

VFI Toolkit: Workshop, Part 2

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- We saw how to setup and solve basic Life-Cycle models.
- d (decision variable), a (endogenous state) and z (exogenous markov state).
- And basic analysis, StationaryDist, AllStats, and AgeConditionalStats.
- VFI Toolkit does much more: i.i.d shocks, panel data, human capital, portfolio-choice, conditional statistics, calibration, GMM estimation.
- Let's take a whirlwind tour!

- All of this is covered in the examples of the Intro to Life-Cycle Models.
- [Intro to Life-Cycle Models](#): pdf of 50 example Life-Cycle models, adding features one at a time. Covers everything we did here, plus much more.

- Rough outline:

- ➊ Exogenous shocks: i.i.d, correlated shocks (e.g., VAR), semi-exogenous state.
- ➋ Permanent Types
- ➌ Analyse model: simulate panel data, conditional stats, faster stats.
- ➍ Calibrate model
- ➎ Alternative preferences.
- ➏ Alternative endogenous states: human capital, riskyasset (portfolio choice).
- ➐ GMM estimation.

Life-Cycle Models: Exogenous shocks

- We saw markov shocks z .
- Can also add i.i.d. exogenous shocks e .
- Beyond the basics, we set things up using *vfoptions* and *simoptions* to explain the model to VFI Toolkit.
- So we would create *vfoptions.n_e*, *vfoptions.e_grid* and *vfoptions.pi_e*, put copies in *simoptions*.
pi_e will be a column vector.
- Action space: $(d, a_{prime}, a, z, e, \dots)$ [e always comes just after z]
- Explain e vars: e vars concept

See Intro to Life-Cycle Models: Life-Cycle Model 11.

Life-Cycle Models: Exogenous shocks

- Can have up to four of each of z and e .
And $semiz$ which are explained next slide.
- Can have grids and probabilities depend on age.
Can also be created by parameterized functions.
- Can use joint-grids, and so can have correlated shocks.
- VFI Toolkit contains many discretization methods: AR(1), AR(1) with gaussian mixture, VAR(1), and versions for 'non-stationary life-cycle' processes.

See Intro to Life-Cycle Models: Appendix A.

Life-Cycle Models: Exogenous shocks

- Semi-exogenous states.
- *aprime* probabilities depend on a decision variable.
- Create *vfoptions.n_semiz*, *vfoptions.semiz_grid*.
- Create *vfoptions.SemiExoStateFn* for the transition probabilities, will be a function of (*semiz*, *semizprime*, *d*, ...).
Put them in *simoptions* as well.
- Uses: endogenous fertility, search labor, years owned house.
- Action space: (*d*, *aprime*, *a*, *semiz*, *e*, ...) [*semiz* always comes after *a*, before *z*]
- Explain *semiz* vars: semiz vars concept

See Intro to Life-Cycle Models: Life-Cycle Model 29, and Appendix A.

Life-Cycle Models: Permanent Types

- Permanent Types: agents with permanent differences.
- Examples
 - Different values of a parameter.
 - Different exogenous shock processes.
 - Different utility functions.
 - Even that one is finite-horizon and one infinite-horizon.
 - Essentially, anything in toolkit can differ between the permanent types.
- Two ways to set up:
 - Number of permanent types: $N_i = 2$, $Params.beta = [0.8, 0.95]$
 - Names of permanent types: $Names_i = 'patient', 'impatient'$
 $Params.beta.impatient = 0.8$; $Params.beta.patient = 0.95$
- VFI Toolkit commands are same in both cases. And all output uses the names. E.g., $V.patient$ and $V.impatient$.
If N_i , names are allocated as $p_{type001}$, $p_{type002}$.
- All model stats reported as aggregate and conditional on ptype.
E.g., $AllStats.earnings.Mean$, $AllStats.earnings.patient.Mean$ and $AllStats.earnings.impatient.Mean$.

See Intro to Life-Cycle Models: Life-Cycle Models 24, 25, 27.

- We saw AllStats and AgeConditionalStats.
- Can simulate panel data: same setup (FnsToEvaluate), different command *SimPanelValues = SimPanelValues_FHorz_Case1()*.
- Conditional stats: *simoptions.conditionalrestrictions* can setup a function in same way we do FnsToEvaluate, then when you call AllStats or LifeCycleProfiles, it will also report moments conditional on this function (function must be binary, e.g. $a > 0$)

See: [explain conditionalrestrictions](#).

- Faster stats: *simoptions.whichstats = ones(7,1)* is default. By setting some zeros, you can e.g., skip lorenz curve and quantiles which take most of the time.

See: [explain whichstats](#).

Really this is for when you want write a custom code that will need to use AllStats or LifeCycleProfiles a large number of times. If you are using them once you probably don't care about the runtime to calculate a few unnecessary stats.

See Intro to Life-Cycle Models: Life-Cycle Models 12.

Life-Cycle Models: Calibrate model

- You can easily choose parameters (*CalibParamNames*) to target moments (*TargetMoments*).
- Setup targets, then just *CalibrateLifeCycleModel()*.
- Targets can be anything in *AllStats* and *AgeConditionalStats*.
Includes conditional restrictions, and both aggregating across or conditional on permanent types.
- Can include, e.g., parametrizing initial agent distribution and/or exogenous shocks with parameters that will be calibrated.

See Intro to Life-Cycle Models: No examples of calibration yet, but if you read Life-Cycle Models 45-50 about GMM estimation, is essentially the same but without Variance-Covariance matrix. Is used in [Huggett, Ventura & Yaron \(2006\) example](#).

Life-Cycle Models: Alternative Preferences

- Can solve Epstein-Zin preferences, Quasi-Hyperbolic discounting, Gul-Pesendorfer preferences, Prospect Theory (loss aversion), and Ambiguity Aversion.
- Most just require a few lines to tell *vfoptions* you want to use them.
- See Life-Cycle Model 12 for Epstein-Zin preferences (and 32 to 35 for EZ prefs in portfolio-choice models).
- See Life-Cycle Models 36-39 for the rest.
- Intro to Life-Cycle Models, Appendix B explains how each of these works.

Once you solve for Policy, alternative preferences are actually irrelevant to StationaryDist and model statistics, so we don't need to tell *simoptions*.

Life-Cycle Models: Alternative Endogenous States

- We can have two endogenous states.
- Set $n_a = [201, 51]$, $a_grid = [a1_grid; a2_grid]$, then action space is $(.., a1prime, a2prime, a1, a2, ..)$.
- In practice, only going to work nicely as of the year 2025 if using $vfoptions.divideandconquer = 1$ and second state is say 51 points.
- E.g., for 'house', 51 points is probably plenty.

See Intro to Life-Cycle Models: Life-Cycle Model 42

Life-Cycle Models: Alternative Endogenous States

- Instead of standard endogenous state, where a is in today's state-space, and we can choose a_{prime} directly (so a_{prime} is also in the action-space)...
- ... risky asset, 'riskyasset', is $a_{prime}(d, u)$, useful for portfolio-choice models.
- ... experience asset, 'expasset' is $a_{prime}(d, a)$, useful for human-capital models.
- ... experience asset u , 'expassetu' is $a_{prime}(d, a, u)$, useful for uncertain human-capital models.
- All require setting up $vfoptions.a_{prime}Fn$ (inputs follow from which we are using)
- All of these mean a_{prime} is no longer in the decision space, so e.g., for expasset, we likely have action-space (d, a, z, \dots) [essentially, write out action-space as usual, but remove a_{prime} for the relevant endogenous state that is of this nature]

E.g., $a_{prime}(d, a)$, means you cannot choose a_{prime} , but you can choose d , and together with between period i.i.d. shocks u , you end up in $a_{prime}(d, u)$. These alternative assets all evaluate $a_{prime}(x)$ on the grids for x , and then linearly interpolate $a_{prime}(x)$ onto the grid on a_{prime} (which in practice is the same grid as a , so a_grid).

Life-Cycle Models: Alternative Endogenous States

- You can have two endogenous states where the first state is a standard endogenous state and the second state is a riskyasset/expasset/expassetu.
- Action space: $(d, a1prime, a1, a2, \dots)$
Notice that we put $a1prime$ for the standard endogenous state, but not for the riskyasset/expasset/expassetu.
- See Intro to Life-Cycle Models: Life-Cycle Model 35 for riskyasset (with housing), Life-Cycle Model 43 for expasset (with savings), Life-Cycle Model 44 for expassetu (with savings).

You can do without a standard endogenous state. See Intro to Life-Cycle Models, Life-Cycle Models 31-34 for 'riskyasset' in portfolio-choice problems. See [Huggett, Ventura & Yaron \(2006\) example](#) for 'expasset'.

- You can easily choose parameters (*EstimParamNames*) to target moments (*TargetMoments*).
This part is identical to calibration.
- Set *WeightingMatrix* and *CovarMatrixDataMoments*.
- Then *EstimateLifeCycleModel_MethodOfMoments()*.
- Outputs: [*EstimParams*, *EstimParamsConflnts*, *estsummary*]
- *estsummary* contains information on identification, sensitivity and more.

See Intro to Life-Cycle Models: Life-Cycle Models 45-50 about GMM estimation. Appendix C explains concepts around GMM estimation of Life-Cycle Models.

Note, permanent types mean you can handle unobserved heterogeneity.

Life-Cycle Models: Combinations?

- Some combinations are coded: e.g., riskyasset with Epstein-Zin preferences.
- Some combination have not been coded yet.
- If you want a combination that is not yet coded, or just want to know if some combo is coded, please ask via discourse.vfitoolkit.com

First time someone asks a combo, decent odds I ignore, but if two or three people are asking then I make it a priority and then it happens :D

- VFI Toolkit can do a lot of models, but equally there are many models it cannot do (yet :P).
- [Intro to Life-Cycle Models](#) explains many of these features by examples.
- Examples implementing models from literature: [Life-Cycle OLG Reading List](#)
- If you want a combination that is not yet coded, or just want to know if some combo is possible, please ask via discourse.vfitoolkit.com

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<https://doi.org/10.5281/zenodo.8136790>.

- Can write Bellman equation like

$$V(a, z, e) = \max_{d, a'} F(d, a', a, z, e) + s_j \beta E[V(a', z', e')|z]$$

- ReturnFn will be a function that plays role of $F(d, a', a, z, e)$
- e is an i.i.d exogenous state (so will have probability distribution pi_e).

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Life-Cycle Model

- Can write Bellman equation like

$$V(a, semiz, z) = \max_{d, a'} F(d, a', a, semiz, z) + s_j \beta E[V(a', semiz', z') | semiz, z, d]$$

- ReturnFn will be a function that plays role of $F(d, a', a, semiz, z)$
- *semi* is an semi-exogenous markov state (so will have probability transitions that depend on (*semiz*, *semizprime*, *d*..)) and a set using *vfoptions.SemiExoStateFn*.
- Roughly, *semiz* is a markov state, but where depending on value of the decision variable, you get different transition matrix.
- I here simplify and have *d* in everything, in practice often you can split *d* in *d1* and *d2*, with both in ReturnFn, but only *d2* determine semi-exo state.

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