

Lecture 5: Putting Algorithms to Work: Solving an Income Fluctuations Problem

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Let's Circle Back to the Beginning

- ▶ We started with the goal of solving dynamic programs.
- ▶ One of the simplest problems you will encounter is the income fluctuation problem:

$$V(k_i, z_j) = \max_{c, k'} \left\{ \frac{c^{1-\gamma}}{1-\gamma} + \beta \mathbb{E} (V(k', z') | z_j) \right\} \quad (1)$$

s.t. $c + k' = (1 + R)k + z$

$$\ln z' = \rho \ln z_j + \eta', \quad k' \geq k_{\min}.$$

Choosing Parameters

MODEL PARAMETERS			
Parameter	Description	Value(s)	Comment
γ	RRA	2 & 10	
ρ	Persist. of earnings	0.90 & 0.98	
σ_z	Uncond. std. dev. of z	0.20	
σ_η	Innovation std. dev.	$0.20 \times \sqrt{1 - \rho^2}$	Gaussian
β	Time discount factor	0.95 & 0.99	
R	Interest rate	Calibrated for $\bar{k}/\bar{z} = 5$	
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CHOICES FOR NUMERICAL SOLUTION			
N	# grid points	Experiment from 20 up	
k_0	Lower bound	$k_{\min} = -(0.6/R)z_{\min}$	
k_N	Upper bound	500	
Support of z	Discrete grid	11-states	Rouwenhorst
θ	Expansion exponent	3, 1.3, 1	
	Optimization method	Brent	
	Interpolation	3-spline or pcws-linear	
V_0	Initial guess	$V_0(k_i, z_j) = U((1+R)k_i + z_j)$	
Stop. criteria	$\max_i((V_i^{n-1} - V_i^n)/(1+ V_i^n))$	$< 10^{-7}$	i indexes grid pts

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- ▶ How to choose k_N ? As discussed, there are two ways:
 - 1 Choose such that $g(k_N, z) < k_N$ for sure. Need a very wide grid and some side calculations.
 - ▶ Inefficiently large grid but safer choice to start
 - ▶ Cannot be used when changing parameters during estimation/calibration
 - 2 Choose a large k_N but be ready to extrapolate.. **with lots of trepidation!**
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 - ▶ If you extrapolate more than $k_N - k_{N-1}$, your grid is too small.
- ▶ Reformulate the utility function so that the Bellman objective reads (recursive CES trick from Lecture 2):

$$\mathcal{V}(k, z) = \max_c \left[(1 - \beta) c^{1-\gamma} + \beta \mathbb{E} \left(\mathcal{V}(k', z')^{1-\gamma} \right) \right]^{1/(1-\gamma)} .$$

Implied Parameters

- ▶ R is calibrated to get $\bar{k}/\bar{z} = 5$. As β is varied, so will R .
- ▶ Since natural borrowing limit depends on R , it will have to be adjusted too.

Table 1: Interest rates and Borrowing Constraint

ρ	β	Interest rate		k_{\min}		min(c)/mean(z)
		2	10	2	10	2 & 10
0.9	0.95	0.050	0.030	-6.34	-10.48	0.212
	0.99	0.010	0.008	-32.56	-42.00	0.212
0.98	0.95	0.051	0.027	-6.29	-11.76	0.212
	0.99	0.009	0.005	-33.91	-61.42	0.212

Caution: Pay Attention to the Lower End

- ▶ Check if the imposed k_{\min} , together with z_{\min} imply a consumption level that is “too low”.
- ▶ Define \underline{c}_{\max} as the maximum consumption feasible in the worst state:
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- ▶ So when maximizing Bellman objective, we will be evaluating $U(c)$ at $c < \underline{c}_{\max}$!
- ▶ For example, above $\underline{c}_{\max}/z_{\min} = 0.21$. If we normalize \bar{z} to 1, and $U(c)$ is a CRRA with $\gamma = 10$, we have $U'(\underline{c}_{\max}) - (0.21)^{-9}/9 \approx 140,000$.

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- ▶ Suppose we are not careful and search down to $c = 0.05$. Then $U'(0.05) \sim 10^{13}$. Every numerical computation as well as interpolation, etc., will be a challenge.
- ▶ So, before we test any lower c , check if $c = \underline{c}_{\max}$ is a solution. Very often it will be.

Stopping Criterion: Issues to Consider

- 1 Using a tolerance based on levels is not a good idea, because:
 - 1 The numerical value of utility has no meaning (10^{-5} , 10^{-8} , etc) since it's an ordinal measure.
 - 2 Because U has wide range of variation, this imposes an “uneven” stringency condition across different parts of value function.

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 - 2 **Modified criterion:**

$$\max_{i,j} \frac{|V^n(k_i, z_j) - V^{n-1}(k_i, z_j)|}{1 + |V^n(k_i, z_j)|} < 4 \times 10^{-7}, \quad (2)$$

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- 3 Ideally, **set criterion based on decision rule deviation**. More on this later.

How To Ensure the Accuracy of Solution

Checking for Accuracy

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 - No. We will see this clearly below.
 - 2 Notice that the **CES formulation avoids** all of these problems at once: Always positive with small slope.
- ▶ Preview of Main Lesson: *There is no silver bullet*. Every test has significant false positives and negatives (or Type I and Type II errors so to speak).
- ▶ That said: there is an **essential checklist** to go through. Collectively they will be informative.

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 - 2** Plot the **derivative** of the functions, which makes problems more visible.

3. **Tighten all the screws** and solve the problem again:

- Increase number of grid points by 3X, 5X, etc.
- **Tighten tolerance criteria** (remember: more in inner loops than outer)
- If you used a less accurate method, like linear interpolation, switch to spline and redo.

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- ▶ N.B. In general, getting **prices** right **requires much higher accuracy** than **quantities**.

Algorithms:

- 1 Plain-vanilla VFI
- 2 VFI + Modified Policy Iteration Algorithm (MPIA)
- 3 VFI + MacQueen-Porteus Bounds (MQP)
- 4 Endogenous Grid Method (EGM)

Checking for Accuracy: Example

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 - $\rho = 0.98, \beta = 0.99$.
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 - $\rho = 0.98, \beta = 0.99$.
 - Choose R as before. Set $k_{\max} = 500$ (times average income).
- ▶ Test **four sets of methods** for solving this problem:
 - (i) CRRA specification, spline interpolation;
 - (ii) CRRA specification, **linear** interpolation,
 - (iii) **CES** specification, spline interpolation; and
 - (iv) **CES** specification, **linear** interpolation.

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- ▶ Take a 20-point grid for k and $\theta = 3$. If convergence fails on this grid \rightarrow increase grid points by 20 until convergence.

Table 2: Convergence Time of VFI Algorithm, Four Methods

Method:	RRA	
	2	10
CES-spline	8.6	15.9
CES-linear	7.8	14.2
CRRA-spline	12.7	55.1*
CRRA-linear	14.8	20.4

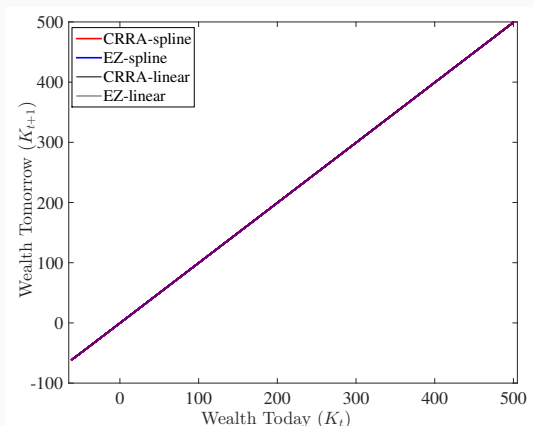
*Note: The baseline number of grid points is 20. This looks extremely low and it is. But we will see how it yields very accurate solutions if the right methods are applied. If convergence is not obtained for this specification we increase grid points by 20 up to 100. *60 point grid was needed. RRA is coded as integer (not real) for efficiency.*

How did CRRA-linear pull off this feat?

- ▶ If it sounds too good to be true, it usually is.
- ▶ But, the convergence criteria was satisfied to 10^{-7} ??
 - Yes, **but on the 20 grid points**. We didn't check off grid points!
- ▶ Suppose, we don't notice that. Let us plot the decision rule.

Plot decision rule: Version 1, $RRA=10$

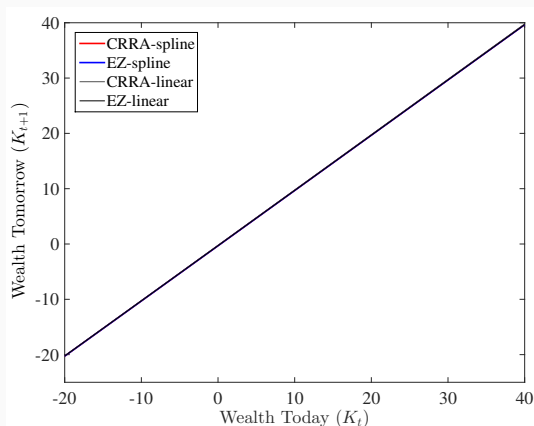
Figure 1: Savings Decision Generated by Four Solutions of the Same Model



- The first reaction to this figure is often that they all look the same. Nothing interesting.. or is there?

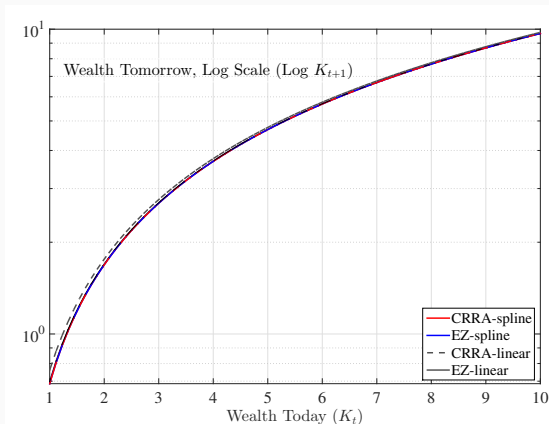
Plot: Version 2, zoom in, RRA=2

Figure 2: Savings Decision Generated by Four Solutions of the Same Model



- ▶ Let's look at the easier case, RRA=2. Zoom in by shrinking x-axis range. Still nothing..

Plot: Version 3, Zoom in Again, $\log(K')$



- Zoom further in to $k_t \in [1, 10]$ and plot log of savings.. A little gap appears at the low end. This gap will increase further at lower k values near limit.

Does this deviation matter? Statistics from Simulation

	Mean	Std. Dev.	Max	Min
	Capital			
CES-S	11.26	28.81	100.24	-33.912
CRRA-S	11.18	28.77	99.76	-33.912
CES-L	13.34	28.90	101.86	-33.912
CRRA-L	31.83	28.66	80.78	-33.912

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Consumption				
CES-S	1.14	0.325	2.16	0.213
CRRA-S	1.13	0.325	2.16	0.213
CES-L	1.15	0.326	2.17	0.213
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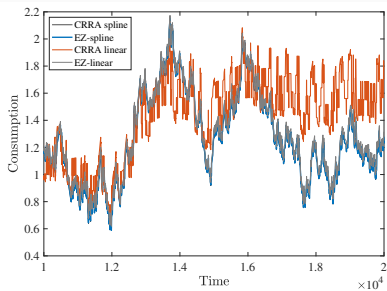
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$K_{t+1} - K_t$				
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CRRA-S	0.001	0.132	0.613	-0.324
CES-L	0.001	0.130	0.609	-0.318
CRRA-L	0.001	0.075	0.566	-0.233

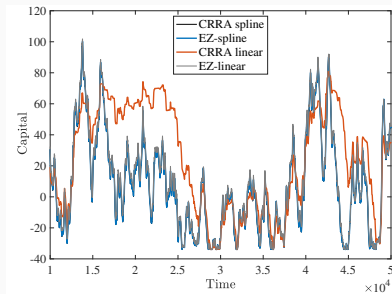
That Doesn't Look Good!

Figure 3: Simulated Paths Generated by Four Solutions Methods, $RRA=2$

(a) Consumption Path



(b) Capital Path



- ▶ CRRA-Linear is way off for thousands of periods at a time. Which is why it generates very different results for some statistics.
- ▶ CES-L (marked EZ) is a little off in extremes but hard to even see. With 40-point grid, all stats within 1% of more accurate CES-spline.

- ▶ Verifying the accuracy of the solution typically requires careful detective work.
 - Checking the convergence criterion is the first of many steps!
 - Plotting the value function and decision rules is essential (esp. scaling up, plotting derivatives, etc.). It is necessary and can reveal very obvious problems but it is nowhere near sufficient.

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- ▶ Then simulate all key variables and compare the statistics that are critical for your work.
- ▶ It is typically **much harder to compute prices accurately** than quantities! We will see examples of this in lectures 6 and 7.