2008 Value and Optimal Savings and Consumption Given Stimulus

This is the example vignette for function: snw_v08_jaeemk from the PrjOptiSNW Package. This is simialr to snw_evuvw20_jaeemk, but for the 2008 Bush stimulus. snw_v08p08_jaeemk already solved for optimal policy and value functions in 2008, given expected unemployment shock in 2009. In this function, given some stimulus amount, we use snw_a4chk_wrk_bisec_vec to compute the updated optimal V and C in 2008 given the stimulus amount, based on the alues for V and C without stimulus computed by snw_v08p08_jaeemk.

Note that snw_a4chk_wrk_bisec_vec computes the adjustment in the savings state that would be equivalent to the increase in stimulus amount (which is not a state variable) to current resources, this is faster than resolving 2008 optimal V and C at specific stimulus check amount levels.

Note snw_evuvw20_jaeemk has EVUVW, but here, we only have V08, because in the 2020 problem, households receive checks ex-post of the COVID MIT shocks in 2020 and the EVUVW is the weighted average in V between the MIT unemployed and non-shock employed state. In 2008, however, there are no shocks yet, the state-space is the same as normal, the only difference is that households might receive stimulus checks from Bush. The Bush stimulus is provided ex-ante of the shock realization. The 2009 shocks due to the great recession is not a MIT shock, but expected shock. The effect of the 2009 shock on consumption, savings is solved by snw_v08p08_jaeemk. The expectation over shock, in another word, for the snw_v08_jaeemk is already included in EV' in 2008 for 2009.

Solve 2008 Value and Policy Function with SNW_V08p08_JAEEMK

Solve for the Value and Policy functions in 2008 given expected unemployment shock that is specific to age and education group in 2009, no stimulus amounts.

First, set various parameters

```
% 1. Paramters
% Parameters
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('beta') = 0.95;
% Control parameters
mp_controls = snw_mp_control('default_test');
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 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') = true;
```

```
Second, solve the steady-state problem, same as employed results in 2009.
 % 2. Solve value steady state (2009 employed)
 [V_VFI_ss, ap_VFI_ss, cons_VFI_ss, mp_valpol_more_ss] = snw_vfi_main_bisec_vec(mp_params, mp_co
 Completed SNW VFI MAIN BISEC VEC; SNW MP PARAM=default docdense; SNW MP CONTROL=default test; time=530.633
 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;
 V_{emp}_{2009} = V_{VFI}_{ss};
 % Solve for probability mass, needed for pre-compute
 [Phi_true] = snw_ds_main_vec(mp_params, mp_controls, ap_VFI_ss, cons_VFI_ss, mp_valpol_more_ss)
 Completed SNW_DS_MAIN_VEC;SNW_MP_PARAM=default_docdense;SNW_MP_CONTROL=default_test;time=1271.649
 % Solve for household head and spouse income and sum to total income.
 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;
Third, solve the unemployment problem in 2009. With 2009 specific unemployment parameters calibrated and
found from data. Using b calibrated by snw calibrate 2009 b.
 % 3. Solve value unemployed 2009
 % Set Unemployment Related Variables
 mp_params('xi') = 0.532;
 % Calibrated by snw_calibrate_2009_b
 mp params('b') = 0.37992;
 mp_params('a2_covidyr') = mp_params('a2_greatrecession_2009');
 mp_params('TR') = 100/fl_p50_hh_income_07; % Value of a stimulus check (can receive multiple check)
 [V unemp 2009] = snw vfi main bisec vec(mp params, mp controls, V VFI ss);
```

Completed SNW VFI MAIN BISEC VEC 1 Period Unemp Shock; SNW MP PARAM=default docdense; SNW MP CONTROL=default test; time

Fourth, solve the 2008 problem, with 2008-specific value and policy functions.

```
% 4. Value and Optimal choice in 2009
[V_2008, ap_2008, cons_2008, ev_empshk_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 Completed SNW_V08P08_JAEEMK;SNW_MP_PARAM=default_docdense;SNW_MP_CONTROL=default_test;time=522.6257

```
% 5. pre-compute
cl_st_precompute_list = {'a', ...
    'inc', 'inc_unemp', 'spouse_inc', 'spouse_inc_unemp', 'ref_earn_wageind_grid'};
mp_controls('bl_print_precompute_verbose') = false;
[mp precompute res] = snw hh precompute(mp params, mp controls, cl st precompute list, ap VFI s
```

Wage quintile cutoffs=0.4645 0.71528 SNW HH PRECOMPUTE: Finished Age Group: 1 of 82, time-this-age: 3.2746

```
SNW_HH_PRECOMPUTE: Finished Age Group:2 of 82, time-this-age:3.2852
SNW HH PRECOMPUTE: Finished Age Group:3 of 82, time-this-age:3.3869
SNW_HH_PRECOMPUTE: Finished Age Group:4 of 82, time-this-age:3.4589
SNW_HH_PRECOMPUTE: Finished Age Group:5 of 82, time-this-age:3.254
SNW_HH_PRECOMPUTE: Finished Age Group:6 of 82, time-this-age:3.3465
SNW HH PRECOMPUTE: Finished Age Group: 7 of 82, time-this-age: 3.4109
SNW_HH_PRECOMPUTE: Finished Age Group:8 of 82, time-this-age:3.4221
SNW_HH_PRECOMPUTE: Finished Age Group:9 of 82, time-this-age:3.4152
SNW HH PRECOMPUTE: Finished Age Group: 10 of 82, time-this-age: 3.4065
SNW_HH_PRECOMPUTE: Finished Age Group:11 of 82, time-this-age:3.1869
SNW_HH_PRECOMPUTE: Finished Age Group:12 of 82, time-this-age:3.3572
SNW_HH_PRECOMPUTE: Finished Age Group:13 of 82, time-this-age:3.4748
SNW_HH_PRECOMPUTE: Finished Age Group:14 of 82, time-this-age:3.3824
SNW HH PRECOMPUTE: Finished Age Group: 15 of 82, time-this-age: 3.4573
SNW_HH_PRECOMPUTE: Finished Age Group:16 of 82, time-this-age:3.3233
SNW HH PRECOMPUTE: Finished Age Group:17 of 82, time-this-age:3.3965
SNW HH PRECOMPUTE: Finished Age Group:18 of 82, time-this-age:3.3613
SNW HH PRECOMPUTE: Finished Age Group:19 of 82, time-this-age:3.3402
SNW HH PRECOMPUTE: Finished Age Group: 20 of 82, time-this-age: 3.479
SNW_HH_PRECOMPUTE: Finished Age Group:21 of 82, time-this-age:3.2899
SNW_HH_PRECOMPUTE: Finished Age Group:22 of 82, time-this-age:3.3222
SNW_HH_PRECOMPUTE: Finished Age Group:23 of 82, time-this-age:3.3025
SNW_HH_PRECOMPUTE: Finished Age Group: 24 of 82, time-this-age: 3.3682
SNW_HH_PRECOMPUTE: Finished Age Group:25 of 82, time-this-age:3.3555
SNW HH PRECOMPUTE: Finished Age Group:26 of 82, time-this-age:3.3701
SNW_HH_PRECOMPUTE: Finished Age Group:27 of 82, time-this-age:3.5453
SNW_HH_PRECOMPUTE: Finished Age Group:28 of 82, time-this-age:3.2218
SNW_HH_PRECOMPUTE: Finished Age Group:29 of 82, time-this-age:3.3947
SNW HH PRECOMPUTE: Finished Age Group: 30 of 82, time-this-age: 3.3024
SNW_HH_PRECOMPUTE: Finished Age Group:31 of 82, time-this-age:3.3699
SNW_HH_PRECOMPUTE: Finished Age Group:32 of 82, time-this-age:3.4931
SNW_HH_PRECOMPUTE: Finished Age Group:33 of 82, time-this-age:3.3584
SNW_HH_PRECOMPUTE: Finished Age Group: 34 of 82, time-this-age: 3.4127
SNW_HH_PRECOMPUTE: Finished Age Group:35 of 82, time-this-age:3.4113
SNW_HH_PRECOMPUTE: Finished Age Group:36 of 82, time-this-age:3.2095
SNW_HH_PRECOMPUTE: Finished Age Group:37 of 82, time-this-age:3.4244
SNW_HH_PRECOMPUTE: Finished Age Group:38 of 82, time-this-age:3.5012
SNW_HH_PRECOMPUTE: Finished Age Group:39 of 82, time-this-age:3.2675
SNW HH PRECOMPUTE: Finished Age Group: 40 of 82, time-this-age: 3.2625
SNW HH PRECOMPUTE: Finished Age Group:41 of 82, time-this-age:3.4011
SNW HH PRECOMPUTE: Finished Age Group:42 of 82, time-this-age:3.2533
SNW_HH_PRECOMPUTE: Finished Age Group:43 of 82, time-this-age:3.5132
SNW_HH_PRECOMPUTE: Finished Age Group:44 of 82, time-this-age:3.4771
SNW_HH_PRECOMPUTE: Finished Age Group:45 of 82, time-this-age:3.3133
SNW_HH_PRECOMPUTE: Finished Age Group:46 of 82, time-this-age:3.4673
SNW_HH_PRECOMPUTE: Finished Age Group:47 of 82, time-this-age:3.1794
SNW_HH_PRECOMPUTE: Finished Age Group:48 of 82, time-this-age:3.1958
SNW_HH_PRECOMPUTE: Finished Age Group:49 of 82, time-this-age:3.4545
SNW HH PRECOMPUTE: Finished Age Group:50 of 82, time-this-age:3.355
SNW_HH_PRECOMPUTE: Finished Age Group:51 of 82, time-this-age:3.2059
SNW HH PRECOMPUTE: Finished Age Group:52 of 82, time-this-age:3.2882
SNW_HH_PRECOMPUTE: Finished Age Group:53 of 82, time-this-age:3.3772
SNW_HH_PRECOMPUTE: Finished Age Group:54 of 82, time-this-age:3.3279
SNW_HH_PRECOMPUTE: Finished Age Group:55 of 82, time-this-age:3.5412
SNW_HH_PRECOMPUTE: Finished Age Group:56 of 82, time-this-age:3.504
SNW_HH_PRECOMPUTE: Finished Age Group:57 of 82, time-this-age:3.4961
SNW_HH_PRECOMPUTE: Finished Age Group:58 of 82, time-this-age:3.3629
SNW_HH_PRECOMPUTE: Finished Age Group:59 of 82, time-this-age:3.4105
SNW_HH_PRECOMPUTE: Finished Age Group:60 of 82, time-this-age:3.3755
SNW_HH_PRECOMPUTE: Finished Age Group:61 of 82, time-this-age:3.4102
SNW HH PRECOMPUTE: Finished Age Group:62 of 82, time-this-age:3.4844
SNW_HH_PRECOMPUTE: Finished Age Group:63 of 82, time-this-age:3.3864
SNW_HH_PRECOMPUTE: Finished Age Group:64 of 82, time-this-age:3.6674
SNW_HH_PRECOMPUTE: Finished Age Group:65 of 82, time-this-age:3.3549
SNW_HH_PRECOMPUTE: Finished Age Group:66 of 82, time-this-age:3.3543
```

```
SNW_HH_PRECOMPUTE: Finished Age Group:67 of 82, time-this-age:3.3775
SNW_HH_PRECOMPUTE: Finished Age Group:68 of 82, time-this-age:3.3574
SNW_HH_PRECOMPUTE: Finished Age Group:69 of 82, time-this-age:3.3706
SNW_HH_PRECOMPUTE: Finished Age Group:70 of 82, time-this-age:3.4841
SNW_HH_PRECOMPUTE: Finished Age Group:71 of 82, time-this-age:3.361
SNW HH PRECOMPUTE: Finished Age Group:72 of 82, time-this-age:3.3692
SNW_HH_PRECOMPUTE: Finished Age Group:73 of 82, time-this-age:3.4296
SNW_HH_PRECOMPUTE: Finished Age Group:74 of 82, time-this-age:3.4363
SNW_HH_PRECOMPUTE: Finished Age Group:75 of 82, time-this-age:3.391
SNW_HH_PRECOMPUTE: Finished Age Group:76 of 82, time-this-age:3.4927
SNW_HH_PRECOMPUTE: Finished Age Group:77 of 82, time-this-age:3.4073
SNW_HH_PRECOMPUTE: Finished Age Group:78 of 82, time-this-age:3.3416
SNW_HH_PRECOMPUTE: Finished Age Group:79 of 82, time-this-age:3.4042
SNW_HH_PRECOMPUTE: Finished Age Group:80 of 82, time-this-age:3.4844
SNW_HH_PRECOMPUTE: Finished Age Group:81 of 82, time-this-age:3.4572
SNW HH PRECOMPUTE: Finished Age Group:82 of 82, time-this-age:3.2411
SNW HH PRECOMPUTE: Finished Age Group:83 of 82, time-this-age:3.3428
Completed SNW_HH_PRECOMPUTE; SNW_MP_PARAM=default_docdense; SNW_MP_CONTROL=default_test; time cost=284.95
```

Solve for 2008 Value and Consumption with 0 and 2 Checks

Now we use the **snw_v08_jaeemk** function, which takes as inputs the 2008-specific value and policy we have already found, to compute the 2008 value and consumption based on different stimulus amounts via asset-equivalent transformation given stimulus amounts.

FIrst, obtain V and C with zero stimulus.

c2

c3

c1

```
% Call Function
welf checks = 0;
[V_2008_check0, C_2008_check0] = snw_v08_jaeemk(...
    welf checks, mp params, mp controls, 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=st_biden_or_trump_undefined;welf_checks=0;TR=0.0018238;SNW_MP_PARAM=c
Completed SNW_V08_JAEEMK;SNW_MP_PARAM=default_docdense;SNW_MP_CONTROL=default_test;timeEUEC=3e-05
-----
CONTAINER NAME: mp outcomes ND Array (Matrix etc)
std
                  i
                      idx
                             ndim
                                     numel
                                               rowN
                                                         colN
                                                                      sum
                                                                                 mean
                                                                                                    coefvai
                                                                                          8.4419
   C_2008_check
                  1
                       1
                              6
                                    4.37e+07
                                                83
                                                       5.265e+05
                                                                   2.3277e+08
                                                                                 5.3267
                                                                                                    1.5848
                       2
                                                83
                                                                                                    -1.4375
   V_2008_check
                  2
                              6
                                    4.37e+07
                                                       5.265e+05
                                                                   -6.6426e+08
                                                                                -15.201
                                                                                           21.85
c1
                                  c3
                                             c4
                                                        c5
                                                                 c526496
                                                                           c526497
                                                                                     c526498
                                                                                               c526499
   r1
          0.036218
                     0.036736
                                0.038184
                                          0.042735
                                                      0.048545
                                                                 12.256
                                                                           12.541
                                                                                     12.835
                                                                                               13.136
   r2
          0.036271
                     0.036736
                                0.038385
                                           0.043404
                                                      0.049852
                                                                 12.491
                                                                           12.778
                                                                                     13.072
                                                                                               13.374
          0.036717
                     0.037251
                                0.039845
                                          0.044907
                                                      0.051515
                                                                 12.744
                                                                           13.032
                                                                                     13.327
   r3
                                                                                               13.628
                                                                 12.989
   r4
          0.038144
                     0.038678
                                0.041269
                                          0.046371
                                                      0.053128
                                                                           13.277
                                                                                     13.573
                                                                                               13.872
                                0.042653
   r5
          0.039534
                     0.040068
                                          0.047793
                                                      0.054687
                                                                 13.224
                                                                           13.513
                                                                                     13.809
                                                                                               14.105
                                                                           37.367
                                                                                     39.414
   r79
           0.2016
                     0.20214
                                0.20586
                                           0.21598
                                                      0.23568
                                                                 35.82
                                                                                               41.705
                                                                 40.755
                                                                           42.955
                                                                                     45.289
                                                                                                47.95
   r80
           0.2016
                     0.20214
                                0.20586
                                           0.21598
                                                      0.23568
   r81
           0.2016
                     0.20214
                                0.20586
                                           0.21598
                                                       0.23568
                                                                 48.912
                                                                           52.041
                                                                                     55.022
                                                                                               57.919
   r82
           0.2016
                     0.20214
                                0.20586
                                           0.21598
                                                       0.23568
                                                                 66.719
                                                                           69.201
                                                                                     72.373
                                                                                               77.005
   r83
           0.2016
                     0.20214
                                0.20586
                                           0.21598
                                                       0.23568
                                                                 116.83
                                                                           122.65
                                                                                     128.67
                                                                                               134.89
xxx TABLE:V 2008 check xxxxxxxxxxxxxxxxxx
```

c5

c526496

c526497

c526498

c526499

c4

r1	-295.66	-295.26	-292.66	-286.62	-277.22	-4.3615	-4.2673	-4.1741	-4.0822
r2	-286.11	-285.71	-283.12	-277.16	-268.03	-4.2548	-4.1641	-4.0744	-3.9858
r3	-276.49	-276.09	-273.59	-267.84	-259.11	-4.1461	-4.0589	-3.9727	-3.8874
r4	-266.77	-266.41	-264.08	-258.7	-250.49	-4.0342	-3.9507	-3.868	-3.7862
r5	-257.99	-257.65	-255.48	-250.43	-242.69	-3.9287	-3.8485	-3.769	-3.6903
r79	-13.356	-13.343	-13.253	-13.025	-12.638	-0.22088	-0.21055	-0.20083	-0.1917
r80	-12.025	-12.012	-11.923	-11.695	-11.308	-0.16977	-0.1618	-0.15428	-0.14721
r81	-10.382	-10.369	-10.28	-10.052	-9.6651	-0.11711	-0.11162	-0.10645	-0.10156
r82	-8.1742	-8.1611	-8.0716	-7.844	-7.457	-0.065329	-0.062239	-0.059357	-0.056632
r83	-4.9602	-4.9471	-4.8576	-4.6301	-4.2431	-0.020966	-0.019971	-0.019037	-0.01816

% Call Function

Second, obtain V and C with two stimulus checks.

```
welf_checks = 2;
[V_2008_check2, C_2008_check2] = snw_v08_jaeemk(...
    welf_checks, mp_params, mp_controls, V_2008, cons_2008, mp_precompute_res);
```

rowN

Solve for V_2008_check for 2 stimulus checks

Completed SNW_A4CHK_WRK_BISEC_VEC;SNW_MP_PARAM=st_biden_or_trump_undefined;welf_checks=2;TR=0.0018238;SNW_MP_PARAM=completed SNW_V08_JAEEMK;SNW_MP_PARAM=default_docdense;SNW_MP_CONTROL=default_test;timeEUEC=2.16e-05

colN

sum

std

mean

coefvai

CONTAINER NAME: mp_outcomes ND Array (Matrix etc)

idx

ndim

numel

_	8_check	1	1	6	4.37e+0		5.265e+05	2.328	3e+08 5	5.3273	8.442
V_200	8_check	2	2	6	4.37e+0	7 83	5.265e+05	-6.6365	5e+08 -1	15.187 2	21.798 -
TABLE	:C_2008_ch	eck	xxxxxxxx	(XXXXXX	xxx						
	c1		c2	,	с3	c4	c 5	c526496	c526497	c526498	c52649
											-
r1	0.037941		0.038148	0.0	39819	0.043807	0.049244	12.256	12.541	12.835	13.136
r2	0.038108		0.038344	0.0	40188	0.044594	0.050571	12.492	12.778	13.073	13.374
r3	0.03941		0.039781	0.0	41664	0.046126	0.052244	12.745	13.032	13.327	13.628
r4	0.040834		0.041205	0.0	43102	0.047618	0.053867	12.989	13.278	13.573	13.872
r5	0.04222		0.042589	0	.0445	0.049065	0.055435	13.224	13.513	13.809	14.105
r79	0.20525		0.20579	0.	20951	0.21963	0.23776	35.821	37.368	39.415	41.707
r80	0.20525		0.20579	0.	20951	0.21963	0.23776	40.756	42.957	45.29	47.952
r81	0.20525		0.20579	0.	20951	0.21963	0.2378	48.914	52.043	55.024	57.92
r82	0.20525		0.20579	0.	20951	0.21963	0.23814	66.72	69.203	72.375	77.007
r83	0.20525		0.20579	0.	20951	0.21963	0.23933	116.84	122.66	128.68	134.89

XXX	TABLE:V	2008	check	XXXXXXXXXXXXXXXXX

	c1	c2	с3	с4	с5	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 = V_2008_check2 - V_2008_check0;
mn_MPC_U_gain_share_check = (C_2008_check2 - C_2008_check0)./(welf_checks*mp_params('TR'));
```

Param Results Define Frames

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

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

1	0	1.0061	0.90365	0.8116	0.72825	0.65329	0.58628	6
2	0.00051498	0.99612	0.89566	0.80504	0.72289	0.64888	0.58264	0.
3	0.0041199	0.84723	0.77113	0.6992	0.63256	0.57176	0.51677	0.
4	0.013905	0.64775	0.59855	0.55009	0.50372	0.46037	0.42041	0.
5	0.032959	0.46258	0.43381	0.40406	0.37453	0.34616	0.31948	0.
6	0.064373	0.33044	0.31319	0.29451	0.27537	0.25656	0.23856	0.

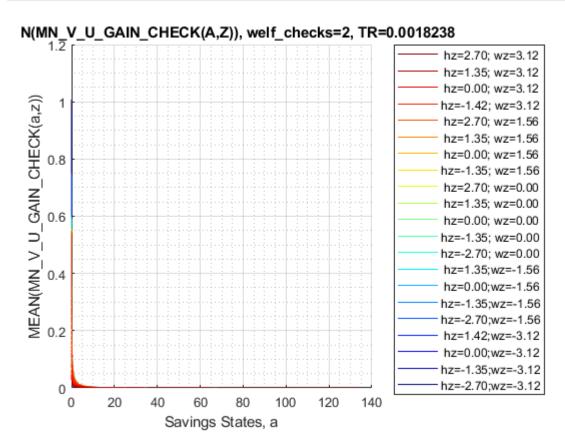
% Consumption

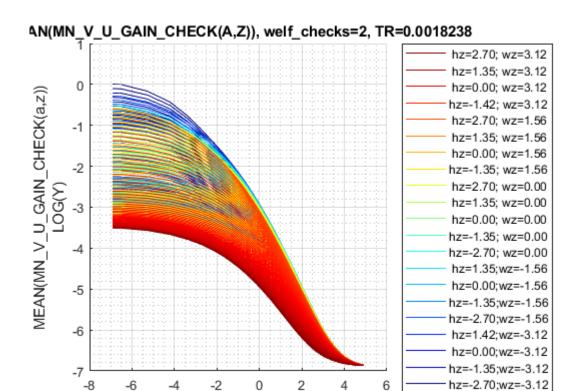
st_title = ['MEAN(MN_MPC_U_GAIN_CHECK(A,Z)), welf_checks=' num2str(welf_checks) ', TR=' num2str
tb_az_c = ff_summ_nd_array(st_title, mn_MPC_U_gain_share_check, true, ["mean"], 4, 1, cl_mp_data

cavinge	mean eta 1			xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx		mean eta 6	mean
0	0.96418	0.92776	0.90436	0.88966	0.87994	0.87411	0.86
0.00051498	0.95984	0.92012	0.8967	0.8773	0.86664	0.86053	0.8
0.0041199	0.87755	0.8775	0.87564	0.87396	0.87313	0.87278	0.87
0.013905	0.79847	0.79688	0.79546	0.79569	0.79653	0.79773	0.79
0.032959	0.71392	0.71064	0.71031	0.71109	0.71248	0.71436	0.7
0.064373	0.63466	0.6344	0.63533	0.63678	0.63877	0.64105	0.64
	0.00051498 0.0041199 0.013905 0.032959	0 0.96418 0.00051498 0.95984 0.0041199 0.87755 0.013905 0.79847 0.032959 0.71392	0 0.96418 0.92776 0.00051498 0.95984 0.92012 0.0041199 0.87755 0.8775 0.013905 0.79847 0.79688 0.032959 0.71392 0.71064	0 0.96418 0.92776 0.90436 0.00051498 0.95984 0.92012 0.8967 0.0041199 0.87755 0.8775 0.87564 0.013905 0.79847 0.79688 0.79546 0.032959 0.71392 0.71064 0.71031	0 0.96418 0.92776 0.90436 0.88966 0.00051498 0.95984 0.92012 0.8967 0.8773 0.0041199 0.87755 0.8775 0.87564 0.87396 0.013905 0.79847 0.79688 0.79546 0.79569 0.032959 0.71392 0.71064 0.71031 0.71109	0 0.96418 0.92776 0.90436 0.88966 0.87994 0.00051498 0.95984 0.92012 0.8967 0.8773 0.86664 0.0041199 0.87755 0.8775 0.87564 0.87396 0.87313 0.013905 0.79847 0.79688 0.79546 0.79569 0.79653 0.032959 0.71392 0.71064 0.71031 0.71109 0.71248	0 0.96418 0.92776 0.90436 0.88966 0.87994 0.87411 0.00051498 0.95984 0.92012 0.8967 0.8773 0.86664 0.86053 0.0041199 0.87755 0.8775 0.87564 0.87396 0.87313 0.87278 0.013905 0.79847 0.79688 0.79546 0.79569 0.79653 0.79773 0.032959 0.71392 0.71064 0.71031 0.71109 0.71248 0.71436

Graph Mean Values:

```
st_title = ['MEAN(MN\_V\_U\_GAIN\_CHECK(A,Z)), welf\_checks=' num2str(welf_checks) ', TR=' num2support_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);
```





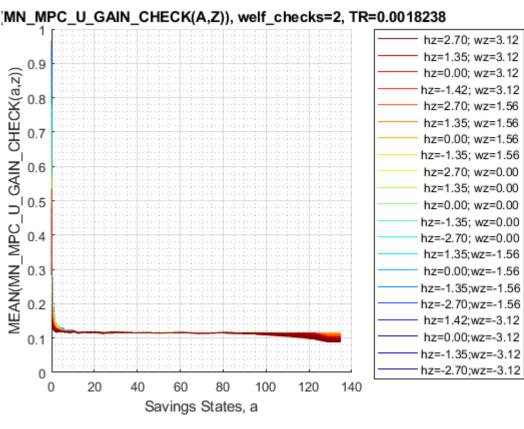
Graph Mean Consumption (MPC: Share of Check Consumed):

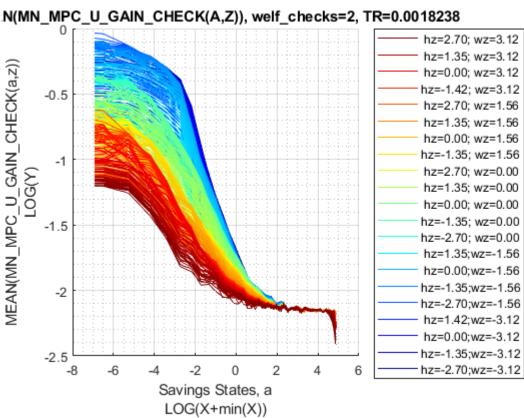
Savings States, a LOG(X+min(X))

-6

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

hz=-2.70;wz=-3.12





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, 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 = V_2008_check2 - V_2008_check0;
mn_C_W_gain_check_use = C_2008_check2 - C_2008_check0;
```

Select Age, Education, Marital, Kids Count:s

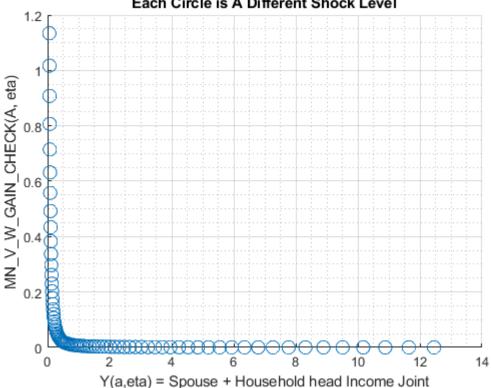
```
% Selections
it age = 21; % +18
it_marital = 1; % 1 = unmarried
it_kids = 1; % 1 = kids is zero
it educ = 1; % 1 = lower education
% Select: NaN(n_jgrid,n_agrid,n_etagrid,n_educgrid,n_marriedgrid,n_kidsgrid);
mn_C W gain check jemk = mn_C W gain check use(it age, :, 1:mp_params('n_eta_H grid'), it_educ,
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;
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

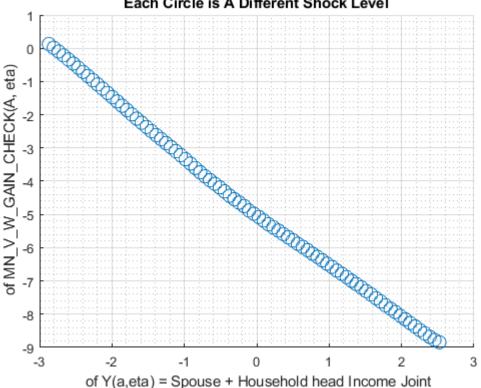
How do shocks and a impact marginal value. First plot one asset level, variation comes only from increasingly higher shocks:

```
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



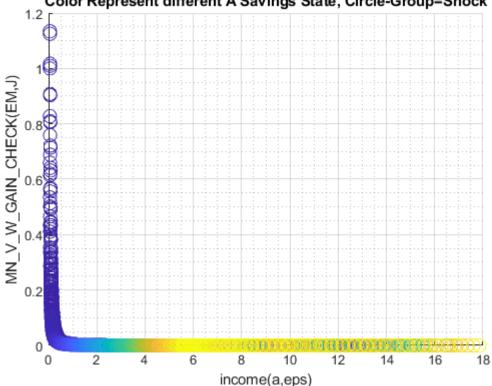
MN_V_W_GAIN_CHECK(Y(A, eta)), Lowest A, J38M0E0K0 Each Circle is A Different Shock Level



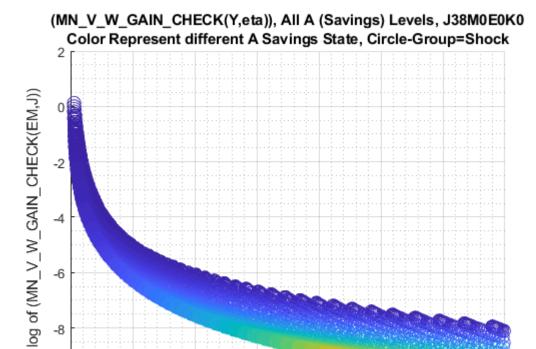
Plot all asset levels:

```
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;
```

(MN_V_W_GAIN_CHECK(Y,eta)), All A (Savings) Levels, J38M0E0K0 Color Represent different A Savings State, Circle-Group=Shock



```
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;
```



3

income(a,eps)

2

-10

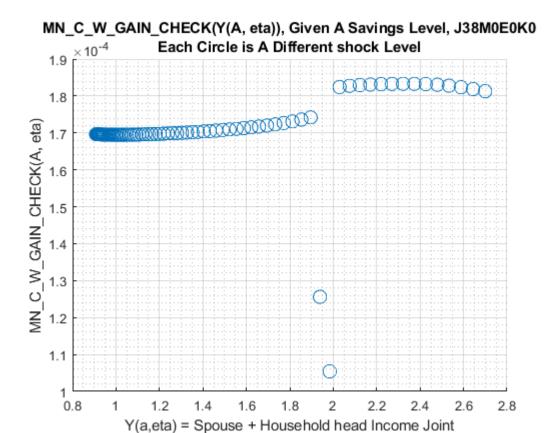
0

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

5

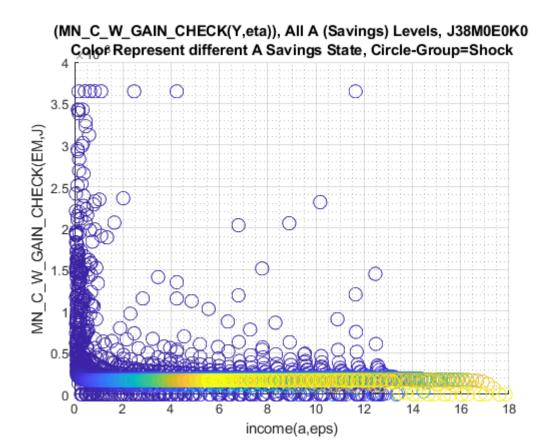
6

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



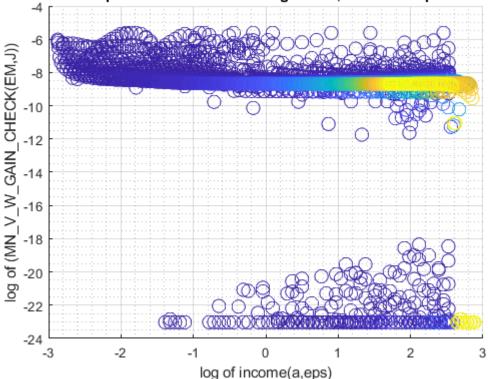
Plot all asset levels:

```
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;
```





Analyze Kids and Marriage and Age

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:

```
% 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_23
1	1	0	0.03116	0.030004	0.028268	0.025846	0.023825	0.022124
2	2	0	0.042925	0.041372	0.038951	0.035539	0.032682	0.03027
3	3	0	0.050042	0.048477	0.045918	0.041935	0.038603	0.035792
4	4	0	0.056814	0.055153	0.05234	0.047822	0.044043	0.040856
5	5	0	0.06224	0.060592	0.057683	0.052755	0.048636	0.045166
6	1	1	0.0089468	0.0085141	0.0080936	0.0073219	0.006674	0.0061254
7	2	1	0.012008	0.011426	0.01086	0.0098202	0.0089462	0.0082092
8	3	1	0.014485	0.01381	0.013144	0.011882	0.010826	0.0099375
9	4	1	0.017392	0.016611	0.015824	0.014325	0.013061	0.011988
10	5	1	0.021156	0.020287	0.019385	0.017573	0.016058	0.014774

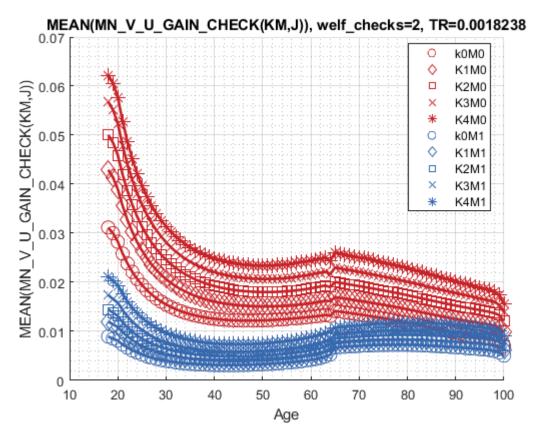
% Consumption Function

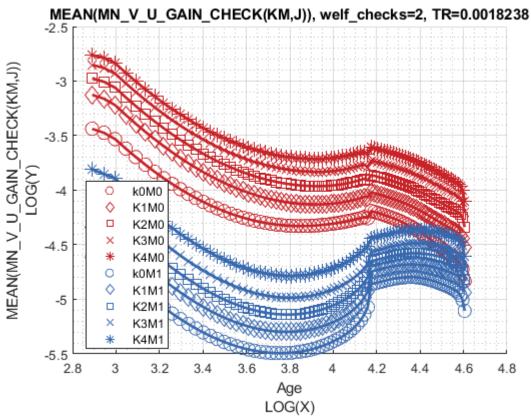
st_title = ['MEAN(MN_MPC_U_GAIN_CHECK(KM,J)), welf_checks=' num2str(welf_checks) ', TR=' num2st
tb_az_c = ff_summ_nd_array(st_title, mn_MPC_U_gain_share_check, true, ["mean"], 3, 1, cl_mp_data

group	kids	marry	mean_age_18	mean_age_19	mean_age_20	mean_age_21	mean_age_22	mean_age_2
1	1	0	0.071209	0.076265	0.082919	0.080484	0.079666	0.078062
2	2	0	0.08025	0.085444	0.092048	0.090298	0.089234	0.088128
3	3	0	0.087972	0.095508	0.10343	0.10119	0.099548	0.097144
4	4	0	0.092255	0.099859	0.10923	0.10651	0.10388	0.10207
5	5	0	0.09665	0.1041	0.11456	0.11152	0.10814	0.10532
6	1	1	0.101	0.10439	0.10978	0.10855	0.10822	0.10834
7	2	1	0.10297	0.10717	0.11242	0.11166	0.10915	0.10816
8	3	1	0.10827	0.11355	0.11922	0.11713	0.11645	0.11572
9	4	1	0.10932	0.11394	0.12031	0.11884	0.11867	0.11712
10	5	1	0.11555	0.12093	0.1289	0.12462	0.12328	0.12196

Graph Mean Values:

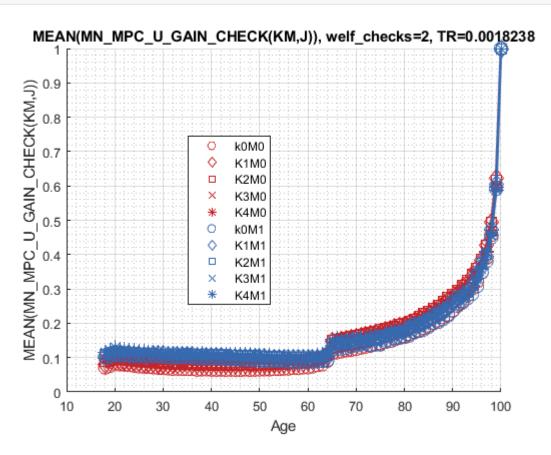
```
st_title = ['MEAN(MN\_V\_U\_GAIN\_CHECK(KM,J)), welf\_checks=' num2str(welf_checks) ', TR=' num
mp_support_graph('cl_st_graph_title') = {st_title};
mp_support_graph('cl_st_ytitle') = {'MEAN(MN\_V\_U\_GAIN\_CHECK(KM,J))'};
ff_graph_grid((tb_az_v{1:end, 4:end}), ar_row_grid, age_grid, mp_support_graph);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

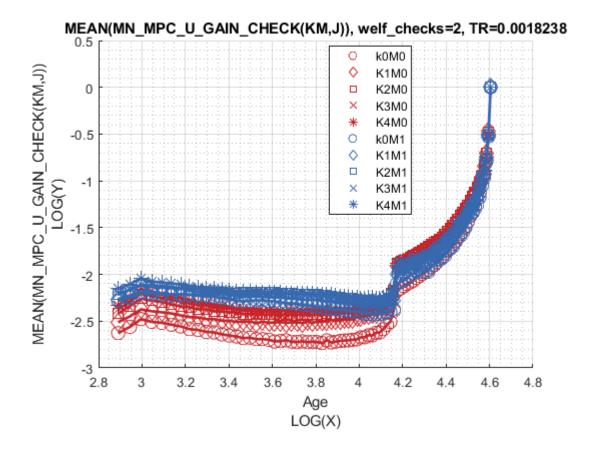




Graph Mean Consumption (MPC: Share of Check Consumed):

```
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))

marry

mean_age_18

Tabulate value and policies:

group

mean_age_20

mean age 21

mean_age_23

mean_age_22

mean_age_19

1	0	0	0.050249	0.049057	0.04709	0.044412	0.042032	0.039911
2	1	0	0.047023	0.045182	0.042174	0.037147	0.033084	0.029772
3	0	1	0.015742	0.015072	0.014412	0.01336	0.012437	0.011628
4	1	1	0.013853	0.013188	0.012511	0.011009	0.0097884	0.0087852

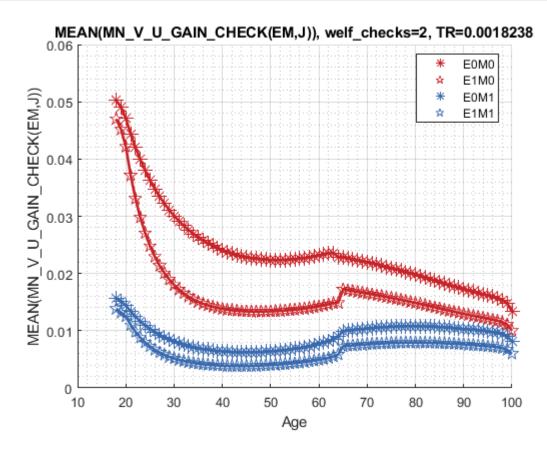
% Consumption

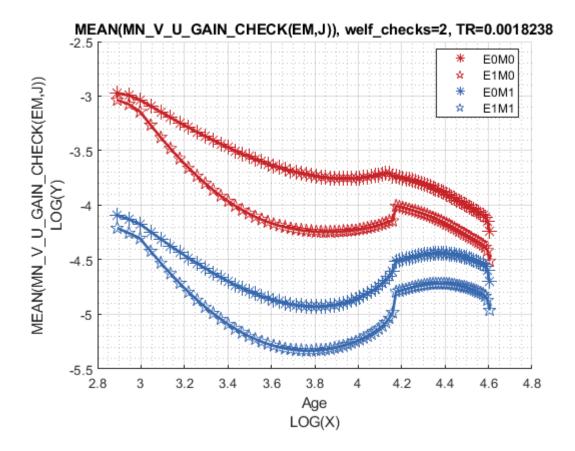
st_title = ['MEAN(MN_MPC_U_GAIN_CHECK(EM,J)), welf_checks=' num2str(welf_checks) ', TR=' num2st
tb_az_c = ff_summ_nd_array(st_title, mn_MPC_U_gain_share_check, true, ["mean"], 3, 1, cl_mp_data

XXX MEA	xxx MEAN(MN_MPC_U_GAIN_CHECK(EM,J)), welf_checks=2, TR=0.0018238 xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx											
grou	up edu	marry	mean_age_18	mean_age_19	mean_age_20	mean_age_21	mean_age_22	mean_age_23				
1	0	0	0.07766	0.081478	0.084248	0.083711	0.083393	0.082988				
2	1	0	0.093675	0.10299	0.11663	0.11229	0.10879	0.1053				
3	0	1	0.099571	0.10294	0.10634	0.10623	0.10602	0.10588				
4	1	1	0.11528	0.12105	0.12991	0.12609	0.12429	0.12264				

Graph Mean Values:

```
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);
```





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);
```

