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 Dense

VFI and Distribution

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_controls = snw_mp_control('default_test');
% set Unemployment Related Variables
xi=0.5; % Proportional reduction in income due to unemployment (xi=0 refers to 0 labor income;
b=0; % Unemployment insurance replacement rate (b=0 refers to no UI benefits; b=1 refers to 100
TR=100/58056; % Value of a welfare check (can receive multiple checks). TO DO: Update with alte
mp_params('xi') = xi;
mp_params('b') = b;
mp_params('TR') = TR;
% Solve for Unemployment Values
mp_controls('bl_print_vfi') = false;
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_dense; SNW_MP_CONTROL=default_test; time=16.9824

```
inc_VFI = mp_valpol_more_ss('inc_VFI');
spouse_inc_VFI = mp_valpol_more_ss('spouse_inc_VFI');
total_inc_VFI = inc_VFI + spouse_inc_VFI;
% Solve unemployment
[V_unemp,~,cons_unemp,~] = 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_dense; SNW_MP_CONTROL=default_test; time=18

```
[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_dense; SNW_MP_CONTROL=default_test; time=61.7368

```
% 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, F
Wage quintile cutoffs=0.49295
                             0.79302
                                         1.3138
                                                    2.1063
Completed SNW HH PRECOMPUTE; SNW MP PARAM=default dense; SNW MP CONTROL=default test; time cost=35.3213
```

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 ss, cons ss, V unemp, cons unemp, mp precompute res);
```

Completed SNW_A4CHK_UNEMP_BISEC_VEC; welf_checks=0; TR=0.0017225; xi=0.5; b=0; SNW_MP_PARAM=default_dense; SNW_MP_CONTROL= Completed SNW_A4CHK_WRK_BISEC_VEC; welf_checks=0; TR=0.0017225; SNW_MP_PARAM=default_dense; SNW_MP_CONTROL=default_test; Completed SNW_EVUVW20_JAEEMK;SNW_MP_PARAM=default_dense;SNW_MP_CONTROL=default_test;timeEUEC=0.67039 Completed SNW_EVUVW19_JAEEMK;SNW_MP_PARAM=default_dense;SNW_MP_CONTROL=default_test;time=8.1727

r79

r80

r81

0.21707

0.21707

0.21707

0.21796

0.21796

0.21796

0.22416

0.22416

0.22416

	i	idx	ndim	numel	rowN	colN	sum	mea	an st	td co	efvari
	_										
_	jaeemk 1			8942e+06	82	23100	8.2855e+				.1971
_	jaeemk 2			9173e+06	83	23100	9.7703e+				.6435
_	jaeemk 3			8942e+06	82	23100	-3.7288e+				11.28
ev20_	jaeemk 4	4	6 1.	9173e+06	83	23100	-4.1377e+	-06 -2.1	L581 22.	.856 -10	0.591
xxx TABLE	:ec19_jaeem	k xxxxxxxx	xxxxxxxxx								
	c1	c2	c 3		c4	c 5	c23096	c23097	c23098	c23099	c2316
r1	0.08064	0.08074	4 0.0825	59 0.0	86958	0.092876	10.7	10.935	11.171	11.402	11.62
r2	0.081432	0.08143	2 0.0846	44 0.0	89174	0.095493	10.86	11.091	11.32	11.543	11.
r3	0.083622	0.08362	2 0.0855	0.0	91223	0.097903	11.024	11.249	11.471	11.683	11.86
r4	0.086619	0.08661	9 0.0885	0.0	94197	0.10085	11.168	11.388	11.601	11.801	11.96
r5	0.089528	0.08952	8 0.0914	31 0.0	97077	0.1037	11.309	11.521	11.724	11.908	12.0
r78	0.21707	0.2170	7 0.217	07 0.	21707	0.22416	26.837	28.052	29.222	30.511	32.0
r79	0.21707	0.2170	7 0.217	07 0.	21707	0.22416	28.992	31.165	32.888	34.236	35.60
r80	0.21707	0.2170	7 0.217	07 0.	21707	0.22416	32.266	33.961	36.121	38.357	40.6
r81	0.21707	0.2170	7 0.217	07 0.	21707	0.22416	38.348	39.931	42.54	46.29	49.2
r82	0.21707	0.2170	7 0.217	07 0.	21707	0.22361	51.027	52.913	57.047	61.392	64.1
xxx TABLE	:ec20_jaeem	k xxxxxxxx	xxxxxxxxx								
	c1	c2	c3		c4	c 5	c23096	c23097	c23098	c23099	c231
r1	0.078786	0.07937	5 0.0803	72 0.0	84318	0.090081	10.601	10.841	11.082	11.321	11.5
r2	0.078786	0.07967	4 0.0815	95 0.0	86622	0.09285	10.755	10.992	11.228	11.46	11.6
r3	0.079575	0.08046	3 0.0845	61 0.0	89582	0.096059	10.912	11.145	11.375	11.598	11.8
r4	0.082616	0.08350	5 0.0875	81 0.0	92591	0.099121	11.281	11.504	11.718	11.91	11.9
r5	0.085578	0.08646	7 0.0905	13 00	95505	0.10201	11.412	11.627	11.83	12.006	12.0

0.241

0.241

0.241

0.26692

0.26692

0.26692

34.627

39.694

47.978

36.056

41.355

51.293

37.896

43.554

53.451

42.95

49.23

58.48

40.382

46.378

55.657

r82 r83	0.21707 0.21707	0.21796 0.21796	0.22416 0.22416				.659 68.5 2.67 119.			79.5 141.
	·ov10 isoom	k xxxxxxxxx								
AAA TABEE	c1	c2	c3	c4	c 5	c23096	c23097	c23098	c23099	c23100
r1	-231.16	-231.1	-230.08	-227.9	-224.17	20.989	21.13	21.27	21.409	21.54
r2	-222.14	-222.14	-221.33	-219.39	-215.84	20.973	21.11	21.246	21.382	21.52
r3	-213.6	-213.6	-213.21	-211.33	-207.93	20.953	21.086	21.219	21.352	21.48
r4	-205.67	-205.67	-205.31	-203.55	-200.36	20.925	21.055	21.185	21.316	21.44
r5	-198.45	-198.45	-198.11	-196.45	-193.44	20.891	21.018	21.145	21.273	21.38
r78	-9.9698	-9.9698	-9.9698	-9.9698	-9.824	2.4708	2.4858	2.5001	2.5134	2.52
r79	-8.9313	-8.9313	-8.9313	-8.9313	-8.7855	2.2408	2.2553	2.2677	2.278	2.287
r80	-7.6669	-7.6669	-7.6669	-7.6669	-7.5211	1.9523	1.9632	1.9722	1.9796	1.986
r81	-5.9967	-5.9967	-5.9967	-5.9967	-5.8508	1.5534	1.5588	1.5654	1.5715	1.576
r82	-3.6113	-3.6113	-3.6113	-3.6113	-3.4767	0.9548	0.95648	0.95963	0.96249	0.9640
xxx TABLE	:ev20_jaeem c1	k xxxxxxxxxx c2	c3	c4	c 5	c23096	c23097	c23098	c23099	c2316
	CI	CZ	CS	C4	CS	C23090	C23097	C23098	C23033	(2316
r1	-235.08	-234.92	-233.84	-231.32	-227.14	21.095	21.236	21.376	21.516	21.6
r2	-226.34	-226.18	-225.14	-222.7	-218.72	21.082	21.219	21.356	21.492	21.6
r3	-217.94	-217.78	-216.77	-214.45	-210.67	21.065	21.198	21.331	21.464	21.
r4	-209.81	-209.67	-208.73	-206.55	-203.01	21.045	21.174	21.304	21.434	21.
r5	-202.41	-202.27	-201.4	-199.35	-196.01	21.014	21.14	21.267	21.394	21.
r79	-9.9634	-9.9447	-9.8178	-9.5061	-8.9993	2.5241	2.5372	2.5494	2.5607	2.5
r80	-8.9252	-8.9064	-8.7795	-8.4679	-7.9637	2.2926	2.3026	2.3118	2.3204	2.3
r81	-7.661	-7.6423	-7.5154	-7.2037	-6.7021	1.9992	2.0061	2.0125	2.0185	2.
r82	-5.9911	-5.9724	-5.8455	-5.5339	-5.0346	1.5919	1.5958	1.5994	1.6027	1.6
r83	-3.6067	-3.588	-3.4611	-3.1494	-2.6526	0.97826	0.97949	0.98063	0.98169	0.98

% Call Function

welf checks = 2;

r5

0.092085

[ev19_jaeemk_check2, ec19_jaeemk_check2, ev20_jaeemk_check2, ec20_jaeemk_check2] = snw_evuvw19_ welf_checks, st_solu_type, mp_params, mp_controls, ... V_ss, cons_ss, V_unemp, cons_unemp, mp_precompute_res);

Completed SNW_A4CHK_UNEMP_BISEC_VEC; welf_checks=2; TR=0.0017225; xi=0.5; b=0; SNW_MP_PARAM=default_dense; SNW_MP_CONTROL= Completed SNW_A4CHK_WRK_BISEC_VEC; welf_checks=2; TR=0.0017225; SNW_MP_PARAM=default_dense; SNW_MP_CONTROL=default_test; Completed SNW_EVUVW20_JAEEMK;SNW_MP_PARAM=default_dense;SNW_MP_CONTROL=default_test;timeEUEC=0.4338 Completed SNW_EVUVW19_JAEEMK;SNW_MP_PARAM=default_dense;SNW_MP_CONTROL=default_test;time=9.4255

rowN

colN

0.10419

sum

11.309

11.521

11.724

11.908

12.04

std

mean

coefvari

CONTAINER NAME: mp_outcomes ND Array (Matrix etc)

idx

0.092085

ndim

numel

0.093374

i

	_										
jaeemk	1	1	6	1.8942e+0	6 82	23100	8.2866e	+06 4	.3747	5.2364	1.197
jaeemk	2	2	6	1.9173e+0	6 83	23100	9.7714e	+06 5	.0964	8.3755	1.6434
jaeemk	3	3	6	1.8942e+0	6 82	23100	-3.7109e	+06 -1	.9591	22.18	-11.321
jaeemk	4	4	6	1.9173e+0	6 83	23100	-4.1188e	+06 -2	.1482	22.829	-10.627
c1		c2	С	3	c4	c 5	c23096	c23097	c2309	8 c2309	9 c2310
C1		CZ		,	C4	CJ	C23030	(23037	(230)	6 (230)	5 (2516
0.081703	3 0	.081806	0.	0836 0	.087675	0.09332	10.7	10.935	11.17	11.40	2 11.62
0.083719	9 6	.083719	0.08	5961 0	.089966	0.095955	10.86	11.091	11.3	2 11.54	3 11.7
0.083719 0.086203		0.083719 0.086201	0.08 0.08		.089966 .092085	0.095955 0.098395	10.86 11.024	11.091 11.25			
	c1	jaeemk 2 jaeemk 3 jaeemk 4 ::ec19_jaeemk xx c1	jaeemk 2 2 jaeemk 3 3 jaeemk 4 4 ::ec19_jaeemk xxxxxxxxxxxxxx c1 c2	jaeemk 2 2 6 jaeemk 3 3 6 jaeemk 4 4 6 ::ec19_jaeemk xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx	jaeemk 2 2 6 1.9173e+0 jaeemk 3 3 6 1.8942e+0 jaeemk 4 4 6 1.9173e+0 ::ec19_jaeemk xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx	jaeemk 2 2 6 1.9173e+06 83 jaeemk 3 3 6 1.8942e+06 82 jaeemk 4 4 6 1.9173e+06 83 ::ec19_jaeemk xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx	jaeemk 2 2 6 1.9173e+06 83 23100 jaeemk 3 3 6 1.8942e+06 82 23100 jaeemk 4 4 6 1.9173e+06 83 23100 ::ec19_jaeemk xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx	jaeemk 2 2 6 1.9173e+06 83 23100 9.7714e jaeemk 3 3 6 1.8942e+06 82 23100 -3.7109e jaeemk 4 4 6 1.9173e+06 83 23100 -4.1188e ::ec19_jaeemk xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx	jaeemk 2 2 6 1.9173e+06 83 23100 9.7714e+06 5 jaeemk 3 3 6 1.8942e+06 82 23100 -3.7109e+06 -1 jaeemk 4 4 6 1.9173e+06 83 23100 -4.1188e+06 -2 ::ec19_jaeemk xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx	jaeemk 2 2 6 1.9173e+06 83 23100 9.7714e+06 5.0964 jaeemk 3 3 6 1.8942e+06 82 23100 -3.7109e+06 -1.9591 jaeemk 4 4 6 1.9173e+06 83 23100 -4.1188e+06 -2.1482 E:ec19_jaeemk xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx	jaeemk 2 2 6 1.9173e+06 83 23100 9.7714e+06 5.0964 8.3755 jaeemk 3 3 6 1.8942e+06 82 23100 -3.7109e+06 -1.9591 22.18 jaeemk 4 4 6 1.9173e+06 83 23100 -4.1188e+06 -2.1482 22.829 E:ec19_jaeemk xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx

0.097931

8 0.22052	0.22052	0.22052				26.838				32.0
		0.22052	0.22052						38.358	
1 0.22052	0.22052	0.22052	0.22052	2 0.22	761	38.349	39.93	3 42.542	46.292	49.2
2 0.22052	0.22052	0.22052	0.22052	2 0.22	706	51.03	52.91	6 57.05	61.396	64.1
BLE:ec20_jaeem	(xxxxxxxxxx	(XXXXXXXX								
c1	c2	c3	c4	c5		c23096	c2309	7 c23098	c23099	c231
										11.5
										11.8
										12.0
										42.9
										58.
3 0.22052	0.2214	0.22/61	0.24444	4 0.27	122	112.67	119.4	3 126.45	133.75	141.
c1	c2	с3	c4	c 5	c236	996 c2	3097	c23098	c23099	c23100
										21.54
										21.51
										21.48
										21.44
										21.38
										2.52
										2.287
										1.986
										1.576
2 -3.5401	-3.5401	-3.5401	-3.5401	-3.4123	0.95	.48 0.	95648	0.95963	0.96249	0.9640
c1	c2	с3	c4	c 5	c236	996 c	23097	c23098	c23099	c2310
										21.6
		-224.64	-222.29					21.356		21.6
	-217.22		-214.05							21.5
										21.5
										21.5
9 -9.8923	-9.8742		-9.4529							2.57
0 -8.8541		-8.7157	-8.4149	-7.9203				2.3118	2.3204	2.32
	-7.5718									2.0
2 -5.9201	-5.9019	-5.7817	-5.4814	-4.9918				1.5994	1.6027	1.60
3 -3.5356	-3.5175	-3.3973	-3.0972	-2.6142	0.97	7826 0	.97949	0.98063	0.98169	0.982
9 6 1 2 E 8 9 6 1 2 E	0.22052 0.22052 0.22052 0.22052 0.22052 0.22052 0.079785 0.080464 0.082148 0.085181 0.088131 0.22052 0.20052 0.20052 0.20052 0.20052 0.20052 0.20052 0.20052 0.20052 0.20052 0.20052 0.20052 0.20052 0.20052 0.20052 0.20052 0	0.22052 0.22052 0.22052 0.22052 0.22052 0.22052 0.22052 0.22052 0.22052 0.22052 0.22052 0.22052 0.22052 0.22052 0.2052 0.22052 0.080464 0.080738 0.082148 0.082733 0.085181 0.085763 0.088131 0.088709 0.22052 0.2214 0.2052 0.2214 0.2052 0.2214 0.2052 0.2214 0.2052 0.2214 0.2052 0.2214 0.2052 0.201	0.22052	0.22052	0.079785	0. 0.22052	0.22052	9 0.22652 0.22652 0.22652 0.22652 0.22652 0.22761 28.993 31.16 10 0.22652 0.22652 0.22652 0.22652 0.22652 0.22761 32.267 33.96 11 0.22652 0.22652 0.22652 0.22652 0.22652 0.22766 51.03 32.96 12 0.22652 0.22652 0.22652 0.22652 0.22766 51.03 32.93 BLE:ec20_jaeemk xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx	0. 0.22052	09 0.22052 0.22052 0.22052 0.22052 0.22052 0.22052 0.22052 0.22053 0.22053 0.22053 0.22053 0.22052 0.20052 0.220052 0.

0.22052 0.22761

26.838

28.053

29.223

30.512

32.06

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(
```

Dense Param Results Define Frames

0.22052

0.22052

r78

0.22052

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
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'}, {'edu', edu_grid});
cl_mp_datasetdesc{6} = 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(metb_az_v = ff_summ_nd_array(st_title, mn_v_U_gain_check, true, ["mean"], 4, 1, cl_mp_datasetdesc
```

oup	savings 	mean_eta_1	mean_eta_2 	mean_eta_3 	mean_eta_4 	mean_eta_5 	mean_eta_6
1	0	0.16422	0.072711	0.044261	0.035185	0.031921	0.030628
2	0.00085734	0.16411	0.072685	0.044254	0.035183	0.031921	0.030628
3	0.0068587	0.16089	0.072203	0.044174	0.03516	0.031906	0.030615
4	0.023148	0.14347	0.068722	0.043233	0.034539	0.031319	0.030033
5	0.05487	0.12489	0.062983	0.040587	0.032437	0.029301	0.028027
6	0.10717	0.10956	0.056872	0.037152	0.029572	0.026538	0.025288
7	0.18519	0.093145	0.050122	0.032855	0.025873	0.022971	0.02175
8	0.29407	0.076208	0.042336	0.027832	0.021566	0.018844	0.017666
9	0.43896	0.063079	0.035973	0.023726	0.018152	0.015638	0.014514
.0	0.625	0.052632	0.030859	0.020414	0.01541	0.013081	0.012018
.1	0.85734	0.044007	0.026566	0.017638	0.013131	0.010971	0.0099637
L2	1.1411	0.036806	0.022912	0.015283	0.011232	0.0092306	0.0082784

13	1.4815	0.030786	0.019772	0.013268	0.0096429	0.0077925	0.0068951	0.0
14	1.8836	0.025765	0.017067	0.011541	0.0083128	0.006606	0.005763	0.0
15	2.3525	0.021575	0.014731	0.010054	0.0071926	0.0056241	0.0048349	0.0
16	2.8935	0.018077	0.012701	0.0087613	0.006238	0.0048017	0.0040654	0.00
17	3.5117	0.015151	0.010936	0.0076339	0.0054202	0.00411	0.0034259	0.00
18	4.2121	0.012711	0.0094052	0.0066501	0.0047191	0.0035306	0.0028969	0.00
19	5	0.010676	0.008083	0.0057903	0.0041163	0.0030444	0.0024584	0.0
20	5.8805	0.008979	0.0069435	0.0050432	0.003597	0.0026352	0.0020946	0.0
21	6.8587	0.0075642	0.0059636	0.0043935	0.0031477	0.002289	0.0017916	0.0
22	7.9398	0.0063835	0.0051226	0.003827	0.0027577	0.001995	0.0015384	0.0
23	9.1289	0.0053974	0.0044023	0.0033352	0.0024197	0.0017445	0.0013264	0.0
24	10.431	0.0045736	0.0037863	0.0029093	0.0021264	0.0015304	0.0011486	0.000
25	11.852	0.0038839	0.0032595	0.0025394	0.0018711	0.0013461	0.00099862	0.000
26	13.396	0.0033057	0.0028093	0.0022181	0.0016485	0.001187	0.00087181	0.000
27	15.069	0.0028204	0.0024245	0.0019392	0.0014539	0.0010493	0.00076433	0.000
28	16.875	0.0024123	0.0020957	0.0016969	0.0012837	0.00092975	0.00067284	0.000
29	18.82	0.0020683	0.0018143	0.0014864	0.0011352	0.00082552	0.00059445	0.000
30	20.91	0.0017776	0.0015732	0.0013032	0.0010051	0.00073426	0.00052692	0.000
31	23.148	0.0015314	0.0013663	0.0011438	0.00089056	0.00065417	0.00046852	0.000
32	25.541	0.0013225	0.0011887	0.001005	0.00079002	0.00058386	0.00041788	0.000
33	28.093	0.0011448	0.0010361	0.00088415	0.00070181	0.00052199	0.00037374	0.000
34	30.81	0.00099337	0.00090469	0.00077879	0.00062417	0.0004674	0.00033511	0.000
35	33.697	0.00086393	0.00079137	0.00068687	0.00055572	0.00041911	0.00030114	0.000
36	36.758	0.00075305	0.0006935	0.0006066	0.00049531	0.0003762	0.00027117	0.000
37	40	0.00065787	0.00060885	0.00053645	0.00044195	0.00033809	0.00024466	0.000
38	43.427	0.00057598	0.0005355	0.00047505	0.00039475	0.00030432	0.00022111	0.000
39	47.044	0.00050536	0.00047182	0.00042126	0.00035297	0.00027428	0.00020014	0.000
40	50.856	0.00044434	0.00041648	0.00037408	0.00031594	0.00024745	0.00018144	0.000
41	54.87	0.00039149	0.00036829	0.00033267	0.00028312	0.00022344	0.00016475	0.000
42	59.089	0.00034563	0.00032626	0.00029627	0.00025398	0.00020201	0.0001498	0.000
43	63.519	0.00030574	0.00028952	0.00026422	0.00022808	0.00018284	0.00013638	9.76
44	68.164	0.00027097	0.00025735	0.00023597	0.00020506	0.00016566	0.00012431	8.92
45	73.032	0.00024062	0.00022915	0.00021104	0.00018456	0.00015024	0.00011344	8.17
46	78.125	0.00021406	0.00020439	0.00018901	0.00016629	0.00013638	0.00010362	7.50
47	83.45	0.00019078	0.00018259	0.00016951	0.00015	0.00012392	9.4736e-05	6.90
48	89.011	0.00017032	0.00016338	0.00015223	0.00013545	0.00011269	8.6725e-05	6.30
49	94.815	0.00015233	0.00014643	0.00013691	0.00012244	0.00010258	7.9488e-05	5.879
50	100.87	0.00013647	0.00013144	0.00012329	0.00011081	9.3454e-05	7.2935e-05	5.448
51	107.17	0.00012246	0.00011817	0.00011118	0.0001004	8.5225e-05	6.6992e-05	5.06
52	113.73	0.00011007	0.0001064	0.0001004	9.1075e-05	7.78e-05	6.1598e-05	4.73
53	120.55	9.9117e-05	9.5971e-05	9.0813e-05	8.2743e-05	7.113e-05	5.6758e-05	4.448
54	127.64	8.9565e-05	8.6875e-05	8.2434e-05	7.5433e-05	6.5251e-05	5.2505e-05	4.21
55	135	8.5718e-05	8.3153e-05	7.8927e-05	7.2281e-05	6.2631e-05	5.0567e-05	4.09

% Consumption

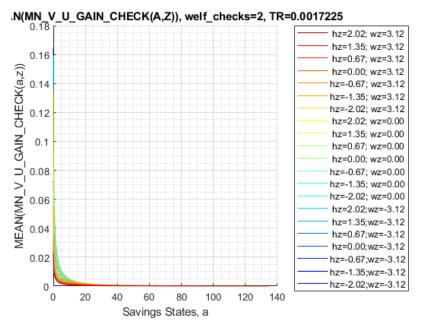
st_title = ['MEAN(MN_MPC_U_GAIN_CHECK(A,Z)), welf_checks=' num2str(welf_checks) ', TR=' num2str
tb_az_c = ff_summ_nd_array(st_title, mn_MPC_U_gain_share_check, true, ["mean"], 4, 1, cl_mp_data

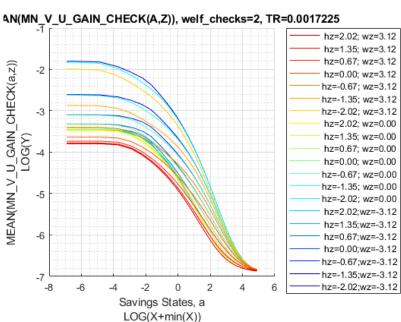
MEAN(N group	MN_MPC_U_GAIN_C savings	<pre>HECK(A,Z)), we mean_eta_1</pre>	elf_checks=2, T mean_eta_2	<pre>FR=0.0017225 > mean_eta_3</pre>	<pre></pre>	xxxxxxxxxxxx mean_eta_5	mean_eta_6	m
								-
1	0	0.74098	0.72618	0.66567	0.62344	0.58432	0.53574	
2	0.00085734	0.7405	0.72615	0.66568	0.62348	0.58437	0.53579	
3	0.0068587	0.72669	0.71733	0.66211	0.62125	0.58261	0.53437	
4	0.023148	0.61478	0.61585	0.59388	0.58469	0.55844	0.51216	
5	0.05487	0.52118	0.52002	0.51447	0.51652	0.50302	0.46639	
6	0.10717	0.4677	0.46139	0.46307	0.46985	0.45963	0.43084	
7	0.18519	0.3973	0.37778	0.38896	0.39697	0.38775	0.36486	
8	0.29407	0.32167	0.31637	0.29863	0.31313	0.31844	0.30063	
9	0.43896	0.25268	0.24569	0.2474	0.24208	0.24978	0.24274	
10	0.625	0.20528	0.20521	0.20529	0.20426	0.20225	0.20456	
11	0.85734	0.1787	0.17548	0.175	0.17606	0.17597	0.17577	
12	1.1411	0.16031	0.1574	0.15581	0.15574	0.15674	0.15888	

13	1.4815	0.14645	0.14407	0.14238	0.14223	0.14334	0.146	0.3
14	1.8836	0.13586	0.1339	0.13226	0.13233	0.13318	0.13504	0.3
15	2.3525	0.12905	0.12748	0.12583	0.12631	0.12672	0.12812	0
16	2.8935	0.1248	0.12332	0.12186	0.12224	0.12237	0.12327	0.3
17	3.5117	0.12316	0.12217	0.12089	0.121	0.12082	0.12193	0.3
18	4.2121	0.1215	0.12046	0.11937	0.11949	0.11943	0.12071	0.3
19	5	0.11643	0.11587	0.1151	0.1147	0.11496	0.11585	0.3
20	5.8805	0.11287	0.11225	0.11191	0.11124	0.11141	0.11229	0.3
21	6.8587	0.112	0.11157	0.11084	0.11056	0.11096	0.11147	0.3
22	7.9398	0.11185	0.11153	0.11102	0.11039	0.11115	0.11162	0
23	9.1289	0.11312	0.1128	0.11262	0.11186	0.11264	0.11291	0.3
24	10.431	0.1119	0.11162	0.11143	0.11076	0.11149	0.11153	0.3
25	11.852	0.10944	0.10926	0.10903	0.10841	0.10905	0.10904	0.3
26	13.396	0.10862	0.10843	0.10826	0.10771	0.10818	0.10845	0.3
27	15.069	0.10932	0.10916	0.109	0.1085	0.10883	0.10922	0.3
28	16.875	0.1099	0.10978	0.10963	0.10944	0.10935	0.10984	0.3
29	18.82	0.10798	0.10788	0.10777	0.10777	0.10737	0.10794	0
30	20.91	0.10705	0.10696	0.10687	0.10661	0.10637	0.10718	0.3
31	23.148	0.10761	0.10755	0.10746	0.10724	0.10705	0.10788	0.3
32	25.541	0.1096	0.10954	0.10947	0.1095	0.10909	0.10997	0.3
33	28.093	0.11004	0.10999	0.10994	0.11	0.10956	0.1104	0.3
34	30.81	0.10764	0.1076	0.10756	0.10759	0.10721	0.10793	0.3
35	33.697	0.10612	0.10609	0.10605	0.10605	0.10566	0.10632	0.3
36	36.758	0.10599	0.10597	0.10593	0.10592	0.10557	0.10611	0
37	40	0.10721	0.1072	0.10716	0.10717	0.10715	0.10722	0.3
38	43.427	0.10865	0.10862	0.10859	0.10859	0.10875	0.10846	0.3
39	47.044	0.10816	0.10813	0.1081	0.10808	0.1081	0.10787	0.3
40	50.856	0.10782	0.10782	0.10778	0.10777	0.10749	0.10758	0.3
41	54.87	0.10726	0.10727	0.10724	0.10722	0.10708	0.10705	0
42	59.089	0.10707	0.10706	0.10706	0.10703	0.10722	0.10686	0.3
43	63.519	0.10759	0.10756	0.10757	0.10755	0.10766	0.10735	0.3
44	68.164	0.1073	0.10729	0.10727	0.10728	0.10734	0.10703	0.3
45	73.032	0.10665	0.10665	0.10662	0.10663	0.10666	0.10632	0.3
46	78.125	0.10663	0.10662	0.10661	0.1066	0.10665	0.10617	0.3
47	83.45	0.10789	0.10788	0.10787	0.10785	0.10789	0.10772	0.3
48	89.011	0.10873	0.10873	0.10872	0.10871	0.10873	0.10887	0.1
49	94.815	0.10779	0.10779	0.10779	0.10778	0.10777	0.10783	0.1
50	100.87	0.10679	0.10679	0.10678	0.10677	0.10675	0.10663	0.3
51	107.17	0.10631	0.1063	0.10629	0.10627	0.10623	0.10561	0.09
52	113.73	0.10697	0.10696	0.10694	0.10689	0.10681	0.10606	0.09
53	120.55	0.1072	0.10716	0.10708	0.10697	0.10673	0.10588	0.09
54	127.64	0.10698	0.10688	0.10673	0.10652	0.10612	0.10498	0.09
55	135	0.10607	0.10591	0.10572	0.10546	0.10504	0.10393	0.08

Graph Mean Values:

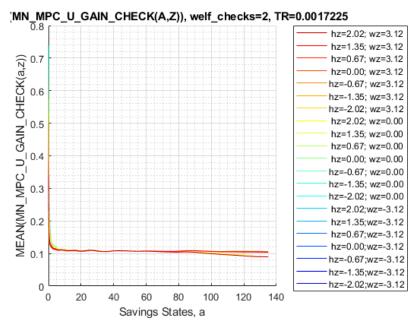
```
st_title = ['MEAN(MN\_V\_U\_GAIN\_CHECK(A,Z)), welf\_checks=' num2str(welf_checks) ', TR='
mp_support_graph('cl_st_graph_title') = {st_title};
mp_support_graph('cl_st_ytitle') = {'MEAN(MN\_V\_U\_GAIN\_CHECK(a,z))'};
ff_graph_grid((tb_az_v{1:end, 3:end}))', ar_st_eta_HS_grid, agrid, mp_support_graph);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

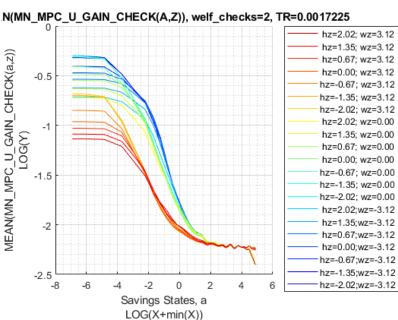




Graph Mean Consumption (*MPC: Share of Check Consumed*):

```
st_title = ['MEAN(MN\_MPC\_U\_GAIN\_CHECK(A,Z)), welf\_checks=' num2str(welf_checks) ', TR=' nump_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);
```





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_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 = 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

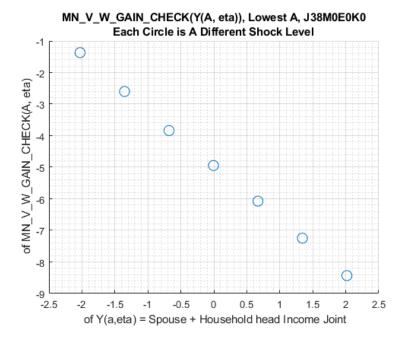
```
% Selections
it age = 21; % +18
it_marital = 1; % 1 = unmarried
it kids = 1; % 1 = kids is zero
it educ = 1; % 1 = lower education
% Select: NaN(n_jgrid,n_agrid,n_etagrid,n_educgrid,n_marriedgrid,n_kidsgrid);
mn C W gain check jemk = mn C W gain check use(it age, :, 1:mp params('n eta H grid'), it 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;</pre>
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;</pre>
% 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;
```

MN_V_W_GAIN_CHECK(Y(A, eta)), Lowest A, J38M0E0K0 Each Circle is A Different Shock Level 0.25 (gg V) 0.25 0.15 0.05 0.05 1 2 3 4 5 6 7 8 Y(a,eta) = Spouse + Household head Income Joint

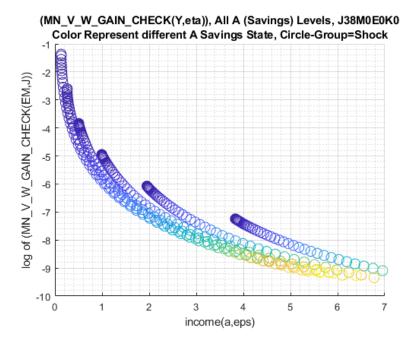


Plot all asset levels:

```
figure();
scatter((mt_total_inc_jemk(:)), (mt_V_W_gain_check_jemk(:)), 100, mt_a(:));
```



```
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;
```



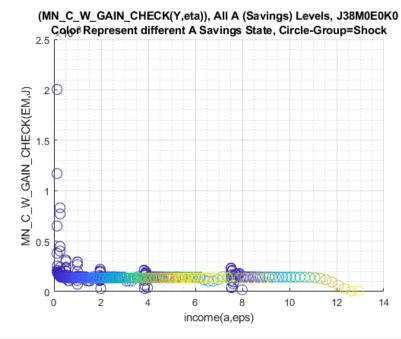
Marginal Consumption Gains, Color as Shock, Conditional on Age, Marital, Kids, and Education

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



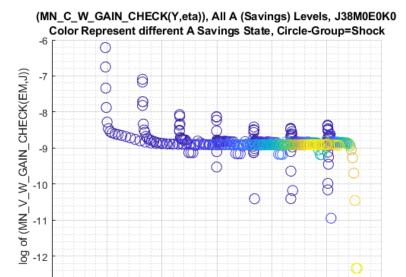
Plot all asset levels:

```
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;
```



log of income(a,eps)

Analyze Kids and Marriage and Age

-2

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

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

Tabulate value and policies:

-13

-3

```
% 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_datasetdesc

group	kids	marry	mean_age_18	mean_age_19	mean_age_20	mean_age_21	mean_age_22	mean_age
			0.040004	0.047300	0.016444	0.045475	0.01414	0.043
1	1	0	0.018081	0.017399	0.016444	0.015175	0.01411	0.013
2	2	0	0.024198	0.023312	0.022022	0.020238	0.018732	0.017
3	3	0	0.027972	0.027119	0.025477	0.023434	0.02171	0.020
4	4	0	0.031555	0.030668	0.028763	0.026469	0.024532	0.022
5	5	0	0.03448	0.033614	0.031423	0.028959	0.026879	0.025
6	1	1	0.0041769	0.0038579	0.0035234	0.0032138	0.0029477	0.0027
7	2	1	0.0057601	0.0053292	0.004867	0.0044266	0.0040531	0.003
8	3	1	0.0068486	0.0063558	0.0058113	0.0052918	0.0048508	0.0044
9	4	1	0.0083574	0.007782	0.0071369	0.006497	0.0059602	0.0054
10	5	1	0.010213	0.0095931	0.0088366	0.0080803	0.0074291	0.0068

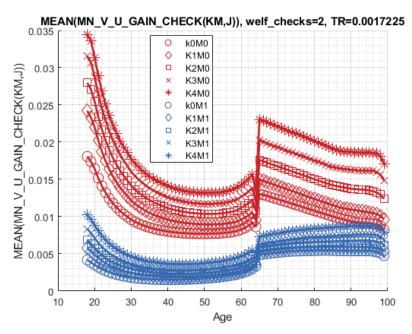
% Consumption Function

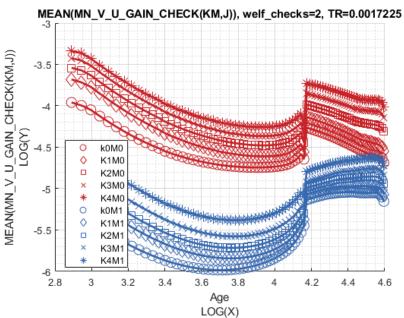
st_title = ['MEAN(MN_MPC_U_GAIN_CHECK(KM,J)), welf_checks=' num2str(welf_checks) ', TR=' num2st
tb_az_c = ff_summ_nd_array(st_title, mn_MPC_U_gain_share_check, true, ["mean"], 3, 1, cl_mp_dat

xxx MEAN(N	MN_MPC_U_	_GAIN_CHEC	CK(KM,J)), welf	_checks=2, TR=0.	.0017225 xxxxxx	xxxxxxxxxxxxx	xxxxxx	
group	kids	marry	mean_age_18	mean_age_19	mean_age_20	mean_age_21	mean_age_22	mean_age_23
1	1	0	0.074031	0.10225	0.11599	0.11147	0.10577	0.095081
2	2	0	0.089094	0.11803	0.13543	0.132	0.12978	0.12523
3	3	0	0.10426	0.14094	0.15528	0.1534	0.15067	0.1437
4	4	0	0.1102	0.15019	0.16182	0.16053	0.1588	0.15606
5	5	0	0.12088	0.15838	0.16724	0.16598	0.16431	0.16225
6	1	1	0.088885	0.10829	0.11259	0.10847	0.10521	0.10685
7	2	1	0.093837	0.11495	0.1194	0.11493	0.11003	0.1082
8	3	1	0.10258	0.126	0.1301	0.12812	0.12491	0.1238
9	4	1	0.10896	0.1303	0.1351	0.13564	0.13402	0.1313
10	5	1	0.12962	0.14112	0.14504	0.14418	0.14327	0.14158

Graph Mean Values:

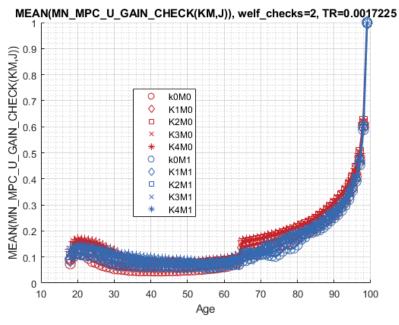
```
st_title = ['MEAN(MN\_V\_U\_GAIN\_CHECK(KM,J)), welf\_checks=' num2str(welf_checks) ', TR=' num
mp_support_graph('cl_st_graph_title') = {st_title};
mp_support_graph('cl_st_ytitle') = {'MEAN(MN\_V\_U\_GAIN\_CHECK(KM,J))'};
ff_graph_grid((tb_az_v{1:end, 4:end}), ar_row_grid, age_grid, mp_support_graph);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

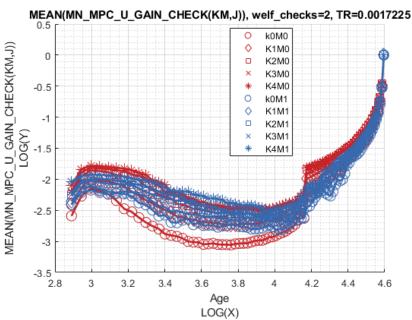




Graph Mean Consumption (MPC: Share of Check Consumed):

```
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(
tb_az_v = ff_summ_nd_array(st_title, mn_v_u_gain_check, true, ["mean"], 3, 1, cl_mp_datasetdeso
group
         edu
               marry
                      mean_age_18
                                 mean_age_19
                                             mean_age_20
                                                         mean_age_21
                                                                    mean_age_22
                                                                                mean_age_23
                       0.028598
                                               0.026816
                                                           0.02537
                                                                      0.024081
                                                                                 0.022927
    1
          0
                0
                                   0.027879
    2
          1
                0
                       0.025917
                                   0.024966
                                               0.022835
                                                           0.02034
                                                                      0.018305
                                                                                  0.01664
    3
          0
                1
                       0.0077546
                                  0.0072639
                                              0.0067503
                                                          0.0062497
                                                                     0.0058142
                                                                                 0.0054304
    4
          1
                1
                       0.0063878
                                  0.0059033
                                              0.0053197
                                                         0.0047542
                                                                     0.0042822
                                                                                 0.0038935
% Consumption
st_title = ['MEAN(MN_MPC_U_GAIN_CHECK(EM,J)), welf_checks=' num2str(welf_checks) ', TR=' num2st
tb_az_c = ff_summ_nd_array(st_title, mn_MPC_U_gain_share_check, true, ["mean"], 3, 1, cl_mp_date
edu
               marry
                      mean_age_18
                                 mean_age_19
   group
                                             mean_age_20
                                                         mean_age_21
                                                                    mean_age_22
                                                                                mean_age_23
```

Graph Mean Values:

1

2

3

4

0

1

0

0

1

1

0.083839

0.096421

0.11555

0.11313

```
st_title = ['MEAN(MN\_V\_U\_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\_U\_GAIN\_CHECK(EM,J))'};
ff_graph_grid((tb_az_v{1:end, 4:end}), ar_row_grid, age_grid, mp_support_graph);
```

0.12214

0.17216

0.11688

0.14001

0.11934

0.17002

0.11343

0.13911

0.11723

0.1665

0.11043

0.13655

0.11285

0.16008

0.11038

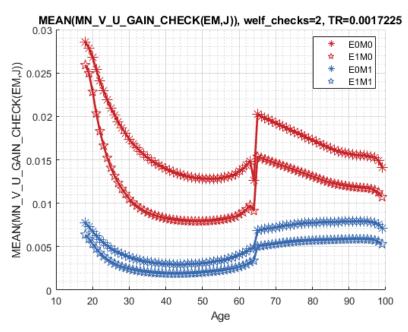
0.13431

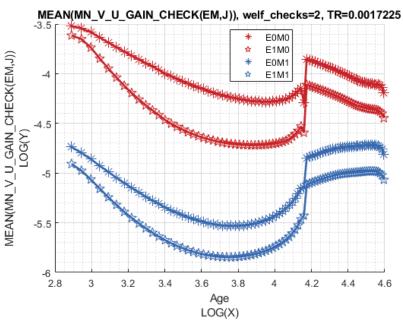
0.10872

0.1592

0.11218

0.13609





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);
```

