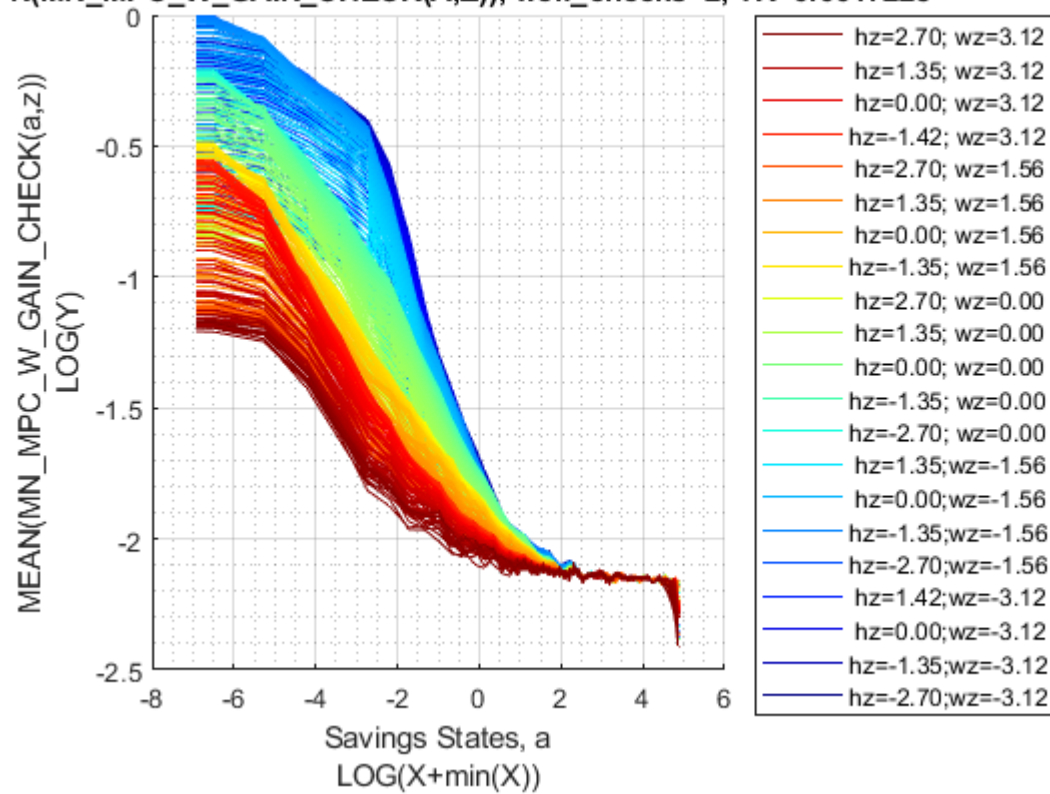
Graph Mean Consumption (**MPC: Share of Check Consumed**):

```
st_title = ['MEAN(MN\MPC\W\GAIN\CHECK(A,Z)), welf_checks=' num2str(welf_checks) ', TR=' num2str(TR)];
mp_support_graph('cl_st_graph_title') = {st_title};
mp_support_graph('cl_st_ytitle') = {'MEAN(MN\MPC\W\GAIN\CHECK(a,z))'};
ff_graph_grid((tb_az_c{1:end, 3:end}),'', ar_st_eta_HS_grid, agrid, mp_support_graph);
```

MN\_MPC\_W\_GAIN\_CHECK(A,Z)), welf\_checks=2, TR=0.0017225



N(MN\_MPC\_W\_GAIN\_CHECK(A,Z)), welf\_checks=2, TR=0.0017225



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', '*', ...
    'o', 'd', 's', 'x', '*'};
mp_support_graph('cl_colors') = {...
    'red', 'red', 'red', 'red', 'red'...
    'blue', '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
st_title = ['MEAN(MN_V_W_GAIN_CHECK(KM,J)), welf_checks=' num2str(welf_checks) ', TR=' num2str(
tb_az_v = ff_summ_nd_array(st_title, mn_V_W_gain_check, true, ["mean"], 3, 1, cl_mp_datasetdesco
```

```
xxx MEAN(MN_V_W_GAIN_CHECK(KM,J)), welf_checks=2, TR=0.0017225 xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx
group kids marry mean_age_18 mean_age_19 mean_age_20 mean_age_21 mean_age_22 mean_age_23
1 1 0 0.028443 0.027382 0.02607 0.023829 0.021959 0.020387
2 2 0 0.039131 0.037712 0.035894 0.032743 0.030106 0.02788
3 3 0 0.04572 0.04432 0.04241 0.038719 0.035631 0.033028
4 4 0 0.051937 0.050449 0.048354 0.044164 0.040661 0.037707
5 5 0 0.056986 0.0555 0.053326 0.04875 0.044927 0.041707
6 1 1 0.008385 0.0079795 0.0075874 0.0068616 0.0062549 0.0057423
7 2 1 0.011253 0.010708 0.010181 0.0092041 0.0083817 0.0076884
8 3 1 0.013554 0.012928 0.012313 0.011136 0.010147 0.0093138
9 4 1 0.016251 0.015529 0.014803 0.013404 0.012225 0.011226
10 5 1 0.019768 0.018969 0.018139 0.016444 0.015026 0.013822
```

```
% Consumption Function
st_title = ['MEAN(MN_MPC_W_GAIN_CHECK(KM,J)), welf_checks=' num2str(welf_checks) ', TR=' num2str(
tb_az_c = ff_summ_nd_array(st_title, mn_MPC_W_gain_share_check, true, ["mean"], 3, 1, cl_mp_data
```

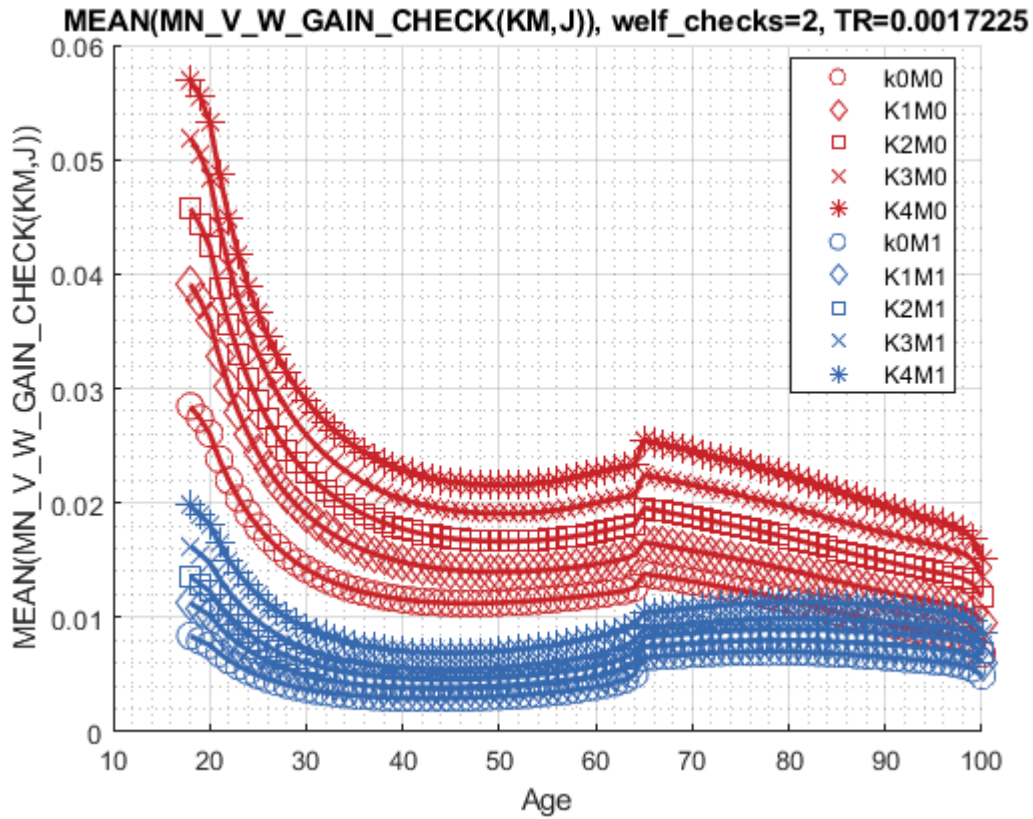
```
xxx MEAN(MN_MPC_W_GAIN_CHECK(KM,J)), welf_checks=2, TR=0.0017225 xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx
group kids marry mean_age_18 mean_age_19 mean_age_20 mean_age_21 mean_age_22 mean_age_23
1 1 0 0.067542 0.074752 0.091075 0.088909 0.086942 0.085404
2 2 0 0.075256 0.083119 0.10165 0.099284 0.097581 0.095623
3 3 0 0.086542 0.095859 0.11593 0.11256 0.10948 0.10827
4 4 0 0.091496 0.10076 0.12129 0.11824 0.11514 0.11272
5 5 0 0.098346 0.10645 0.12728 0.12409 0.12073 0.11757
6 1 1 0.10277 0.10672 0.1125 0.11137 0.11019 0.10941
```

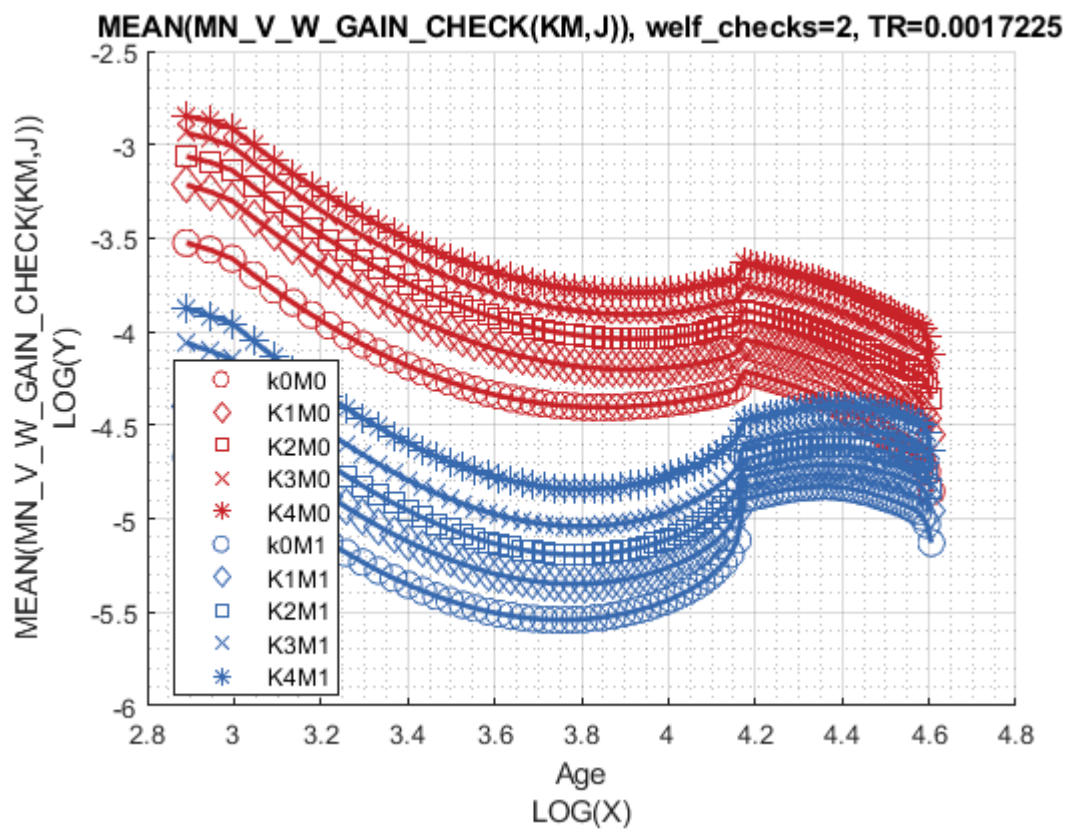


7	2	1	0.10343	0.1077	0.11433	0.11354	0.11208	0.11076
8	3	1	0.10875	0.11374	0.12309	0.11975	0.11861	0.1182
9	4	1	0.11014	0.11556	0.12324	0.12217	0.12228	0.11986
10	5	1	0.1166	0.1232	0.13246	0.13017	0.12661	0.12494

Graph Mean Values:

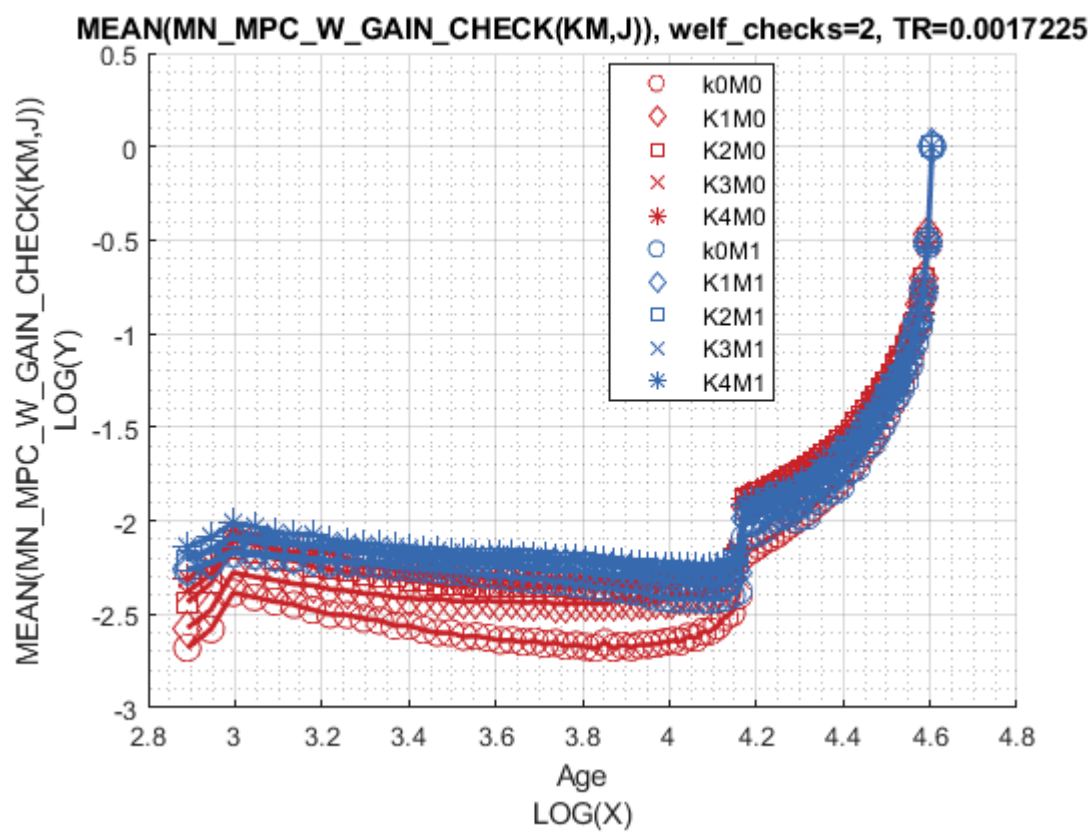
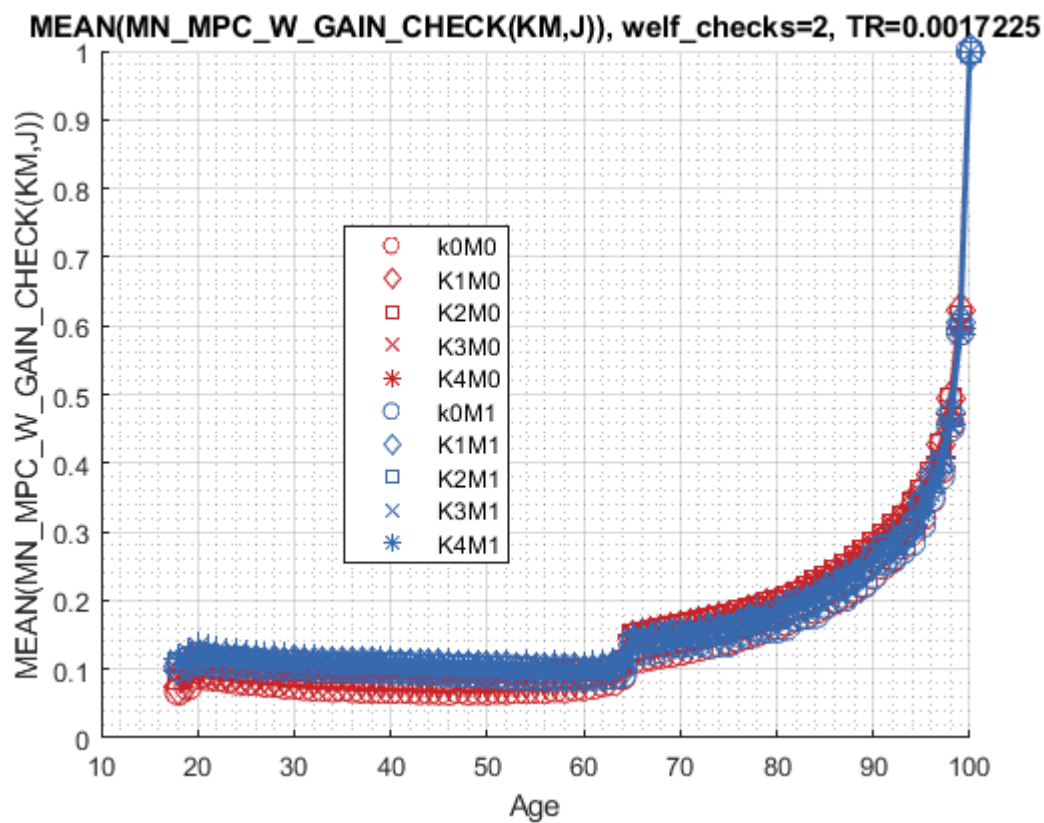
```
st_title = ['MEAN(MN\V\W\_GAIN\_CHECK(KM,J)), welf\_checks=' num2str(welf_checks) ', TR=' num2str(TR)];
mp_support_graph('cl_st_graph_title') = {st_title};
mp_support_graph('cl_st_ytitle') = {'MEAN(MN\V\W\_GAIN\_CHECK(KM,J))'};
ff_graph_grid((tb_az_v{1:end, 4:end}), ar_row_grid, age_grid, mp_support_graph);
```





Graph Mean Consumption (**MPC: Share of Check Consumed**):

```
st_title = ['MEAN(MN\MPC\W\_GAIN\_CHECK(KM,J)), welf\_checks=' num2str(welf_checks) ', TR=' r
mp_support_graph('cl_st_graph_title') = {st_title};
mp_support_graph('cl_st_ytitle') = {'MEAN(MN\MPC\W\_GAIN\_CHECK(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:

```
% Set
% NaN(n_jgrid,n_agrid,n_etagrid,n_educgrid,n_marriedgrid,n_kidsgrid);
ar_permute = [2,3,6,1,4,5];
% Value Function
st_title = ['MEAN(MN_V_W_GAIN_CHECK(EM,J)), welf_checks=' num2str(welf_checks) ', TR=' num2str(
tb_az_v = ff_summ_nd_array(st_title, mn_V_W_gain_check, true, ["mean"], 3, 1, cl_mp_datasetdesc
```

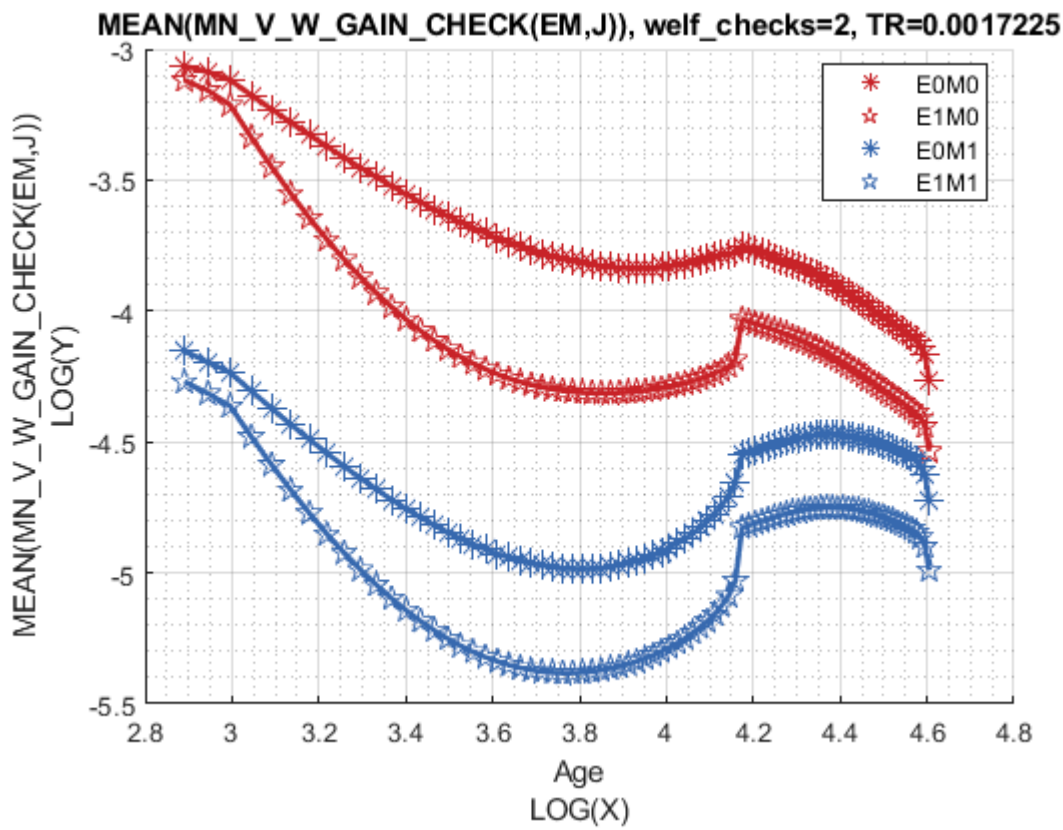
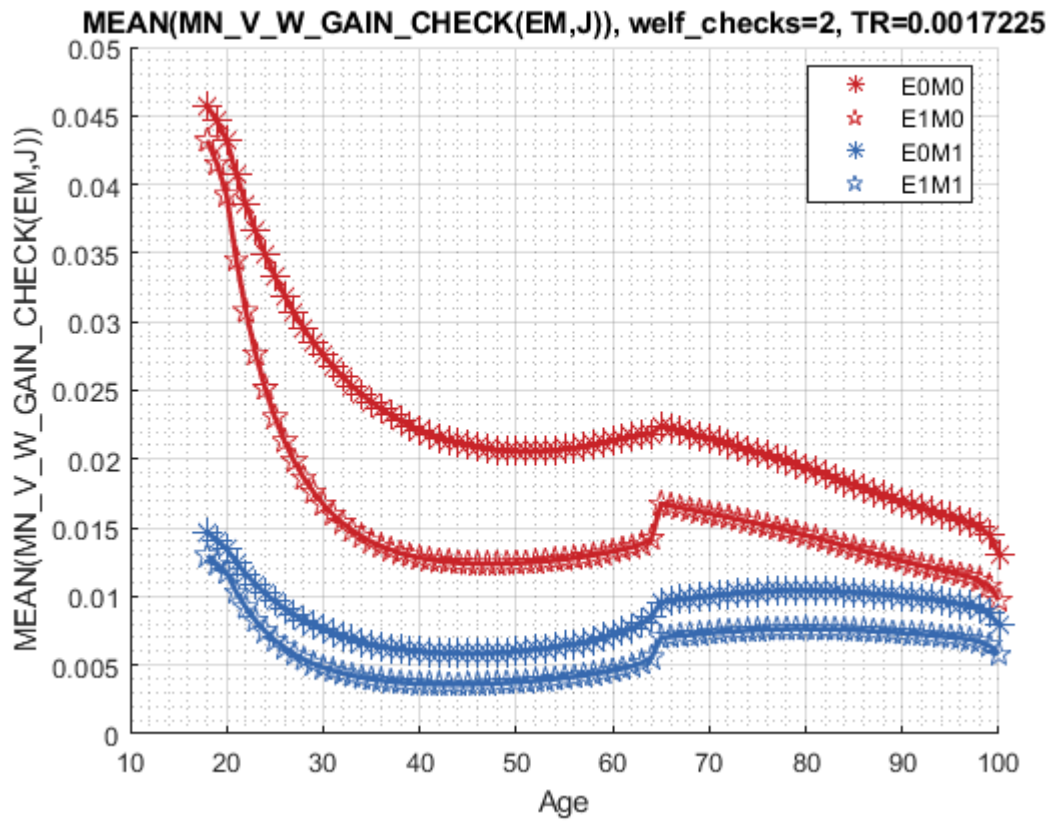
xxx	MEAN(MN_V_W_GAIN_CHECK(EM,J)), welf_checks=2, TR=0.0017225	xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx							
group	edu	marry	mean_age_18	mean_age_19	mean_age_20	mean_age_21	mean_age_22	mean_age_23	
1	0	0	0.045692	0.044619	0.043207	0.040746	0.03856	0.036612	
2	1	0	0.043194	0.041526	0.039215	0.034536	0.030754	0.027671	
3	0	1	0.014697	0.014079	0.01347	0.012491	0.01163	0.010875	
4	1	1	0.012987	0.012367	0.011739	0.010329	0.009184	0.008242	

```
% Consumption
st_title = ['MEAN(MN_MPC_W_GAIN_CHECK(EM,J)), welf_checks=' num2str(welf_checks) ', TR=' num2str(
tb_az_c = ff_summ_nd_array(st_title, mn_MPC_W_gain_share_check, true, ["mean"], 3, 1, cl_mp_data
```

xxx	MEAN(MN_MPC_W_GAIN_CHECK(EM,J)), welf_checks=2, TR=0.0017225	xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx							
group	edu	marry	mean_age_18	mean_age_19	mean_age_20	mean_age_21	mean_age_22	mean_age_23	
1	0	0	0.075296	0.080407	0.092505	0.091671	0.091522	0.091354	
2	1	0	0.092377	0.10397	0.13038	0.12556	0.12042	0.11648	
3	0	1	0.099842	0.10362	0.10816	0.10847	0.10824	0.10764	
4	1	1	0.11684	0.12315	0.13408	0.13033	0.12767	0.12562	

Graph Mean Values:

```
st_title = ['MEAN(MN_V_W_GAIN_CHECK(EM,J)), welf_checks=' num2str(welf_checks) ', TR=' num
mp_support_graph('cl_st_graph_title') = {st_title};
mp_support_graph('cl_st_ytitle') = {'MEAN(MN_V_W_GAIN_CHECK(EM,J))'};
ff_graph_grid((tb_az_v{1:end, 4:end}), ar_row_grid, age_grid, mp_support_graph);
```

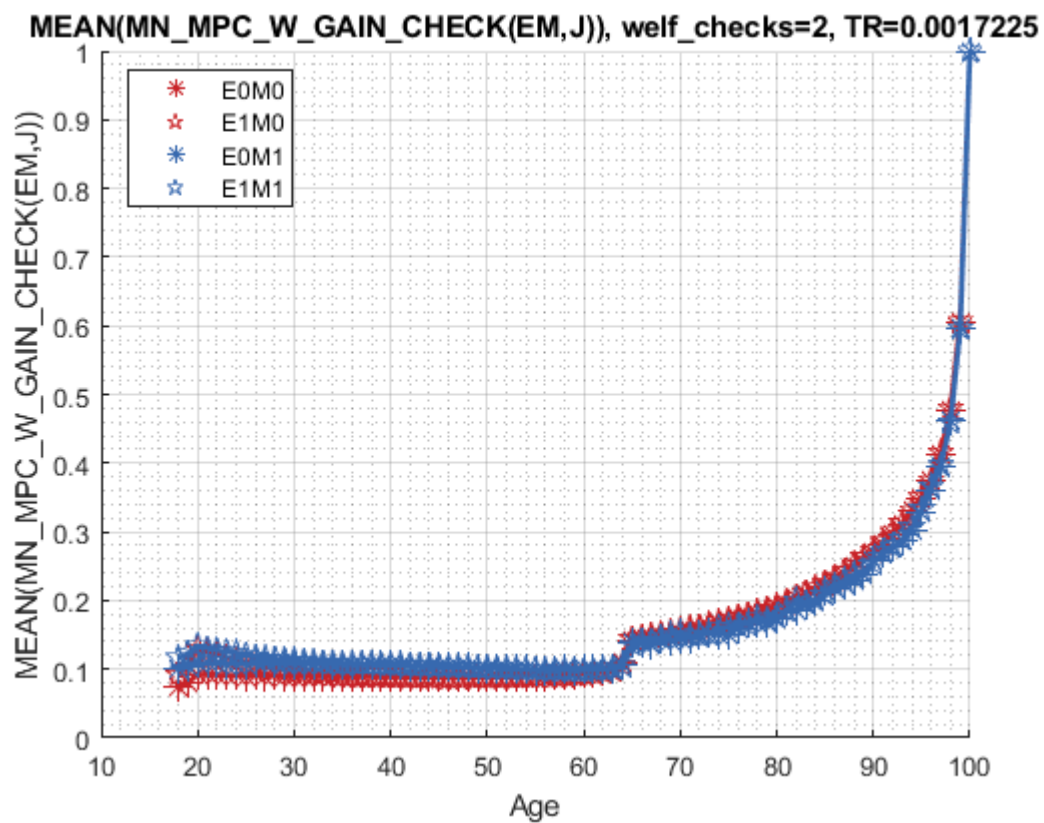


Graph Mean Consumption (**MPC: Share of Check Consumed**):

```

st_title = ['MEAN(MN_MPC_W_GAIN_CHECK(EM,J)), welf_checks=' num2str(welf_checks) ', TR=' r
mp_support_graph('cl_st_graph_title') = {st_title};
mp_support_graph('cl_st_ytitle') = {'MEAN(MN_MPC_W_GAIN_CHECK(EM,J))'};
ff_graph_grid((tb_az_c{1:end, 4:end}), ar_row_grid, age_grid, mp_support_graph);

```



MEAN(MN\_MPC\_W\_GAIN\_CHECK(EM,J)), welf\_checks=2, TR=0.0017225

