

# 2019 Full States EV and EC of One Check

This is the example vignette for function: [snw\\_evuvw19\\_jaeemk](#) from the [PrjOptiSNW Package](#). 2019 integrated over VU and VW, given optimal savings choices, unemployment shocks and various expectations.

## Test SNW\_EVUVW19\_JAEEMK Defaults

Call the function with defaults.

```
clear all;
st_solu_type = 'bisec_vec';

% Solve the VFI Problem and get Value Function
mp_params = snw_mp_param('default_docdense');
mp_params('beta') = 0.95;
% mp_params = snw_mp_param('default_dense');
mp_controls = snw_mp_control('default_test');

% set Unemployment Related Variables
mp_params('a2_covidyr') = mp_params('a2_covidyr_manna_heaven');
% mp_params('a2_covidyr') = mp_params('a2_covidyr_tax_fully_pay');

% Solve for Unemployment Values
mp_controls('bl_print_vfi') = false;
mp_controls('bl_print_vfi_verbose') = true;
mp_controls('bl_print_ds') = false;
mp_controls('bl_print_ds_verbose') = false;
mp_controls('bl_print_precompute') = false;
mp_controls('bl_print_precompute_verbose') = false;
mp_controls('bl_print_a4chk') = false;
mp_controls('bl_print_a4chk_verbose') = false;
mp_controls('bl_print_evuvw20_jaeemk') = false;
mp_controls('bl_print_evuvw20_jaeemk_verbose') = false;

% Solve the Model to get V working and unemployed
[V_ss,ap_ss,cons_ss,mp_valpol_more_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=488.7831

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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	-1.2728e+08	-2.9126	20.655	-7.0915
ap_VFI	2	2	6	4.37e+07	83	5.265e+05	1.3962e+09	31.95	36.423	1.14
cons_VFI	3	3	6	4.37e+07	83	5.265e+05	2.3374e+08	5.3487	8.4439	1.5787

xxx TABLE:V\_VFI XXXXXXXXXXXXXXXXXXXXXXX

	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	c526500
	—	—	—	—	—	—	—	—	—	—
r1	-274.81	-274.42	-271.94	-266.29	-257.26	14.439	14.533	14.626	14.718	14.808
r2	-265.29	-264.9	-262.43	-256.84	-248.12	14.494	14.585	14.674	14.763	14.854
r3	-255.77	-255.38	-252.93	-247.53	-239.24	14.55	14.636	14.723	14.808	14.895





r4	-254.66	-254.03	-250.17	-243.06	-234.04	14.565	14.65	14.734	14.817	14.89
r5	-245.67	-245.08	-241.48	-234.76	-226.24	14.611	14.693	14.774	14.854	14.93
r79	-9.6662	-9.655	-9.5783	-9.3823	-9.0457	2.4688	2.4792	2.489	2.4982	2.506
r80	-8.7031	-8.6919	-8.6152	-8.4192	-8.0826	2.2523	2.2603	2.2679	2.275	2.281
r81	-7.5138	-7.5026	-7.4258	-7.2298	-6.8933	1.9743	1.9799	1.9851	1.99	1.994
r82	-5.9155	-5.9043	-5.8275	-5.6315	-5.295	1.5817	1.5848	1.5878	1.5905	1.593
r83	-3.5892	-3.578	-3.5012	-3.3052	-2.9687	0.97895	0.97995	0.98089	0.98178	0.9826

```
xxx TABLE:ap_VFI xxxxxxxxxxxxxxxxxxxx
```

	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	c526500
r1	0	0	0	0.0011815	0.013905	109.97	115.52	121.25	127.17	133.28
r2	0	0	0	0.00090277	0.013905	109.94	115.49	121.22	127.14	133.25
r3	0	0	0	0.00051498	0.013905	109.9	115.44	121.17	127.09	133.2
r4	0	0	0	0.00051498	0.013905	110.33	115.88	121.6	127.52	133.64
r5	0	0	0	0.00048777	0.013905	110.78	116.32	122.05	127.97	134.09
r79	0	0	0	0	0	80.977	84.854	88.823	92.746	96.839
r80	0	0	0	0	0	75.625	79.248	82.93	86.483	90.439
r81	0	0	0	0	0	67.452	70.146	73.182	76.669	81.091
r82	0	0	0	0	0	50.126	53.467	55.817	57.4	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	c526500
r1	0.027723	0.028258	0.031999	0.040974	0.048028	11.996	12.272	12.557	12.851	13.136
r2	0.027723	0.028258	0.031999	0.041253	0.048028	12.23	12.508	12.794	13.089	13.374
r3	0.027723	0.028258	0.031999	0.041641	0.048028	12.483	12.762	13.05	13.345	13.63
r4	0.028805	0.029339	0.033081	0.042722	0.049108	12.728	13.008	13.297	13.593	13.878
r5	0.029859	0.030394	0.034135	0.043802	0.050161	12.963	13.245	13.534	13.83	14.115
r79	0.2179	0.21844	0.22216	0.23228	0.25197	35.453	37.4	39.448	41.74	44.03
r80	0.2179	0.21844	0.22216	0.23228	0.25197	40.785	42.986	45.321	47.983	50.639
r81	0.2179	0.21844	0.22216	0.23228	0.25197	48.942	52.071	55.052	57.78	60.49
r82	0.2179	0.21844	0.22216	0.23228	0.25197	66.254	68.736	72.404	77.036	81.668
r83	0.2179	0.21844	0.22216	0.23228	0.25197	116.37	122.19	128.21	134.43	140.65

```
[Phi_true] = snw_ds_main(mp_params, mp_controls, ap_ss, cons_emp_2020, mp_valpol_more_ss);
```

```
Completed SNW_DS_MAIN;SNW_MP_PARAM=default_docdense;SNW_MP_CONTROL=default_test;time=1911.1684
```

```
% Get Matrixes
```

```
cl_st_precompute_list = {'a', ...
    'inc', 'inc_unemp', 'spouse_inc', 'spouse_inc_unemp', 'ref_earn_wageind_grid',...
    'ap_idx_lower_ss', 'ap_idx_higher_ss', 'ap_idx_lower_weight_ss'};
mp_controls('bl_print_precompute_verbose') = false;
[mp_precompute_res] = snw_hh_precompute(mp_params, mp_controls, cl_st_precompute_list, ap_ss, P
```

```
Wage quintile cutoffs=0.4645    0.71528    1.0335    1.5632
Completed SNW_HH_PRECOMPUTE;SNW_MP_PARAM=default_docdense;SNW_MP_CONTROL=default_test;time cost=358.215
```

## Solve for 2019 Evuvw With 0 and 2 Checks

```
% Call Function
```

```
welf_checks = 0;
[ev19_jaeemk_check0, ec19_jaeemk_check0, ev20_jaeemk_check0, ec20_jaeemk_check0] = snw_evuvw19(
    welf_checks, st_solu_type, mp_params, mp_controls, ...
    V_emp_2020, cons_emp_2020, V_unemp_2020, cons_unemp_2020, mp_precompute_res);
```

```
Completed SNW_A4CHK_WRK_BISEC_VEC;welf_checks=0;TR=0.0017225;SNW_MP_PARAM=default_docdense;SNW_MP_CONTROL=default_test;time=1911.1684
```

Completed SNW\_A4CHK\_UNEMP\_BISEC\_VEC;welf\_checks=0;TR=0.0017225;xi=0.5;b=0.5;SNW\_MP\_PARAM=default\_docdense;SNW\_MP\_CON  
Completed SNW\_EVUVW20\_JAEEMK;SNW\_MP\_PARAM=default\_docdense;SNW\_MP\_CONTROL=default\_test;timeEUEC=8.2469  
Completed SNW\_EVUVW19\_JAEEMK;SNW\_MP\_PARAM=default\_docdense;SNW\_MP\_CONTROL=default\_test;time=5058.1482

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

	i	idx	ndim	numel	rowN	colN	sum	mean	std	coefva
ec19_jaeemk	1	1	6	4.3173e+07	82	5.265e+05	1.9762e+08	4.5774	5.3272	1.163
ec20_jaeemk	2	2	6	4.37e+07	83	5.265e+05	2.3357e+08	5.3448	8.4447	1.5
ev19_jaeemk	3	3	6	4.3173e+07	82	5.265e+05	-1.2119e+08	-2.8072	20.003	-7.125
ev20_jaeemk	4	4	6	4.37e+07	83	5.265e+05	-1.2937e+08	-2.9604	20.785	-7.02

xxx TABLE:ec19_jaeemk	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	c526500
r1	0.036494	0.036494	0.037029	0.041925	0.048857	12.024	12.296	12.576	12.864	13.152
r2	0.036494	0.036494	0.037029	0.041745	0.049665	12.268	12.541	12.822	13.111	13.399
r3	0.037912	0.037912	0.038127	0.041994	0.050655	12.503	12.777	13.06	13.349	13.636
r4	0.039293	0.039293	0.039401	0.043382	0.052052	12.761	13.036	13.319	13.607	13.894
r5	0.040635	0.040635	0.04064	0.044725	0.053494	13.01	13.286	13.569	13.855	14.142
r78	0.2179	0.2179	0.2179	0.2179	0.2179	27.797	28.793	29.808	30.995	32.182
r79	0.2179	0.2179	0.2179	0.2179	0.2179	30.454	31.684	32.756	33.984	35.214
r80	0.2179	0.2179	0.2179	0.2179	0.2179	33.715	35.537	37.399	38.98	40.701
r81	0.2179	0.2179	0.2179	0.2179	0.2179	40.14	41.425	43.212	45.644	47.866
r82	0.2179	0.2179	0.2179	0.2179	0.2179	52.118	55.559	58.496	60.127	62.759

xxx TABLE:ec20_jaeemk	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	c526500
r1	0.033462	0.033996	0.037408	0.043215	0.048855	12.249	12.534	12.827	13.127	13.427
r2	0.033462	0.033996	0.037408	0.043883	0.049712	12.485	12.771	13.065	13.366	13.666
r3	0.033462	0.033996	0.037604	0.045059	0.050801	12.739	13.026	13.321	13.621	13.921
r4	0.034763	0.035298	0.038972	0.046376	0.052249	12.985	13.273	13.568	13.866	14.166
r5	0.036032	0.036566	0.040303	0.047654	0.053654	13.221	13.51	13.805	14.101	14.401
r79	0.2179	0.21844	0.22216	0.23228	0.25197	35.858	37.4	39.448	41.74	44.04
r80	0.2179	0.21844	0.22216	0.23228	0.25197	40.785	42.986	45.321	47.983	50.283
r81	0.2179	0.21844	0.22216	0.23228	0.25197	48.942	52.071	55.052	57.95	60.85
r82	0.2179	0.21844	0.22216	0.23228	0.25197	66.755	69.238	72.404	77.036	80.568
r83	0.2179	0.21844	0.22216	0.23228	0.25197	116.87	122.69	128.71	134.92	141.14

xxx TABLE:ev19_jaeemk	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	c526500
r1	-264.92	-264.92	-264.47	-260.13	-252.27	14.364	14.458	14.551	14.643	14.729
r2	-254.9	-254.9	-254.45	-250.7	-243.18	14.418	14.509	14.599	14.688	14.776
r3	-244.75	-244.75	-244.59	-241.74	-234.51	14.473	14.56	14.646	14.732	14.819
r4	-235.6	-235.6	-235.52	-232.8	-226.09	14.529	14.612	14.695	14.777	14.86
r5	-227.32	-227.32	-227.32	-224.7	-218.45	14.576	14.656	14.736	14.814	14.894
r78	-9.6725	-9.6725	-9.6725	-9.6725	-9.6725	2.4176	2.4297	2.441	2.4518	2.4629
r79	-8.7092	-8.7092	-8.7092	-8.7092	-8.7092	2.2043	2.215	2.2241	2.2322	2.2403
r80	-7.5196	-7.5196	-7.5196	-7.5196	-7.5196	1.9308	1.938	1.9447	1.9507	1.9566
r81	-5.9209	-5.9209	-5.9209	-5.9209	-5.9209	1.5463	1.55	1.5539	1.558	1.562
r82	-3.5937	-3.5937	-3.5937	-3.5937	-3.5937	0.95581	0.95855	0.96061	0.96167	0.96333

xxx TABLE:ev20_jaeemk	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	c526500
r1	-278.01	-277.51	-274.43	-268.08	-258.6	14.436	14.53	14.623	14.715	14.806

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r1	-262.28	-262.28	-261.88	-258.26	-250.89	14.364	14.458	14.551	14.643	14.725
r2	-252.28	-252.28	-251.88	-248.84	-241.85	14.418	14.509	14.599	14.688	14.768
r3	-242.33	-242.33	-242.18	-239.92	-233.24	14.473	14.56	14.646	14.732	14.812
r4	-233.33	-233.33	-233.27	-231.09	-224.88	14.529	14.612	14.695	14.777	14.857
r5	-225.21	-225.21	-225.2	-223.1	-217.31	14.576	14.656	14.736	14.814	14.894
r78	-9.6013	-9.6013	-9.6013	-9.6013	-9.6013	2.4176	2.4297	2.441	2.4518	2.462
r79	-8.6379	-8.6379	-8.6379	-8.6379	-8.6379	2.2043	2.215	2.2241	2.2322	2.2403
r80	-7.4483	-7.4483	-7.4483	-7.4483	-7.4483	1.9308	1.938	1.9447	1.9507	1.9567
r81	-5.8497	-5.8497	-5.8497	-5.8497	-5.8497	1.5463	1.55	1.5539	1.558	1.562
r82	-3.5225	-3.5225	-3.5225	-3.5225	-3.5225	0.95582	0.95855	0.96061	0.96167	0.963
xxx TABLE:ev20_jaeemk xxxxxxxxxxxxxxxxxxxxxx										
	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	c526500
r1	-275.12	-274.68	-272.28	-266.43	-257.32	14.436	14.53	14.623	14.715	14.807
r2	-265.6	-265.16	-262.78	-257.01	-248.18	14.491	14.582	14.671	14.76	14.848
r3	-256.09	-255.66	-253.32	-247.71	-239.31	14.546	14.633	14.72	14.805	14.888
r4	-246.56	-246.15	-243.96	-238.69	-230.79	14.603	14.686	14.769	14.85	14.932
r5	-237.94	-237.56	-235.5	-230.54	-223.08	14.65	14.73	14.81	14.888	14.968
r79	-9.595	-9.584	-9.5115	-9.3234	-8.9958	2.4698	2.4801	2.4898	2.4989	2.507
r80	-8.6319	-8.6209	-8.5484	-8.3603	-8.0332	2.253	2.261	2.2685	2.2756	2.282
r81	-7.4426	-7.4316	-7.3591	-7.171	-6.8443	1.9749	1.9803	1.9855	1.9904	1.995
r82	-5.8443	-5.8333	-5.7608	-5.5727	-5.2463	1.582	1.5851	1.588	1.5907	1.593
r83	-3.518	-3.507	-3.4345	-3.2464	-2.9207	0.97904	0.98004	0.98097	0.98185	0.9826

Differences between Checks in Expected Value and Expected Consumption

```
mn_V_U_gain_check = ev19_jaeemk_check2 - ev19_jaeemk_check0;
mn_MPC_U_gain_share_check = (ec19_jaeemk_check2 - ec19_jaeemk_check0)./(welf_checks*mp_params('welf_checks'));
```

## 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 = 18:99;
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_eduagrid,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_U_GAIN_CHECK(A,Z)), welf_checks=' num2str(welf_checks) ', TR=' num2str(m
tb_az_v = ff_summ_nd_array(st_title, mn_V_U_gain_check, true, ["mean"], 4, 1, cl_mp_datasetdesc
```

group	savings	mean_eta_1	mean_eta_2	mean_eta_3	mean_eta_4	mean_eta_5	mean_eta_6	mean_eta_7
1	0	0.5996	0.55567	0.50718	0.45861	0.41293	0.3714	0.33000
2	0.00051498	0.5996	0.55567	0.50718	0.45861	0.41293	0.3714	0.33000
3	0.0041199	0.59858	0.55412	0.50516	0.45663	0.41125	0.37004	0.32800
4	0.013905	0.49775	0.46232	0.42566	0.38891	0.35398	0.32177	0.28677
5	0.032959	0.39069	0.36715	0.34122	0.31498	0.28971	0.26606	0.24228
6	0.064373	0.31749	0.30013	0.28105	0.26146	0.24228	0.22405	0.20582

% Consumption

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

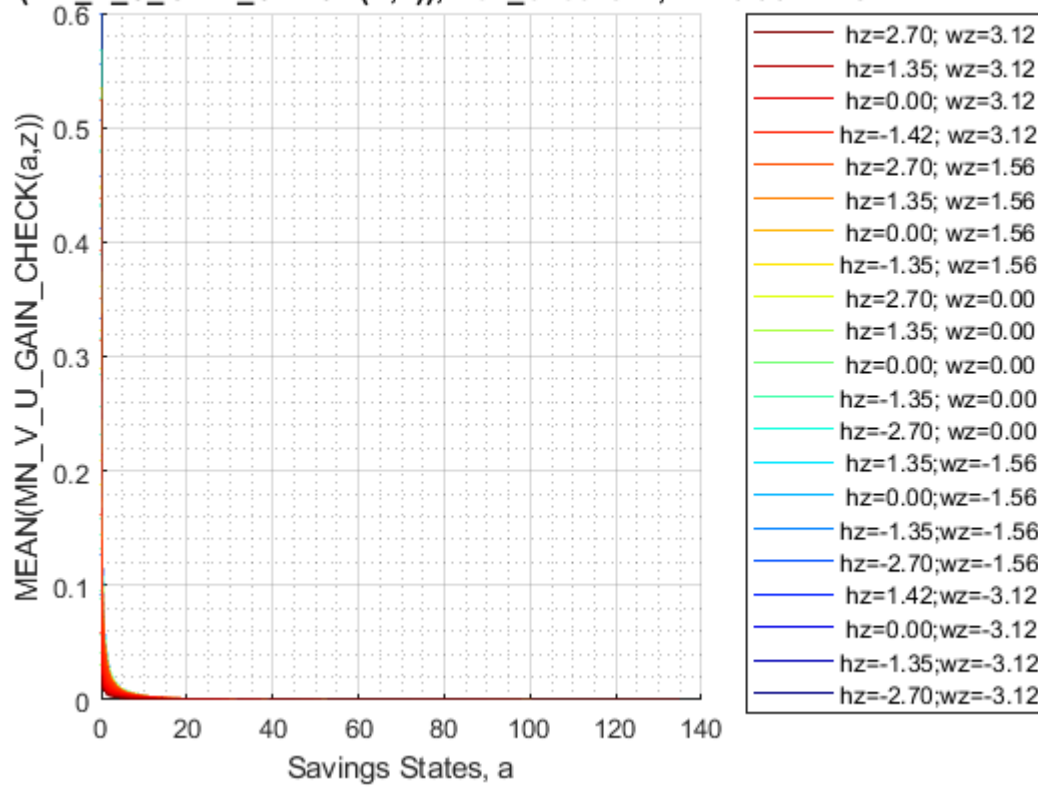
group	savings	mean_eta_1	mean_eta_2	mean_eta_3	mean_eta_4	mean_eta_5	mean_eta_6	mean_eta_7
1	0	0.81458	0.81464	0.81469	0.81468	0.81453	0.81424	0.81395
2	0.00051498	0.81458	0.81464	0.81469	0.81468	0.81453	0.81424	0.81395
3	0.0041199	0.81443	0.81422	0.81395	0.81376	0.81357	0.81328	0.81300
4	0.013905	0.74362	0.73968	0.73889	0.73872	0.73889	0.73924	0.73968
5	0.032959	0.66095	0.66107	0.6621	0.66413	0.66676	0.66957	0.67238
6	0.064373	0.56543	0.56521	0.56569	0.56689	0.56869	0.57092	0.57315

Graph Mean Values:

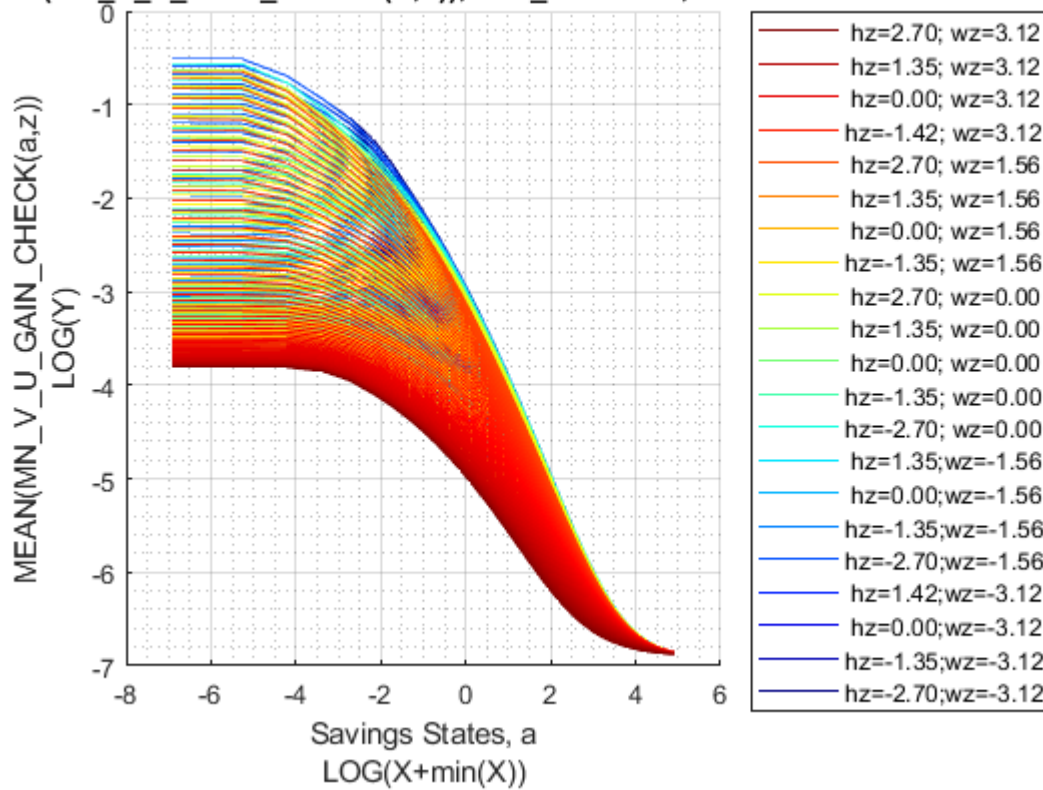
```
st_title = ['MEAN(MN_V_U_GAIN_CHECK(A,Z)), welf_checks=' num2str(welf_checks) ', TR=' num2str(m
mp_support_graph('cl_st_graph_title') = {st_title};
mp_support_graph('cl_st_ytitle') = {'MEAN(MN_V_U_GAIN_CHECK(a,z))'};
ff_graph_grid((tb_az_v{1:end, 3:end}'), ar_st_eta_HS_grid, agrid, mp_support_graph);
```



$N(MN\_V\_U\_GAIN\_CHECK(A,Z))$ , welf\_checks=2, TR=0.0017225



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

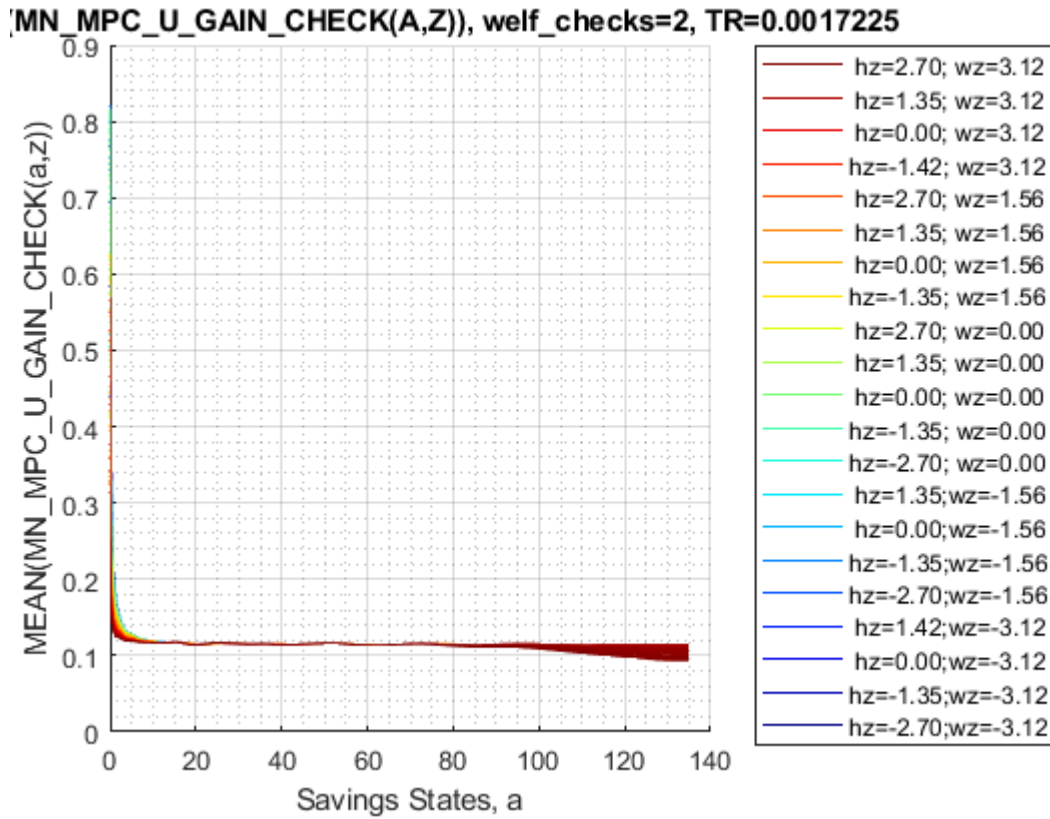


Graph Mean Consumption (**MPC: Share of Check Consumed**):

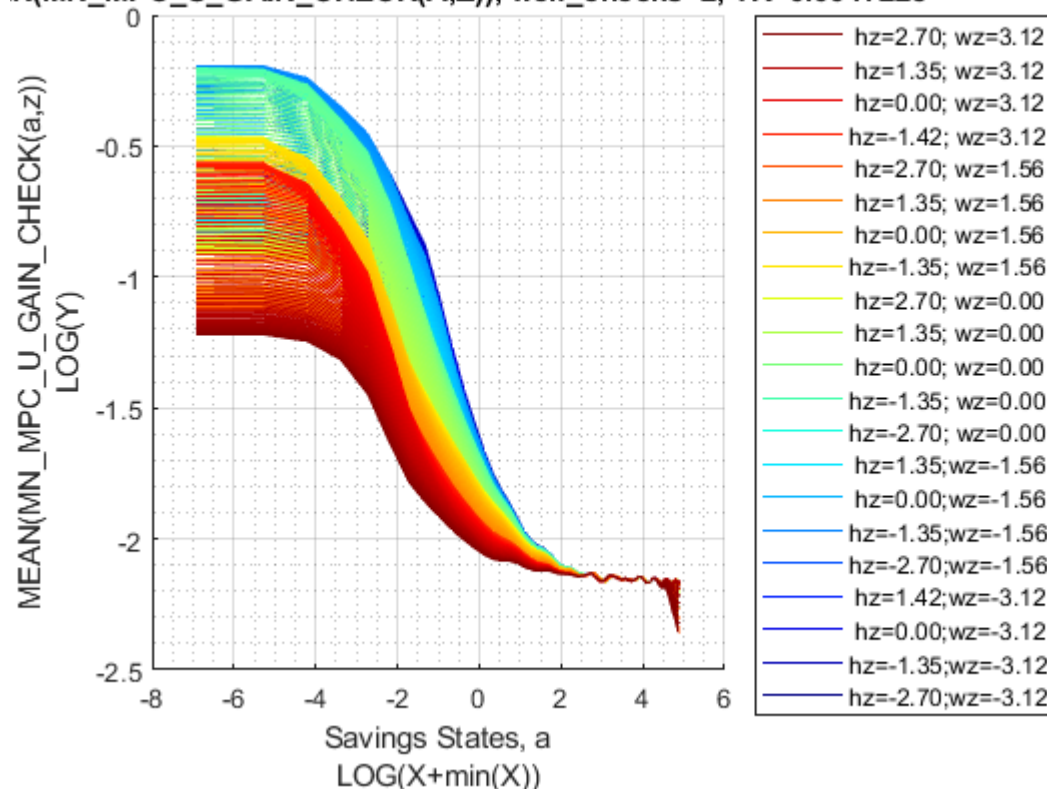
```

st_title = ['MEAN(MN\_MPC\_U\_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\_U\_GAIN\_CHECK(a,z))'};
ff_graph_grid((tb_az_c{1:end, 3:end})', ar_st_eta_HS_grid, agrid, mp_support_graph);

```



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



## Analyze Marginal Value and MPC over $Y(a, \eta)$ , Conditional On Kids, Marry, Age, Education

Income is generated by savings and shocks, what are the income levels generated by all the shock and savings points conditional on kids, marital status, age and educational levels. Plot on the Y axis MPC, and plot on the X axis income levels, use colors to first distinguish between different  $a$  levels, then use colors to distinguish between different  $\eta$  levels.

Set Up date, Select Age 37vn

, unmarried, no kids, lower education:

```
% NaN(n_jgrid,n_agrid,n_etagrid,n_eduagrid,n_marriedgrid,n_kidsgrid);
% 38 year old, unmarried, no kids, lower educated
% Only Household Head Shock Matters so select up to 'n_eta_H_grid'
mn_total_inc_jemk = total_inc_VFI(19, :, 1:mp_params('n_eta_H_grid'), 1, 1, 1);
mn_V_W_gain_check_use = ev19_jaeemk_check2 - ev19_jaeemk_check0;
mn_C_W_gain_check_use = ec19_jaeemk_check2 - ec19_jaeemk_check0;
```

Select Age, Education, Marital, Kids Count:s

```
% Selections
it_age = 21; % +18
it_marital = 1; % 1 = unmarried
it_kids = 1; % 1 = kids is zero
it_educ = 1; % 1 = lower education
% Select: NaN(n_jgrid,n_agrid,n_etagrid,n_eduagrid,n_marriedgrid,n_kidsgrid);
```

```

mn_C_W_gain_check_jemk = mn_C_W_gain_check_use(it_age, :, 1:mp_params('n_eta_H_grid'), it_educ,
mn_V_W_gain_check_jemk = mn_V_W_gain_check_use(it_age, :, 1:mp_params('n_eta_H_grid'), it_educ,
% Reshape, so shock is the first dim, a is the second
mt_total_inc_jemk = permute(mn_total_inc_jemk,[3,2,1]);
mt_C_W_gain_check_jemk = permute(mn_C_W_gain_check_jemk,[3,2,1]);
mt_C_W_gain_check_jemk(mt_C_W_gain_check_jemk<=1e-10) = 1e-10;
mt_V_W_gain_check_jemk = permute(mn_V_W_gain_check_jemk,[3,2,1]);
mt_V_W_gain_check_jemk(mt_V_W_gain_check_jemk<=1e-10) = 1e-10;
% Generate meshed a and shock grid
[mt_eta_H, mt_a] = ndgrid(eta_H_grid(1:mp_params('n_eta_H_grid')), agrid);

```

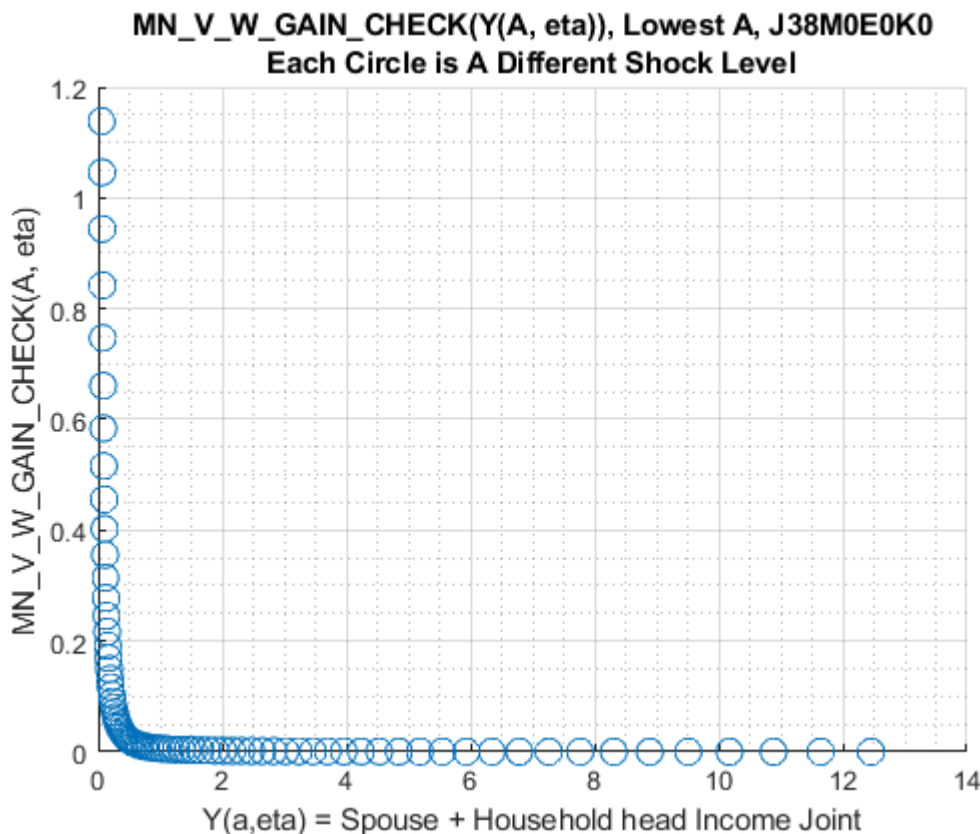
## Marginal Value Gains, Color as Shock, Conditional on Age, Marital, Kids, and Education

How do shocks and a impact marginal value. First plot one asset level, variation comes only from increasingly higher shocks:

```

figure();
it_a = 1;
scatter((mt_total_inc_jemk(:,it_a)), (mt_V_W_gain_check_jemk(:,it_a)), 100);
title({'MN\_V\_W\_GAIN\_CHECK(Y(A, eta)), Lowest A, J38M0E0K0', ...
      'Each Circle is A Different Shock Level'});
xlabel('Y(a,eta) = Spouse + Household head Income Joint');
ylabel('MN\_V\_W\_GAIN\_CHECK(A, eta)');
grid on;
grid minor;

```

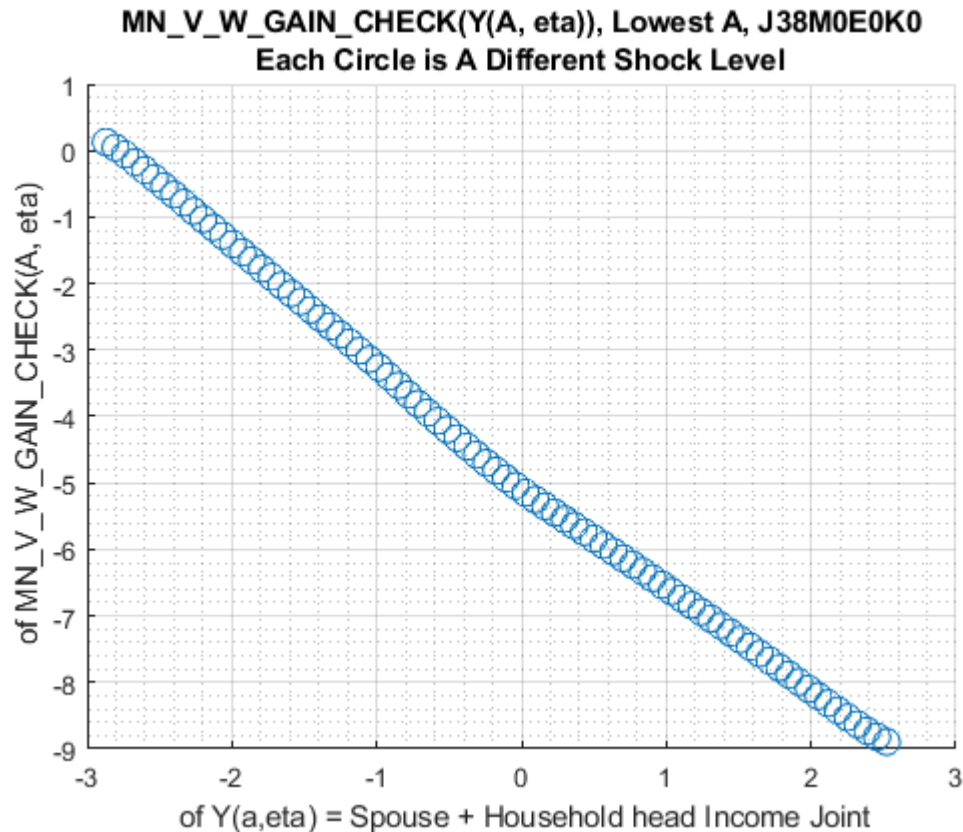


```
figure();
```

```

it_shock = 1;
scatter(log(mt_total_inc_jemk(:,it_a)), log(mt_V_W_gain_check_jemk(:,it_a)), 100);
title({'MN\_V\_W\_GAIN\_CHECK(Y(A, eta)), Lowest A, J38M0E0K0', ...
      'Each Circle is A Different Shock Level'});
xlabel(' of Y(a,eta) = Spouse + Household head Income Joint');
ylabel(' of MN\_V\_W\_GAIN\_CHECK(A, eta)');
grid on;
grid minor;

```

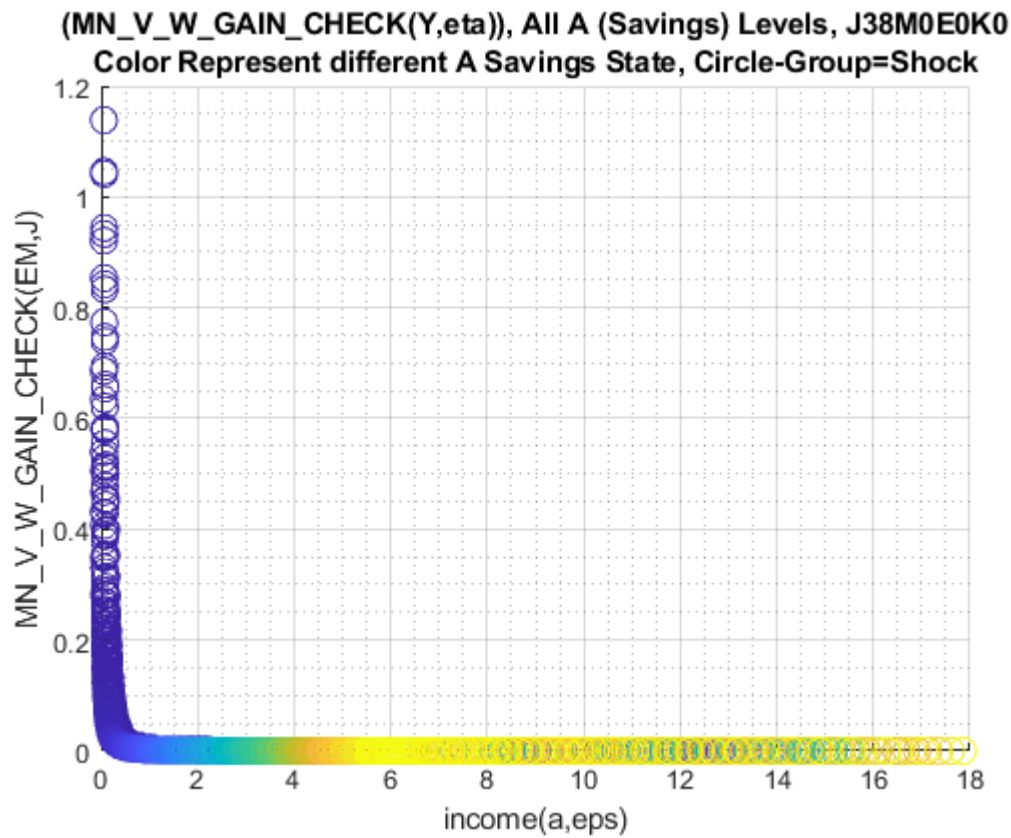


Plot all asset levels:

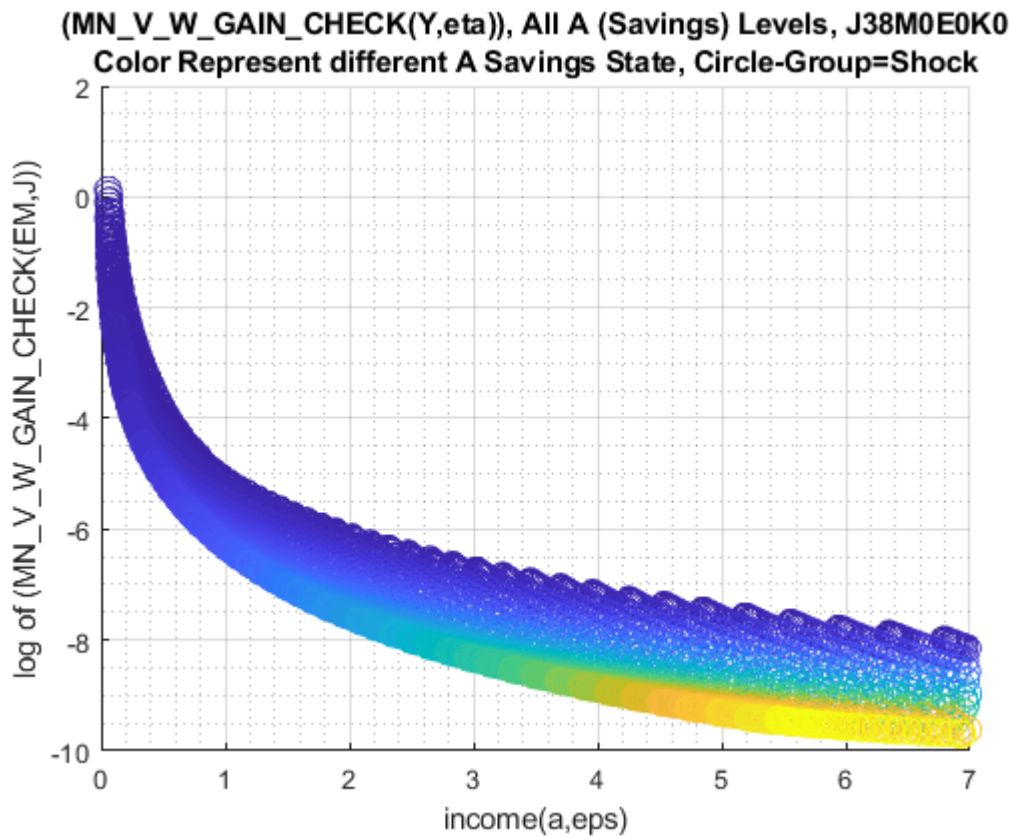
```

figure();
scatter((mt_total_inc_jemk(:)), (mt_V_W_gain_check_jemk(:)), 100, mt_a(:));
title({'(MN\_V\_W\_GAIN\_CHECK(Y,eta)), All A (Savings) Levels, J38M0E0K0', ...
      'Color Represent different A Savings State, Circle-Group=Shock'});
xlabel('income(a,eps)');
ylabel('MN\_V\_W\_GAIN\_CHECK(EM,J)');
grid on;
grid minor;

```



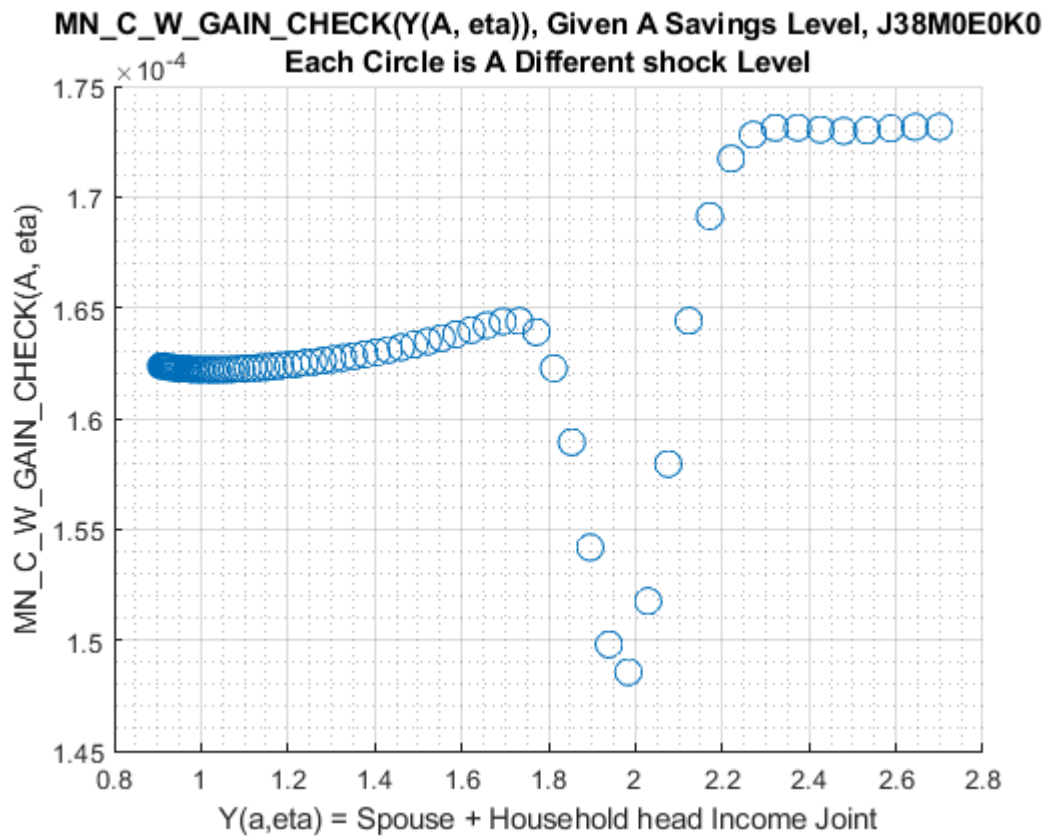
```
figure();
scatter((mt_total_inc_jemk(:)), log(mt_V_W_gain_check_jemk(:)), 100, mt_a(:));
title({'(MN_V_W_GAIN_CHECK(Y,eta)), All A (Savings) Levels, J38M0E0K0', ...
      'Color Represent different A Savings State, Circle-Group=Shock'});
xlabel('income(a,eps)');
ylabel('log of (MN_V_W_GAIN_CHECK(EM,J))');
xlim([0,7]);
grid on;
grid minor;
```



## Marginal Consumption Gains, Color as Shock, Conditional on Age, Marital, Kids, and Education

How do shocks and a impact marginal value. First plot one asset level, variation comes only from increasingly higher shocks:

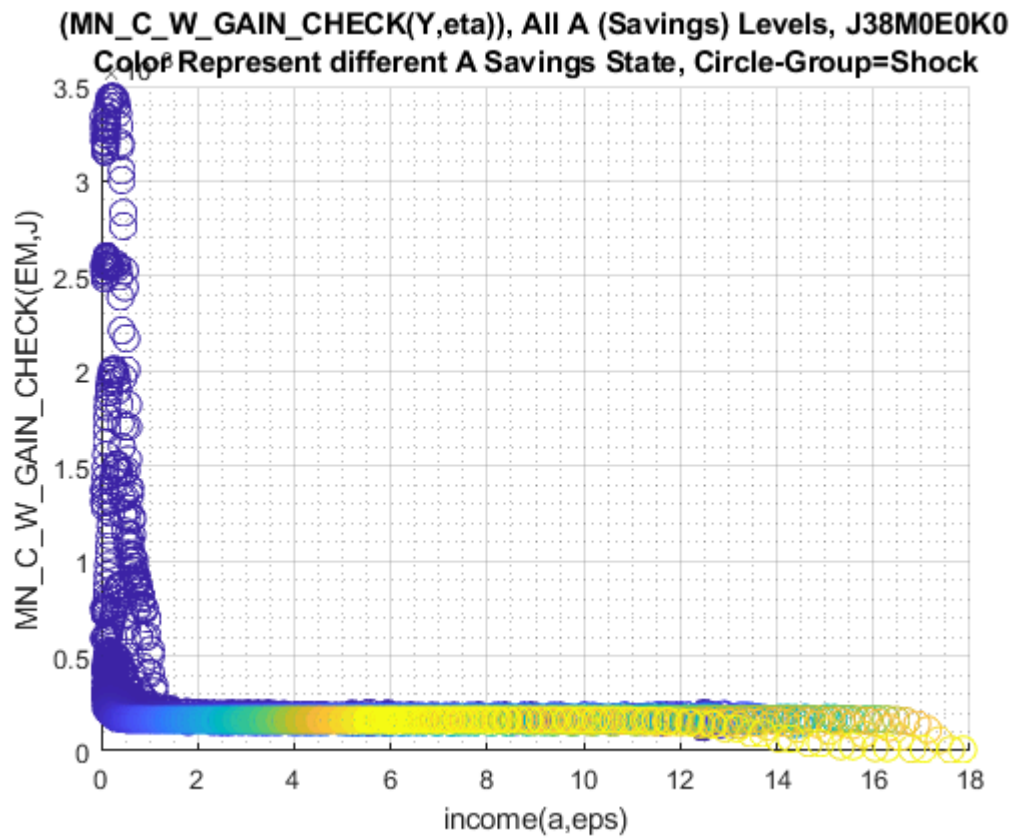
```
figure();
it_a = 50;
scatter(log(mt_total_inc_jemk(:,it_a)), mt_C_W_gain_check_jemk(:,it_a), 100);
title({'MN\C\W\GAIN\CHECK(Y(A, eta)), Given A Savings Level, J38M0E0K0', ...
      'Each Circle is A Different shock Level'});
xlabel('Y(a,eta) = Spouse + Household head Income Joint');
ylabel('MN\C\W\GAIN\CHECK(A, eta)');
grid on;
grid minor;
```



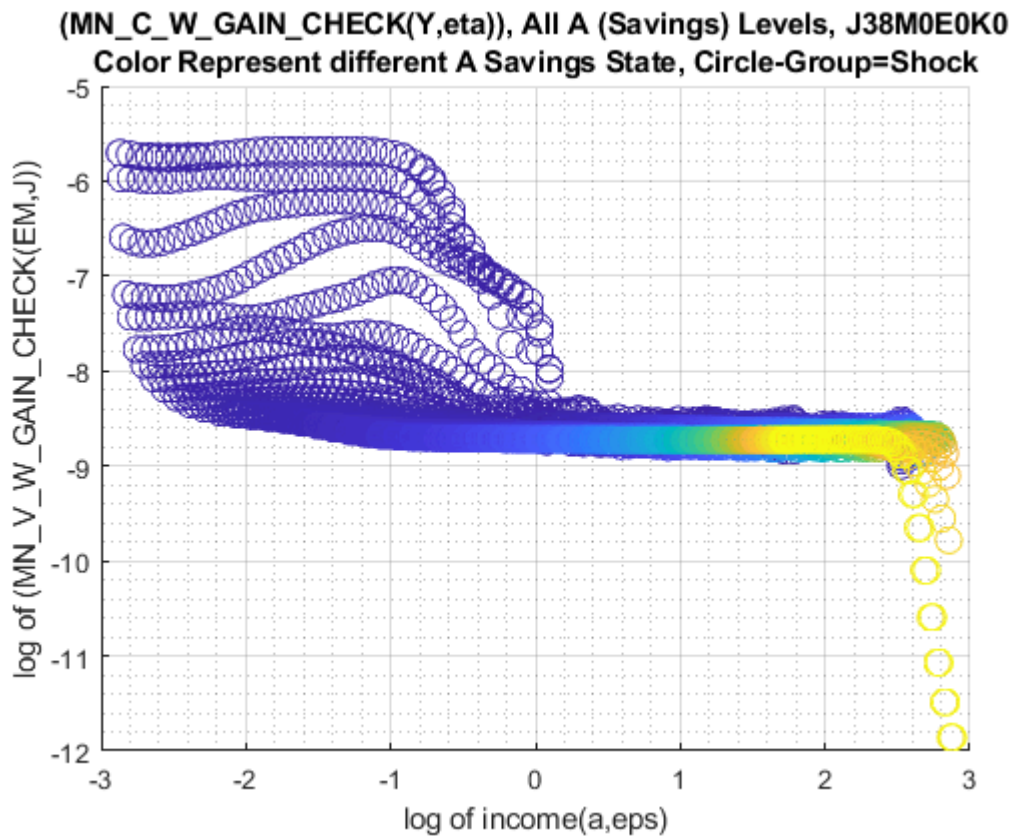
Plot all asset levels:

```
figure();
scatter((mt_total_inc_jemk(:)), (mt_C_W_gain_check_jemk(:)), 100, mt_a(:));
title({'(MN\C_W_GAIN_CHECK(Y,eta)), All A (Savings) Levels, J38M0E0K0', ...
      'Color Represent different A Savings State, Circle-Group=Shock'});
xlabel('income(a,eps)');
ylabel('MN\C_W_GAIN_CHECK(EM,J)');
grid on;
grid minor;
```





```
figure();
scatter(log(mt_total_inc_jemk(:)), log(mt_C_W_gain_check_jemk(:)), 100, mt_a(:));
title({'(MN\C_W_GAIN_CHECK(Y,eta)), All A (Savings) Levels, J38M0E0K0', ...
      'Color Represent different A Savings State, Circle-Group=Shock'});
xlabel('log of income(a,eps)');
ylabel('log of (MN\V_W_GAIN_CHECK(EM,J))');
grid on;
grid minor;
```



## 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_U_GAIN_CHECK(KM,J)), welf_checks=' num2str(welf_checks) ', TR=' num2str(
tb_az_v = ff_summ_nd_array(st_title, mn_V_U_gain_check, true, ["mean"], 3, 1, cl_mp_datasetdeso
```

```
xxx MEAN(MN_V_U_GAIN_CHECK(KM,J)), welf_checks=2, TR=0.0017225 xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx
group kids marry mean_age_18 mean_age_19 mean_age_20 mean_age_21 mean_age_22 mean_age_23
1 1 0 0.032313 0.031538 0.029203 0.026718 0.024639 0.022886
2 2 0 0.044673 0.043637 0.04037 0.036851 0.033899 0.0314
3 3 0 0.052552 0.051453 0.047283 0.043234 0.03984 0.036967
4 4 0 0.059787 0.058589 0.053843 0.049266 0.045426 0.042177
5 5 0 0.06554 0.064322 0.059162 0.054211 0.05006 0.046547
6 1 1 0.0059295 0.0054975 0.0049865 0.0045164 0.0041185 0.0037811
7 2 1 0.0083787 0.0077803 0.0070563 0.0063837 0.0058177 0.0053346
8 3 1 0.010146 0.0094404 0.0085806 0.0077709 0.0070881 0.0065113
9 4 1 0.012661 0.011814 0.01075 0.0097404 0.0088898 0.0081607
10 5 1 0.015891 0.014939 0.013651 0.012419 0.011375 0.010485
```

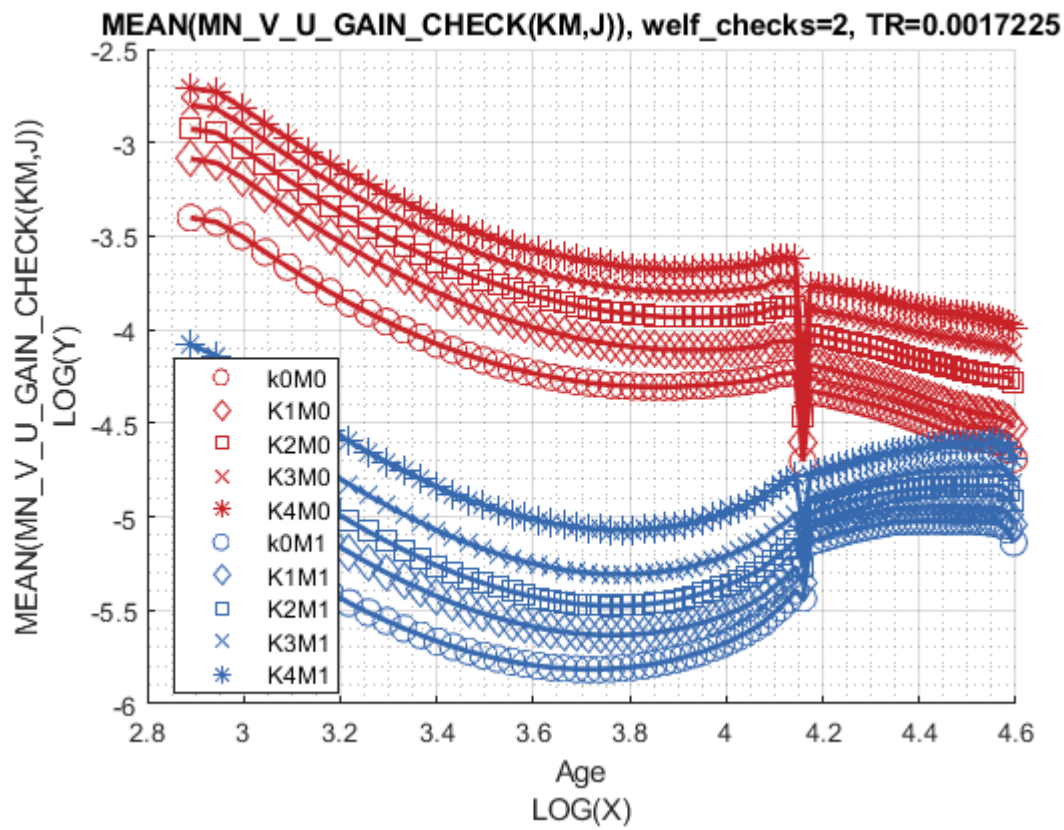
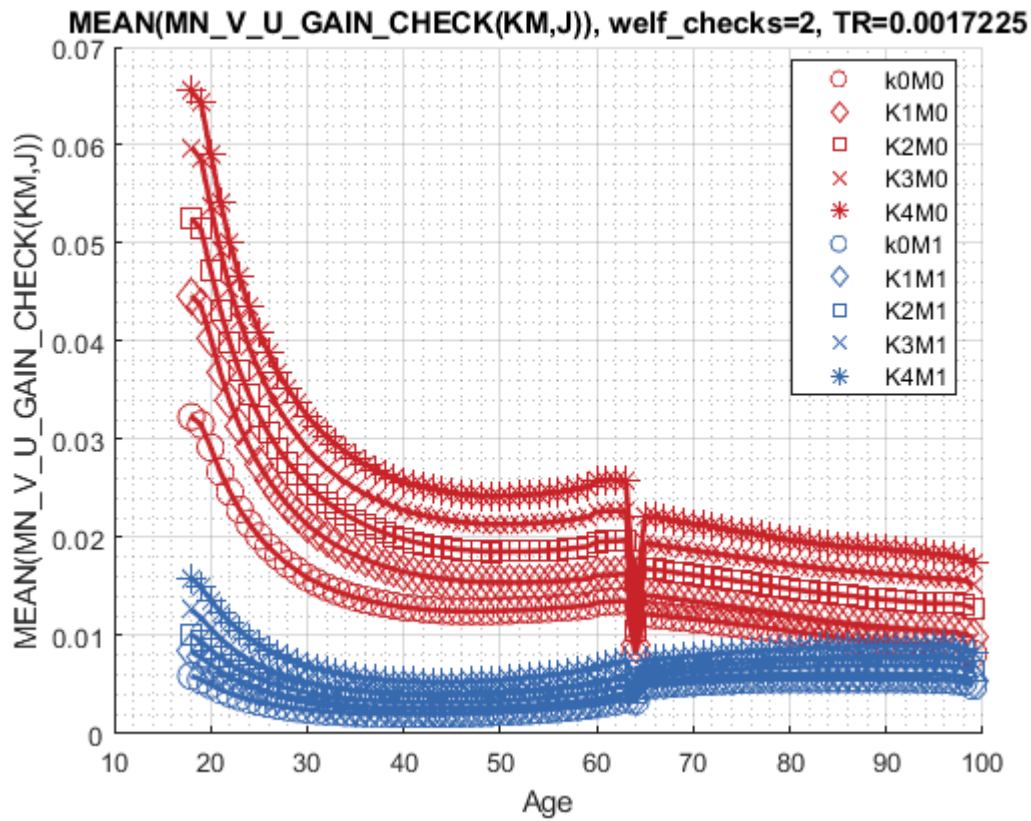
### % Consumption Function

```
st_title = ['MEAN(MN_MPC_U_GAIN_CHECK(KM,J)), welf_checks=' num2str(welf_checks) ', TR=' num2str(
tb_az_c = ff_summ_nd_array(st_title, mn_MPC_U_gain_share_check, true, ["mean"], 3, 1, cl_mp_dat
```

```
xxx MEAN(MN_MPC_U_GAIN_CHECK(KM,J)), welf_checks=2, TR=0.0017225 xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx
group kids marry mean_age_18 mean_age_19 mean_age_20 mean_age_21 mean_age_22 mean_age_23
1 1 0 0.084565 0.099857 0.10794 0.10516 0.10247 0.099966
2 2 0 0.096126 0.11139 0.12136 0.11819 0.11553 0.11307
3 3 0 0.1078 0.12631 0.13473 0.13138 0.12786 0.12514
4 4 0 0.114 0.13339 0.14178 0.13811 0.13446 0.13124
5 5 0 0.11992 0.14069 0.14855 0.14469 0.14011 0.13651
6 1 1 0.096646 0.10442 0.10672 0.10443 0.10194 0.099587
7 2 1 0.10031 0.1093 0.11166 0.10931 0.10836 0.10496
8 3 1 0.10594 0.11757 0.11904 0.11731 0.11604 0.11334
9 4 1 0.11209 0.12204 0.12456 0.12263 0.12006 0.11837
10 5 1 0.12333 0.13314 0.13686 0.13165 0.12869 0.12899
```

Graph Mean Values:

```
st_title = ['MEAN(MN_V_U_GAIN_CHECK(KM,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_U_GAIN_CHECK(KM,J))'};
ff_graph_grid((tb_az_v{1:end, 4:end}), ar_row_grid, age_grid, mp_support_graph);
```

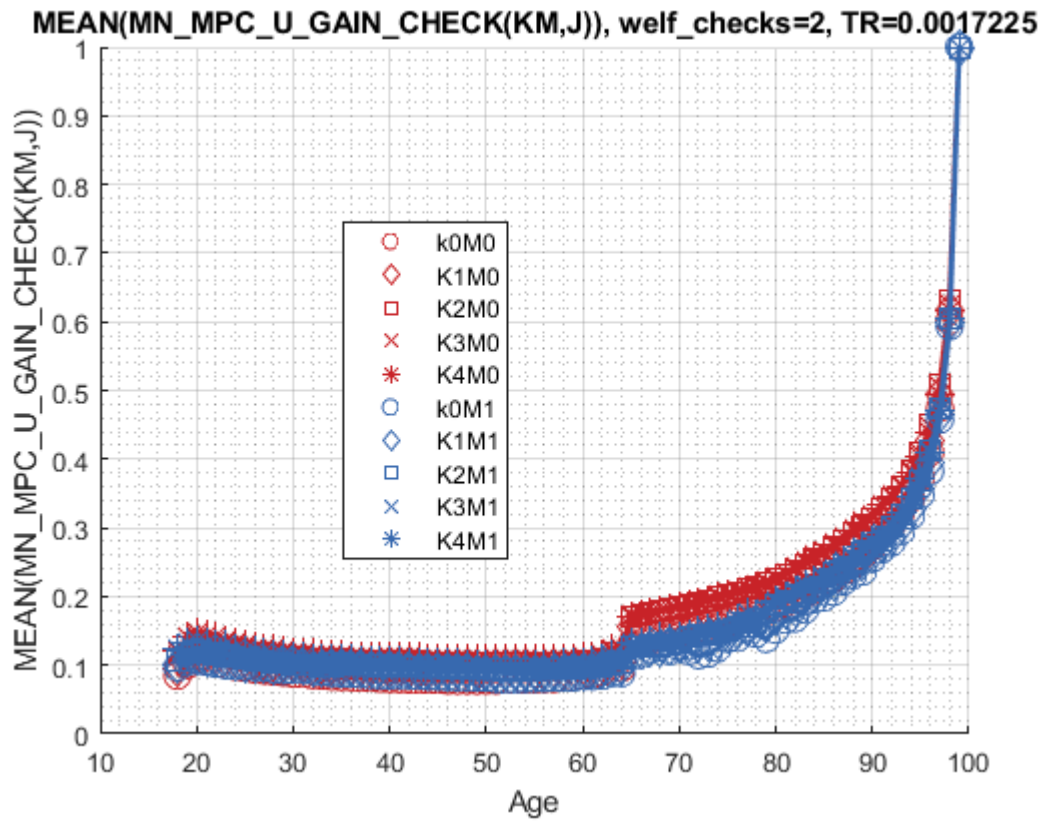


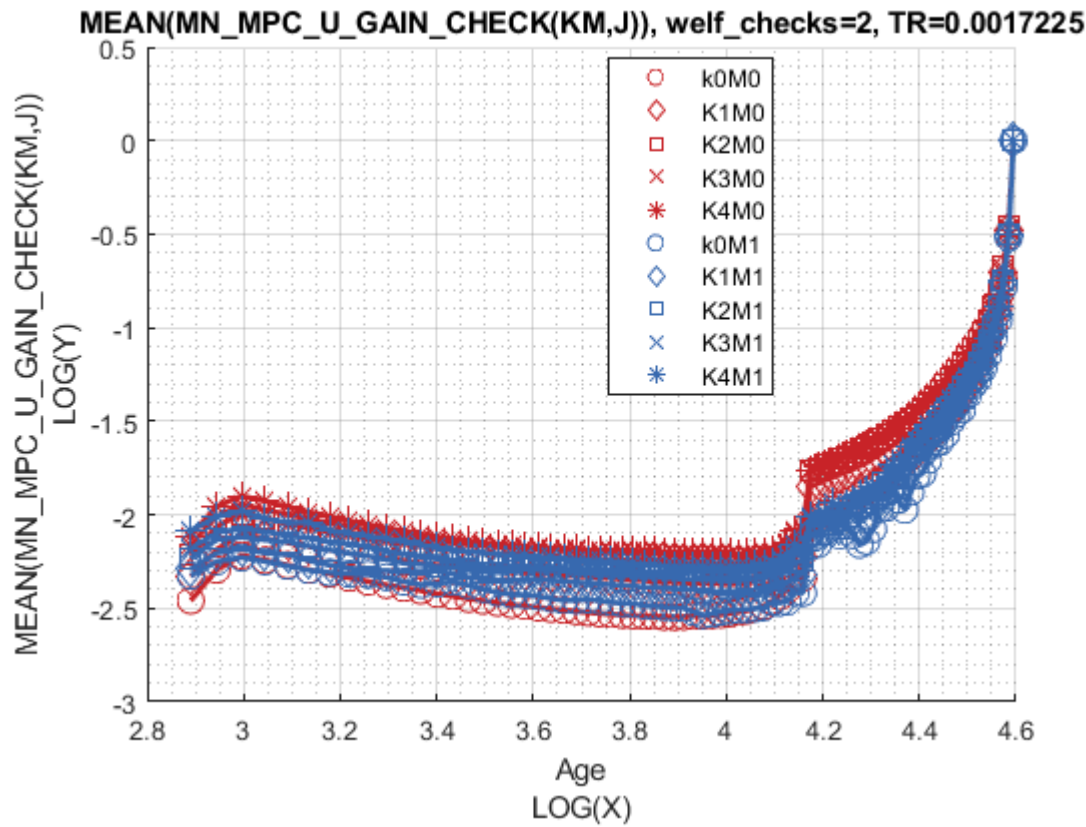
Graph Mean Consumption (**MPC: Share of Check Consumed**):

```

st_title = ['MEAN(MN_MPC_U_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_U_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_U_GAIN_CHECK(EM,J)), welf_checks=' num2str(welf_checks) ', TR=' num2str(TR)];
tb_az_v = ff_summ_nd_array(st_title, mn_V_U_gain_check, true, ["mean"], 3, 1, cl_mp_datasetdesc);
```

```
xxx MEAN(MN_V_U_GAIN_CHECK(EM,J)), welf_checks=2, TR=0.0017225 xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx
group   edu   marry   mean_age_18   mean_age_19   mean_age_20   mean_age_21   mean_age_22   mean_age_23
```



1	0	0	0.052161	0.051299	0.048759	0.045984	0.043519	0.041317
2	1	0	0.049785	0.048517	0.043186	0.038128	0.034027	0.030674
3	0	1	0.011544	0.010825	0.010006	0.0092235	0.0085439	0.0079503
4	1	1	0.0096585	0.0089631	0.0080043	0.0071088	0.006372	0.0057587

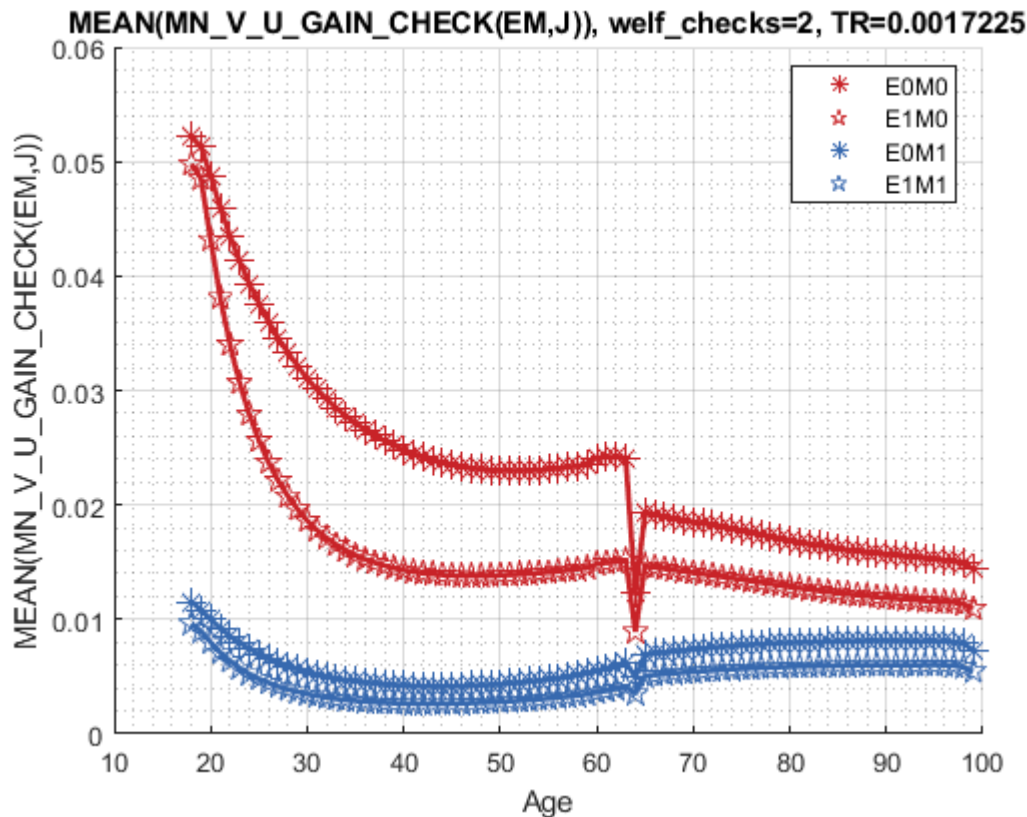
### % Consumption

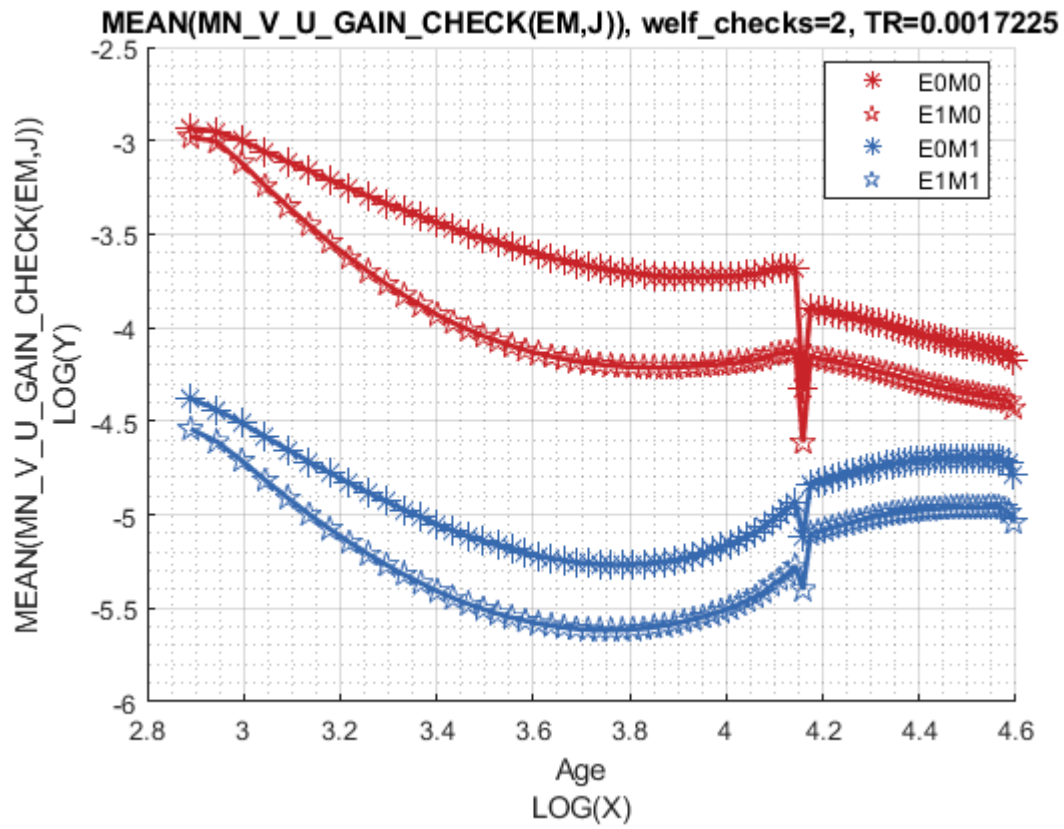
```
st_title = ['MEAN(MN_MPC_U_GAIN_CHECK(EM,J)), welf_checks=' num2str(welf_checks) ', TR=' num2str(TR)'];
tb_az_c = ff_summ_nd_array(st_title, mn_MPC_U_gain_share_check, true, ["mean"], 3, 1, cl_mp_data);
```

xxx	group	edu	marry	mean_age_18	mean_age_19	mean_age_20	mean_age_21	mean_age_22	mean_age_23
1	0	0	0.09247	0.10277	0.10886	0.10804	0.10737	0.10669	
2	1	0	0.11649	0.14189	0.15288	0.14697	0.1408	0.13569	
3	0	1	0.09821	0.10337	0.10589	0.10515	0.10424	0.1031	
4	1	1	0.11712	0.13122	0.13364	0.12898	0.12579	0.12299	

Graph Mean Values:

```
st_title = ['MEAN(MN_V_U_GAIN_CHECK(EM,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_U_GAIN_CHECK(EM,J))'};
ff_graph_grid((tb_az_v{1:end, 4:end}), ar_row_grid, age_grid, mp_support_graph);
```



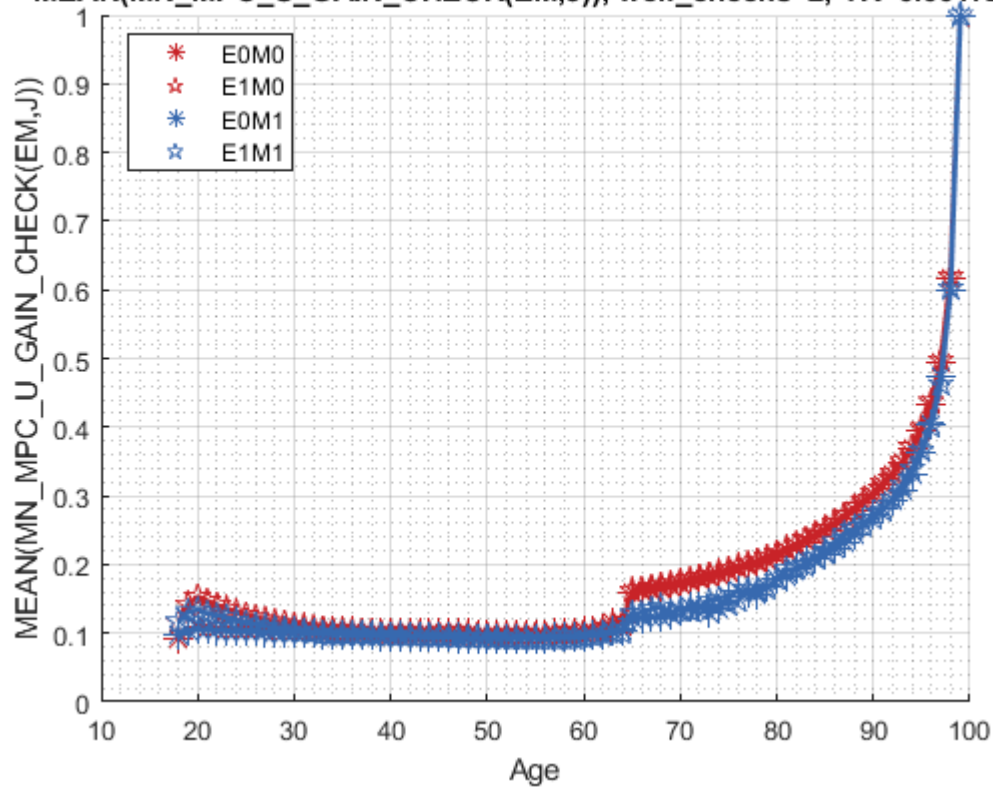


Graph Mean Consumption (**MPC: Share of Check Consumed**):

```
st_title = ['MEAN(MN\MPC\U\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\U\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\_U\_GAIN\_CHECK(EM,J)), welf\_checks=2, TR=0.0017225



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

