

2007 (Bush 2008 Stimulus) Full States EV and EC of Two Checks

This is the example vignette for function: `snw_evuvw19_jaeemk_foc` from the [PrjOptiSNW Package](#). 2008 integrated over VU and VW, given optimal savings choices, unemployment shocks and various expectations.

Given 2008 policy and value functions (given expectation of 2009 crisis unemployment shocks), call `SNW_V08_JAEEMK` to solve for value and consumption given stimulus checks. And then integrate over 08 JAEEMK states given 07 JAEEMK states (age, endogenous savings, education, income shock, marital status, kids count). The stimulus will be provisioned based on 07 JAEEMK states. Note that `snw_evuvw19_jaeemk` does not solve the 07/08 problem.

Despite the name, this function supports solving the 2019 looking into 2020 as well as the 2007 looking into 2008 problems. The idea is that the planner only has information from 2019 and from 2007, and must allocate using those information. Stimulus, however, is given in 2020 and in 2008. So the planner needs to consider expected values in consumption or welfare given the transition probabilities of states in 2007 to 2008 and in 2019 to 2020. The `snw_evuvw19_jmky` file then aggregates the full state-space results to just JMKY state-space, which is the extend of information available to the planner.

Test SNW_EVUVW19_JAEEMK Defaults for 2019

Call the function with defaults. First, set up some parameters.

```
clear all;
% Solution types
st_biden_or_trump = 'bushchck';

% Solve the VFI Problem and get Value Function
mp_controls = snw_mp_control('default_test');

% Solve for Unemployment Values
mp_controls('bl_print_a4chk') = false;
mp_controls('bl_print_a4chk_verbose') = false;
mp_controls('bl_print_vfi') = false;
mp_controls('bl_print_vfi_verbose') = false;
mp_controls('bl_print_ds') = false;
mp_controls('bl_print_ds_verbose') = false;
mp_controls('bl_print_precompute') = false;
mp_controls('bl_print_evuvw20_jaeemk') = false;
mp_controls('bl_print_evuvw20_jaeemk_verbose') = false;
mp_controls('bl_print_v08p08_jaeemk') = false;
mp_controls('bl_print_v08p08_jaeemk_verbose') = false;
mp_controls('bl_print_v08_jaeemk') = true;
mp_controls('bl_print_v08_jaeemk_verbose') = false;
```

Second, run initializing functions.

```
% 1. generate MP_PARAMS specific to 2008 stimulus
% Use non-default values for Bush Stimulus
mp_more_inputs = containers.Map('KeyType','char', 'ValueType','any');
mp_more_inputs('fl_ss_non_college') = 0.225;
mp_more_inputs('fl_ss_college') = 0.271;
```

```

fl_p50_hh_income_07 = 54831;
mp_more_inputs('fl_scaleconvertor') = fl_p50_hh_income_07;
% st_param_group = 'default_small';
st_param_group = 'default_dense';
st_param_group = 'default_docdense';
mp_params = snw_mp_param(st_param_group, false, 'tauchen', false, 8, 8, mp_more_inputs);
mp_params('st_biden_or_trump') = st_biden_or_trump;
mp_params('beta') = 0.95;
% 2. Solve value steady state (2009 employed)
[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=495.65

```

V_emp_2009 = V_ss;
inc_ss = mp_valpol_more_ss('inc_VFI');
spouse_inc_ss = mp_valpol_more_ss('spouse_inc_VFI');
total_inc_ss = inc_ss + spouse_inc_ss;
% 3. Solve value unemployed 2009
mp_params('xi') = 0.532;
mp_params('b') = 0.37992;
mp_params('a2_covidyr') = mp_params('a2_greatrecession_2009');
mp_params('TR') = 100/fl_p50_hh_income_07;
[V_unemp_2009] = 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=497.3876

```

% 4. Value and Optimal choice in 2008
[V_2008, ap_2008, cons_2008, ev_emphsk_2009] = ...
    snw_v08p08_jaeemk(mp_params, mp_controls, V_emp_2009, V_unemp_2009);

```

Completed SNW_VFI_MAIN_BISEC_VEC 1 Period Unemp Shock;SNW_MP_PARAM=default_docdense;SNW_MP_CONTROL=default_test;time=497.3876
Completed SNW_V08P08_JAEEMK;SNW_MP_PARAM=default_docdense;SNW_MP_CONTROL=default_test;time=497.3876

```

% 5. matrixes to pre-compute
% Only using the SNW_A4CHK_WRK_BISEC_VEC function, no unemployment
% related matrixes needed Also don't need REF_EARN_WAGEIND_GRID,
% become unemployment not conditional on wage in 2009.
cl_st_precompute_list = {'a', ...
    'inc', 'inc_unemp', 'spouse_inc',...
    'ar_z_ctr_amz'};
% Shared: Steady-State distribution
[Phi_true] = snw_ds_main(mp_params, mp_controls, ap_ss, cons_ss, mp_valpol_more_ss);

```

Completed SNW_DS_MAIN;SNW_MP_PARAM=default_docdense;SNW_MP_CONTROL=default_test;time=1775.4241

```

% Shared: precompute, get Matrixes
% note, the mp_params inputs are based on unemployed in 2020 (MIT) or unemployed in 2009 (Expect)
% note, however, for the 2008/9 problem, only will use inc, inc_unemp, spouse_inc
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=254.7405

Solve for 0 and 2 checks, by finding the increase to asset state-space that is equivalent to the check increase, so that the problem can be solved without increasing the state-space.

```
welf_checks = 0;
[ev07_jaeemk_check0, ec07_jaeemk_check0, ev08_jaeemk_check0, ec08_jaeemk_check0] = snw_evuvvw19_
    welf_checks, mp_params, mp_controls, ...
    ap_ss, V_2008, cons_2008, mp_precompute_res);
```

Solve for V_2008_check for 0 stimulus checks
Completed SNW_A4CHK_WRK_BISEC_VEC;SNW_MP_PARAM=bushchck;welf_checks=0;TR=0.0018238;SNW_MP_PARAM=default_docdense;SNW_MP_CONTROL=default_test;timeEUEC=6.99e-05
Completed SNW_V08_JAEEMK;SNW_MP_PARAM=default_docdense;SNW_MP_CONTROL=default_test;timeEUEC=6.99e-05
Completed SNW_EVUVW19_JAEEMK_FOC;st_biden_or_trump=bushchck;SNW_MP_PARAM=default_docdense;SNW_MP_CONTROL=default_test;timeEUEC=6.99e-05

```
XXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXX
CONTAINER NAME: mp_outcomes ND Array (Matrix etc)
XXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXXX
```

	i	idx	ndim	numel	rowN	colN	sum	mean	std	coefvar
	—	—	—	—	—	—	—	—	—	—
ec07_jaeemk	1	1	6	4.3173e+07	82	5.265e+05	1.9685e+08	4.5597	5.3244	1.167
ec08_jaeemk	2	2	6	4.37e+07	83	5.265e+05	2.3277e+08	5.3267	8.4419	1.584
ev07_jaeemk	3	3	6	4.3173e+07	82	5.265e+05	-6.4618e+08	-14.967	21.06	-1.407
ev08_jaeemk	4	4	6	4.37e+07	83	5.265e+05	-6.6426e+08	-15.201	21.85	-1.432

xxx TABLE:ec07_jaeemk xxxxxxxxxxxxxxxxxxxx										
	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	c526500
r1	0.039543	0.039543	0.039978	0.042271	0.04854	12.009	12.281	12.561	12.849	13.137
r2	0.039889	0.039889	0.040323	0.043305	0.049735	12.251	12.524	12.806	13.095	13.383
r3	0.041432	0.041432	0.041608	0.043991	0.050734	12.485	12.759	13.042	13.331	13.619
r4	0.042935	0.042935	0.043023	0.045459	0.052199	12.742	13.017	13.3	13.588	13.866
r5	0.044395	0.044395	0.044399	0.046895	0.053615	12.99	13.266	13.548	13.834	14.12
r78	0.2016	0.2016	0.2016	0.2016	0.20214	27.775	28.774	29.785	30.965	32.21
r79	0.2016	0.2016	0.2016	0.2016	0.2016	30.43	31.663	32.736	33.958	35.24
r80	0.2016	0.2016	0.2016	0.2016	0.2016	33.68	35.501	37.368	38.956	40.605
r81	0.2016	0.2016	0.2016	0.2016	0.2016	40.118	41.397	43.175	45.605	48.094
r82	0.2016	0.2016	0.2016	0.2016	0.2016	52.1	55.541	58.464	60.094	62.866

xxx TABLE:ec08_jaeemk xxxxxxxxxxxxxxxxxxxx										
	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	c526500
r1	0.036218	0.036736	0.038184	0.042735	0.048545	12.256	12.541	12.835	13.136	13.437
r2	0.036271	0.036736	0.038385	0.043404	0.049852	12.491	12.778	13.072	13.374	13.675
r3	0.036717	0.037251	0.039845	0.044907	0.051515	12.744	13.032	13.327	13.628	13.929
r4	0.038144	0.038678	0.041269	0.046371	0.053128	12.989	13.277	13.573	13.872	14.171
r5	0.039534	0.040068	0.042653	0.047793	0.054687	13.224	13.513	13.809	14.105	14.401
r79	0.2016	0.20214	0.20586	0.21598	0.23568	35.82	37.367	39.414	41.705	44.196
r80	0.2016	0.20214	0.20586	0.21598	0.23568	40.755	42.955	45.289	47.95	50.861
r81	0.2016	0.20214	0.20586	0.21598	0.23568	48.912	52.041	55.022	57.919	60.73
r82	0.2016	0.20214	0.20586	0.21598	0.23568	66.719	69.201	72.373	77.005	81.638
r83	0.2016	0.20214	0.20586	0.21598	0.23568	116.83	122.65	128.67	134.89	141.31

xxx TABLE:ev07_jaeemk xxxxxxxxxxxxxxxxxxxx									
	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499
r1	-282.59	-282.59	-282.23	-278.38	-270.8	-4.3833	-4.2888	-4.1954	-4.1032
r2	-272.49	-272.49	-272.13	-268.93	-261.76	-4.2754	-4.1844	-4.0945	-4.0056

r3	-262.36	-262.36	-262.23	-259.89	-253.05	-4.1651	-4.0777	-3.9913	-3.9057
r4	-253.22	-253.22	-253.16	-250.92	-244.6	-4.0512	-3.9675	-3.8846	-3.8025
r5	-244.95	-244.95	-244.95	-242.81	-236.93	-3.9436	-3.8633	-3.7837	-3.7047
r78	-13.362	-13.362	-13.362	-13.362	-13.349	-0.27313	-0.26104	-0.24971	-0.2389
r79	-12.032	-12.032	-12.032	-12.032	-12.032	-0.21855	-0.20781	-0.19875	-0.19057
r80	-10.388	-10.388	-10.388	-10.388	-10.388	-0.16126	-0.15407	-0.14734	-0.14134
r81	-8.1801	-8.1801	-8.1801	-8.1801	-8.1801	-0.10114	-0.097396	-0.093438	-0.089426
r82	-4.9651	-4.9651	-4.9651	-4.9651	-4.9651	-0.044201	-0.041462	-0.039412	-0.038348

r1	0.037941	0.038148	0.039819	0.043807	0.049244	12.256	12.541	12.835	13.136	1
r2	0.038108	0.038344	0.040188	0.044594	0.050571	12.492	12.778	13.073	13.374	1
r3	0.03941	0.039781	0.041664	0.046126	0.052244	12.745	13.032	13.327	13.628	1
r4	0.040834	0.041205	0.043102	0.047618	0.053867	12.989	13.278	13.573	13.872	1
r5	0.04222	0.042589	0.0445	0.049065	0.055435	13.224	13.513	13.809	14.105	1
r79	0.20525	0.20579	0.20951	0.21963	0.23776	35.821	37.368	39.415	41.707	5
r80	0.20525	0.20579	0.20951	0.21963	0.23776	40.756	42.957	45.29	47.952	5
r81	0.20525	0.20579	0.20951	0.21963	0.2378	48.914	52.043	55.024	57.92	6
r82	0.20525	0.20579	0.20951	0.21963	0.23814	66.72	69.203	72.375	77.007	8
r83	0.20525	0.20579	0.20951	0.21963	0.23933	116.84	122.66	128.68	134.89	1

xxx TABLE:ev07_jaeemk xxxxxxxxxxxxxxxxxxxx

	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	
r1	-280.23	-280.23	-279.88	-276.5	-269.35	-4.3833	-4.2887	-4.1954	-4.1031	
r2	-270.21	-270.21	-269.89	-267.08	-260.35	-4.2753	-4.1844	-4.0944	-4.0055	
r3	-260.26	-260.26	-260.13	-258.1	-251.71	-4.165	-4.0777	-3.9912	-3.9057	
r4	-251.26	-251.26	-251.2	-249.25	-243.33	-4.0511	-3.9674	-3.8846	-3.8025	
r5	-243.12	-243.12	-243.12	-241.24	-235.73	-3.9436	-3.8633	-3.7837	-3.7047	
r78	-13.274	-13.274	-13.274	-13.274	-13.262	-0.27312	-0.26103	-0.2497	-0.23889	
r79	-11.944	-11.944	-11.944	-11.944	-11.944	-0.21854	-0.2078	-0.19875	-0.19056	
r80	-10.3	-10.3	-10.3	-10.3	-10.3	-0.16125	-0.15406	-0.14733	-0.14134	
r81	-8.0921	-8.0921	-8.0921	-8.0921	-8.0921	-0.10113	-0.097391	-0.093433	-0.089423	
r82	-4.8771	-4.8771	-4.8771	-4.8771	-4.8771	-0.044198	-0.04146	-0.03941	-0.038346	

xxx TABLE:ev08_jaeemk xxxxxxxxxxxxxxxxxxxx

	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	
r1	-293.09	-292.72	-290.49	-284.88	-275.86	-4.3615	-4.2672	-4.1741	-4.0821	
r2	-283.55	-283.18	-280.98	-275.47	-266.73	-4.2548	-4.164	-4.0743	-3.9857	
r3	-274.01	-273.65	-271.52	-266.23	-257.88	-4.146	-4.0589	-3.9726	-3.8874	
r4	-264.47	-264.13	-262.14	-257.19	-249.33	-4.0341	-3.9506	-3.8679	-3.7861	
r5	-255.84	-255.53	-253.66	-249	-241.59	-3.9286	-3.8484	-3.769	-3.6903	
r79	-13.268	-13.255	-13.171	-12.954	-12.578	-0.22088	-0.21054	-0.20082	-0.1917	
r80	-11.937	-11.924	-11.841	-11.623	-11.248	-0.16976	-0.16179	-0.15428	-0.1472	
r81	-10.294	-10.281	-10.198	-9.9804	-9.6057	-0.11711	-0.11162	-0.10644	-0.10156	
r82	-8.0862	-8.0735	-7.9895	-7.7724	-7.3981	-0.065327	-0.062237	-0.059355	-0.056631	
r83	-4.8723	-4.8595	-4.7755	-4.5584	-4.1854	-0.020965	-0.01997	-0.019036	-0.018159	

Differences between Checks in Expected Value and Expected Consumption

```
mn_V_U_gain_check_07 = ev07_jaeemk_check2 - ev07_jaeemk_check0;
mn_MPC_U_gain_share_check_07 = (ec07_jaeemk_check2 - ec07_jaeemk_check0)./(welf_checks*mp_param
```

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_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_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_07, true, ["mean"], 4, 1, cl_mp_datasetc

```

xxx	MEAN(MN_V_U_GAIN_CHECK(A,Z)), welf_checks=2, TR=0.0018238	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	0.52526	0.48873	0.44813	0.40715	0.3683	0.33276	0.30644	0.27025
2	0.00051498	0.52526	0.48873	0.44813	0.40715	0.3683	0.33276	0.30644	0.27025
3	0.0041199	0.52469	0.48792	0.44711	0.40616	0.36748	0.3321	0.30644	0.27025
4	0.013905	0.47091	0.43897	0.40471	0.37007	0.337	0.30644	0.27025	0.23037
5	0.032959	0.39774	0.3737	0.34722	0.32037	0.29449	0.27025	0.23037	0.20037
6	0.064373	0.32479	0.30724	0.28795	0.26815	0.24878	0.23037	0.20037	0.17037

```

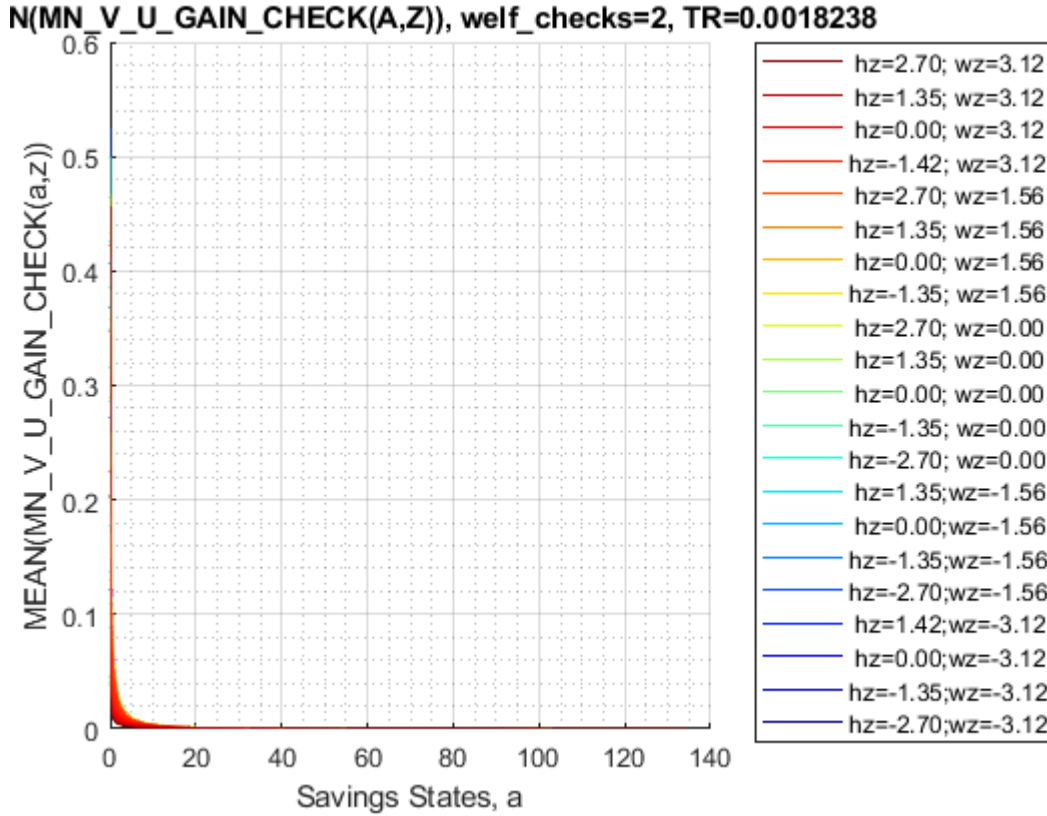
% Consumption
st_title = ['MEAN(MN_MPC_U_GAIN_CHECK(A,Z)), welf_checks=' num2str(welf_checks) ', TR=' num2str(m
tb_az_c = ff_summ_nd_array(st_title, mn_MPC_U_gain_share_check_07, true, ["mean"], 4, 1, cl_mp_datasetc

```

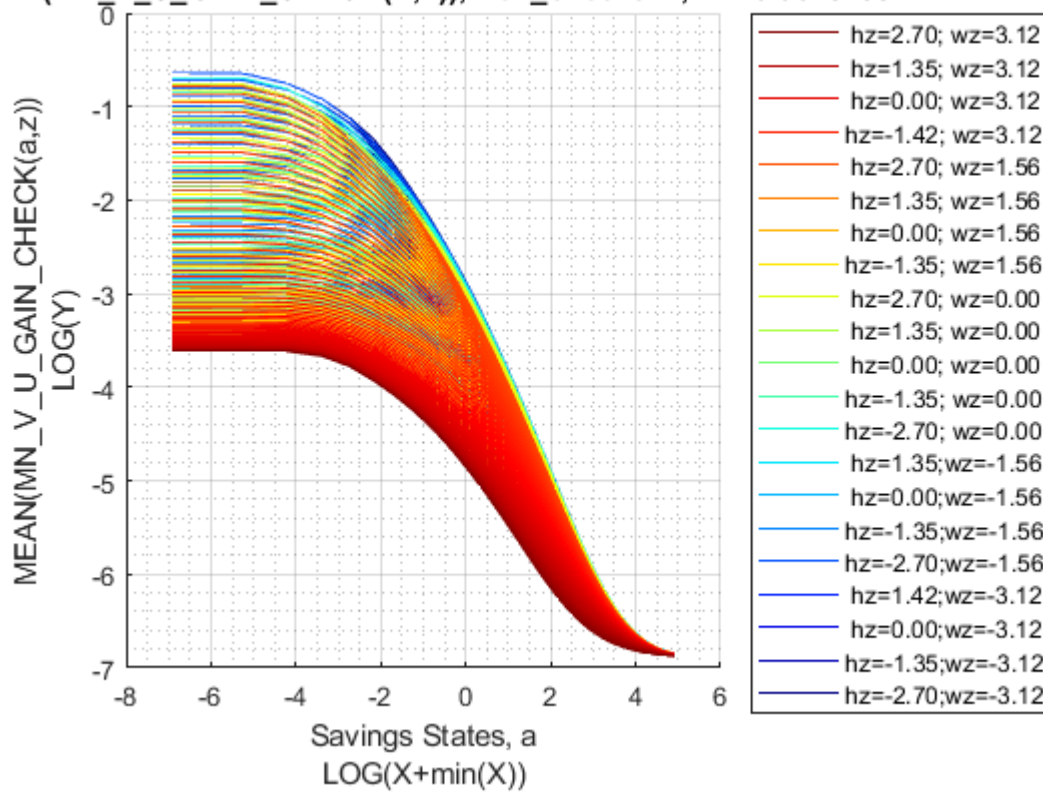
xxx	MEAN(MN_MPC_U_GAIN_CHECK(A,Z)), welf_checks=2, TR=0.0018238	xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx							
group	savings	mean_eta_1	mean_eta_2	mean_eta_3	mean_eta_4	mean_eta_5	mean_eta_6	mean_eta_7	mean_eta_8
1	0	0.74182	0.72635	0.71164	0.69952	0.69052	0.68421	0.67793	0.67164
2	0.00051498	0.74182	0.72635	0.71164	0.69952	0.69052	0.68421	0.67793	0.67164
3	0.0041199	0.74131	0.72479	0.70912	0.69655	0.68793	0.68245	0.67616	0.66987
4	0.013905	0.69437	0.6887	0.68682	0.6856	0.68485	0.68443	0.68401	0.68359
5	0.032959	0.62728	0.6255	0.62438	0.62423	0.62492	0.62615	0.62677	0.62739
6	0.064373	0.55114	0.54963	0.54908	0.54953	0.55084	0.55275	0.55466	0.55657

Graph Mean Values:

```
st_title = ['MEAN(MN\_V\_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\_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);
```



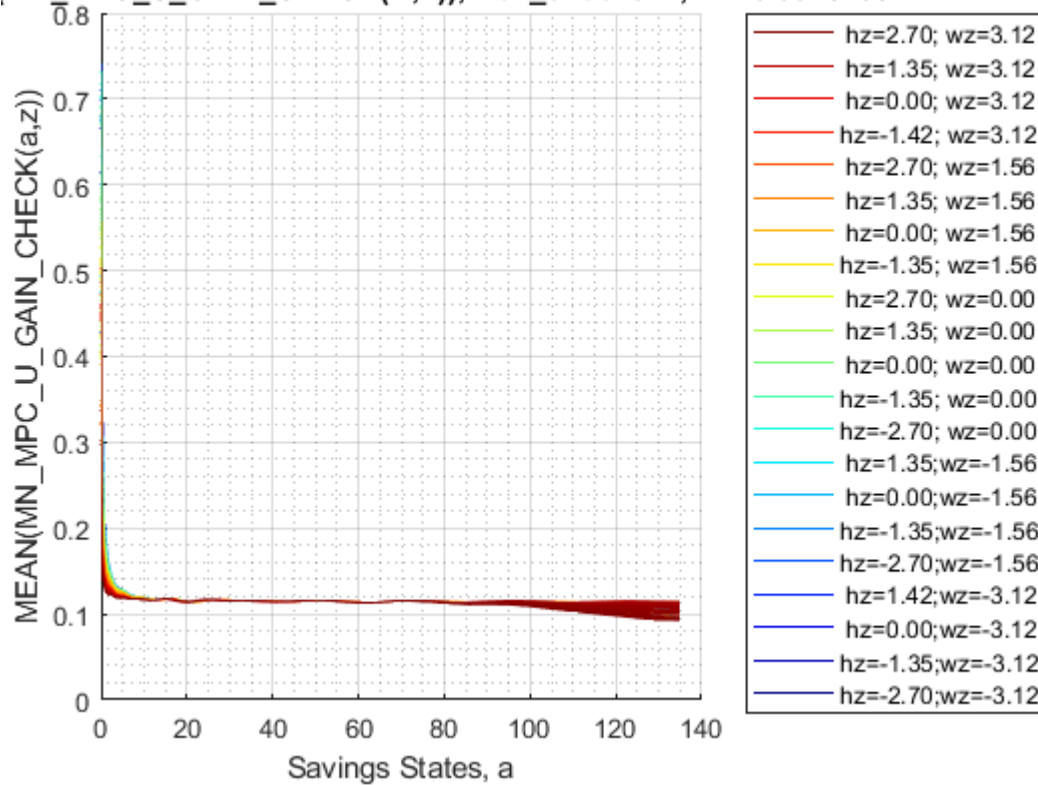
AN(MN_V_U_GAIN_CHECK(A,Z)), welf_checks=2, TR=0.0018238



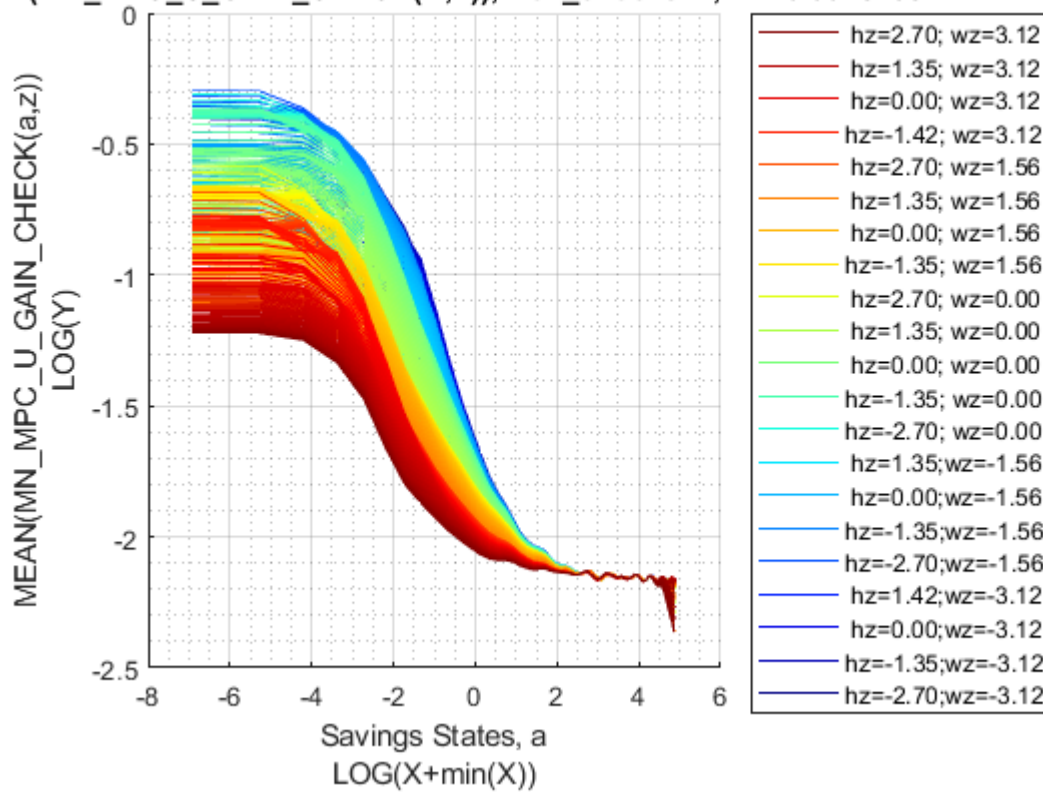
Graph Mean Consumption (**MPC: Share of Check Consumed**):

```
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);
```


MN_MPC_U_GAIN_CHECK(A,Z)), welf_checks=2, TR=0.0018238



N(MN_MPC_U_GAIN_CHECK(A,Z)), welf_checks=2, TR=0.0018238



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 levles.

Set Up date, Select Age 37vn

, unmarried, no kids, lower education:

```
% NaN(n_jgrid,n_agrid,n_etagrid,n_educgrid,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_ss = total_inc_ss(19, :, 1:mp_params('n_eta_H_grid'), 1, 1, 1);
mn_V_W_gain_check_use_07 = ev07_jaeemk_check2 - ev07_jaeemk_check0;
mn_C_W_gain_check_use_07 = ec07_jaeemk_check2 - ec07_jaeemk_check0;
```

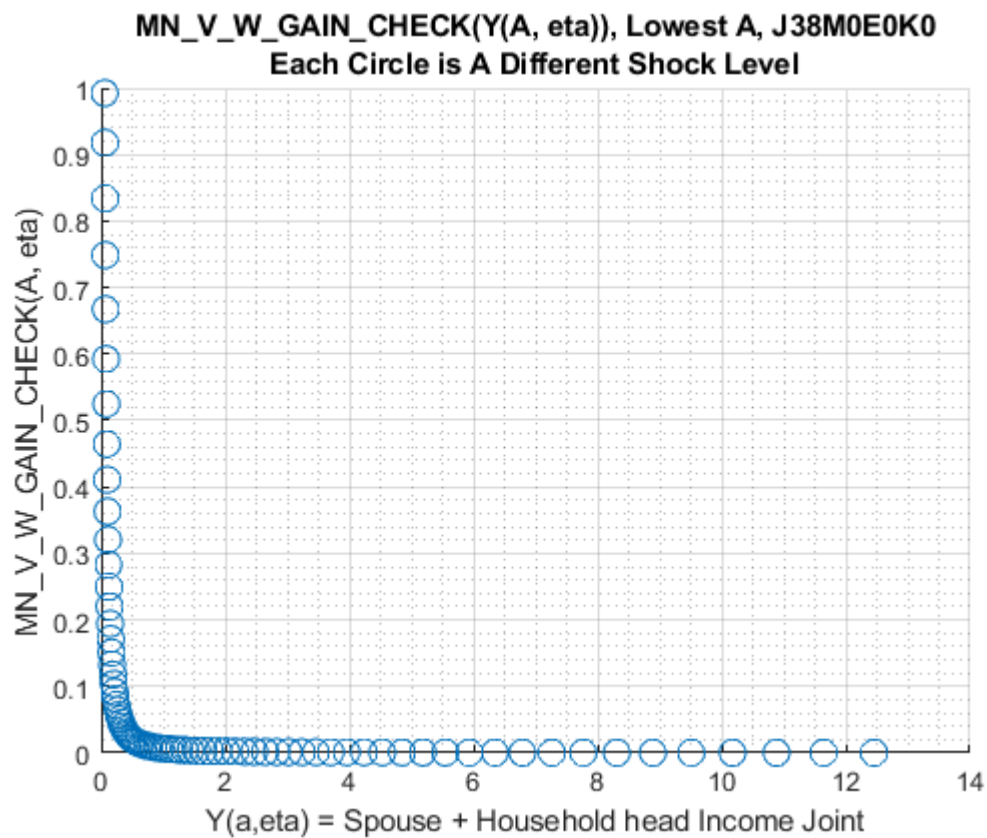
Select Age, Education, Marital, Kids Count:s

```
% 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_educgrid,n_marriedgrid,n_kidsgrid);
mn_C_W_gain_check_jemk_07 = mn_C_W_gain_check_use_07(it_age, :, 1:mp_params('n_eta_H_grid'), it_educ, it_marital, it_kids);
mn_V_W_gain_check_jemk_07 = mn_V_W_gain_check_use_07(it_age, :, 1:mp_params('n_eta_H_grid'), it_educ, it_marital, it_kids);
% Reshape, so shock is the first dim, a is the second
mt_total_inc_jemk = permute(mn_total_inc_jemk_ss, [3, 2, 1]);
mt_C_W_gain_check_jemk_07 = permute(mn_C_W_gain_check_jemk_07, [3, 2, 1]);
mt_C_W_gain_check_jemk_07(mt_C_W_gain_check_jemk_07 <= 1e-10) = 1e-10;
mt_V_W_gain_check_jemk_07 = permute(mn_V_W_gain_check_jemk_07, [3, 2, 1]);
mt_V_W_gain_check_jemk_07(mt_V_W_gain_check_jemk_07 <= 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);
```

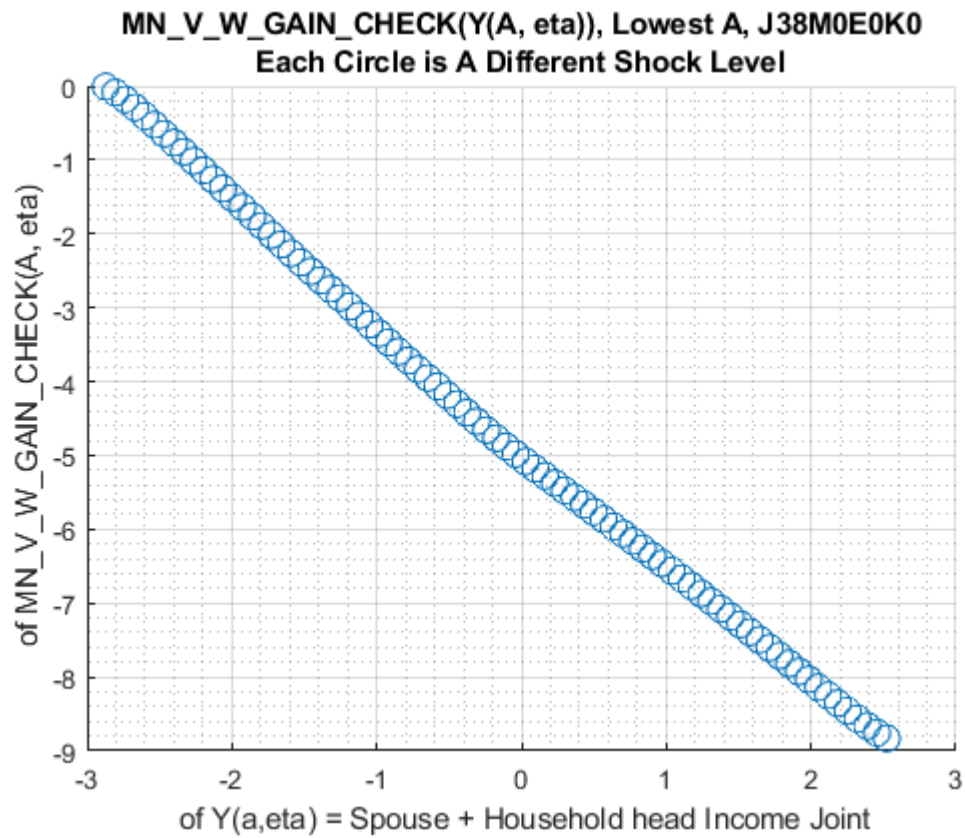
Marginal Value Gains, Color as Shock, Conditional on Age, Marital, Kids, and Education

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_07(:, 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;
```

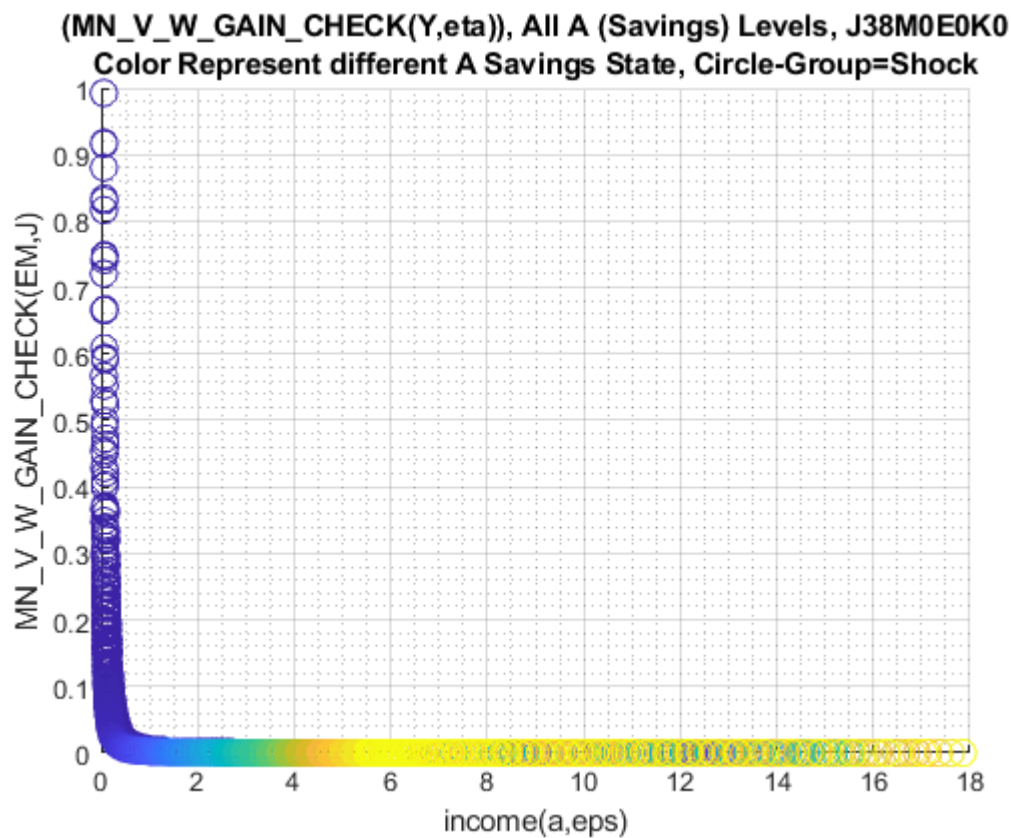


```
figure();
it_shock = 1;
scatter(log(mt_total_inc_jemk(:,it_a)), log(mt_V_W_gain_check_jemk_07(:,it_a)), 100);
title({'MN_V_W_GAIN_CHECK(Y(A, eta)), Lowest A, J38M0E0K0', ...
      'Each Circle is A Different Shock Level'});
xlabel(' of Y(a,eta) = Spouse + Household head Income Joint');
ylabel(' of MN_V_W_GAIN_CHECK(A, eta)');
grid on;
grid minor;
```

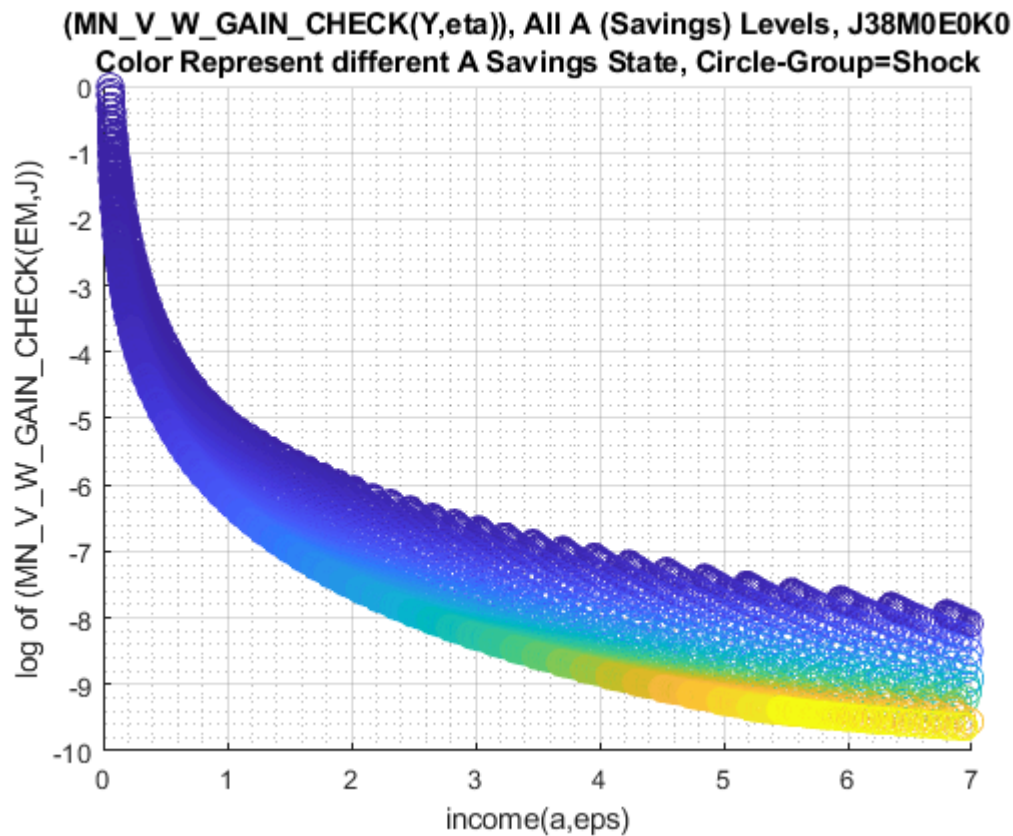


Plot all asset levels:

```
figure();
scatter((mt_total_inc_jemk(:)), (mt_V_W_gain_check_jemk_07(:)), 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;
```



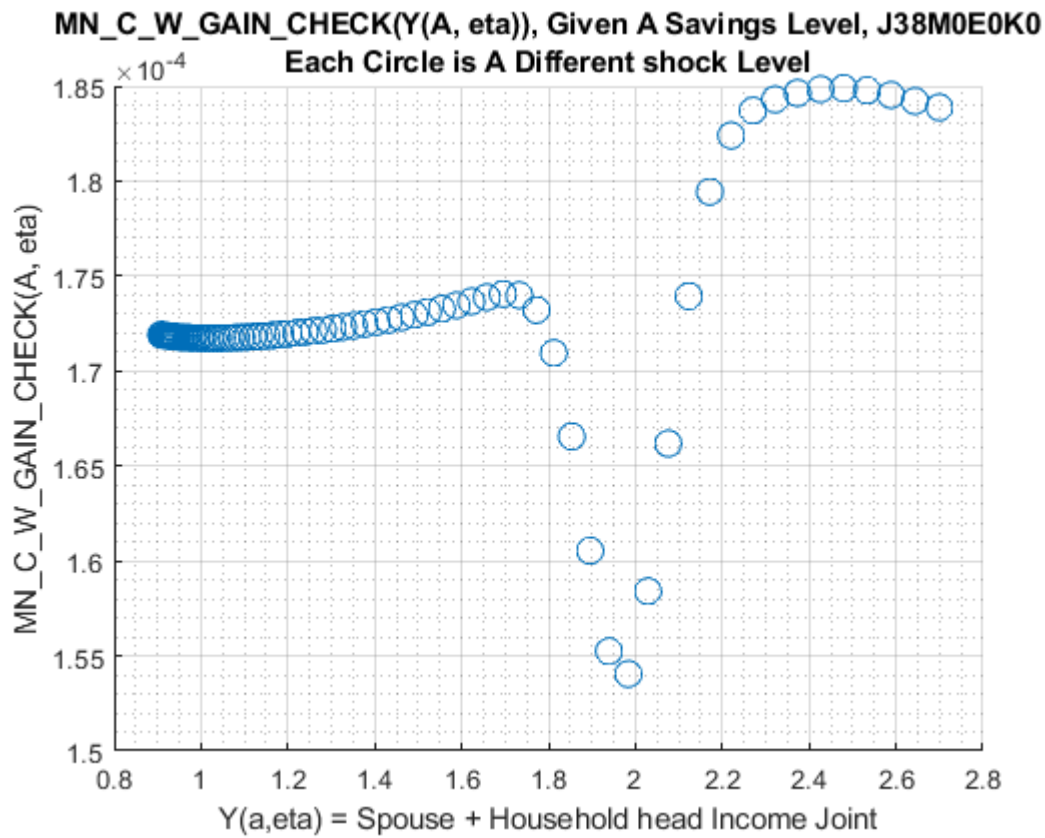
```
figure();
scatter((mt_total_inc_jemk(:)), log(mt_V_W_gain_check_jemk_07(:)), 100, mt_a(:));
title({'(MN\_V\_W\_GAIN\_CHECK(Y,eta)), All A (Savings) Levels, J38M0E0K0', ...
      'Color Represent different A Savings State, Circle-Group=Shock'});
xlabel('income(a,eps)');
ylabel('log of (MN\_V\_W\_GAIN\_CHECK(EM,J))');
xlim([0,7]);
grid on;
grid minor;
```



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

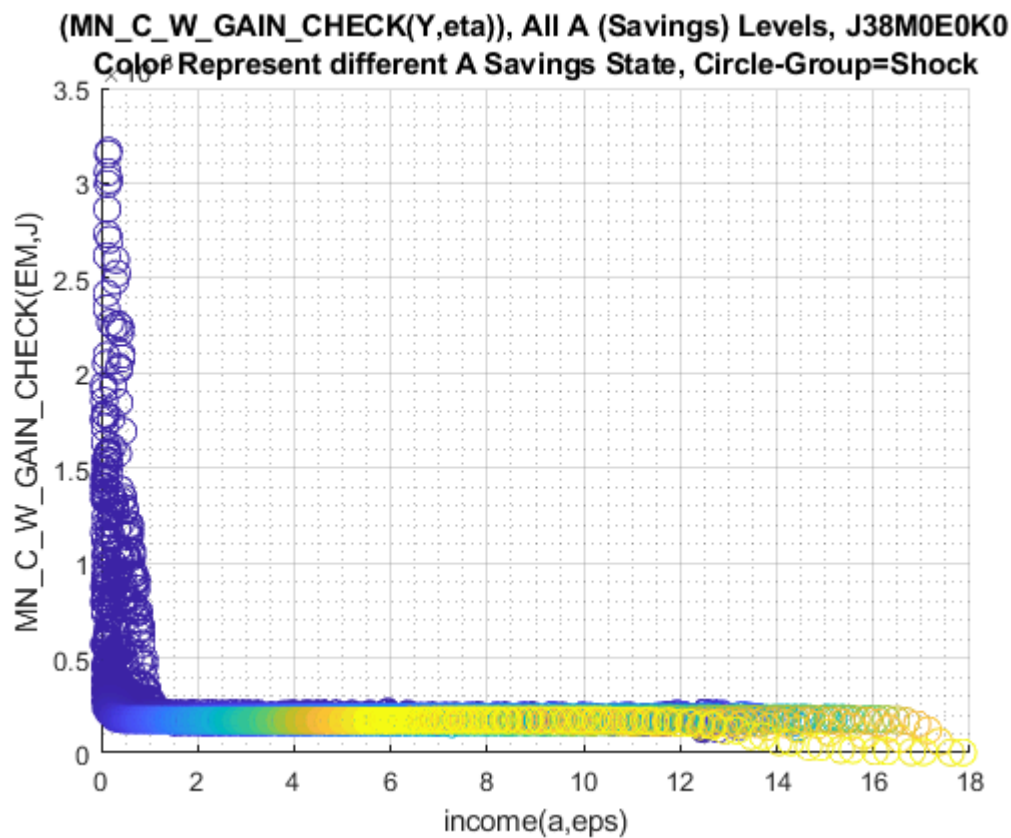
How do shocks and a impact marginal value. First plot one asset level, variation comes only from increasingly higher shocks:

```
figure();
it_a = 50;
scatter(log(mt_total_inc_jemk(:,it_a)), mt_C_W_gain_check_jemk_07(:,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;
```

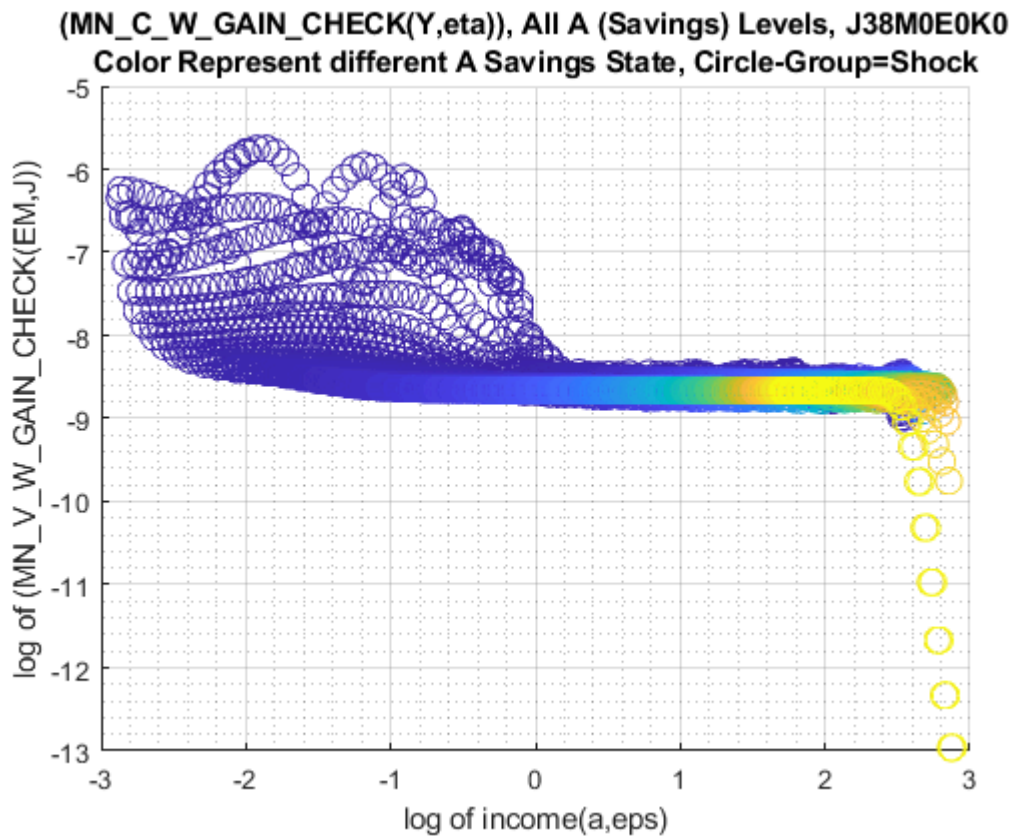


Plot all asset levels:

```
figure();
scatter((mt_total_inc_jemk(:)), (mt_C_W_gain_check_jemk_07(:)), 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_07(:)), 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_07, true, ["mean"], 3, 1, cl_mp_datasetc
```

```
xxx MEAN(MN_V_U_GAIN_CHECK(KM,J)), welf_checks=2, TR=0.0018238 xxxxxxxxxxxxxxxxxxxxxxxxxxxxxx
group kids marry mean_age_18 mean_age_19 mean_age_20 mean_age_21 mean_age_22 mean_age_23
1 1 0 0.031908 0.030377 0.028045 0.025708 0.023751 0.0221
2 2 0 0.043916 0.04186 0.038596 0.035292 0.032514 0.030161
3 3 0 0.051104 0.04891 0.044943 0.04117 0.038001 0.035317
4 4 0 0.057927 0.055555 0.051064 0.046812 0.043241 0.040216
5 5 0 0.06325 0.060842 0.055977 0.051401 0.047557 0.044303
6 1 1 0.005762 0.0053279 0.0048412 0.0043955 0.0040141 0.0036902
7 2 1 0.0081705 0.0075587 0.0068473 0.0061989 0.005658 0.0051977
8 3 1 0.0099025 0.0091765 0.0083275 0.0075396 0.0068765 0.0063167
9 4 1 0.012339 0.011465 0.01043 0.0094589 0.008631 0.0079256
10 5 1 0.015374 0.014388 0.013133 0.01196 0.010966 0.01013
```

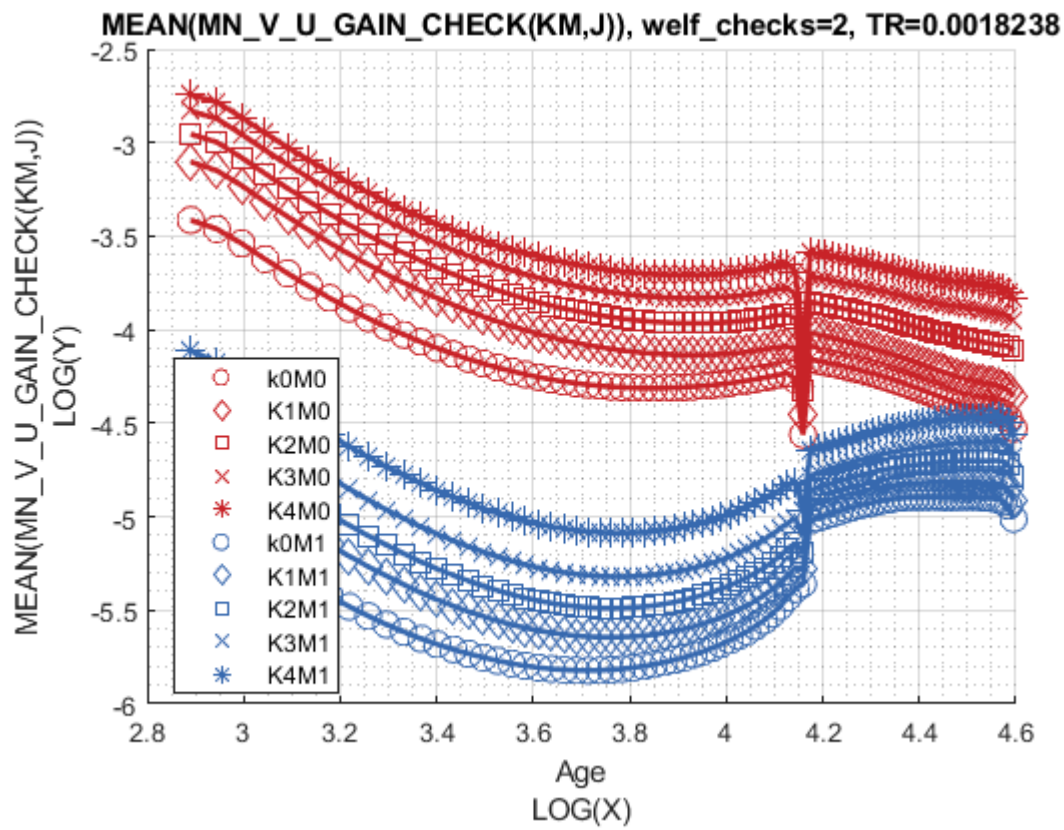
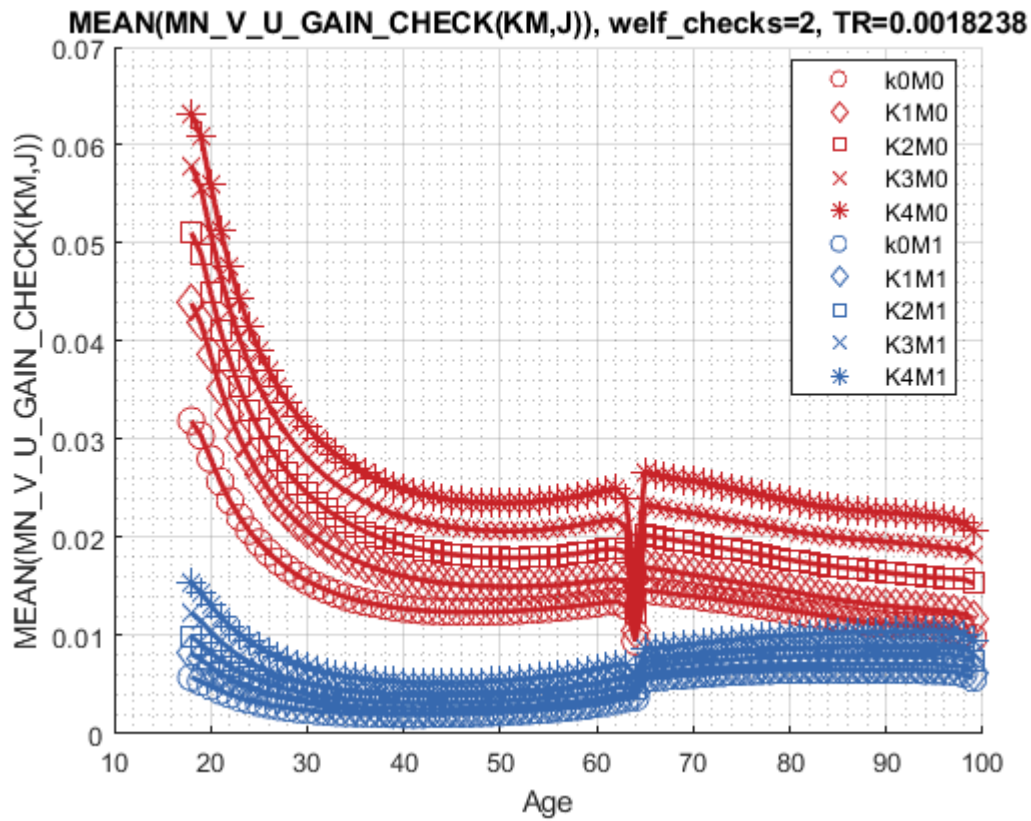
% Consumption Function

```
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_07, true, ["mean"], 3, 1, cl_mp
```

```
xxx MEAN(MN_MPC_U_GAIN_CHECK(KM,J)), welf_checks=2, TR=0.0018238 xxxxxxxxxxxxxxxxxxxxxxxxxxxxxx
group kids marry mean_age_18 mean_age_19 mean_age_20 mean_age_21 mean_age_22 mean_age_23
1 1 0 0.078069 0.086102 0.087377 0.085109 0.082559 0.080767
2 2 0 0.088091 0.096475 0.099705 0.096801 0.094582 0.092454
3 3 0 0.096309 0.10647 0.11111 0.10818 0.10451 0.10207
4 4 0 0.10137 0.11245 0.11806 0.11424 0.11063 0.1079
5 5 0 0.10569 0.11834 0.12441 0.11996 0.11559 0.11262
6 1 1 0.091639 0.097614 0.095796 0.09274 0.094336 0.09373
7 2 1 0.098891 0.10394 0.10413 0.10026 0.097592 0.097894
8 3 1 0.10228 0.10811 0.10782 0.10839 0.10652 0.10731
9 4 1 0.10459 0.11283 0.11208 0.11053 0.10913 0.10771
10 5 1 0.11468 0.12506 0.12335 0.12035 0.11807 0.11752
```

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);
```

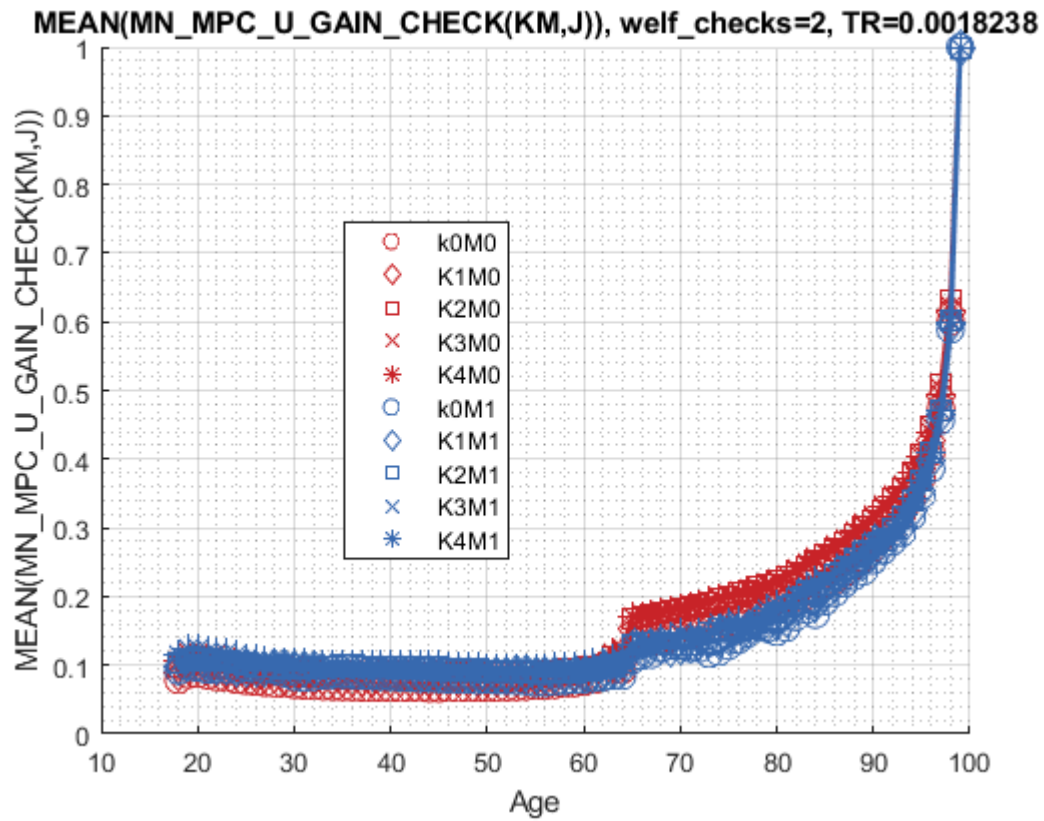


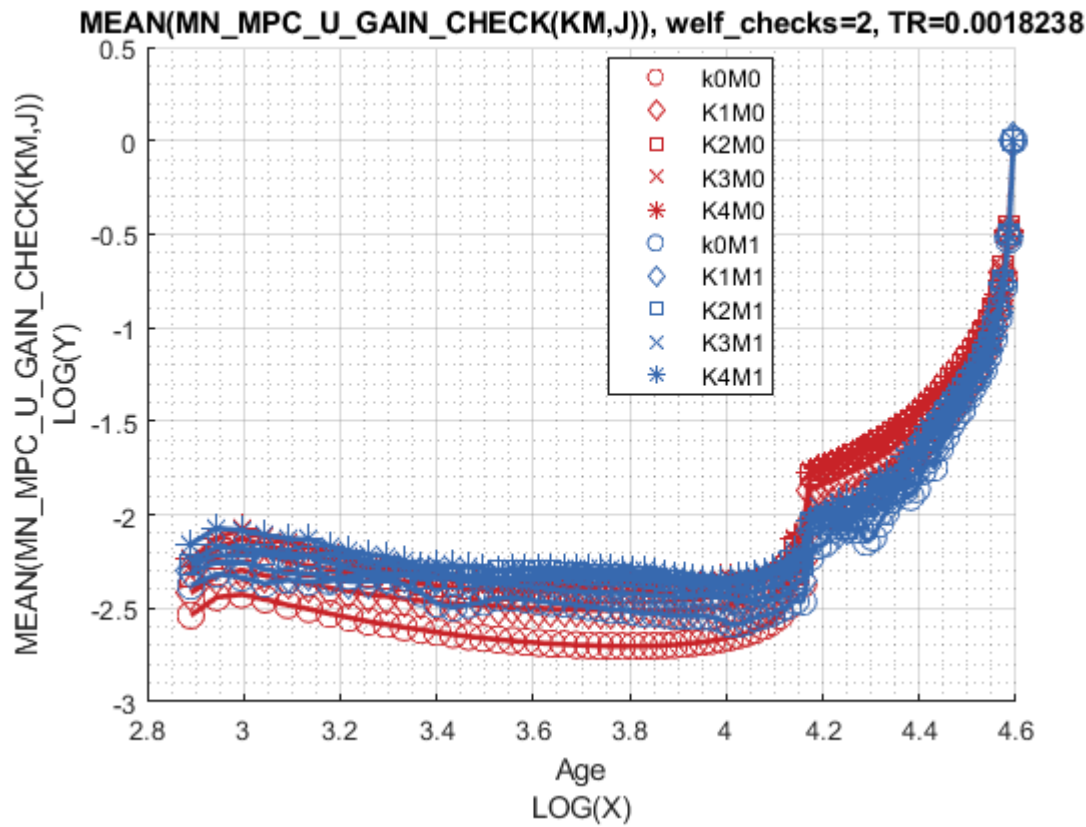
Graph Mean Consumption (**MPC: Share of Check Consumed**):

```

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_07, true, ["mean"], 3, 1, cl_mp_datasets);
```

```
xxx MEAN(MN_V_U_GAIN_CHECK(EM,J)), welf_checks=2, TR=0.0018238 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	0.051256	0.049485	0.046969	0.044331	0.041985	0.039891
2	1	0	0.047986	0.045532	0.040481	0.035822	0.032041	0.028948
3	0	1	0.01129	0.010552	0.0097415	0.0089796	0.0083154	0.0077436
4	1	1	0.0093296	0.0086141	0.0076904	0.0068414	0.0061429	0.0055604

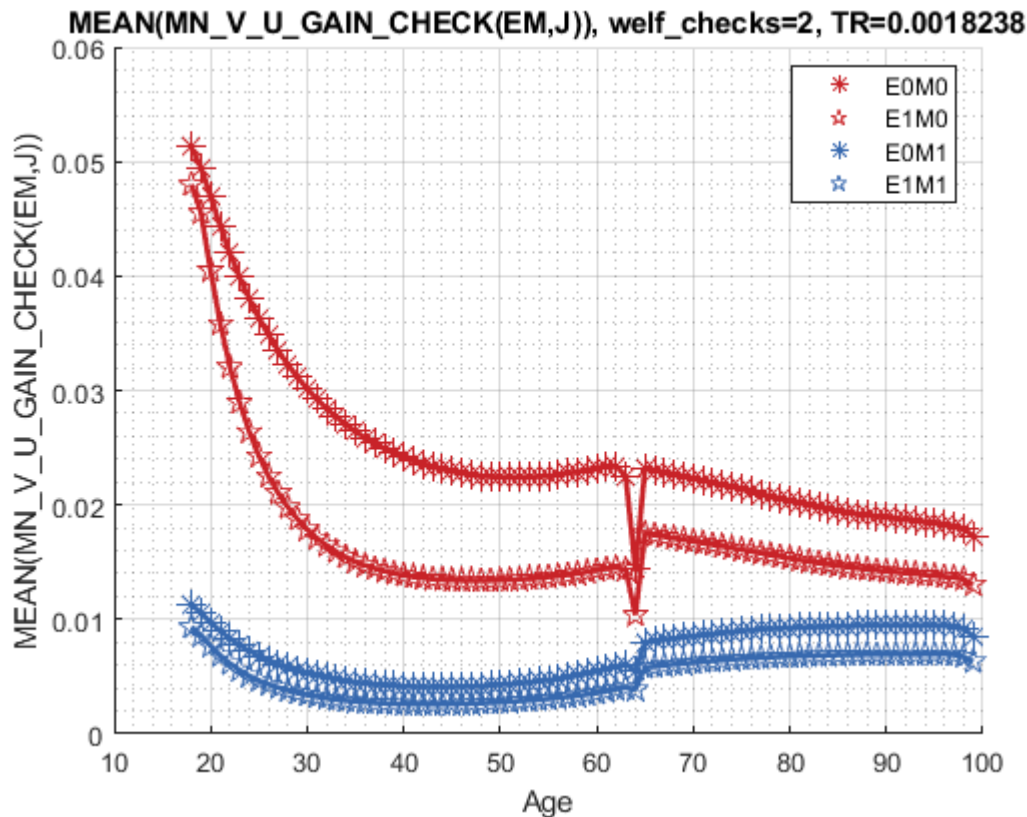
% Consumption

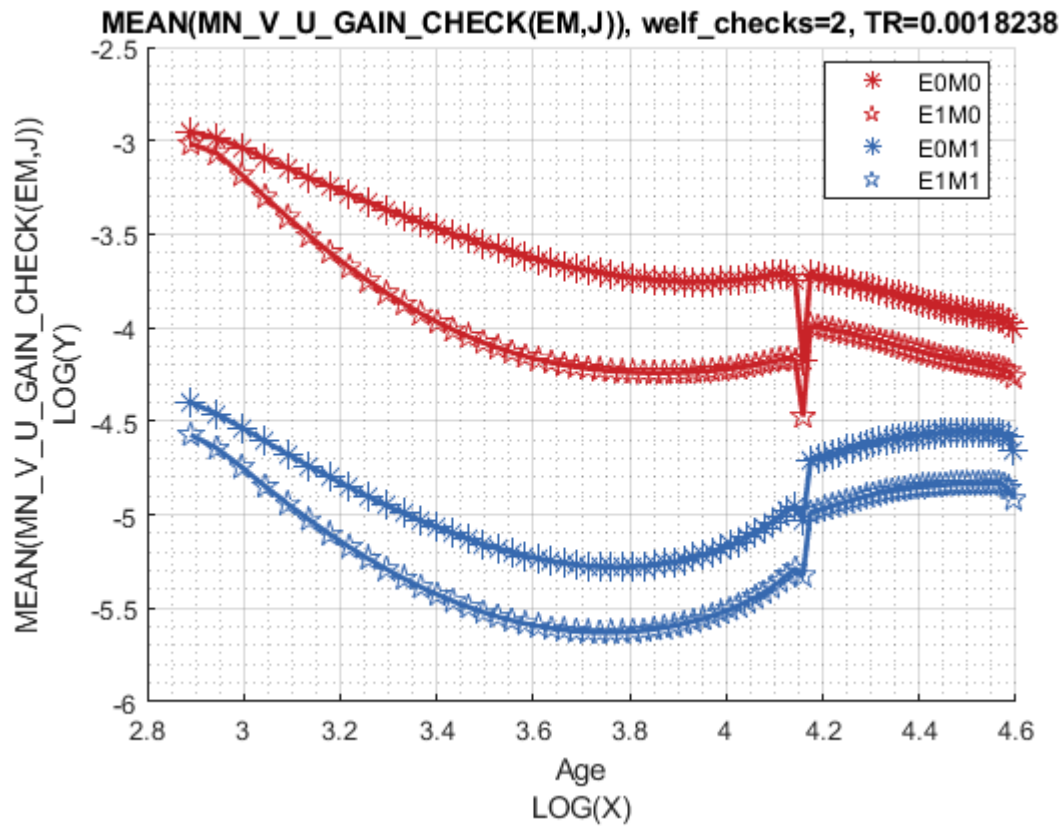
```
st_title = ['MEAN(MN_MPC_U_GAIN_CHECK(EM,J)), welf_checks=' num2str(welf_checks) ', TR=' num2str(TR)'];
tb_az_c = ff_summ_nd_array(st_title, mn_MPC_U_gain_share_check_07, true, ["mean"], 3, 1, cl_mp_07);
```

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	0.081794	0.085652	0.08723	0.08665	0.085947	0.085754
	2	1	0	0.10602	0.12228	0.12903	0.12307	0.1172	0.11257
	3	0	1	0.093649	0.097104	0.096702	0.095946	0.095996	0.096866
	4	1	1	0.11118	0.12192	0.12057	0.11696	0.11426	0.1128

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);
```

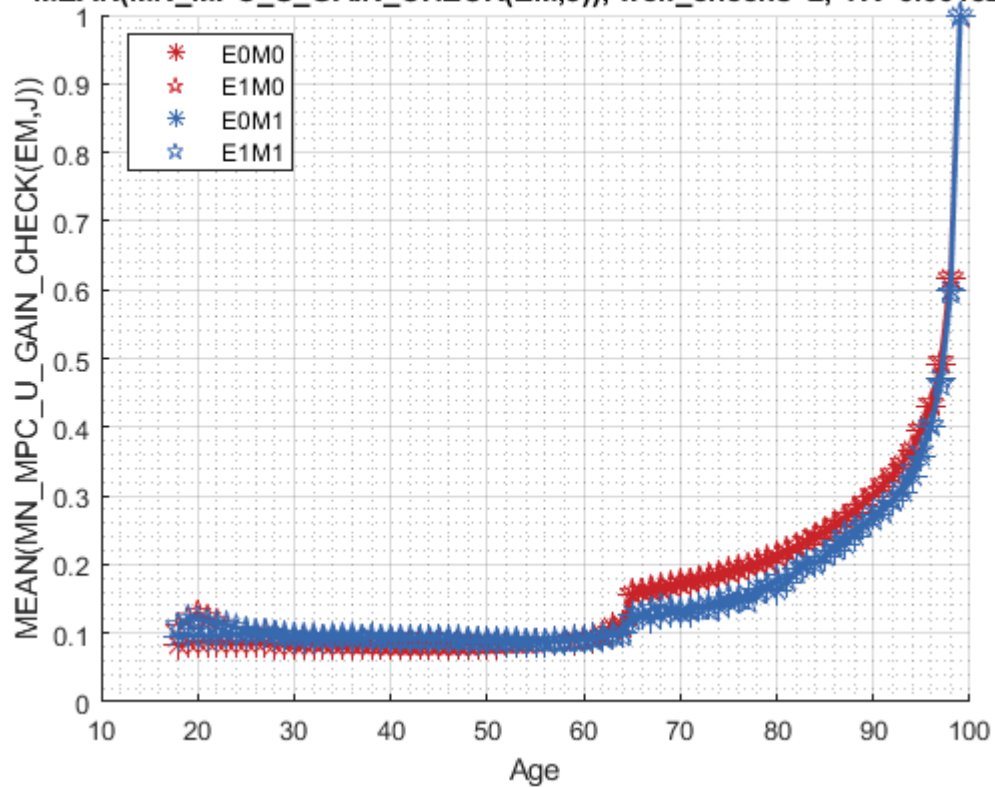




Graph Mean Consumption (**MPC: Share of Check Consumed**):

```
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=0.0018238



MEAN(MN_MPC_U_GAIN_CHECK(EM,J)), welf_checks=2, TR=0.0018238

