

Distribution Exact Savings Choices

This is the example vignette for function: [snw_ds_main](#) from the [PrjOptiSNW Package](#). This function solves for vfi and gets distribution induced by policy functions and exogenous distributions. Looped to get distribution, but uses bisect vec for VFI.

Test SNW_DS_MAIN Defaults

Call the function with testing defaults.

```
mp_params = snw_mp_param('default_docdense');
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(mp_params, mp_controls);
```

Completed SNW_VFI_MAIN_BISEC_VEC;SNW_MP_PARAM=default_docdense;SNW_MP_CONTROL=default_test;time=551.0862

XX

CONTAINER NAME: mp_outcomes ND Array (Matrix etc)

XX

	i	idx	ndim	numel	rowN	colN	sum	mean	std	coefvari
	—	—	—	—	—	—	—	—	—	—
V_VFI	1	1	6	4.37e+07	83	5.265e+05	-8.6673e+08	-19.834	28.177	-1.4206
ap_VFI	2	2	6	4.37e+07	83	5.265e+05	1.4164e+09	32.412	36.8	1.1354
cons_VFI	3	3	6	4.37e+07	83	5.265e+05	2.131e+08	4.8764	8.3268	1.7076

xxx TABLE:V_VFI XXXXXXXXXXXXXXXXXXXXXXX

	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499
	—	—	—	—	—	—	—	—	—
r1	-376.05	-375.66	-373.17	-367.4	-358.05	-6.68	-6.5297	-6.3792	-6.2274
r2	-363.8	-363.41	-360.93	-355.25	-346.25	-6.4892	-6.3437	-6.1974	-6.0495
r3	-351.75	-351.36	-348.9	-343.44	-334.9	-6.2948	-6.1538	-6.0116	-5.8671
r4	-339.81	-339.45	-337.16	-332.06	-324.04	-6.095	-5.9584	-5.82	-5.6786
r5	-328.99	-328.65	-326.51	-321.72	-314.17	-5.9054	-5.7725	-5.6372	-5.4986
r79	-14.033	-14.02	-13.926	-13.689	-13.287	-0.22848	-0.21775	-0.20768	-0.19824
r80	-12.564	-12.55	-12.457	-12.22	-11.818	-0.17427	-0.16611	-0.15842	-0.15117
r81	-10.778	-10.764	-10.671	-10.434	-10.032	-0.11927	-0.11368	-0.10843	-0.10346
r82	-8.4226	-8.4089	-8.3155	-8.0786	-7.6766	-0.06597	-0.06284	-0.059924	-0.057184
r83	-5.0665	-5.0529	-4.9595	-4.7226	-4.3206	-0.020968	-0.019972	-0.019038	-0.018161

xxx TABLE:ap_VFI XXXXXXXXXXXXXXXXXXXXXXX

	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	c526500
	—	—	—	—	—	—	—	—	—	—
r1	0	0	0.0005656	0.0075134	0.022901	114.76	120.42	126.29	132.39	138.81
r2	0	0	0.00051498	0.0065334	0.021549	114.87	120.54	126.42	132.55	138.97
r3	0	0	0.00051498	0.0049294	0.019875	114.98	120.67	126.57	132.72	139.13
r4	0	0	0.00051498	0.0047937	0.019672	115.74	121.44	127.36	133.52	139.94
r5	0	0	0.00048517	0.0046683	0.019484	116.51	122.22	128.16	134.34	140.76
r79	0	0	0	0	0.00051498	81.091	85.68	90.325	94.371	98.41
r80	0	0	0	0	0	76.669	80.55	84.292	88.029	91.682
r81	0	0	0	0	0	68.313	71.52	74.459	77.816	81.096
r82	0	0	0	0	0	50.126	53.467	56.953	58.728	60.587
r83	0	0	0	0	0	0	0	0	0	0

xxx TABLE:cons_VFI XXXXXXXXXXXXXXXXXXXXXXX

	c1	c2	c3	c4	c5	c526496	c526497	c526498	c526499	c
r1	0.036717	0.037251	0.040426	0.04363	0.048012	9.6396	9.8066	9.9533	10.06	
r2	0.036717	0.037251	0.040477	0.04461	0.049364	9.8014	9.9571	10.088	10.177	1
r3	0.036717	0.037251	0.040477	0.046214	0.051039	9.9664	10.108	10.22	10.287	1
r4	0.038144	0.038678	0.041903	0.047776	0.052666	10.118	10.244	10.339	10.388	1
r5	0.039534	0.040068	0.043323	0.04929	0.054241	10.258	10.369	10.446	10.483	1
r79	0.19737	0.19791	0.20163	0.21175	0.23093	35.811	37.046	38.418	40.587	4
r80	0.19737	0.19791	0.20163	0.21175	0.23145	40.207	42.15	44.426	46.904	4
r81	0.19737	0.19791	0.20163	0.21175	0.23145	48.541	51.158	54.236	57.094	6
r82	0.19737	0.19791	0.20163	0.21175	0.23145	66.71	69.193	71.724	76.164	
r83	0.19737	0.19791	0.20163	0.21175	0.23145	116.82	122.65	128.66	134.88	1

Completed SNW_DS_MAIN;SNW_MP_PARAM=default_docdense;SNW_MP_CONTROL=default_test;time=2153.0961

```
% [Phi_true,Phi_adj] = snw_ds_main(mp_params, mp_controls);
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	p
a_ss	4.3602	2228	6.8796	1.5778	0.6755	0	135	0
ap_ss	4.4621	5.3216e+08	6.9169	1.5501	0.67638	0	163.73	0
cons_ss	1.0635	5.0787e+07	0.6938	0.65237	0.33936	0.036717	141.61	
v_ss	-36.615	-4.0773e+08	24.55	-0.67049	-0.33945	-615.77	-0.0071775	
n_ss	2.3554	21	1.4375	0.61029	0.3128	1	6	
y_all	1.4189	8.353e+07	1.4929	1.0521	0.47667	0	50.873	0
y_head_inc	1.1081	1.9253e+06	1.013	0.91419	0.42164	0.038108	24.357	
y_head_earn	0.88655	19732	0.92804	1.0468	0.53121	0	18.957	
y_spouse_inc	0.35797	4.827e+05	0.95437	2.6661	0.85269	0	26.627	0
yshr_interest	0.12865	3.8438e+06	0.17577	1.3663	0.65781	0	0.99299	0
yshr_wage	0.77402	8.8881e+06	0.33679	0.43512	0.2062	0	1	0
yshr_SS	0.097329	29012	0.2266	2.3282	0.91382	0	1	
yshr_tax	0.17833	2.8338e+06	0.035661	0.19998	0.11386	0.036506	0.2552	
yshr_nttxss	0.080996	2.8048e+06	0.24691	3.0485	1.2592	-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;'));
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

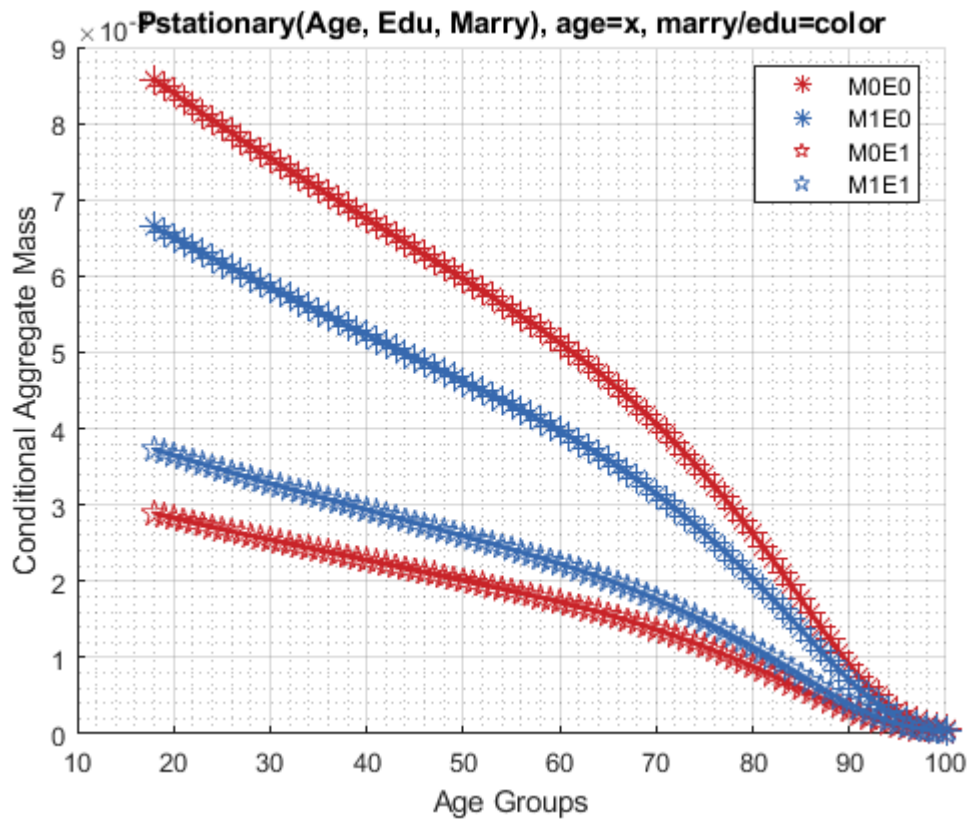
```

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);

```



Kids and Marry By Age Mass

```
% NaN(n_jgrid,n_agrid,n_etagrid,n_educgrid,n_marriedgrid,n_kidsgrid);
ar_permute = [2,3,4,1,6,5];
% Value Function
tb_prob_amarrykids = ff_summ_nd_array("P(Age, Kids, Marry)", Phi_true, true, ["sum"], 3, 1, cl)
```

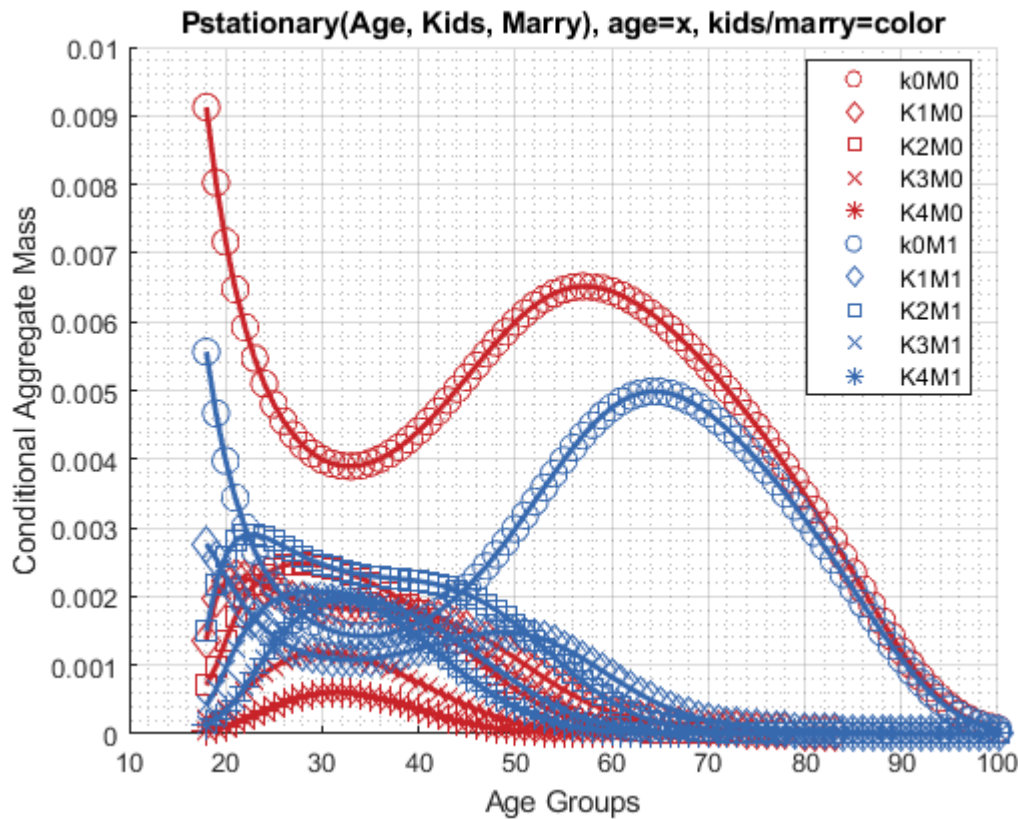
xxx	P(Age, Kids, Marry))	xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx								
group	kids	marry	sum_age_18	sum_age_19	sum_age_20	sum_age_21	sum_age_22	sum_age_23	su	
1	1	0	0.0091249	0.0080278	0.0071652	0.0064765	0.0059205	0.0054683	0	
2	2	0	0.0013699	0.0019743	0.0022187	0.0022858	0.0022687	0.0022149	0	
3	3	0	0.00071266	0.00098425	0.0013537	0.0016929	0.0019639	0.0021645	0	
4	4	0	0.00020622	0.00027865	0.00037326	0.00049476	0.00062818	0.00075864	0	
5	5	0	5.0761e-05	7.8715e-05	0.000113	0.00015485	0.00020534	0.00026306	0	
6	1	1	0.0055624	0.0046679	0.0039774	0.0034368	0.0030088	0.0026667	0	
7	2	1	0.0027682	0.0025539	0.0023005	0.0020611	0.0018525	0.0016773	0	
8	3	1	0.0014982	0.0021823	0.0025943	0.0028096	0.002896	0.0029031	0	
9	4	1	0.00041197	0.00064648	0.00095224	0.0012491	0.0015009	0.0016975	0	
10	5	1	0.00013221	0.0002132	0.00033097	0.00049097	0.00068255	0.0008901	0	

```
mp_support_graph('cl_st_graph_title') = {'Pstationary(Age, Kids, Marry), age=x, kids/marry=col...
mp_support_graph('cl_st_ytitle') = {'Conditional Aggregate Mass'};
ar_row_grid = [...
    "k0M0", "k1M0", "k2M0", "k3M0", "k4M0", ...
    "k0M1", "k1M1", "k2M1", "k3M1", "k4M1"];
mp_support_graph('cl_scatter_shapes') = {...
    'o', 'd', 's', 'x', '*', ...
```

```

'o', 'd', 's', 'x', '*'};
mp_support_graph('cl_colors') = {...
'red', 'red', 'red', 'red', 'red'...
'blue', 'blue', 'blue', 'blue', 'blue'};
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

```

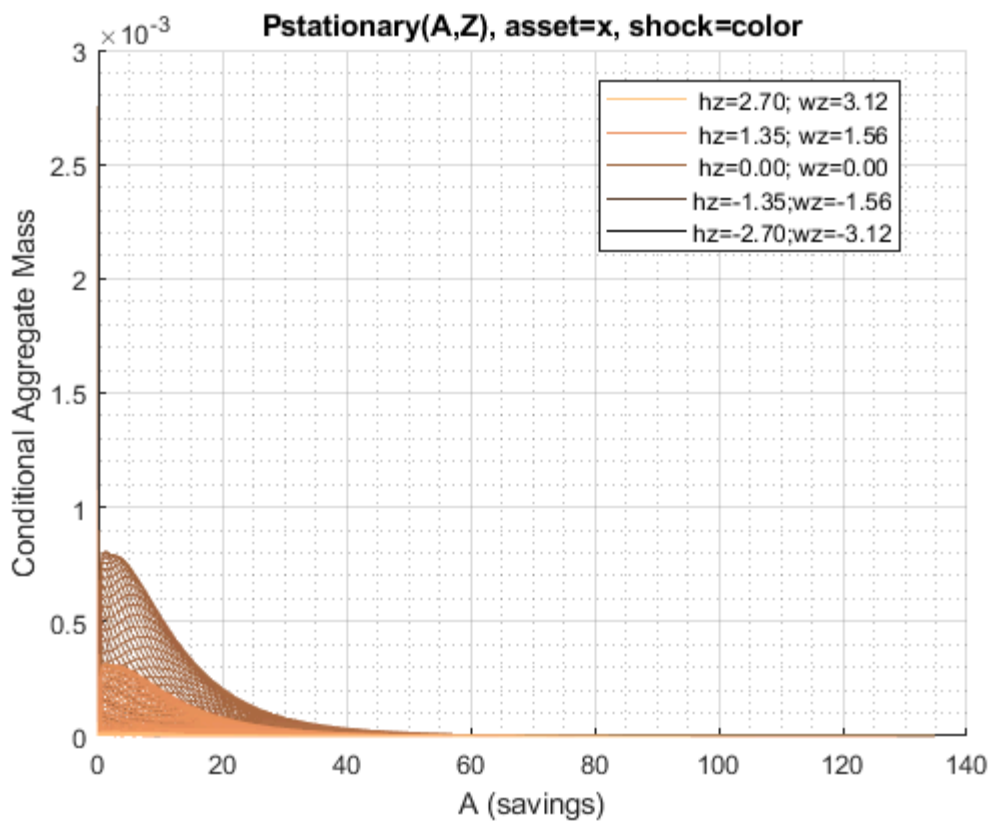
xxx P(A,Z)	xxxxxxxxxxxxxxxxxxxxxxxxxxxxxxxx								
group	savings	sum_eta_1	sum_eta_2	sum_eta_3	sum_eta_4	sum_eta_5	sum_eta_6	sum_eta_7	sum_eta_8

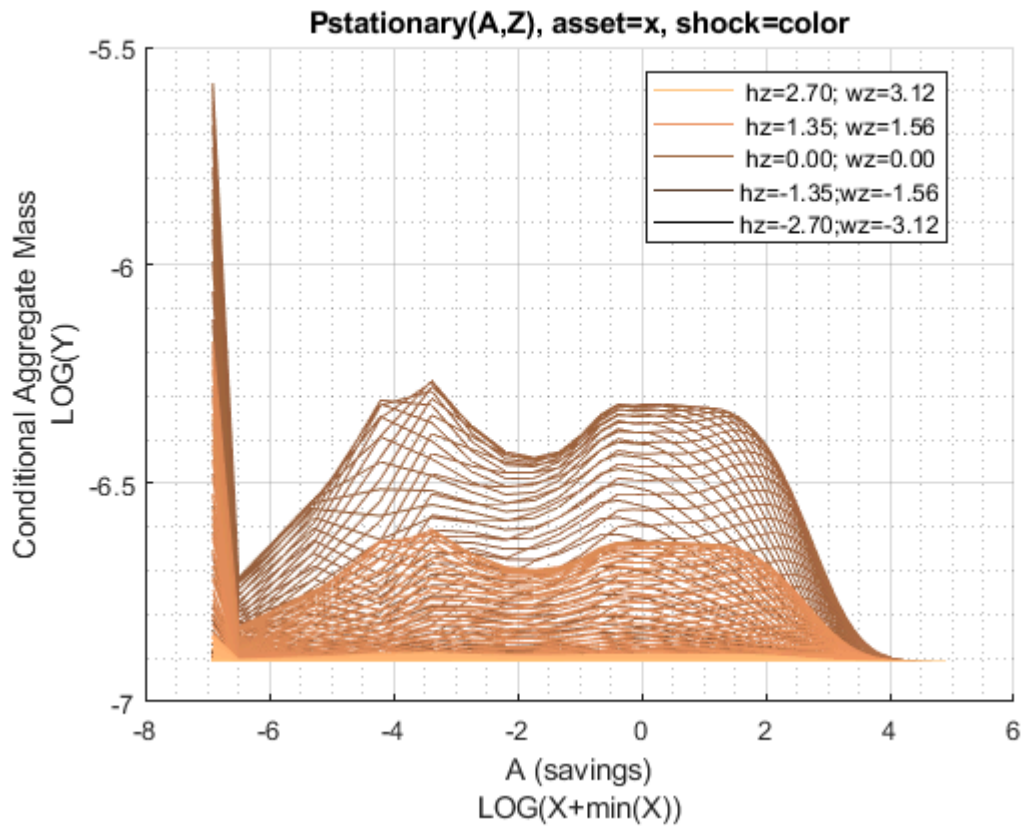
1	0	1.6729e-07	1.4316e-07	2.1759e-07	3.1665e-07	4.509e-07	6.3359e-07	8.797
2	0.00051498	3.7493e-10	3.6049e-10	6.2649e-10	1.1273e-09	2.1799e-09	4.4826e-09	9.374
3	0.0041199	7.5745e-10	6.6694e-10	1.0474e-09	1.5976e-09	2.4182e-09	3.6548e-09	5.511
4	0.013905	1.6314e-09	1.4169e-09	2.1927e-09	3.2778e-09	4.8429e-09	7.1372e-09	1.051
5	0.032959	5.5034e-09	4.7405e-09	7.269e-09	1.0722e-08	1.5557e-08	2.2396e-08	3.204
6	0.064373	6.5761e-09	5.6858e-09	8.729e-09	1.2871e-08	1.8634e-08	2.6723e-08	3.803

```

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_age = ff_summ_nd_array("P(A,Z)", Phi_true, true, ["sum"], 4, 1, cl_mp_datasetdesc, ar
```

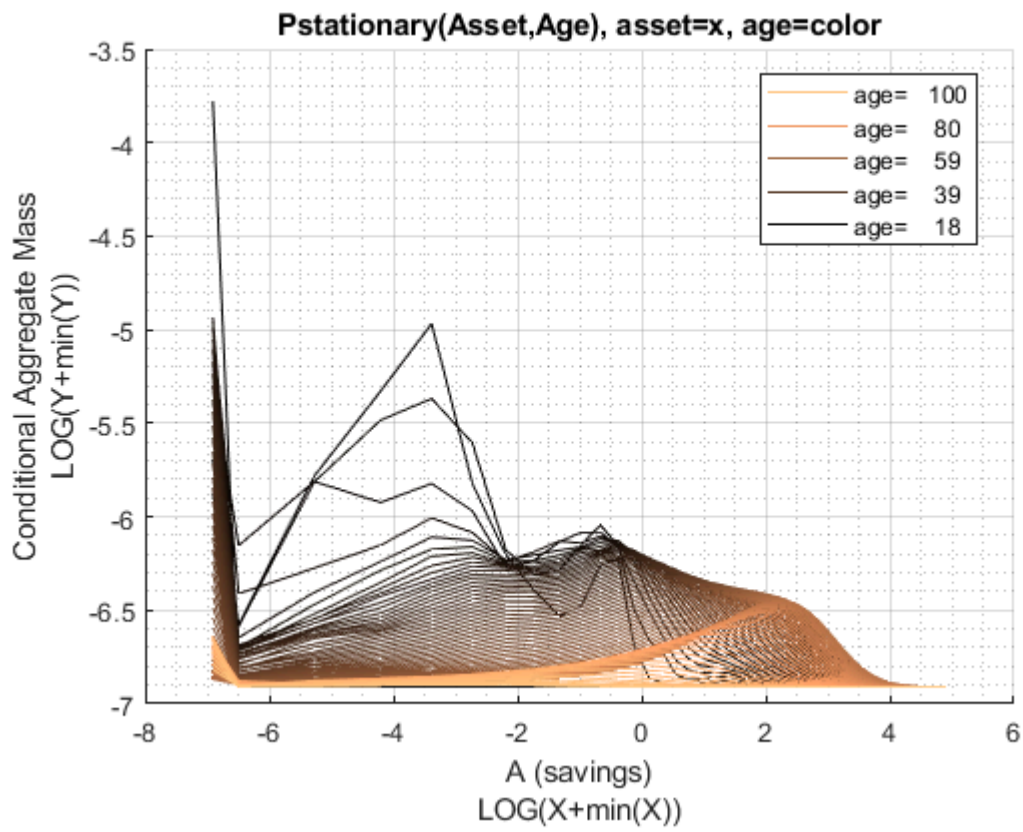
xxx	P(A,Z)	xxxxxxxxxxxxxxxxxxxxxxxxxxxx							
group	savings	sum_age_18	sum_age_19	sum_age_20	sum_age_21	sum_age_22	sum_age_23	sum_a	
1	0	0.021837	0.0023506	0.0017989	0.0039364	0.0058286	0.0061897	0.00	
2	0.00051498	0	0.00039478	0.00037898	0.0011264	0.00064408	0.00030008	0.00	
3	0.0041199	0	0.0020814	0.0019874	0.0019922	0.00088345	0.00065467	0.00	
4	0.013905	0	0.0038512	0.0031528	0.0016753	0.0011288	0.00094985	0.00	
5	0.032959	0	0.0059559	0.0036616	0.0019605	0.0014625	0.0012232	0.00	
6	0.064373	0	0.0019624	0.0026806	0.0015589	0.0012782	0.0011853	0.00	
7	0.11124	0	0.0010231	0.0010755	0.0008941	0.0009494	0.0009688	0.00	
8	0.17664	0	0.00067266	0.00082141	0.0009652	0.0010565	0.0010188	0.00	
9	0.26367	0	0.00045811	0.00086333	0.0011647	0.0011767	0.0010716	0.00	
10	0.37542	0	0.00053656	0.001129	0.0012812	0.0011528	0.0011099	0.00	
11	0.51498	0	0.00090841	0.0013729	0.0012822	0.0012345	0.001202	0.00	
12	0.68544	0	0.00097635	0.0011151	0.0011024	0.001139	0.0011822	0.00	
13	0.88989	0	0.00023558	0.00050592	0.00075138	0.00094942	0.001009	0.00	
14	1.1314	0	4.591e-05	0.00027667	0.00049327	0.00061366	0.00072435	0.00	
15	1.4131	0	1.7547e-05	0.00019582	0.00030348	0.00040728	0.00051335	0.00	
16	1.7381	0	8.2268e-06	6.7346e-05	0.00015086	0.00025852	0.00035749	0.00	
17	2.1094	0	6.1503e-06	3.6395e-05	9.6376e-05	0.00016375	0.00023657	0.00	
18	2.5301	0	1.345e-05	3.725e-05	7.4002e-05	0.00012113	0.00017621	0.00	
19	3.0034	0	2.2526e-05	4.8231e-05	7.8125e-05	0.00011085	0.00014554	0.00	
20	3.5323	0	2.9888e-05	5.5596e-05	8.1028e-05	0.00010487	0.00012962	0.00	
21	4.1199	0	3.0433e-05	5.4594e-05	7.2792e-05	9.1925e-05	0.00011221	0.00	

22	4.7693	0	2.0409e-05	3.7846e-05	5.5558e-05	7.2536e-05	8.9826e-05	0.000
23	5.4836	0	5.1452e-06	1.8425e-05	3.2883e-05	4.8468e-05	6.5255e-05	8.271
24	6.2658	0	7.3282e-07	5.3334e-06	1.4182e-05	2.7049e-05	4.1491e-05	5.716
25	7.1191	0	1.062e-07	1.2922e-06	4.9633e-06	1.2247e-05	2.3108e-05	3.599
26	8.0466	0	1.7779e-08	5.0549e-07	2.0442e-06	5.3225e-06	1.1275e-05	2.02
27	9.0514	0	3.0263e-09	3.0488e-07	1.0981e-06	2.7972e-06	5.7624e-06	1.071
28	10.136	0	1.2227e-10	1.6491e-07	5.5442e-07	1.5926e-06	3.3079e-06	6.054
29	11.305	0	0	4.8394e-08	2.2296e-07	8.1497e-07	1.9096e-06	3.606
30	12.56	0	0	9.3997e-09	7.016e-08	3.1478e-07	1.0131e-06	2.122
31	13.905	0	0	1.808e-09	2.0992e-08	9.9385e-08	4.7981e-07	1.196
32	15.342	0	0	4.1404e-10	6.2716e-09	3.4866e-08	1.9071e-07	6.339
33	16.875	0	0	9.9687e-11	1.6909e-09	1.3108e-08	6.2084e-08	3.079
34	18.507	0	0	2.1381e-11	4.7515e-10	4.3065e-09	2.1839e-08	1.274
35	20.241	0	0	8.897e-13	1.3917e-10	1.2185e-09	8.3899e-09	4.17
36	22.08	0	0	0	2.966e-11	3.6653e-10	2.8415e-09	1.411
37	24.027	0	0	0	3.6991e-12	1.1419e-10	8.4508e-10	5.449
38	26.085	0	0	0	7.8046e-13	2.4029e-11	2.7382e-10	1.848
39	28.258	0	0	0	1.7968e-13	4.0593e-12	8.1361e-11	5.863
40	30.548	0	0	0	8.7684e-15	1.0642e-12	1.7711e-11	1.896
41	32.959	0	0	0	0	1.9771e-13	3.6622e-12	5.469
42	35.493	0	0	0	0	1.5011e-14	9.2977e-13	1.217
43	38.154	0	0	0	0	2.3721e-15	1.525e-13	2.847
44	40.945	0	0	0	0	3.0185e-16	1.7487e-14	6.67
45	43.868	0	0	0	0	6.4297e-18	3.2969e-15	1.069
46	46.928	0	0	0	0	0	3.4855e-16	1.612
47	50.126	0	0	0	0	...		

```

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_aage{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
```

```

end
end
mp_cl_mt_xyz_of_s = ff_simu_stats(Phi_true_age(:)/sum(Phi_true_age,'all'), mp_cl_ar_xyz

```

age =18

xxx tb_outcomes: all stats xxx

OriginalVariableNames	ap_ss	c_ss	v_ss	n_ss	y_head_inc	y_spouse
{'mean' }	0.13209	0.63361	-59.72	1.9854	0.71265	0.22832
{'unweighted_sum' }	1.0935e+07	8.5257e+05	-1.1176e+07	21	15541	5033.3
{'sd' }	0.34847	0.37861	29.967	1.0848	0.54567	0.56949
{'coefofvar' }	2.6381	0.59754	-0.50178	0.54639	0.76569	2.4943
{'gini' }	0.7705	0.3109	-0.25126	0.268	0.36259	0.84016
{'min' }	0	0.036717	-615.77	1	0.038108	0
{'max' }	145.08	10.204	-3.7499	6	13.784	10.368
{'pYis0' }	0.10805	0	0	0	0	0.52499
{'pYls0' }	0	0	1	0	0	0
{'pYgr0' }	0.89195	1	0	1	1	0.47501
{'pYisMINY' }	0.10805	1.3288e-05	5.8837e-08	0.41786	2.5312e-05	0.52499
{'pYisMAXY' }	0	0	0	0.0060544	0	3.9814e-08
{'p0_01' }	0	0.047727	-352.03	1	0.046651	0
{'p10' }	0	0.24819	-96.425	1	0.23528	0
{'p25' }	0.012214	0.36957	-70.656	1	0.35258	0
{'p50' }	0.032959	0.55272	-52.866	2	0.56523	0
{'p75' }	0.076248	0.80075	-39.739	3	0.90612	0.24502
{'p90' }	0.4782	1.1197	-31.147	4	1.3579	0.84753
{'p99_99' }	5.4534	3.6548	-10.999	6	6.8484	8.2655
{'fl_cov_ap_ss' }	0.12143	0.055156	2.4756	0.02663	0.050357	0.18246
{'fl_cor_ap_ss' }	1	0.41805	0.23707	0.070443	0.26483	0.91943
{'fl_cov_c_ss' }	0.055156	0.14335	8.0725	0.076682	0.18653	0.071672
{'fl_cor_c_ss' }	0.41805	1	0.7115	0.1867	0.90288	0.33241
{'fl_cov_v_ss' }	2.4756	8.0725	898	0.45095	10.05	3.4951
{'fl_cor_v_ss' }	0.23707	0.7115	1	0.013872	0.61462	0.2048
{'fl_cov_n_ss' }	0.02663	0.076682	0.45095	1.1768	-4.12e-18	0.13323
{'fl_cor_n_ss' }	0.070443	0.1867	0.013872	1	-6.96e-18	0.21565
{'fl_cov_y_head_inc' }	0.050357	0.18653	10.05	-4.12e-18	0.29776	0.010455
{'fl_cor_y_head_inc' }	0.26483	0.90288	0.61462	-6.96e-18	1	0.033645
{'fl_cov_y_spouse' }	0.18246	0.071672	3.4951	0.13323	0.010455	0.32432
{'fl_cor_y_spouse' }	0.91943	0.33241	0.2048	0.21565	0.033645	1
{'fl_cov_yshr_wage' }	7.6315e-33	1.7081e-32	-2.0646e-31	3.5437e-31	6.4579e-33	2.907e-33
{'fl_cor_yshr_wage' }	9.8628e-17	2.0318e-16	-3.1028e-17	1.4712e-15	5.3299e-17	2.2988e-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' }	0.0057593	0.011163	0.86319	0.007516	0.01319	0.008688
{'fl_cor_yshr_nttxss' }	0.48714	0.86903	0.84902	0.20421	0.71249	0.44966
{'fracByP0_01' }	0	7.1734e-06	0.00073274	0.21046	7.788e-06	0
{'fracByP10' }	0	0.030664	0.21367	0.21046	0.027495	0
{'fracByP25' }	0.0067284	0.10372	0.4286	0.21046	0.092606	0
{'fracByP50' }	0.046851	0.29072	0.67444	0.53024	0.26377	0
{'fracByP75' }	0.13176	0.54795	0.87012	0.77109	0.5245	0.12959
{'fracByP90' }	0.35932	0.76949	0.95644	0.92834	0.74403	0.33886
{'fracByP99_99' }	0.99576	0.99938	0.99998	1	0.99912	0.99627

age =59

xxx tb_outcomes: all stats xxx

OriginalVariableNames	ap_ss	c_ss	v_ss	n_ss	y_head_inc	y_spouse
{'mean' }	9.6978	1.2003	-27.032	1.7239	1.6127	0.44918
{'unweighted_sum' }	1.1254e+07	1.0744e+06	-5.7273e+06	21	45380	9578.2
{'sd' }	9.5091	0.76817	15.51	0.90777	1.276	1.113
{'coefofvar' }	0.98054	0.64	-0.57376	0.52659	0.79122	2.4779
{'gini' }	0.47956	0.33158	-0.29874	0.23461	0.38177	0.83796
{'min' }	0	0.05663	-230.76	1	0.059541	0

{ 'max' }	158.47	12.271	-1.694	6	23.47	20.112	
{ 'pYis0' }	0.004589	0	0	0	0	0.52499	
{ 'pYls0' }	0	0	1	0	0	0	
{ 'pYgr0' }	0.99541	1	0	1	1	0.47501	
{ 'pYisMINY' }	0.004589	9.8045e-06	2.9682e-09	0.48835	9.869e-06	0.52499	
{ 'pYisMAXY' }	9.1885e-09	2.1301e-11	5.3537e-07	0.0036816	1.4932e-06	3.6378e-08	
{ 'p0_01' }	0	0.07838	-123.12	1	0.08341	0	
{ 'p10' }	1.2229	0.40584	-47.779	1	0.49527	0	
{ 'p25' }	3.196	0.6516	-33.261	1	0.77993	0	
{ 'p50' }	7.0976	1.0499	-23.135	2	1.2719	0	
{ 'p75' }	13.089	1.547	-16.274	2	2.0397	0.48062	
{ 'p90' }	21.159	2.1475	-11.776	3	3.1029	1.7714	
{ 'p99_99' }	112.62	8.4781	-2.7295	6	15.937	16.033	
{ 'fl_cov_ap_ss' }	90.423	6.9267	101.69	0.81683	10.484	2.2203	
{ 'fl_cor_ap_ss' }	1	0.94827	0.68949	0.094628	0.86408	0.20978	
{ 'fl_cov_c_ss' }	6.9267	0.59008	8.839	0.23092	0.85409	0.24459	
{ 'fl_cor_c_ss' }	0.94827	1	0.74189	0.33116	0.87137	0.28607	
{ 'fl_cov_v_ss' }	101.69	8.839	240.55	2.5586	13.062	3.6105	
{ 'fl_cor_v_ss' }	0.68949	0.74189	1	0.18173	0.66001	0.20915	
{ 'fl_cov_n_ss' }	0.81683	0.23092	2.5586	0.82404	0.055266	0.27625	
{ 'fl_cor_n_ss' }	0.094628	0.33116	0.18173	1	0.047713	0.27342	
{ 'fl_cov_y_head_inc' }	10.484	0.85409	13.062	0.055266	1.6281	0.1162	
{ 'fl_cor_y_head_inc' }	0.86408	0.87137	0.66001	0.047713	1	0.08182	
{ 'fl_cov_y_spouse' }	2.2203	0.24459	3.6105	0.27625	0.1162	1.2388	
{ 'fl_cor_y_spouse' }	0.20978	0.28607	0.20915	0.27342	0.08182	1	
{ 'fl_cov_yshr_wage' }	-0.53029	-0.034916	-0.87809	0.0021706	-0.036142	0.021749	
{ 'fl_cor_yshr_wage' }	-0.55378	-0.45137	-0.56221	0.023745	-0.28128	0.19404	
{ 'fl_cov_yshr_SS' }	0	0	0	0	0	0	
{ 'fl_cor_yshr_SS' }	NaN	NaN	NaN	NaN	NaN	NaN	
{ 'fl_cov_yshr_nttxss' }	0.19504	0.017838	0.43624	0.0074857	0.026908	0.013251	-0.000000
{ 'fl_cor_yshr_nttxss' }	0.67474	0.76393	0.92529	0.27128	0.69373	0.39167	
{ 'fracByP0_01' }	0	6.875e-06	0.00049633	0.28329	5.7641e-06	0	3.000000
{ 'fracByP10' }	0.0057779	0.026049	0.23158	0.28329	0.022494	0	
{ 'fracByP25' }	0.040074	0.091752	0.44951	0.28329	0.082256	0	
{ 'fracByP50' }	0.16977	0.26933	0.70658	0.72028	0.24076	0	
{ 'fracByP75' }	0.42173	0.53549	0.88743	0.72028	0.48935	0.13542	
{ 'fracByP90' }	0.67785	0.76094	0.96591	0.85389	0.72071	0.34015	
{ 'fracByP99_99' }	0.99869	0.99925	0.99999	1	0.99889	0.99665	

age =100

xxx tb_outcomes: all stats xxx

OriginalVariableNames	ap_ss	c_ss	v_ss	n_ss	y_head_inc	y_spouse	yshr_w
{ 'mean' }	0	0.33019	-4.3147	1.4797	0.2355	0.09988	0.100000
{ 'unweighted_sum' }	0	1.2179e+05	-3630.5	21	209.83	33.1	82.000000
{ 'sd' }	0	0.23351	1.1931	0.50567	0.021892	0.2444	0.200000
{ 'coefofvar' }	NaN	0.7072	-0.27653	0.34173	0.092957	2.4469	1.000000
{ 'gini' }	NaN	0.2934	-0.14462	0.12034	0.043711	0.78724	0.700000
{ 'min' }	0	0.19737	-12.197	1	0.22	0	
{ 'max' }	0	141.61	-0.0071775	6	5.666	3.0753	0.900000
{ 'pYis0' }	1	0	0	0	0	0.52499	0.500000
{ 'pYls0' }	0	0	1	0	0	0	
{ 'pYgr0' }	0	1	0	1	1	0.47501	0.400000
{ 'pYisMINY' }	1	0.35707	1.4848e-10	0.5232	0.50347	0.52499	0.500000
{ 'pYisMAXY' }	1	0	0	4.2206e-08	0	1.0335e-08	5.1171e-08
{ 'p0_01' }	0	0.19737	-8.038	1	0.22	0	
{ 'p10' }	0	0.19737	-5.0665	1	0.22	0	
{ 'p25' }	0	0.19737	-5.0665	1	0.22	0	
{ 'p50' }	0	0.23607	-5.0103	1	0.22	0	
{ 'p75' }	0	0.34876	-3.7889	2	0.266	0.10166	0.300000
{ 'p90' }	0	0.58892	-2.3842	2	0.26656	0.48427	0.600000
{ 'p99_99' }	0	2.8508	-0.49608	4	0.31126	2.9082	0.900000
{ 'fl_cov_ap_ss' }	0	0	0	0	0	0	
{ 'fl_cor_ap_ss' }	NaN	NaN	NaN	NaN	NaN	NaN	
{ 'fl_cov_c_ss' }	0	0.054526	0.22623	0.059682	0.0015233	0.050691	0.000000

{'fl_cor_c_ss' }	NaN	1	0.812	0.50545	0.298	0.88823	0.7
{'fl_cov_v_ss' }	0	0.22623	1.4236	0.21735	0.011521	0.18354	0.1
{'fl_cor_v_ss' }	NaN	0.812	1	0.36026	0.44106	0.62943	0.6
{'fl_cov_n_ss' }	0	0.059682	0.21735	0.2557	0.0018733	0.052581	0.08
{'fl_cor_n_ss' }	NaN	0.50545	0.36026	1	0.16923	0.42546	0.7
{'fl_cov_y_head_inc' }	0	0.0015233	0.011521	0.0018733	0.00047925	0.00064761	0.000
{'fl_cor_y_head_inc' }	NaN	0.298	0.44106	0.16923	1	0.12104	0.1
{'fl_cov_y_spouse' }	0	0.050691	0.18354	0.052581	0.00064761	0.059731	0.04
{'fl_cor_y_spouse' }	NaN	0.88823	0.62943	0.42546	0.12104	1	0.7
{'fl_cov_yshr_wage' }	0	0.040339	0.19022	0.087809	0.00068213	0.043248	0.05
{'fl_cor_yshr_wage' }	NaN	0.75656	0.69819	0.76049	0.13646	0.77497	
{'fl_cov_yshr_SS' }	0	-0.041495	-0.19728	-0.089315	-0.00073742	-0.04328	-0.05
{'fl_cor_yshr_SS' }	NaN	-0.77343	-0.71963	-0.76875	-0.14661	-0.77074	-0.9
{'fl_cov_yshr_nttxss' }	0	0.045639	0.21697	0.096294	0.00089018	0.04769	0.05
{'fl_cor_yshr_nttxss' }	NaN	0.78185	0.72746	0.76178	0.16266	0.78059	0.9
{'fracByP0_01' }	NaN	0.21345	0.00043423	0.35357	0.47033	0	
{'fracByP10' }	NaN	0.21345	0.51425	0.35357	0.47033	0	
{'fracByP25' }	NaN	0.21345	0.51425	0.35357	0.47033	0	
{'fracByP50' }	NaN	0.33309	0.6114	0.35357	0.47033	0	
{'fracByP75' }	NaN	0.53778	0.8525	0.99419	0.87579	0.19248	0.3
{'fracByP90' }	NaN	0.74393	0.95864	0.99419	0.8936	0.62777	0.5
{'fracByP99_99' }	NaN	0.99922	0.99999	0.99999	0.99991	0.9996	0.9