

# VFI Toolkit: Workshop, Part 3

[vfitoolkit.com](http://vfitoolkit.com)

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- We have seen Life-Cycle Models, which were partial equilibrium.
- Now for OLG models, which add general equilibrium.
- You can add general eqm to any of the models we have seen so far!

- Basic idea: *GeneralEqmEqns*, which must evaluate to zero in general eqm.

E.g., instead of  $r = \alpha K^{\alpha-1} L^{1-\alpha}$  we will write  $r - \alpha K^{\alpha-1} L^{1-\alpha}$ .

- Internally VFI Toolkit uses optimization routines to choose 'GEPriceParamNames' to minimize the sum-of-squares of the *GeneralEqmEqns*

E.g.  $GEPriceParamNames = \{r'\}$

- Let's solve a simple OLG model.
- Code: WorkshopOLGModel1.m (and WorkshopOLGModel1\_ReturnFn.m)

- Continuum of mass one of households, solving 'household problem'.
- Continuum of mass one of firms, solving 'firm problem'.
- Labor & Capital markets: labor & assets of household=labor & assets of firm.
- Assume perfectly competitive markets.

- Household problem

$$\begin{aligned} V(a, z, j) = & \max_{h, c, a_{\text{prime}}} \frac{c^{1-\sigma}}{1-\sigma} - \psi \frac{h^{1+\eta}}{1+\eta} + s_j \beta E[V(a_{\text{prime}}, z_{\text{prime}}, j+1) | z] \\ & \text{if } j < J_r : c + a_{\text{prime}} = (1+r)a + w\kappa_j h \exp(z) \\ & \text{if } j \geq J_r : c + a_{\text{prime}} = (1+r)a + \text{pension} \\ & 0 \leq h \leq 1, a_{\text{prime}} \geq 0 \\ & z' = \rho_z z + \epsilon', \quad \epsilon \sim N(0, \sigma_{z, \epsilon}) \end{aligned}$$

- Note, is just our WorkshopModel3 again!

# Basic OLG Model 1

- Representative Firm problem

$$\begin{aligned} \max_{K^f, L^f} Y - rK^f - wL^f - \delta K^f \\ \text{s.t. } Y = (K^f)^\alpha (L^f)^{1-\alpha} \end{aligned}$$

- Profit maximization (with price normalized to one), for firm with Cobb-Douglas production fn.
- Simplifies to two FOCs,

$$\begin{aligned} r &= \alpha (K^f)^{\alpha-1} (L^f)^{1-\alpha} - \delta \\ w &= (1 - \alpha) (K^f)^{\alpha-1} (L^f)^{-\alpha} \end{aligned}$$

- Trick: combine to get  $w = (1 - \alpha) \left( \frac{r + \delta}{\alpha} \right)^{\frac{\alpha}{1-\alpha}}$ .

This trick is standard and nothing to do with toolkit per se.

Our continuum of firms with constant-returns-to-scale production functions can just be replaced by a representative firm.

- Define aggregates, household labor supply and household assets

$$L^h = \int \kappa_j h \exp(z) d\mu$$

$$K^h = \int a d\mu$$

- $\mu$  is the agent distribution of the households. Integral basically says 'add up across households'.



- General eqm, is about  $L^h = L^f$  and  $K^h = K^f$ . But we can substitute these into firm problem and instead get general eqm eqns,

$$r = \alpha(K^h)^{\alpha-1}(L^h)^{1-\alpha} - \delta$$

- and also the eqn we derived earlier  $w = (1 - \alpha) \left( \frac{r+\delta}{\alpha} \right)^{\frac{\alpha}{1-\alpha}} \alpha^{-1}$ .

- Definition of Stationary General Equilibrium,

- Okay, let's write the code.
- Most is easy as is just copy-paste of `WorkshopModel3.m` where we already solved the household problem.
- One change here is `WorkshopModel3_ReturnFn`, which used to have  $w$  as input, but now remove this and instead use

$$w = (1 - \alpha) \left( \frac{r + \delta}{\alpha} \right)^{\frac{\alpha}{\alpha - 1}} \text{ inside the return fn.}$$

This is a standard trick and not about toolkit, we could just solve for  $r$  and  $w$  as general eqm parameters, and use both FOCs of firm problem as general eqm eqns, but would be slower. If you ever coded Aiyagari model you probably used this same trick so general eqm is only about  $r/K$ , and not also  $w/L$ .

Also add alpha and delta as inputs to `ReturnFn`. Code also removes  $w$  from `Params`, and creates alpha and delta in `Params`.

- We need to add *FnsToEvaluate* for  $L^h$  (we already have one for  $K^h$ , although let's rename it to  $K$ ).

```
1 %% Set up FnsToEvaluate
2 FnsToEvaluate.L=@(h,aprime,a,z,kappa_j) kappa_j*
    h*exp(z); % effective labor supply
3 FnsToEvaluate.K=@(h,aprime,a,z) a; % assets
```

# Basic OLG Model 1

- Set up *GEPriceParamNames* as the names of parameters to find in general eqm, and *GeneralEqmEqns* as the equations that must evaluate to zero in general eqm.

```
1 %% General Eqm
2 GEPriceParamNames={'r'};
3 % note, Params.r we set earlier was an initial
  guess
4
5 GeneralEqmEqns.capitalmarket=@(r, alpha, delta, K, L
    ) r - (alpha * (K^(alpha - 1)) * (L^(1 - alpha)) - delta)
  ;
6 % GeneralEqmEqn inputs must be either Params or
  AggVars (AggVars is like AllStats, but just
  the Mean)
7 % So here: r, alpha, delta will be taken from
  Params, and K, L will be taken from AggVars
```

- Done setting up, now solve the Stationary General Equilibrium

```
1 %% Solve for stationary general eqm
2 heteroagentoptions.verbose=1; % just use
  defaults
3 [p_eqm,~,GEcondns]=
    HeteroAgentStationaryEqm_Case1_FHorz(
      jequaloneDist, AgeWeightParamNames, n_d, n_a,
      n_z, N_j, [], pi_z, d_grid, a_grid, z_grid,
      ReturnFn, FnsToEvaluate, GeneralEqmEqns,
      Params, DiscountFactorParamNames, [], [], [],
      GEPriceParamNames, heteroagentoptions,
      simoptions, vfoptions);
4 % Done, the general eqm prices are in p_eqm
5 % GEcondns tells us the values of the
  GeneralEqmEqns, should be near zero
```

- While solving, it repeatedly gives feedback

```
1 Current GE prices :  
2     r:    0.0533  
3 Current aggregate variables :  
4     L:    0.6701  
5     K:    4.6979  
6 Current GeneralEqmEqns :  
7     capitalmarket:  -0.0002
```

- Optimization: Given prices, evaluate FnsToEvaluate, evaluate GeneralEqmEqns. Repeat with updated prices. Different optimization routines are just different ways to 'update prices'.

- Analyzing the general eqm is easy.
- Just put general eqm price into Params:  $Params.r = p_{eqm}.r$
- Then just do  $V$ ,  $Policy$ ,  $StationaryDist$ ,  $AllStats$ ,  $AgeConditionalStats$  like we did for life-cycle models.



# Basic OLG Model 1

- Done!
- Code: WorkshopOLGModel1.m (and WorkshopOLGModel1\_ReturnFn.m)

- You can have  $n$  general eqm prices/parameters, for  $n$  general eqm eqns.

They don't have to be prices, e.g., might choose gov. spending  $G$  to satisfy government budget constraint as one of our general eqm eqns.

- If you have a calibration target, like  $K/Y = 3$ , you can just set this up like another general eqm eqn.

Only works if the calibration target can be expressed in terms of aggregates and parameters. (This is 'joint-optimization' of calibration and general eqm.)

- *heteroagentoptions.multiGEweights* can be used to give relative weights (default is equal weights).

E.g. if you have lots of calibration targets you probably want to give them smaller weights so as not to interfere with the general eqm.

- *heteroagentoptions.GEtype* can be used to name general eqm eqns to evaluate conditional on permanent type.

- By default, it is using *heteroagentoptions.fminalgo* = 1, which is Matlab's *fminsearch*().
- You can set different optimization routines (like *lsqnonlin*, *CMA – ES*, and *shooting*) just by changing *heteroagentoptions.fminalgo* (=8,4,5 respectively).

*lsqnonlin* requires Matlab Optimization Toolbox and implements Levenberg-Marquardt, it can be very fast. *CMA – ES* is Covariance-Matrix Adaptation Evolutionary Strategy, it is slow but very robust. *shooting* is very very fast, but requires substantially more user input about how to update. See Appendix to [Intro to OLG Models](#).

- Control solution tolerance using *heteroagentoptions*.  
heteroagentoptions.toleranceGEprices=10<sup>-4</sup>; % Accuracy of general eqm prices  
heteroagentoptions.toleranceGEcondns=10<sup>-4</sup>; % Accuracy of general eqm eqns
- You can use two algorithms, one after the other, with different tolerance. Can help get good mix of speed (first algo) and robust accuracy (second algo).

E.g., set heteroagentoptions.fminalgo=[8,1] and heteroagentoptions.toleranceGEcondns=[10<sup>-3</sup>,10<sup>-5</sup>]

- Anything we saw with Life-Cycle Models, can be done with OLG models!
- [Intro to OLG Models](#) has some examples.

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