

Distribution Grid Search

This is the example vignette for function: [snw_ds_main_grid_search](#) from the [PrjOptiSNW Package](#). This function solves for vfi and gets distribution induced by policy functions and exogenous distributions. Grid Search for VFI and Grid Search also for Distribution. The results are illustrative of the differences between using grid search and exact solution. The grid search solution here is not fully vectorized but loops over the state-space.

Test SNW_DS_MAIN_GRID_SEARCH Defaults More Dense

Rather than solving for "docdense", this solves for "moredense", which has fewer shocks, in order to save time given the relatively slow speed of this algorithm.

```
mp_params = snw_mp_param('default_moredense');
mp_controls = snw_mp_control('default_test');
mp_controls('bl_print_vfi') = false;
mp_controls('bl_print_ds') = false;
mp_controls('bl_print_ds_verbose') = false;
[Phi_true,Phi_adj,A_agg,Y_inc_agg,it,mp_dsvfi_results] = snw_ds_main_grid_search(mp_params, mp_controls);
```

```
Elapsed time is 13045.688258 seconds.
Completed SNW_VFI_MAIN_GRID_SEARCH;SNW_MP_PARAM=default_moredense;SNW_MP_CONTROL=default_test
Elapsed time is 13841.078692 seconds.
Completed SNW_DS_MAIN;SNW_MP_PARAM=default_moredense;SNW_MP_CONTROL=default_test
```

```
Phi_true = Phi_true/sum(Phi_true(:));
```

Show All Info in mp_dsvfi_results More Dense

```
mp_cl_mt_xyz_of_s = mp_dsvfi_results('mp_cl_mt_xyz_of_s');
disp(mp_cl_mt_xyz_of_s('tb_outcomes'))
```

	mean	unweighted_sum	sd	coefofvar	gini	min	max	pVis0
a_ss	4.2955	5130.2	8.2965	1.9314	0.74079	0	135	0.1766
ap_ss	33.948	11476	25.584	0.75362	0.43382	1	151	
cons_ss	1.1795	1.585e+07	1.0182	0.86318	0.40848	0.035637	141.61	
v_ss	-19.79	-1.1145e+07	35.654	-1.8016	-0.774	-868.79	25.518	
n_ss	2.3554	21	1.4375	0.61029	0.3128	1	6	
y_all	1.6272	2.3969e+07	1.8982	1.1665	0.50121	0.038108	50.873	
y_head_inc	1.2682	5.6172e+05	1.5441	1.2175	0.50432	0.038108	24.357	
y_head_earn	1.0492	2628.2	1.4242	1.3574	0.60462	0	18.957	0.201
y_spouse_inc	0.35895	55552	0.96039	2.6755	0.85307	0	26.627	0.5249
yshr_interest	0.11509	1.0971e+06	0.17681	1.5363	0.70728	0	0.99299	0.1766
yshr_wage	0.78433	2.4004e+06	0.34004	0.43354	0.19505	0	1	0.1058
yshr_SS	0.10058	67295	0.23745	2.3608	0.91583	0	1	0.798
yshr_tax	0.17694	7.7853e+05	0.040535	0.22909	0.13026	0.036506	0.2552	
yshr_nttxss	0.076363	7.1123e+05	0.25868	3.3875	1.4024	-0.89715	0.2552	

More Dense Param Results Define Frames

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

```
% Grids:
```

```

age_grid = 18:100;
agrid = mp_params('agrid');
eta_H_grid = mp_params('eta_H_grid');
eta_S_grid = mp_params('eta_S_grid');
ar_st_eta_HS_grid = string(cellstr([num2str(eta_H_grid', 'hz=%3.2f;'), num2str(eta_S_grid', 'wz=%3.2f;')], 'wz=%3.2f;'), 'hz=%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 Probability Mass Along Age Dimensions

Where are the mass at? Analyze mass given state space components.

```

% Get the Joint distribution over all states
% Define Graph Inputs
mp_support_graph = containers.Map('KeyType', 'char', 'ValueType', 'any');
mp_support_graph('st_legend_loc') = 'best';
mp_support_graph('bl_graph_logy') = false; % do not log

```

Exogenous Permanent States Mass: Life Cycle, Edu and Marraige

Tabulate value and policies along savings and shocks:

```

% NaN(n_jgrid,n_agrid,n_etagrid,n_educgrid,n_marriedgrid,n_kidsgrid);
ar_permute = [2,3,6,1,5,4];
% Value Function
tb_prob_aem = ff_summ_nd_array("P(Age, EDU, MARRY)", Phi_true, true, ["sum"], 3, 1, cl_mp_data

```

xxx	P(Age, EDU, MARRY))	xxxxxxxxxxxxxxxxxxxxxxxxxxxx								
group	marry	edu	sum_age_18	sum_age_19	sum_age_20	sum_age_21	sum_age_22	sum_age_23	sum	
1	0	0	0.0085768	0.0084866	0.0083969	0.0083078	0.0082194	0.0081317	0.	
2	1	0	0.0066438	0.0065739	0.0065044	0.0064354	0.0063669	0.006299	0.0	
3	0	1	0.0028875	0.0028571	0.002827	0.002797	0.0027672	0.0027377	0.0	
4	1	1	0.0037292	0.0036899	0.0036509	0.0036122	0.0035738	0.0035356	0.0	

```

mp_support_graph('cl_st_graph_title') = {'Pstationary(Age, Edu, Marry), age=x, marry/edu=color'};
mp_support_graph('cl_st_ytitle') = {'Conditional Aggregate Mass'};
ar_row_grid = ["M0E0", "M1E0", "M0E1", "M1E1"];
mp_support_graph('cl_st_xtitle') = {'Age Groups'};
mp_support_graph('cl_scatter_shapes') = {'*', '*', 'p', 'p'};
mp_support_graph('cl_colors') = {'red', 'blue', 'red', 'blue'};
ff_graph_grid((tb_prob_aem{1:end, 4:end}), ar_row_grid, age_grid, mp_support_graph);

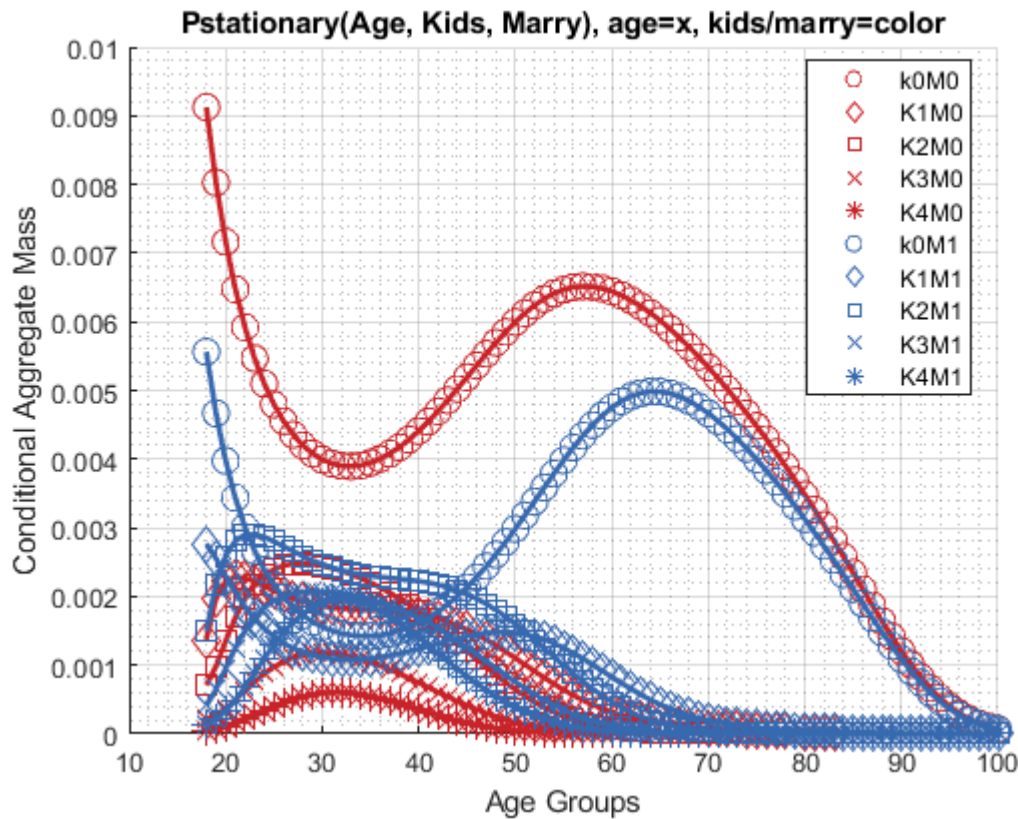
```



```

'o', 'd', 's', 'x', '*'};
mp_support_graph('cl_colors') = {...
'red', 'red', 'red', 'red', 'red'...
'blue', 'blue', 'blue', 'blue', 'blue'};
mp_support_graph('cl_st_xtitle') = {'Age Groups'};
ff_graph_grid((tb_prob_amarrykids{1:end, 4:end}), ar_row_grid, age_grid, mp_support_graph);

```



Analyze Probability Mass Asset and Shock Dimensions

Where are the mass at?

```

% Define Graph Inputs
mp_support_graph = containers.Map('KeyType', 'char', 'ValueType', 'any');
mp_support_graph('st_legend_loc') = 'best';
mp_support_graph('bl_graph_logy') = false; % do not log

```

Asset and Shock Mass

```

% NaN(n_jgrid,n_agrid,n_etagrid,n_educgrid,n_marriedgrid,n_kidsgrid);
ar_permute = [1,4,5,6,3,2];
% Value Function
tb_prob_az = ff_summ_nd_array("P(A,Z)", Phi_true, true, ["sum"], 4, 1, cl_mp_datasetdesc, ar_p

```

```

xxx P(A,Z)) xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx
group savings sum_eta_1 sum_eta_2 sum_eta_3 sum_eta_4 sum_eta_5 sum_eta_6 sum_eta_

```

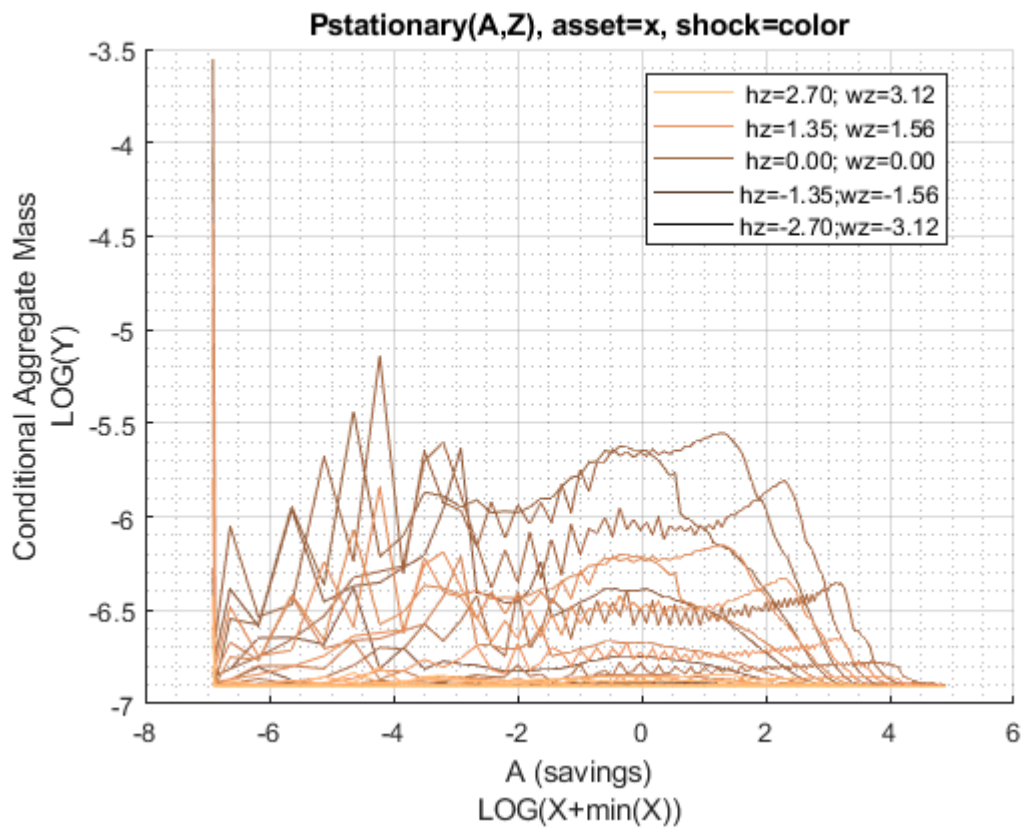
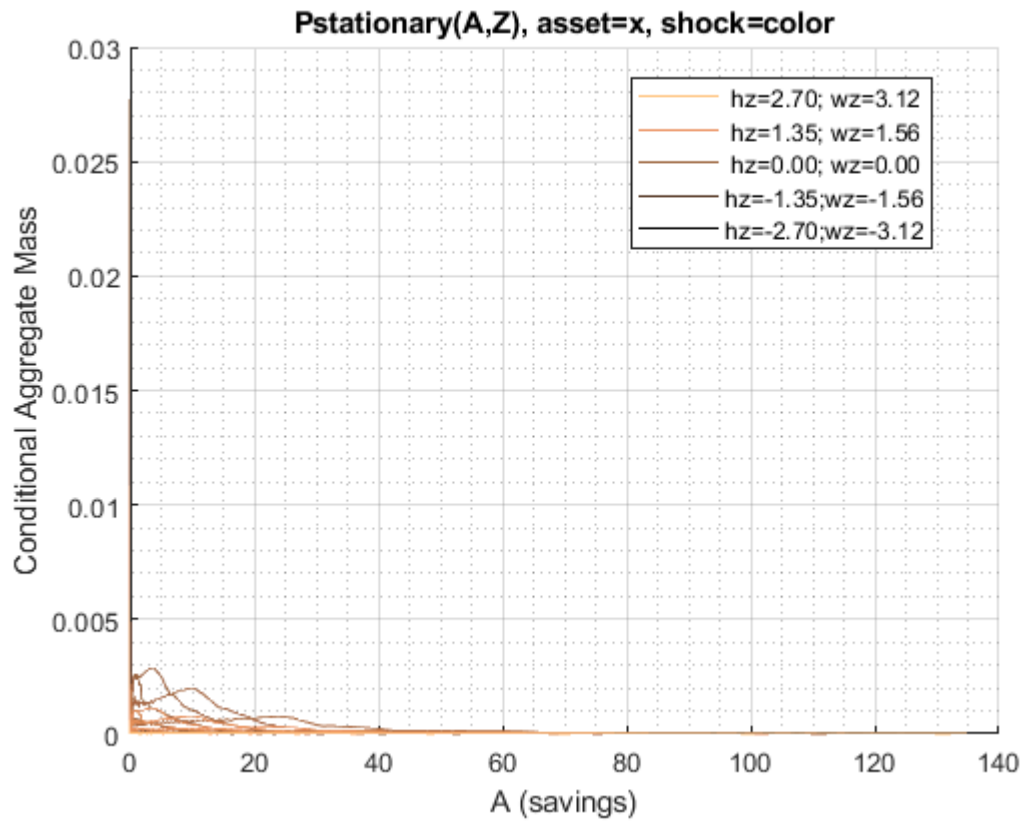
1	0	1.7754e-05	0.00011362	0.00037708	0.00061913	0.00056244	0.00033244	0.000115
2	4e-05	2.8763e-07	1.3442e-06	4.0195e-06	3.6795e-07	6.5205e-06	1.2822e-06	2.0438e-06
3	0.00032	8.5896e-07	2.1837e-06	1.5288e-05	9.7352e-06	3.0233e-05	2.2281e-06	7.9033e-06
4	0.00108	2.4168e-06	6.6439e-06	8.5514e-06	8.806e-06	1.0167e-05	3.836e-06	3.4378e-06
5	0.00256	7.7644e-07	6.7137e-06	3.5185e-05	3.6273e-05	1.2262e-05	6.4167e-06	5.2691e-06
6	0.005	1.615e-07	5.6988e-06	1.2805e-05	1.6079e-05	5.4165e-05	1.0241e-05	1.1273e-05
7	0.00864	2.9022e-07	1.5618e-05	1.5907e-05	7.5335e-05	2.1309e-05	1.7692e-05	2.649e-05
8	0.01372	9.2058e-08	2.2512e-06	1.6629e-05	2.2447e-05	0.00010906	1.9013e-05	5.201e-05
9	0.02048	1.493e-07	5.139e-06	1.987e-05	2.7684e-05	1.8495e-05	2.0213e-05	4.9447e-05
10	0.02916	2.2636e-07	2.1911e-06	2.2674e-05	4.0829e-05	5.6366e-05	5.2344e-05	8.8742e-05
11	0.04	3.6847e-07	3.2622e-06	3.5384e-05	3.9524e-05	3.7428e-05	6.0069e-05	6.175e-05
12	0.05324	2.7682e-07	2.8958e-06	5.7484e-05	3.6034e-05	3.501e-05	4.0214e-05	9.433e-05
13	0.06912	3.353e-07	2.6767e-06	1.4572e-05	3.8273e-05	2.474e-05	2.563e-05	1.3972e-05
14	0.08788	3.0394e-07	2.2742e-06	1.4027e-05	3.4255e-05	3.7729e-05	1.5417e-05	5.5589e-05
15	0.10976	2.5858e-07	1.9422e-06	1.2587e-05	3.4742e-05	2.7058e-05	2.4059e-05	3.9792e-05
16	0.135	2.7326e-07	1.9079e-06	1.2852e-05	3.438e-05	3.6783e-05	1.6552e-05	1.4835e-05
17	0.16384	2.7552e-07	1.9749e-06	1.5066e-05	3.5948e-05	3.0858e-05	2.8702e-05	5.0937e-05
18	0.19652	3.0673e-07	2.1238e-06	2.128e-05	3.877e-05	3.75e-05	1.7398e-05	7.7674e-05
19	0.23328	3.2755e-07	2.1478e-06	1.0054e-05	3.8731e-05	3.1329e-05	2.6872e-05	1.3415e-05
20	0.27436	3.5257e-07	2.2995e-06	1.0938e-05	4.2758e-05	4.3572e-05	1.8163e-05	1.1471e-05
21	0.32	4.0641e-07	2.5518e-06	1.1635e-05	4.5919e-05	3.416e-05	3.1068e-05	6.508e-05
22	0.37044	4.4361e-07	2.8814e-06	1.2784e-05	4.688e-05	4.9116e-05	2.2826e-05	1.2243e-05
23	0.42592	4.7158e-07	3.2535e-06	1.3383e-05	5.1616e-05	4.104e-05	2.9579e-05	8.6872e-05
24	0.48668	5.1923e-07	3.4305e-06	1.432e-05	5.2283e-05	5.1981e-05	2.8417e-05	1.3112e-05
25	0.55296	5.437e-07	3.6017e-06	1.5469e-05	5.6439e-05	5.2035e-05	3.2783e-05	9.0771e-05
26	0.625	5.5347e-07	3.8425e-06	1.5479e-05	5.5965e-05	5.6994e-05	2.7605e-05	1.2764e-05
27	0.70304	5.568e-07	3.7614e-06	1.5095e-05	5.8432e-05	5.4503e-05	3.5853e-05	1.0157e-05
28	0.78732	5.7466e-07	3.7497e-06	1.502e-05	5.7903e-05	5.6109e-05	2.9314e-05	1.5137e-05
29	0.87808	5.7563e-07	3.8005e-06	1.4892e-05	5.6949e-05	5.3853e-05	3.2364e-05	8.6997e-05
30	0.97556	5.7132e-07	3.8705e-06	1.5095e-05	5.6906e-05	5.5758e-05	2.9176e-05	1.1682e-05
31	1.08	5.726e-07	3.8645e-06	1.5349e-05	5.6413e-05	5.4339e-05	3.1568e-05	9.0865e-05
32	1.1916	5.6625e-07	3.7872e-06	1.484e-05	5.448e-05	5.7108e-05	2.7934e-05	1.2906e-05
33	1.3107	5.4238e-07	3.6971e-06	1.4269e-05	5.2724e-05	5.4906e-05	3.2967e-05	8.5976e-05
34	1.4375	5.3278e-07	3.5724e-06	1.3626e-05	4.9663e-05	5.6325e-05	2.6723e-05	1.1993e-05
35	1.5722	5.0996e-07	3.485e-06	1.3686e-05	4.6691e-05	5.5236e-05	3.069e-05	8.7119e-05
36	1.715	5.0398e-07	3.4238e-06	1.3137e-05	4.7646e-05	5.6783e-05	2.8995e-05	1.2248e-05
37	1.8662	4.767e-07	3.3615e-06	1.2844e-05	3.4086e-05	5.6434e-05	3.0261e-05	8.7292e-05
38	2.0261	4.684e-07	3.2063e-06	1.2531e-05	3.2602e-05	5.8615e-05	2.8719e-05	1.0928e-05
39	2.1949	4.5256e-07	3.1314e-06	1.2176e-05	3.2249e-05	5.8445e-05	3.0339e-05	8.5903e-05
40	2.3728	4.2248e-07	3.093e-06	1.2003e-05	3.0763e-05	6.0571e-05	2.7865e-05	1.1857e-05
41	2.56	4.0678e-07	2.958e-06	1.1808e-05	3.0172e-05	6.0706e-05	3.0549e-05	8.5883e-05
42	2.7568	3.9214e-07	2.8541e-06	1.1601e-05	2.9569e-05	6.2065e-05	2.9269e-05	1.1257e-05
43	2.9635	3.6841e-07	2.7289e-06	1.1074e-05	2.8387e-05	6.2402e-05	3.0802e-05	9.4896e-05
44	3.1803	3.4717e-07	2.6124e-06	1.0683e-05	2.5913e-05	6.3791e-05	2.948e-05	1.1039e-05
45	3.4074	3.2367e-07	2.453e-06	1.0352e-05	2.5365e-05	6.4051e-05	3.198e-05	9.2027e-05
46	3.645	3.0387e-07	2.3261e-06	9.9363e-06	2.5054e-05	6.4312e-05	3.0105e-05	1.1997e-05
47	3.8934	2.8153e-07	2.1577e-06	9.4355e-06	2.4465e-05	6.3912e-05	3.2461e-05	9.3007e-05
48	4.1529	2.5915e-07	2.0525e-06	8.9454e-06	2.4289e-05	6.2333e-05	3.3681e-05	1.1739e-05
49	4.4237	2.3743e-07	1.8857e-06	8.3087e-06	2.2777e-05	6.0276e-05	3.3448e-05	1.015e-05
50	4.706	2.1918e-07	1.7288e-06	7.9955e-06	2.2245e-05	5.8169e-05	3.4403e-05	1.1532e-05
51	5	1.9952e-07	1.5877e-06	7.4469e-06	2.1203e-05	5.5849e-05	3.516e-05	1.0208e-05
52	5.306	1.8097e-07	1.466e-06	6.9877e-06	2.0273e-05	5.2403e-05	3.5974e-05	1.2261e-05
53	5.6243	1.6229e-07	1.3297e-06	6.4332e-06	1.9708e-05	4.7628e-05	3.6187e-05	1.0049e-05
54	5.9551	1.4898e-07	1.2561e-06	6.0843e-06	1.9174e-05	4.4076e-05	3.7855e-05	1.2105e-05
55	6.2986	1.3486e-07	1.128e-06	5.585e-06	1.771e-05	4.0303e-05	3.7916e-05	1.1196e-05
56	6.655	1.2261e-07	1.0323e-06	5.2044e-06	1.6634e-05	3.8461e-05	3.8144e-05	1.1796e-05
57	7.0246	1.0954e-07	9.419e-07	4.8168e-06	1.5965e-05	3.7804e-05	4.0109e-05	1.0986e-05
58	7.4077	9.7875e-08	8.5816e-07	4.4452e-06	1.4933e-05	3.5109e-05	4.0919e-05	1.2814e-05
59	7.8045	8.546e-08	7.7452e-07	4.0603e-06	1.4176e-05	3.2431e-05	4.1509e-05	1.1002e-05
60	8.2152	7.3682e-08	6.8673e-07	3.7298e-06	1.328e-05	2.9425e-05	4.2677e-05	1.3136e-05
61	8.64	6.477e-08	6.1848e-07	3.4011e-06	1.2396e-05	2.7206e-05	4.266e-05	1.1463e-05
62	9.0792	5.531e-08	5.4066e-07	3.0693e-06	1.1543e-05	2.5521e-05	4.3977e-05	1.3086e-05
63	9.5331	4.7541e-08	4.7103e-07	2.763e-06	1.0681e-05	2.4224e-05	4.4506e-05	1.1644e-05
64	10.002	4.0248e-08	4.0902e-07	2.4495e-06	9.8848e-06	2.2907e-05	4.4746e-05	1.3433e-05

65	10.486	3.4304e-08	3.5272e-07	2.1931e-06	9.3275e-06	2.2397e-05	4.3939e-05	1.1663e-
66	10.985	2.8818e-08	3.0078e-07	1.8744e-06	8.5509e-06	2.0402e-05	4.185e-05	1.359e-
67	11.5	2.381e-08	2.5613e-07	1.6478e-06	7.8158e-06	1.8976e-05	4.0119e-05	1.2096e-
68	12.031	1.9337e-08	2.1722e-07	1.4123e-06	7.0946e-06	1.7273e-05	3.7509e-05	1.3534e-
69	12.577	1.5777e-08	1.8339e-07	1.2258e-06	6.3824e-06	1.6049e-05	3.4531e-05	1.2334e-
70	13.14	1.2851e-08	1.5121e-07	1.0425e-06	5.5804e-06	1.4547e-05	3.2353e-05	1.3968e-
71	13.72	1.0265e-08	1.2303e-07	8.8363e-07	4.9327e-06	1.3365e-05	2.8115e-05	1.2302e-
72	14.316	7.9662e-09	9.9864e-08	7.4205e-07	4.3605e-06	1.2054e-05	2.5938e-05	1.4114e-
73	14.93	6.2761e-09	8.0975e-08	6.2212e-07	3.8095e-06	1.0943e-05	2.5285e-05	1.2588e-
74	15.561	4.9435e-09	6.4741e-08	5.1199e-07	3.2737e-06	9.9099e-06	2.404e-05	1.4222e-
75	16.209	3.9203e-09	5.1535e-08	4.1842e-07	2.8741e-06	8.6393e-06	2.2512e-05	1.2829e-
76	16.875	3.0965e-09	4.1106e-08	3.4266e-07	2.4563e-06	7.837e-06	2.0909e-05	1.4707e-
77	17.559	2.4345e-09	3.245e-08	2.7852e-07	2.0728e-06	6.9943e-06	1.9281e-05	1.4198e-
78	18.261	1.8995e-09	2.5906e-08	2.2453e-07	1.8078e-06	6.1702e-06	1.838e-05	1.5235e-
79	18.982	1.4113e-09	2.0505e-08	1.7857e-07	1.4705e-06	5.4235e-06	1.6869e-05	1.4901e-
80	19.722	1.0589e-09	1.5945e-08	1.4273e-07	1.2186e-06	4.7175e-06	1.4272e-05	1.5306e-
81	20.48	7.4099e-10	1.2449e-08	1.1516e-07	1.0016e-06	4.0402e-06	1.2422e-05	1.5529e-
82	21.258	5.3869e-10	9.3255e-09	9.1737e-08	8.0834e-07	3.454e-06	1.0801e-05	1.6191e-
83	22.055	3.8509e-10	7.0555e-09	7.294e-08	6.5411e-07	2.9309e-06	9.4775e-06	1.6242e-
84	22.871	2.8297e-10	5.0721e-09	5.6197e-08	5.2093e-07	2.5225e-06	8.5883e-06	1.664e-
85	23.708	2.0045e-10	3.6381e-09	4.3661e-08	4.1573e-07	2.1339e-06	7.5613e-06	1.615e-
86	24.565	1.4573e-10	2.6214e-09	3.2964e-08	3.3317e-07	1.8413e-06	7.1167e-06	1.6518e-
87	25.442	1.0412e-10	1.9298e-09	2.4937e-08	2.6889e-07	1.5139e-06	6.1692e-06	1.6074e-
88	26.34	7.2935e-11	1.3869e-09	1.8357e-08	2.1245e-07	1.2556e-06	5.9791e-06	1.5443e-

```

mp_support_graph('cl_st_graph_title') = {'Pstationary(A,Z), asset=x, shock=color'};
mp_support_graph('cl_st_ytitle') = {'Conditional Aggregate Mass'};
mp_support_graph('cl_st_xtitle') = {'A (savings)'};
mp_support_graph('st_rowvar_name') = 'z=';
mp_support_graph('it_legend_select') = 5;
mp_support_graph('st_rounding') = '6.2f';
mp_support_graph('bl_graph_logy') = true;
mp_support_graph('cl_colors') = 'copper';
ff_graph_grid((tb_prob_az{1:end, 3:end}),' ar_st_eta_HS_grid, agrid, mp_support_graph);% Consum

```



Asset Mass by Age


```
% NaN(n_jgrid,n_agrid,n_etagrid,n_educgrid,n_marriedgrid,n_kidsgrid);
ar_permute = [3,4,5,6,1,2];
% Value Function
tb_prob_aage = ff_summ_nd_array("P(A,Z)", Phi_true, true, ["sum"], 4, 1, cl_mp_datasetdesc, ar
```

xxx	P(A,Z))	xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx							
	group	savings	sum_age_18	sum_age_19	sum_age_20	sum_age_21	sum_age_22	sum_age_23	sum_age_24
1	0	0.021837	0.002388	0.0018389	0.0064409	0.0087881	0.010534	0.0104	
2	4e-05	0	2.3862e-06	2.8257e-06	1.5227e-05	0.0005064	7.2295e-07	4.2993e-	
3	0.00032	0	3.749e-05	3.8393e-05	0.00067452	0.0013202	3.5912e-05	3.0603e-	
4	0.00108	0	0.00031485	0.0003134	0.00027522	6.8839e-05	0.00011766	1.0826e-	
5	0.00256	0	0.0012853	0.0012851	0.0015569	8.2445e-05	6.9654e-05	7.7599e-	
6	0.005	0	0.00034215	0.00051426	0.0020793	0.00015903	0.00015172	8.3842e-	
7	0.00864	0	0.0028722	0.0026464	0.00031254	0.00033461	0.00019967	0.00015	
8	0.01372	0	0.003431	0.003249	0.00031772	0.00028917	0.00016973	0.000192	
9	0.02048	0	0.00028503	0.00067598	0.00043826	0.00042815	0.00033024	0.000223	
10	0.02916	0	0.004274	0.0016076	0.00077901	0.00039339	0.00032964	0.000329	
11	0.04	0	0.0024741	0.0016863	0.0015146	0.0012551	0.00031843	0.000240	
12	0.05324	0	0.00012193	0.0017565	0.00025805	0.00022615	0.0011226	0.000258	
13	0.06912	0	0.00044563	0.00062939	0.00029172	0.00029416	0.0002238	0.00106	
14	0.08788	0	2.7692e-05	0.00011258	0.00015714	0.00018719	0.00024969	0.00023	
15	0.10976	0	6.2377e-05	8.9179e-06	7.7238e-05	0.00018316	0.00019603	0.00022	
16	0.135	0	0.00067668	0.00016484	0.00010206	0.00022055	0.00021446	0.000203	
17	0.16384	0	5.8231e-06	5.0833e-05	0.00019203	0.00023346	0.00025489	0.000201	
18	0.19652	0	3.2338e-05	4.7739e-05	0.00021932	0.00024025	0.00027886	0.000268	
19	0.23328	0	2.7827e-05	0.00062964	0.00031839	0.00035773	0.00022081	0.000230	
20	0.27436	0	3.3098e-06	0.00010242	0.00072231	0.0003499	0.00032682	0.000254	
21	0.32	0	4.0326e-05	0.00030725	0.00040297	0.00025423	0.00032626	0.000357	
22	0.37044	0	0.00023294	0.00035244	0.00045546	0.00075563	0.00031229	0.000285	
23	0.42592	0	0.00029162	0.00046138	0.00031538	0.00034379	0.0003491	0.000318	
24	0.48668	0	0.0002901	0.00049108	0.00051896	0.00038724	0.0006911	0.000389	
25	0.55296	0	0.00034886	0.00054564	0.0003598	0.00041642	0.00040996	0.000381	
26	0.625	0	0.00050916	0.00043448	0.00029056	0.00033958	0.00039168	0.000684	
27	0.70304	0	0.00039586	0.00037749	0.00035324	0.00035324	0.00032519	0.000394	
28	0.78732	0	0.00020681	0.00035156	0.00038514	0.00031557	0.00036163	0.00031	
29	0.87808	0	1.4297e-05	5.5411e-05	0.00015561	0.00033422	0.00029275	0.00031	
30	0.97556	0	1.5592e-05	6.288e-05	0.00029113	0.00022006	0.00028829	0.000292	
31	1.08	0	2.009e-06	8.4115e-05	0.00014086	0.00019115	0.00020396	0.000243	
32	1.1916	0	2.1045e-05	0.00010101	0.00015504	0.00012678	0.00017981	0.000202	
33	1.3107	0	1.4435e-06	6.9572e-05	5.1203e-05	0.00020997	0.00013959	0.000170	
34	1.4375	0	5.1689e-07	4.651e-05	7.846e-05	7.725e-05	0.00011616	0.000138	
35	1.5722	0	4.7793e-07	4.484e-06	2.0179e-05	8.927e-05	0.00010084	0.000127	
36	1.715	0	2.3446e-06	4.8601e-06	2.089e-05	4.4483e-05	0.00017191	9.4146e-	
37	1.8662	0	2.6545e-07	5.0217e-06	2.8241e-05	3.3543e-05	5.1577e-05	0.000107	
38	2.0261	0	5.4286e-07	3.5584e-06	1.983e-05	3.2623e-05	4.7562e-05	7.1725e-	
39	2.1949	0	1.5332e-06	2.2585e-05	9.6032e-06	3.0291e-05	4.164e-05	0.000130	
40	2.3728	0	4.1159e-06	1.2542e-05	1.5038e-05	2.1111e-05	4.7981e-05	4.8526e-	
41	2.56	0	4.9992e-06	9.9161e-06	2.4244e-05	2.6564e-05	3.3403e-05	5.1081e-	
42	2.7568	0	7.7981e-06	1.5369e-05	2.0348e-05	3.8876e-05	3.2557e-05	4.6239e-	
43	2.9635	0	1.0694e-05	1.9867e-05	2.6661e-05	3.3652e-05	4.0382e-05	4.2981e-	
44	3.1803	0	1.3309e-05	1.8776e-05	4.5294e-05	3.4882e-05	4.4248e-05	4.3244e-	
45	3.4074	0	1.3226e-05	2.3002e-05	3.5636e-05	3.4395e-05	4.1287e-05	4.2401e-	
46	3.645	0	3.533e-06	2.5709e-05	3.2868e-05	4.0509e-05	3.8809e-05	4.4435e-	
47	3.8934	0	1.8503e-05	2.4946e-05	2.6065e-05	3.9174e-05	3.9923e-05	4.3463e-	

```
mp_support_graph('cl_st_graph_title') = {'Pstationary(Asset,Age), asset=x, age=color'};
mp_support_graph('cl_st_ytitle') = {'Conditional Aggregate Mass'};
mp_support_graph('cl_st_xtitle') = {'A (savings)'};
mp_support_graph('st_rowvar_name') = 'age=';
mp_support_graph('it_legend_select') = 5;
```

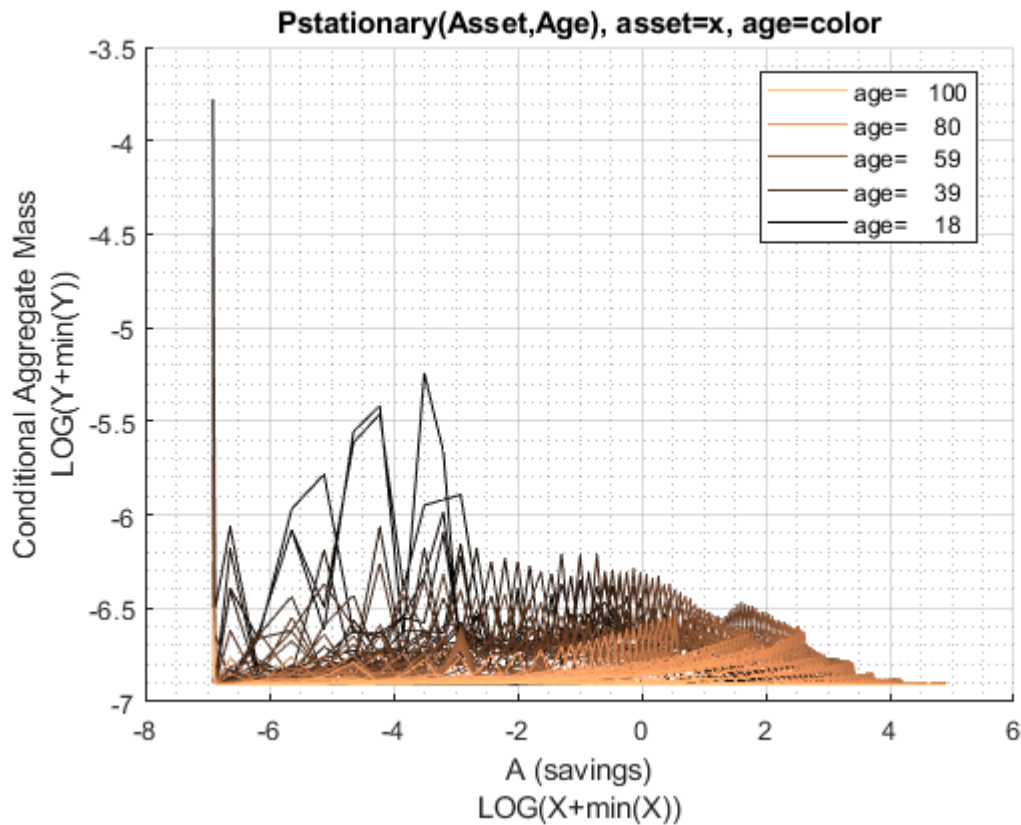


```

mp_support_graph('st_rounding') = '6.0f';
mp_support_graph('bl_graph_logy') = true;
mp_support_graph('cl_colors') = 'copper';
ff_graph_grid((tb_prob_age{1:end, 3:end}),'', age_grid, agrid, mp_support_graph);% Consumption C

```





Probability Statistics A, C and V Conditional on Ages

Where are the mass at?

```

ap_ss = mp_dsvfi_results('ap_ss');
c_ss = mp_dsvfi_results('cons_ss');
v_ss = mp_dsvfi_results('v_ss');
n_ss = mp_dsvfi_results('n_ss');

y_head_inc = mp_dsvfi_results('y_head_inc_ss');
y_spouse_inc = mp_dsvfi_results('y_spouse_inc_ss');

yshr_wage = mp_dsvfi_results('yshr_wage_ss');
yshr_SS = mp_dsvfi_results('yshr_SS_ss');
yshr_nttxss = mp_dsvfi_results('yshr_nttxss_ss');

for it_ctr=1:size(ap_ss, 1)
    if (ismember(it_ctr, round(linspace(1, size(ap_ss, 1), 3))))
        display(['age = ' num2str(age_grid(it_ctr))]);

        % construct input data
        Phi_true_age = Phi_true(it_ctr, :, :, :, :);
        ap_ss_age = ap_ss(it_ctr, :, :, :, :);
        c_ss_age = c_ss(it_ctr, :, :, :, :);
        v_ss_age = v_ss(it_ctr, :, :, :, :);
        n_ss_age = n_ss(it_ctr, :, :, :, :);

        y_head_inc_age = y_head_inc(it_ctr, :, :, :, :);

```

```

y_spouse_inc_age = y_spouse_inc(it_ctr, :, :, : ,: ,:);
yshr_wage_age = yshr_wage(it_ctr, :, :, : ,: ,:);
yshr_SS_age = yshr_SS(it_ctr, :, :, : ,: ,:);
yshr_nttxss_age = yshr_nttxss(it_ctr, :, :, : ,: ,:);

mp_cl_ar_xyz_of_s = containers.Map('KeyType','char','ValueType','any');
mp_cl_ar_xyz_of_s('ap_ss') = {ap_ss_age(:), zeros(1)};
mp_cl_ar_xyz_of_s('c_ss') = {c_ss_age(:), zeros(1)};
mp_cl_ar_xyz_of_s('v_ss') = {v_ss_age(:), zeros(1)};
mp_cl_ar_xyz_of_s('n_ss') = {n_ss_age(:), zeros(1)};
mp_cl_ar_xyz_of_s('y_head_inc') = {y_head_inc_age(:), zeros(1)};
mp_cl_ar_xyz_of_s('y_spouse') = {y_spouse_inc_age(:), zeros(1)};
mp_cl_ar_xyz_of_s('yshr_wage') = {yshr_wage_age(:), zeros(1)};
mp_cl_ar_xyz_of_s('yshr_SS') = {yshr_SS_age(:), zeros(1)};
mp_cl_ar_xyz_of_s('yshr_nttxss') = {yshr_nttxss_age(:), zeros(1)};
mp_cl_ar_xyz_of_s('ar_st_y_name') = ["ap_ss", "c_ss", "v_ss", "n_ss",...
    "y_head_inc", "y_spouse", "yshr_wage", "yshr_SS", "yshr_nttxss"];

% controls
mp_support = containers.Map('KeyType','char','ValueType','any');
mp_support('ar_fl_percentiles') = [0.01 10 25 50 75 90 99.99];
mp_support('bl_display_final') = true;
mp_support('bl_display_detail') = false;
mp_support('bl_display_drvm2outcomes') = false;
mp_support('bl_display_drvstats') = false;
mp_support('bl_display_drvm2covcor') = false;

% Call Function
mp_cl_mt_xyz_of_s = ff_simu_stats(Phi_true_age(:)/sum(Phi_true_age,'all'), mp_cl_ar_xyz_of_s);
end
end

```

age =18

xxx tb_outcomes: all stats xxx

OriginalVariableNames		ap_ss	c_ss	v_ss	n_ss	y_head_inc	y_spouse
{'mean'}	}	10.116	0.75737	-37.592	1.9854	0.84341	0.22902
{'unweighted_sum'}	}	11476	2.4405e+05	-7.8955e+05	21	4422.1	561.99
{'sd'}	}	6.9537	0.67774	55.748	1.0848	0.90505	0.5733
{'coefofvar'}	}	0.68742	0.89486	-1.483	0.54639	1.0731	2.5032
{'gini'}	}	0.32034	0.41117	-0.64344	0.268	0.41353	0.83721
{'min'}	}	1	0.035637	-868.79	1	0.038108	0
{'max'}	}	151	18.059	25.518	6	13.784	10.368
{'pYis0'}	}	0	0	0	0	0	0.52499
{'pYls0'}	}	0	0	0.8166	0	0	0
{'pYgr0'}	}	1	1	0.1834	1	1	0.47501
{'pYisMINY'}	}	0.11052	0.0014188	7.8342e-06	0.41786	0.0033703	0.52499
{'pYisMAXY'}	}	0	0	0	0.0060544	0	5.3013e-06
{'p0_01'}	}	1	0.035637	-746.63	1	0.038108	0
{'p10'}	}	1	0.24578	-86.517	1	0.14676	0
{'p25'}	}	7	0.3161	-50.751	1	0.28802	0
{'p50'}	}	9	0.51551	-25.389	2	0.56523	0
{'p75'}	}	11	0.88958	-5.527	3	1.1092	0.23956
{'p90'}	}	23	1.5797	6.0744	4	2.1768	0.8323
{'p99_99'}	}	52	6.8857	23.692	6	8.3836	8.6488
{'fl_cov_ap_ss'}	}	48.354	1.9167	116.57	0.29345	1.7747	3.1074
{'fl_cov_ap_ss'}	}	1	0.4067	0.3007	0.038901	0.28199	0.77947
{'fl_cov_c_ss'}	}	1.9167	0.45934	20.369	0.067217	0.59824	0.081697

{'fl_cor_c_ss' }	0.4067	1	0.5391	0.091423	0.9753	0.21026
{'fl_cov_v_ss' }	116.57	20.369	3107.8	2.9005	24.615	4.9476
{'fl_cor_v_ss' }	0.3007	0.5391	1	0.047962	0.48787	0.15481
{'fl_cov_n_ss' }	0.29345	0.067217	2.9005	1.1768	-1.236e-17	0.13364
{'fl_cor_n_ss' }	0.038901	0.091423	0.047962	1	-1.2589e-17	0.21488
{'fl_cov_y_head_inc' }	1.7747	0.59824	24.615	-1.236e-17	0.81911	0.021751
{'fl_cor_y_head_inc' }	0.28199	0.9753	0.48787	-1.2589e-17	1	0.04192
{'fl_cov_y_spouse' }	3.1074	0.081697	4.9476	0.13364	0.021751	0.32867
{'fl_cor_y_spouse' }	0.77947	0.21026	0.15481	0.21488	0.04192	1
{'fl_cov_yshr_wage' }	3.7471e-30	2.4421e-31	-9.1828e-31	1.0754e-30	8.1847e-31	7.0393e-32
{'fl_cor_yshr_wage' }	4.0447e-16	2.7046e-16	-1.2364e-17	7.4411e-16	6.788e-16	9.2163e-17
{'fl_cov_yshr_SS' }	0	0	0	0	0	0
{'fl_cor_yshr_SS' }	NaN	NaN	NaN	NaN	NaN	NaN
{'fl_cov_yshr_nttxss' }	0.16611	0.021334	1.8609	0.0077776	0.025219	0.0090401
{'fl_cor_yshr_nttxss' }	0.58487	0.77071	0.81728	0.17554	0.68223	0.38607
{'fracByP0_01' }	0.010925	6.6761e-05	0.0030452	0.21046	0.00015228	0
{'fracByP10' }	0.010925	0.050401	0.44014	0.21046	0.019229	0
{'fracByP25' }	0.148	0.072459	0.71161	0.21046	0.096342	0
{'fracByP50' }	0.28531	0.21889	0.94673	0.53024	0.29663	0
{'fracByP75' }	0.60536	0.47077	1.0363	0.77109	0.59361	0.13003
{'fracByP90' }	0.758	0.70215	1.0323	0.92834	0.84502	0.34306
{'fracByP99_99' }	0.99975	0.99993	1	1	1	0.99814

age =59

xxx tb_outcomes: all stats xxx

OriginalVariableNames	ap_ss	c_ss	v_ss	n_ss	y_head_inc	y_spouse	y
{'mean' }	55.659	1.287	-12.919	1.7239	1.8545	0.45057	
{'unweighted_sum' }	11476	2.6894e+05	-1.1138e+05	21	13268	1069.5	
{'sd' }	23.095	1.0938	20.385	0.90777	2.0429	1.1205	
{'coefofvar' }	0.41494	0.84994	-1.5779	0.52659	1.1016	2.4867	
{'gini' }	0.22938	0.40011	-0.80515	0.23461	0.47957	0.83345	
{'min' }	1	0.055882	-235.34	1	0.059541	0	
{'max' }	151	32.48	14.759	6	23.47	20.112	
{'pYis0' }	0	0	0	0	0	0.52499	
{'pYls0' }	0	0	0.74277	0	0	0	
{'pYgr0' }	1	1	0.25723	1	1	0.47501	
{'pYisMINY' }	0.0037896	2.9499e-05	3.9537e-07	0.48835	9.9096e-05	0.52499	
{'pYisMAXY' }	4.9199e-06	2.3292e-08	0	0.0036816	2.0186e-06	4.8438e-06	
{'p0_01' }	1	0.05663	-137.64	1	0.059554	0	
{'p10' }	28	0.31379	-41.113	1	0.39098	0	
{'p25' }	41	0.59299	-18.867	1	0.6458	0	
{'p50' }	55	1.065	-7.2226	2	1.1351	0	
{'p75' }	70	1.6559	0.35778	2	2.1525	0.48062	
{'p90' }	85	2.4892	6.453	3	4.19	1.7443	
{'p99_99' }	146	15.179	14.69	6	22.847	16.777	
{'fl_cov_ap_ss' }	533.38	21.832	417.21	2.9474	37.948	5.9801	
{'fl_cor_ap_ss' }	1	0.86423	0.88619	0.14059	0.80428	0.2311	
{'fl_cov_c_ss' }	21.832	1.1965	14.391	0.23796	2.0766	0.27801	
{'fl_cor_c_ss' }	0.86423	1	0.64539	0.23965	0.92925	0.22684	
{'fl_cov_v_ss' }	417.21	14.391	415.54	3.8082	23.854	4.7175	
{'fl_cor_v_ss' }	0.88619	0.64539	1	0.2058	0.5728	0.20654	
{'fl_cov_n_ss' }	2.9474	0.23796	3.8082	0.82404	0.062177	0.2771	
{'fl_cor_n_ss' }	0.14059	0.23965	0.2058	1	0.033527	0.27244	
{'fl_cov_y_head_inc' }	37.948	2.0766	23.854	0.062177	4.1736	0.1726	
{'fl_cor_y_head_inc' }	0.80428	0.92925	0.5728	0.033527	1	0.075404	
{'fl_cov_y_spouse' }	5.9801	0.27801	4.7175	0.2771	0.1726	1.2554	
{'fl_cor_y_spouse' }	0.2311	0.22684	0.20654	0.27244	0.075404	1	
{'fl_cov_yshr_wage' }	-1.3386	-0.041425	-1.1118	-0.0063156	-0.054208	0.014235	
{'fl_cor_yshr_wage' }	-0.6493	-0.42426	-0.61101	-0.07794	-0.29725	0.14232	
{'fl_cov_yshr_SS' }	0	0	0	0	0	0	
{'fl_cor_yshr_SS' }	NaN	NaN	NaN	NaN	NaN	NaN	
{'fl_cov_yshr_nttxss' }	0.76653	0.028255	0.72181	0.0084863	0.047728	0.014539	
{'fl_cor_yshr_nttxss' }	0.88794	0.69107	0.94731	0.2501	0.62501	0.34714	
{'fracByP0_01' }	6.8085e-05	5.4514e-06	0.0013821	0.28329	4.1425e-06	0	

{'fracByP10' }	0.031672	0.019903	0.46352	0.28329	0.013283	0
{'fracByP25' }	0.1219	0.075667	0.79865	0.28329	0.054393	0
{'fracByP50' }	0.34388	0.22765	1.0532	0.72028	0.1806	0
{'fracByP75' }	0.62522	0.48161	1.1105	0.72028	0.41873	0.15283
{'fracByP90' }	0.8294	0.71297	1.0711	0.85389	0.65167	0.3418
{'fracByP99_99' }	0.9998	0.99881	1	1	0.99935	0.99834
age =100						
xxx tb_outcomes: all stats xxx						
OriginalVariableNames	ap_ss	c_ss	v_ss	n_ss	y_head_inc	y_spouse
<hr/>						
{'mean' }	1	0.33746	-3.2579	1.4797	0.23579	0.09988
{'unweighted_sum' }	1	2.8049e+05	789.51	21	483.8	33.1
{'sd' }	1.088e-14	0.23923	1.2254	0.50567	0.022052	0.2444
{'coefofvar' }	1.088e-14	0.70891	-0.37615	0.34173	0.093527	2.4469
{'gini' }	0	0.29996	-0.20031	0.12034	0.044484	0.78724
{'min' }	1	0.19737	-11.197	1	0.22	0
{'max' }	1	141.61	0.99282	6	5.666	3.0753
{'pYis0' }	0	0	0	0	0	0.52499
{'pYls0' }	0	0	0.99204	0	0	0
{'pYgr0' }	1	1	0.007965	1	1	0.47501
{'pYisMINY' }	1	0.34474	1.4552e-10	0.5232	0.48869	0.52499
{'pYisMAXY' }	1	0	0	4.2206e-08	0	1.0335e-08
{'p0_01' }	1	0.19737	-7.038	1	0.22	0
{'p10' }	1	0.19737	-4.0665	1	0.22	0
{'p25' }	1	0.19737	-4.0665	1	0.22	0
{'p50' }	1	0.23607	-3.7707	1	0.2202	0
{'p75' }	1	0.36676	-2.6758	2	0.266	0.10166
{'p90' }	1	0.59408	-1.2803	2	0.26717	0.48427
{'p99_99' }	1	2.9028	0.51281	4	0.31843	2.9082
{'fl_cov_ap_ss' }	1.1838e-28	4.4139e-31	3.5754e-30	4.121e-29	2.8489e-30	-3.2619e-31
{'fl_cor_ap_ss' }	1	1.6958e-16	2.6816e-16	7.4904e-15	1.1874e-14	-1.2267e-16
{'fl_cov_c_ss' }	4.4139e-31	0.057229	0.23842	0.059118	0.0016668	0.050594
{'fl_cor_c_ss' }	1.6958e-16	1	0.81327	0.48871	0.31595	0.86534
{'fl_cov_v_ss' }	3.5754e-30	0.23842	1.5017	0.20689	0.012148	0.17973
{'fl_cor_v_ss' }	2.6816e-16	0.81327	1	0.33387	0.44951	0.60012
{'fl_cov_n_ss' }	4.121e-29	0.059118	0.20689	0.2557	0.0018516	0.052581
{'fl_cor_n_ss' }	7.4904e-15	0.48871	0.33387	1	0.16604	0.42546
{'fl_cov_y_head_inc' }	2.8489e-30	0.0016668	0.012148	0.0018516	0.0004863	0.00064389
{'fl_cor_y_head_inc' }	1.1874e-14	0.31595	0.44951	0.16604	1	0.11947
{'fl_cov_y_spouse' }	-3.2619e-31	0.050594	0.17973	0.052581	0.00064389	0.059731
{'fl_cor_y_spouse' }	-1.2267e-16	0.86534	0.60012	0.42546	0.11947	1
{'fl_cov_yshr_wage' }	-6.2277e-32	0.040102	0.18489	0.087766	0.00067273	0.043238
{'fl_cor_yshr_wage' }	-2.5076e-17	0.73439	0.66096	0.76038	0.13365	0.77505
{'fl_cov_yshr_SS' }	-3.036e-30	-0.041567	-0.19392	-0.089099	-0.00074396	-0.043211
{'fl_cor_yshr_SS' }	-1.2162e-15	-0.75734	-0.68973	-0.768	-0.14704	-0.77064
{'fl_cov_yshr_nttxss' }	-3.8383e-30	0.045721	0.21331	0.096069	0.00089733	0.047619
{'fl_cor_yshr_nttxss' }	-1.4131e-15	0.76558	0.69726	0.76104	0.163	0.78049
{'fracByP0_01' }	1	0.20164	0.00049502	0.35357	0.45597	0
{'fracByP10' }	1	0.20164	0.5347	0.35357	0.45597	0
{'fracByP25' }	1	0.20164	0.5347	0.35357	0.45597	0
{'fracByP50' }	1	0.31034	0.64676	0.35357	0.46775	0
{'fracByP75' }	1	0.52813	0.89014	0.99419	0.87014	0.19248
{'fracByP90' }	1	0.73972	0.97784	0.99419	0.88578	0.62777
{'fracByP99_99' }	1	0.9992	1	0.99999	0.99992	0.9996