Julia Mono<br/>[ Extension = .ttf, Path = ./, Scale = 0.87, Contextuals = Alternate, Upright<br/>Font = \*-Light, BoldFont = \*-SemiBold, ItalicFont = \*-Lightitalic, BoldItalicFont = \*-SemiBolditalic, ]

This problem set was completed by Danny Edgel, Mitchell Valdes Bobes, Ryan Mather, and Yobin Timilsena.

I. Consider the same environment as Huggett (1993, JEDC) except assume that there are enforceable insurance markets regarding the idiosyncratic shocks to earnings and that there are no initial asset holdings. Solve for a competitive equilibrium. What are prices? What is the allocation? (Hint: think about the planner's problem and then decentralize).

<u>Answer:</u> Under the assumptions of enforceable insurance markets + locally non-satiated preferences, the basic first and second welfare theorems hold. Hence, we will solve the planner's problem for allocations and then decentralize by setting asset prices that support the allocations as a CE.

The planner's problem can be written as

$$\mathbb{E} \sum_{t=0}^{\infty} \beta^{t} [\pi(e)u(c_{e,t}) + \pi(u)u(c_{u,t})] \quad \text{s.t.} \quad \pi(e)c_{e,t} + \pi(u)c_{u,t} \leq \pi(e)y_{t}(e) + \pi(u)y_{t}(u), \ \forall t \in \mathbb{E}$$

The first order conditions are

$$[c_{e,t}]: \beta^t \pi(e) u'(c_{e,t}) = \lambda \pi(e)$$
  $[c_{u,t}]: \beta^t \pi(u) u'(c_{u,t}) = \lambda \pi(u)$ 

Combined, we have that  $u'(c_{e,t}) = u'(c_{u,t}) \Leftrightarrow c_{e,t} = c_{u,t} = \bar{c}$ .

Plugging into the BC, we get  $\bar{c} = y_t(u) + \pi(e)[y_t(e) - y_t(u)]$ . For instance, in Part II, we are given  $\pi(e) = 0.94$ ,  $y_t(e) = 1$  and  $y_t(u) = 0.5$ , which would imply  $\bar{c} = 0.97$ .

The decentralized EE for an individual i is  $\beta^t u'(c_t^i) = \lambda q_t \Leftrightarrow q_{t+1} = \beta q_t = \beta^{t+1} q_0$ .

- II. Now compute Huggett (1993, JEDC) with incomplete markets. The following takes you through the steps of solving a simple general equilibrium model that generates an endogenous steady state wealth distribution. The basic algorithm is to: 1) taking a price of discount bonds  $q \in [0,1]$  as given , solve the agent's dynamic programming problem for her decision rule  $a' = g_{\theta}(a,s;q)$  where  $a \in A$  are asset holdings,  $s \in S \subset R_{++}$  is exogenous earnings, and  $\theta$  is a parameter vector; 2) given the decision rule and stochastic process for earnings, solve for the invariant wealth distribution  $\mu^*(A,S;q)$ ; 3) given  $\mu^*$ , check whether the asset market clears at q (i.e.  $\int_{A,S} g_{\theta}(a,s;q)\mu^*(da,ds;q)=0$ ). If it is, we are done. If not (i.e. it is not within an acceptable tolerance), then bisect [0,1] in the direction that clears the market (e.g. if  $\int_{A,S} a'\mu^*(da,ds;q)>0$ ), then choose a new price  $\hat{q}=q+[1-q]/2$  and go to step 1.
  - 4. After finding fixed points of the T and  $T^{st}$  operators, answer the following questions:
    - a. Plot the policy function g(a,s) over a for each s to verify that there exist  $\hat{a}$  where  $g(\hat{a},s)<\hat{a}$  as in Figure 1 of Huggett. (Recall this condition establishes an upper bound on the set A necessary to obtain an invariant distribution).

**Answer:** The policy function is graphed below in Figure 1. As can be seen, there does exist an  $\hat{a}$  beyond which modeled agents always, whether employed or unemployed, dissave on net.

b. What is the equilibrium bond price? Plot the cross-sectional distribution of wealth for those employed and those unemployed on the same graph.

**Answer:** The equilibrium bond price is q=0.9943074, and the resulting cross-sectional distribution of wealth is shown in Figure 2.

c. Plot a Lorenz curve. What is the gini index for your economy? Compare them to the data. For this problem set, define wealth as current earnings (think of this as direct deposited into your bank, so it is your cash holdings) plus net assets. Since market clearing implies aggregate assets equal zero, this wealth definition avoids division by zero in computing the Gini and Lorenz curve.

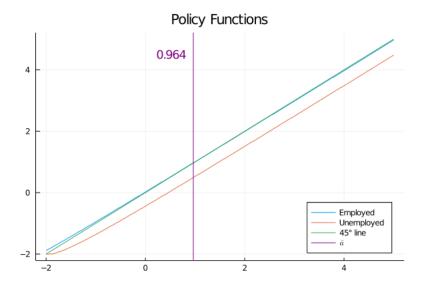


Figure 1: Problem 4(a)

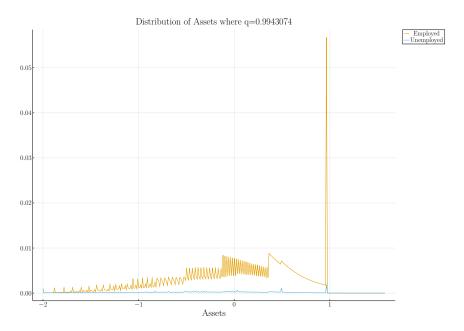


Figure 2: Problem 4(b)

Answer: The Lorenz curve is shown in Figure 3, and implies a Gini coefficient of 0.29969592. The true Income Gini coefficient for the US economy in 2020 was .458, so if we take this to be the relevant statistic against which to compare our model results, our model explains about 65% of inequality as captured by the Gini Coefficient.

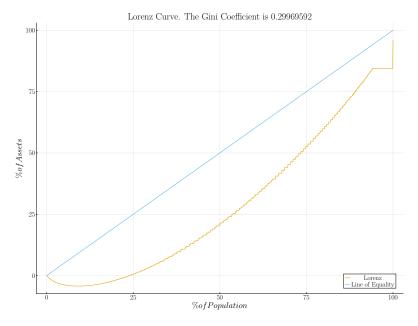


Figure 3: Problem 4(c)

III. (a) Plot  $\lambda(a, s)$  across a for both s = e and s = u in the same graph.

## Answer:

•

(b) What is  $W^{FB}$ ? What is  $W^{INC} = \sum_{(a,s) \in A \times S} \mu(a,s) \nu(a,s)$ ? What is WG?

## Answer:

•

(c) What fraction of the population would favor changing to complete markets? That is  $\sum_{(a,s)\in A\times S}\mathbbm{1}_{\lambda(a,s)\geq 0}(a,s)\mu(a,s)$ .

## Answer:

.

## **Code Appendix**

This is the "computation" code that does most of the numerical work involved for the solutions above:

<sup>&</sup>lt;sup>1</sup>U.S. Census Bureau, Current Population Survey, 1968 to 2021 Annual Social and Economic Supplements (CPS ASEC), Table F-4

```
n
 #strutur tht hols mol rsults
 Ovrywhr mut 1 strut Rsults
     vl_un::Arry{Flot, } #vlu untion pol_un::Arry{Flot, } #poliny untion
    q::Flot
    q_Bouns::Arry{Flot , }
 #untion or initilizing mol primitivs n rsults
 untion Initiliz()
    prim = Primitivs() #initiliz primtiivs
    vl_un = zros(prim.n,) #initil vlu untion guss
pol_un = zros(prim.n,) #zros(prim.nk,) #initil poliy untion guss
    q = (prim.\beta+)/
     q_Bouns=[prim.\beta, ]
     rs = Rsults(vl_un, pol_un, q, q_Bouns) #initiliz rsults strut
    prim, rs #rturn livrls
 \hbox{\tt\#Mking} \qquad \hbox{\tt untion or th innr loop}
 @vrywhr moul IL
     untion Fin_p(S_inx,_inx,rs,prim)
        #unp k mo 1 primitivs _gri , \beta, \alpha, n , \Pi, S_gri = prim. _gri , prim.\beta, prim.\alpha,
                prim.n , prim.\Pi, prim.S_gri #unp k mo l primitivs
         #Utility Funtion
             untion U(x)
                i x<
                    rturn -In
                 ls
                    rturn (x^( -\alpha)- )/( -\alpha)
        #Exploiting Monotoniity o V
         u g t=S_gri[S_in x] + _gri[_in x];
#S r h or vl_mx in th oun intrvl.
         m \times v l, m \times p = -In,
         or p_in x=:n
             v1=U( u g t - rs.q*_gri[p_in x]) +

β*(trnspos(Π[S_in x,:])*[ rs.vl_un[p_in x, ] ; rs.vl_un[p_in x, ]])
             i v1>mx_v1
                m \times v l = v l;
             m x_p=p_in x;
ls i v1<mx_v1
                 r k #Th Vlu untion is now lining
        rturn [mx_v1, mx_p]
     n
 n
 #Bllmn Oprtor
 untion Bllmn(prim::Primitivs,rs::Rsults)
     @unp k n , _gri = prim
v_nxt = zros(n , ) #nxt guss o vlu untion to ill
     or S_inx=:
        out=pmp(_in x -> IL.Fin_p(S_in x, _in x,rs,prim), :n)
         or _in x=:n #Unpking th pmp rsults
v_nxt[_in x,S_in x]=out[_in x][]
            rs.pol_un[_in x,S_in x]=out[_in x][]
     n
     v_nxt #rturn nxt guss o vlu untion
 #Vlu untion itrtion
 untion V_itrt(prim::Primitivs, rs::Rsults; tol::Flot = e-4, rr::Flot =
    .0)
    n = #ountr
     whil rr>tol # gin itrtion
        v_nxt = Bllmn(prim, rs) #spit out nw v tors
         rr = s.(mximum(v_nxt.-rs.vl_un))/ s(v_nxt[prim.n,]) #rst rror lvl
         rs.vl_un = .8*v_nxt+.2*rs.vl_un #up t vlu untion
        n+=
        i mo(n, )==
```

```
println("Vlu Funtion itrtion $(n), Error $(rr)")
    n
   println("Vlu untion onvrg in", n, "itrtions.")
#Mrkt lring or ssts/sts
                               n w a
untion MC_ssts(prim,rs; ist_tol::Flot = e-6, ist_rr::Flot = .0, ES_tol= e-2, Don= ls)
   Qunp k \Pi, n, gri = prim TrnsMt=zros(*n,*n) #Th irst n points r or mploy olks, n th nxt n r or unmploy
    or _in x = :n
       TrnsMt[Int (rs.pol_un[_in x, ])+n,_in x]=\Pi[, ] #Svings hoi or thos moving rom mp->unmp
TrnsMt[Int (rs.pol_un[_in x, ]),_in x+n]=\Pi[, ] #Svings hoi or thos moving rom unmp->mp
       TrnsMt[Int (rs.pol_un[_in x, ])+n,_in x+n]=II[,] #Svings hoi or thos moving rom unmp->mp
   Dist=ons(*n)*(/(*n)) #
   Dist_nw=opy(Dist)
   whil ist_rr>ist_tol
       or i=: #Itrt until w r h th st y-stt istriution
          Dist nw=TrnsMt*Dist nw
       ist_rr= s.(mximum(Dist_nw.-Dist))
       Dist=opy(Dist_nw)
   #Fin Exss Supply n rst q
       ExssSupply=trnspos(Dist)*v t(_gri,_gri)
         s(ExssSupply)>ES_tol
           #Do vrint o Bistion Mtho
           i ExssSupply<
              rs.q_Bouns[]=rs.q
              #Wight slightly towr ol q to voi wil lututions
              rs.q=rs.q_Bouns[]*.3+ rs.q_Bouns[]*.7
              rs.q_Bouns[]=rs.q
              rs.q=rs.q_Bouns[]*.7 +rs.q_Bouns[]*.3
           print("Ex ss Supply: $(ExssSupply), q:$(rs.q)")
          Don=tru
        n
       rturn Don
#solv th mol
untion Solv_mol() #prim::Primitivs, rs::Rsults)
   prim, rs = Initiliz()
    onvrg = ls
   Outr_loop_Itr=
   whil ~ onvrg && Outr_loop_Itr<
       println("Bginning Asst Clring Loop $(Outr_loop_Itr)")
       V_itrt(prim, rs)
        onvrg =MC_ssts(prim,rs)
       Outr_loop_Itr+=
    n
   rturn prim, rs
#Gt Distriution or Plotting
untion FinDist_ForPlot(prim, rs; ist_tol::Flot = e-6, ist_rr::Flot =
   .0.)
   Qunp k \Pi, n , _gri = prim TrnsMt=zros(*n , *n) #Th irst n points r or mploy olks, n th nxt n r or unmploy
    or _in x = :n
       TrnsMt[Int (rs.pol_un[_in x, ]),_in x]=N[,, ] #Svings hoi or thos moving rom mp->mp
       TrnsMt[Int (rs.pol_un[_in x,])+n,_in x]=II[,] #Svings hoi or thos moving rom mp->unmp
TrnsMt[Int (rs.pol_un[_in x,]),_in x+n]=II[,] #Svings hoi or thos moving rom unmp->mp
TrnsMt[Int (rs.pol_un[_in x,])+n,_in x+n]=II[,] #Svings hoi or thos moving rom unmp->mp
   Dist=ons( *n)*( /( *n )) #
   Dist_nw=opy(Dist)
   whil ist_rr>ist_tol
       or i=: #Itrt until w r h th st y-stt istriution
          Dist_nw=TrnsMt*Dist_nw
       ist_rr= s.(mximum(Dist_nw.-Dist))
       Dist=opy(Dist_nw)
   rturn Dist, Dist[ :n].+Dist[(n+): *n]
```

This code calls the "computation" code above and then prints some figures:

```
#Gtting th Prlll R y
       using Distriut #, Shr Arrys
       #R -initilizing th workrs
               rmpros(workrs())
                 pros()
        @vrywhr using Prmtrs
#Sving Dtils
       in lu ("Comput_Dr t.jl")
#Solv th Mol
       #initiliz primitiv n rsults struts
        @tim out_primitivs, out_rsults = Solv_mol() #solv th mol!
        @unp k vl_un , pol_un = out_rsults
        @unp k _gri , n , S_gri = out_primitivs
#Plotting rsults
using Plots, LTXStrings #import th lirris w wnt
Plots.plot(_gri, vl_un[:,], titl="Vlu Funtion", l l="Employ")
    plot!(_gri, vl_un[:,], l l="Unmploy")
        Plots.svig("Vlu_Funtions.png")
       #Plotting Poliy untions
               untion PoliyPolots()
                         _h t=
                         or i= :n
                                i _gri[Int .(pol_un[i,])] <= _gri[i]</pre>
                                       _h t = [ _gri [ i]];
                                         r k
                       Plots.plot(_gri, _gri[Int.(pol_un[:,])], titl="Poliy Funtions", 1 1="Employ")
                              plot!(_gri, _gri [Int .(pol_un[:, ])], 1 1="Unmploy")
plot!(_gri, _gri, 1 1=" lin", 1 g n =:ottomright)
vlin!(_ht, 1 1=L"\ht{}",olor=:purpl)
                                nnott!(_ht[]-.1 , .5, txt("$(roun(_ht[],igits=))", :purpl, :right, ))
                               Plots.svig("Poliy_Funtions.png")
               PoliyPolots()
        #Plotting Distriution
                untion DistPlots()
                       TS_Distriution, SS_WlthDistriution=FinDist_ForPlot(out_primitivs,out_rsults)
                       MxNonZro=
                       ForDistPlot=opy(TS_Distriution)
                        or i= :n
                              i ForDistPlot[i]==
                                      ForDistPlot[i]=NN
                               i ForDistPlot[n +i]==
                                      ForDistPlot[n +i]=N N
                                n
                               i SS_WlthDistriution[i]!=
                                      MxNonZro=opy(i)
                         n
                       Plots.plot(_gri[:MxNonZro], ForDistPlot[:MxNonZro], titl="Distriution o Assts
whr q=$(roun(out_rsults.q,igits=))",
                               1 1 = "Employ")
                               plot!(_gri[:MxNonZro], ForDistPlot[(n + ):n +MxNonZro], 1 1="Unmploy", x1 1="Assts")
                              Plots.s v ig("Distriution.png")
                       #Lornz Curv
                       n lornz=
                       Lornz=zros(n_lornz, )
                              Lornz[:, ]= oll t(rng(,lngth=n_lornz,)) #First olumn is pr nt o popultion
                              i=
                                        _in x = :n
                                       i sum(SS_WlthDistriution[ : _in x]) <= Lornz[i, ]</pre>
                                               \label{lornz} Lornz[i, ] = Lornz[i, ] + TS_Distriution[ \_in x] * (\_gri[ \_in x] + S_gri[ ]) + TS_Distriution[ \_in x] + T
                                                      TS_Distriution[n + _ in x]*(_gri[_in x]+S_gri[]) #S on olumn is umultiv ssts
                                               whil sum(SS_WlthDistriution[ : _in x])>Lornz[i, ]
                                                      Lornz[i, ]=Lornz[i- , ]+; #opy ovr th prvious umultiv wlth
```