
Thinking Big:

Determinacy and Large-Scale Solutions in the Sequence Space

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Solving het-agent models to first order

- ❖ Two key considerations:
 - ❖ Size of idiosyncratic state space S
 - ❖ Number of endogenous aggregate variables N
- ❖ **State-space approach:** costly when S large. Has determinacy criterion.
[Reiter, Ahn-Kaplan-Moll-Winberry-Wolf, Bayer-Luetticke, ...]
- ❖ **Sequence-space approach:** fast when S large, costly when N large.
[Boppart-Krusell-Mitman, Auclert-Bardoczy-Rognlie-Straub, ...]
- ❖ How do we solve models when both S and N are large?

Introducing... SSJ 2.0!

- ❖ Obtain a **structure theorem** for sequence-space Jacobians
- ❖ When het-agent model is *stationary*, Jacobians are *quasi-Toeplitz*:

$$\mathbf{J} = \mathbf{T}(\mathbf{j}) + \mathbf{E}$$

i.e. sum of a Toeplitz operator $\mathbf{T}(\mathbf{j})$ and a compact operator \mathbf{E}

- ❖ Exploit this structure (generalizing to $N > 1$ case) in many ways:
 - ❖ **Winding number criterion** on \mathbf{j} for **determinacy** & existence
 - ❖ More accurate computations working directly with \mathbf{j} and \mathbf{E}
 - ❖ Using $\mathbf{T}(\mathbf{j}^{-1})$ as guess for \mathbf{J}^{-1} gives **rapid iterative solution**, even when N large

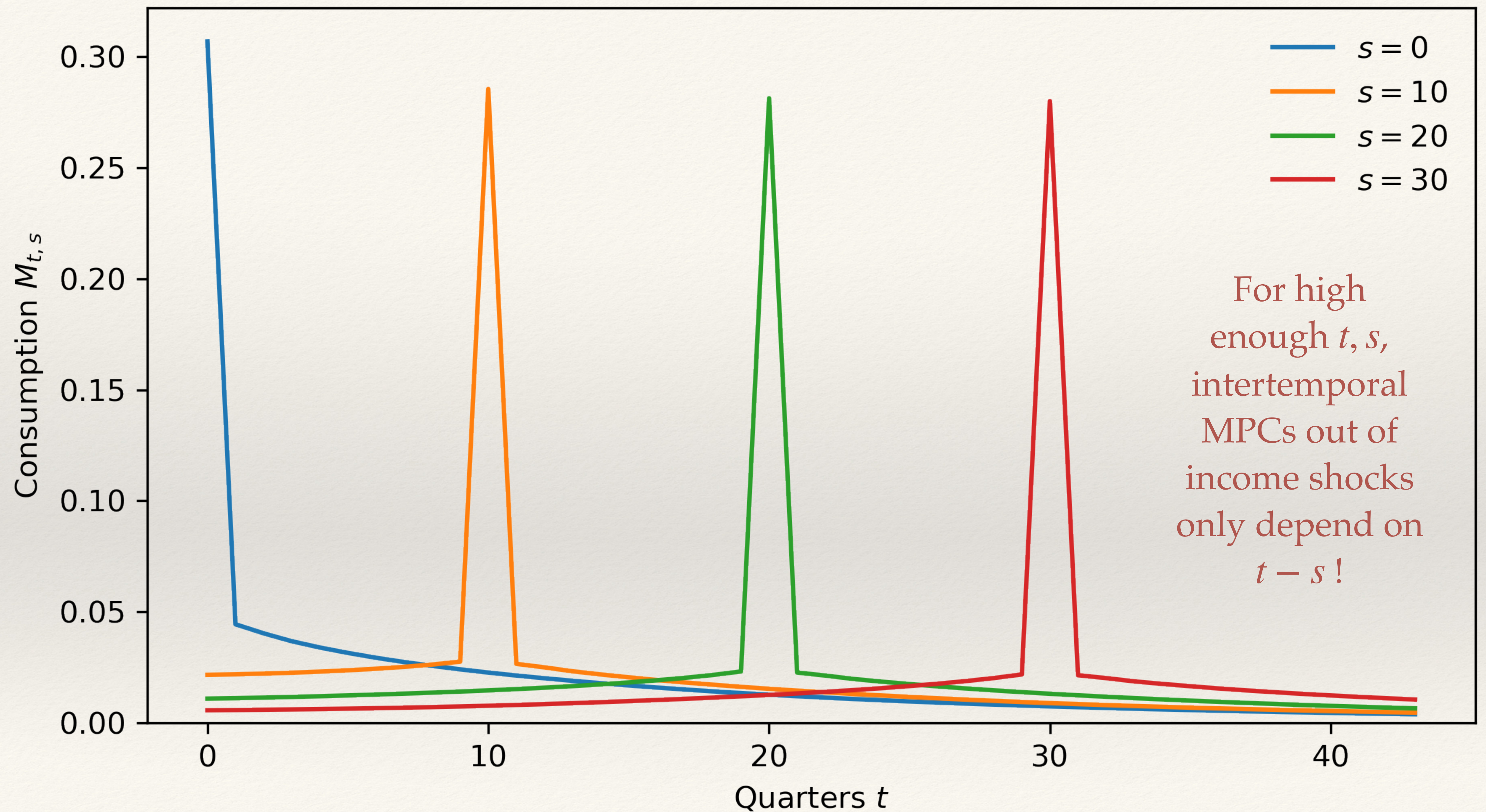
(When represented as matrix, “Toeplitz” means “each diagonal is constant.”)

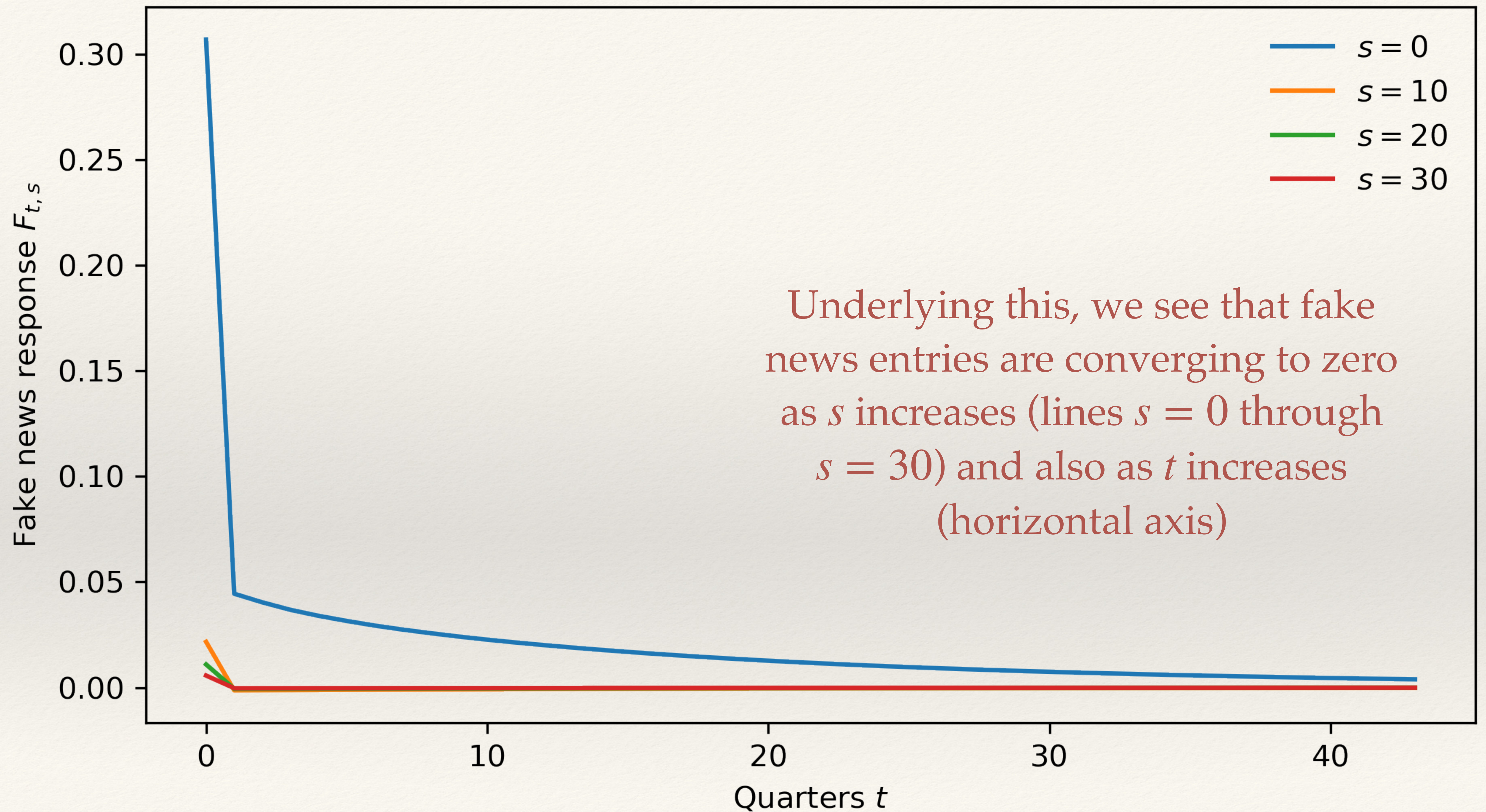
Structure theorem

Building block: the fake news matrix

- ❖ Recall that Jacobian is cumulative diagonal sum of fake news matrix, i.e. $J_{2,3} = F_{0,1} + F_{1,2} + F_{2,3}$
- ❖ If $F_{t,s} \rightarrow 0$ fast enough as $t, s \rightarrow \infty$, then each diagonal of \mathbf{J} will converge to a constant j_{t-s}
- ❖ Interpretation: if shock anticipated far enough in advance, only position $t - s$ vs. shock matters

$$\begin{pmatrix} F_{00} & F_{01} & F_{02} & \cdots \\ F_{10} & F_{11} & F_{12} & \cdots \\ F_{20} & F_{21} & F_{22} & \cdots \\ \vdots & \vdots & \vdots & \ddots \end{pmatrix}$$





Why would fake news entries converge to zero?

- ❖ $F_{t,s}$ is the effect at date t of having thought at date 0 that there would be shock at date s
- ❖ Plausible that:
 - ❖ Effect on date-0 policy decays as horizon s increases
 - ❖ Persistent effect from date-0 policy decays as t increases
- ❖ If both, we say model is **stationary**

$$\begin{pmatrix} F_{00} & F_{01} & F_{02} & \cdots \\ F_{10} & F_{11} & F_{12} & \cdots \\ F_{20} & F_{21} & F_{22} & \cdots \\ \vdots & \vdots & \vdots & \ddots \end{pmatrix}$$

Fake news matrix of stationary het-agent models

- ❖ For $t > 0$, SSJ paper shows that $F_{t,s}$ takes the form:

$$F_{t,s} = \mathcal{E}'_{t-1} \mathcal{D}_s$$

- ❖ \mathcal{D}_s is effect of anticipating date- s shock at date 0 on the date-1 distribution
- ❖ \mathcal{E}_{t-1} is the *expectation function*, giving expected quantity of interest (e.g. consumption, assets) in $t - 1$ periods, if steady-state policy followed
- ❖ Eventual decay at rate of at least $\mathcal{D}_s \sim \beta^s$ from discounting as $s \rightarrow \infty$ (Note $\mathbf{1}'\mathcal{D}_s = 0$, since distribution mass preserved.)
- ❖ If Λ_{ss} ergodic, \mathcal{E}_t approaches multiple of $\mathbf{1}$ at rate γ^t , with γ largest eigenvalue < 1
- ❖ If both: $|F_{t,s}| \leq K\beta^s\gamma^t \leq K\Delta^{s+t}$, for $\Delta \equiv \max(\beta, \gamma) < 1$, we call model “stationary”

Using this to characterize Jacobian

- ❖ If $|F_{t,s}| \leq K\Delta^{s+t}$, then we can define and guarantee convergence for:

$$j_u = F_{u,0} + F_{u+1,1} + F_{u+2,2} + \dots$$

i.e. the sum of all entries in \mathbf{F} on the u th lower diagonal (analogous for $u < 0$)

- ❖ We then have $J_{t+i,s+i} \rightarrow j_{t-s}$ as $i \rightarrow \infty$ for any t, s [“asymptotically Toeplitz”]

- ❖ j_{t-s} is response $t - s$ periods after shock, if shock infinitely well-anticipated

- ❖ We also have $J_{t,s} - j_{t-s} \equiv E_{t,s} = -(F_{t+1,s+1} + F_{t+2,s+2} + \dots) \leq \frac{K}{1 - \Delta^2} \Delta^{t+s}$

[E is “correction” vs. exact Toeplitz, with terms on diagonal not yet summed]

Quasi-Toeplitz form

- ❖ Summing up: for any Jacobian \mathbf{J} of a stationary het-agent model, we have $J_{t,s} - j_{t-s} = E_{t,s}$, where $|E_{t,s}|$ bounded by multiple of Δ^{t+s}
- ❖ Can write as sum of Toeplitz operator $T(\mathbf{j})$ and **compact** “correction” operator \mathbf{E} :

$$\mathbf{J} = \underbrace{\begin{pmatrix} j_0 & j_{-1} & j_{-2} & & \\ j_1 & j_0 & j_{-1} & \ddots & \\ j_2 & j_1 & j_0 & \ddots & \\ & \ddots & \ddots & \ddots & \end{pmatrix}}_{\equiv T(\mathbf{j})} + \mathbf{E}$$

(“Compact” on ℓ^2 means limit of finite-rank operators, behaves similarly to a finite-dimensional matrix. $|E_{t,s}| \leq K'\Delta^{t+s}$ readily implies this.)

- ❖ This is called a **quasi-Toeplitz operator**, and has many nice properties!

Why might this representation be useful?

- ❖ Quasi-Toeplitz operators are **closed** under addition, multiplication, etc., and even inversion, assuming an inverse exists [e.g. Bini, Massi, Robol 2019]
 - ❖ so we can chain along DAG, solve for unknowns, and stay quasi-Toeplitz!
 - ❖ (simple aggregate equations already have Toeplitz Jacobians)
- ❖ Toeplitz has nice theory for **existence & uniqueness** of solutions; this mostly extends to quasi-Toeplitz
- ❖ In practice, **E** often very well-approximated by **low-rank** matrix
 - ❖ So can represent and work with **J** a lot more efficiently than ordinary $T \times T$

Existence and uniqueness of solutions

The winding number

- ❖ Recall: j_k is the entry on the k th lower diagonal of Toeplitz $T(\mathbf{j})$
- ❖ $\{j_k\}_{k=-\infty}^{\infty}$ is a two-sided sequence, and we say its **symbol** is the Laurent series

$$j(z) \equiv \sum_{k=-\infty}^{\infty} j_k z^k$$

- ❖ The **winding number** $\text{wind}(j)$ is # of times $j(z)$ rotates counterclockwise around 0 as z goes counterclockwise around the unit circle
- ❖ Standard result: $T(\mathbf{j})$ is invertible iff its winding number is zero!
 - ❖ If $\text{wind}(j) < 0$, then surjective but not injective; vice versa if $\text{wind}(j) > 0$

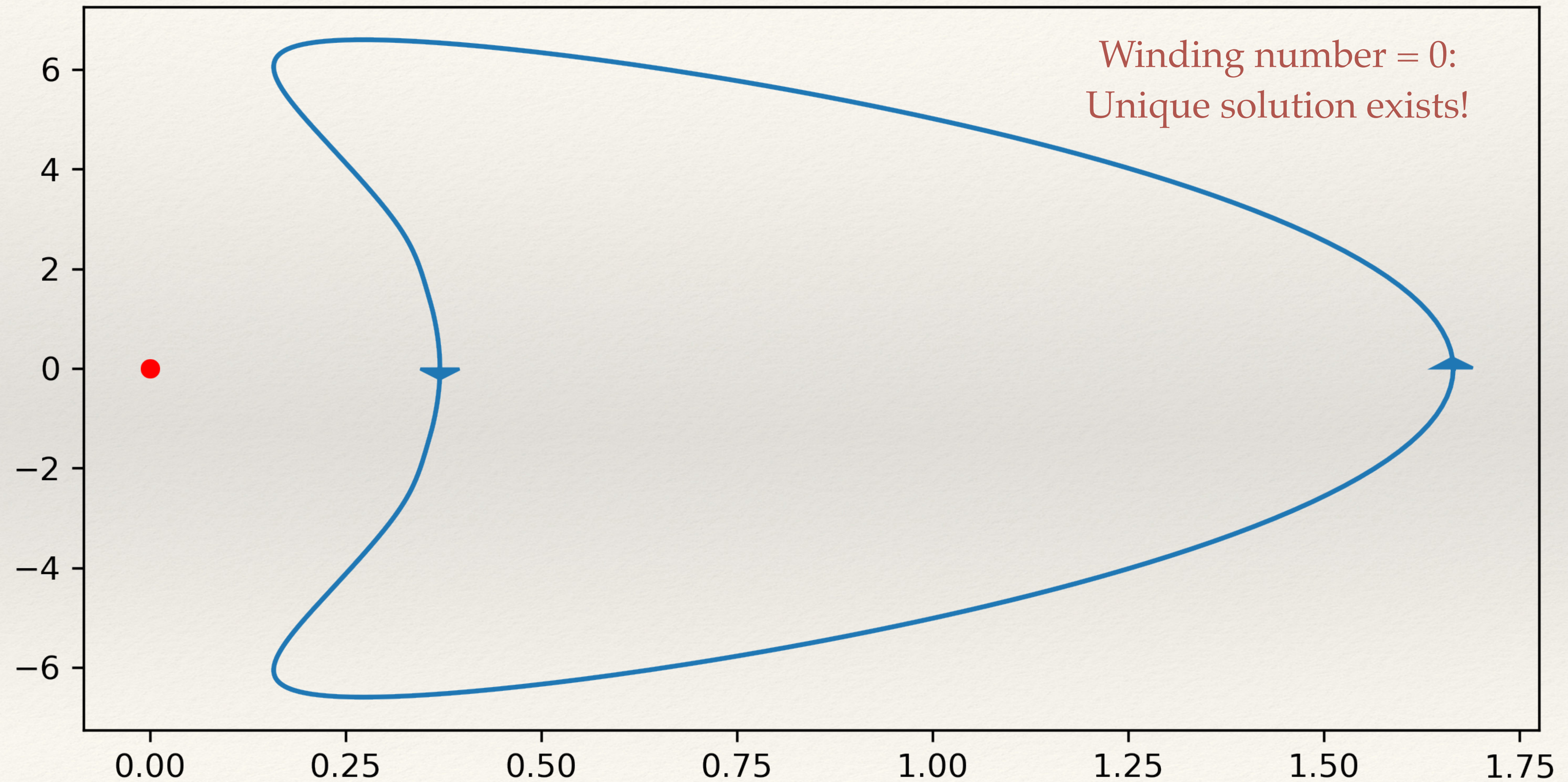
Simple examples of winding number

- ❖ Lag operator $\mathbf{L} : (x_0, x_1, \dots) \mapsto (0, x_0, x_1, \dots)$ is injective but not surjective
 - ❖ symbol $j(z) = z$, with winding number 1 (j goes counterclockwise once)
- ❖ Lead operator $\mathbf{F} : (x_0, x_1, \dots) \mapsto (x_1, x_2, \dots)$ is surjective but not injective
 - ❖ symbol $j(z) = z^{-1}$, with winding number -1 (j goes clockwise instead)
- ❖ More general result: winding number $n > 0$ implies n dimensions missing from range, $n < 0$ implies null-space of dimension $-n$ [consider examples \mathbf{L}^n and \mathbf{F}^n]
 - ❖ loosely, winding number of n says “this Toeplitz similar to taking n lags”

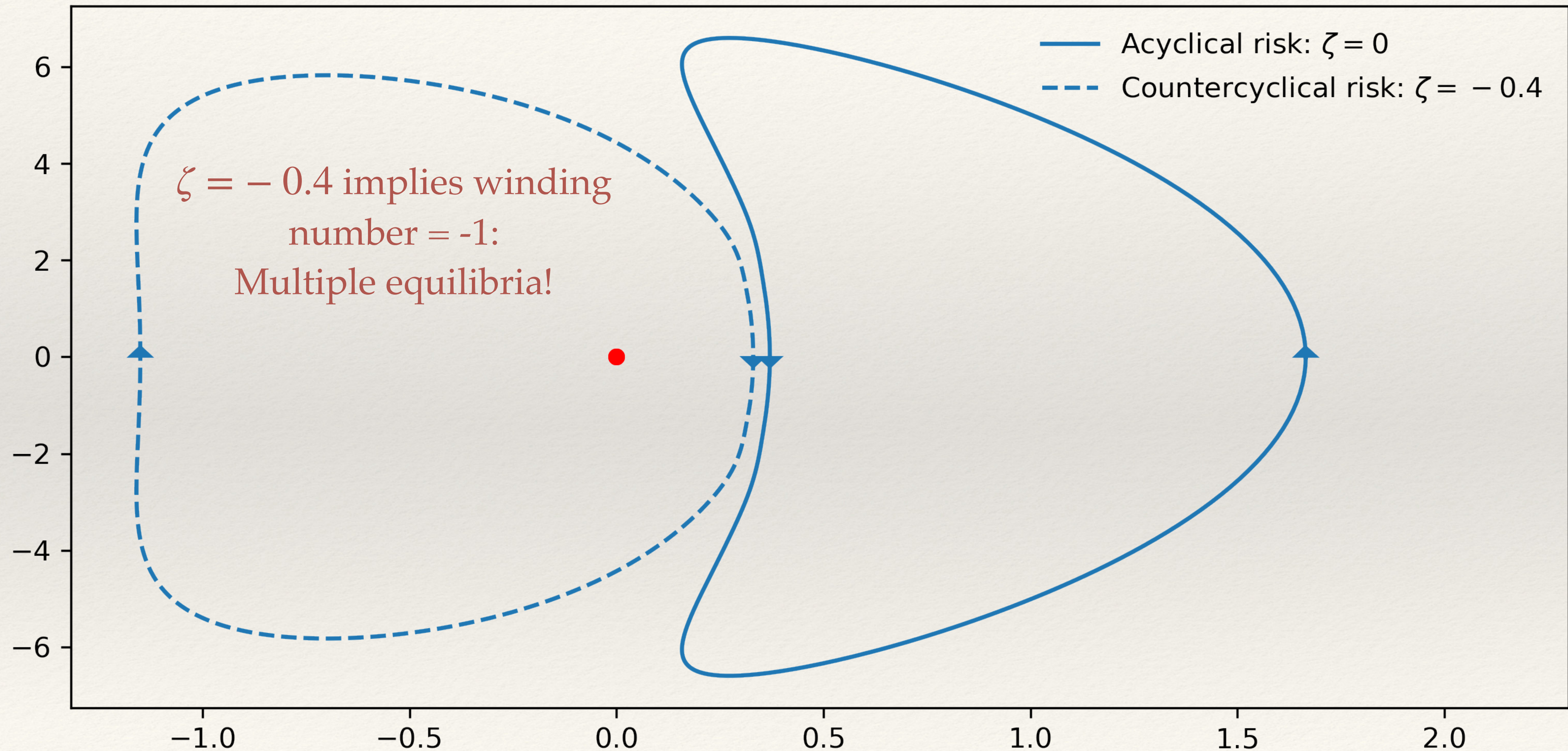
Extending to quasi-Toeplitz

- ❖ Same result doesn't always hold for quasi-Toeplitz, but holds “generically” on an open and dense set of \mathbf{E} (i.e. almost all \mathbf{E}):
 - ❖ If $\text{wind}(j) = 0$, then \mathbf{J} is generically invertible
 - ❖ If $\text{wind}(j) < 0$, then \mathbf{J} is not injective, but generically surjective
 - ❖ If $\text{wind}(j) > 0$, then \mathbf{J} is not surjective, but generically injective
- ❖ Solving asset market IKC $\mathbf{A}(d\mathbf{Y} - d\mathbf{T}) = d\mathbf{B}$, then (generically) exists unique solution if $\text{wind}(a) = 0$, indeterminacy if $\text{wind}(a) < 0$, nonexistence if > 0

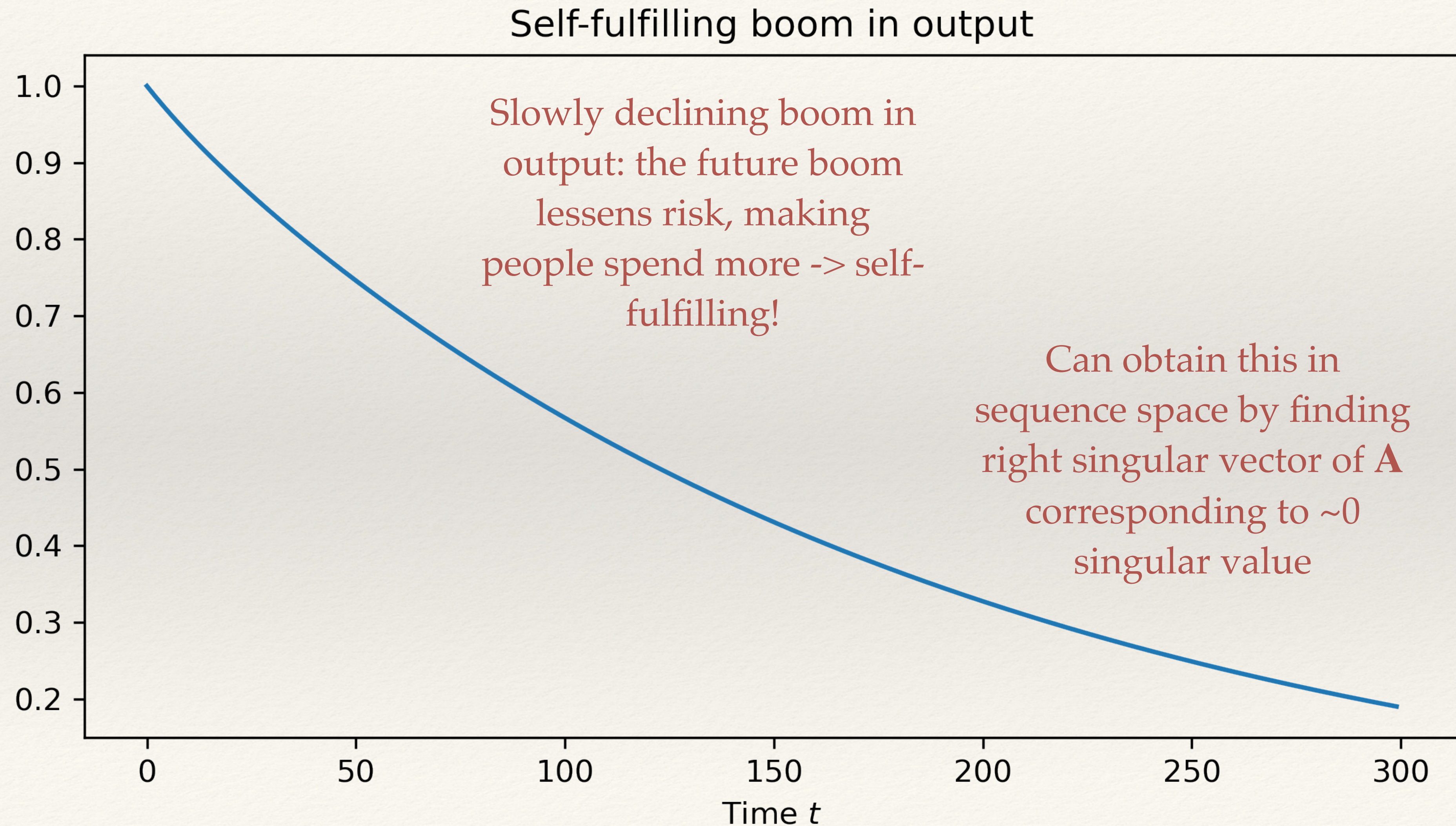
Winding number plot for A with standard calibration



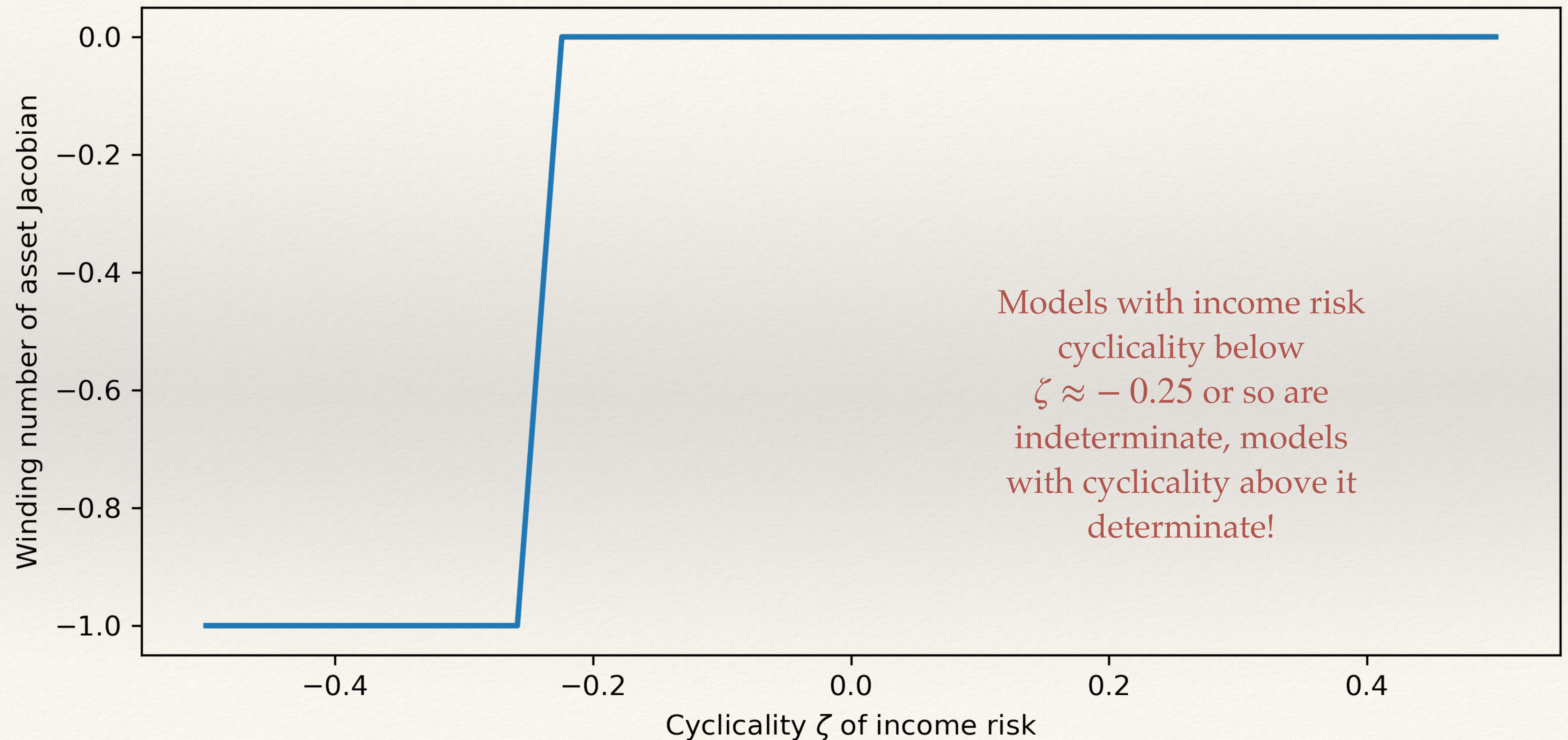
Compare to case with countercyclical risk



Can find shape of multiplicity in sequence space



How does winding number vary with ζ in general?



Block quasi-Toeplitz case

- ❖ Say we have N^2 quasi-Toeplitz matrices from N unknowns to one of N targets
- ❖ Can think of this as being one **block quasi-Toeplitz operator**, like a quasi-Toeplitz but where entries are each $N \times N$ blocks
- ❖ Then $\{j_k\}_{k=-\infty}^{\infty}$ is two-sided sequence of $N \times N$ matrices, so *matrix-valued* $j(z)$:

$$j(z) \equiv \sum_{k=-\infty}^{\infty} j_k z^k$$

- ❖ Winding number test still holds generically, now for $\text{wind}(\det j)$
- ❖ Important case in practice, we're still working out details [see also Onatski 2006]

Operations with quasi-Toeplitz operators:
No more truncation!

Directly use quasi-Toeplitz form

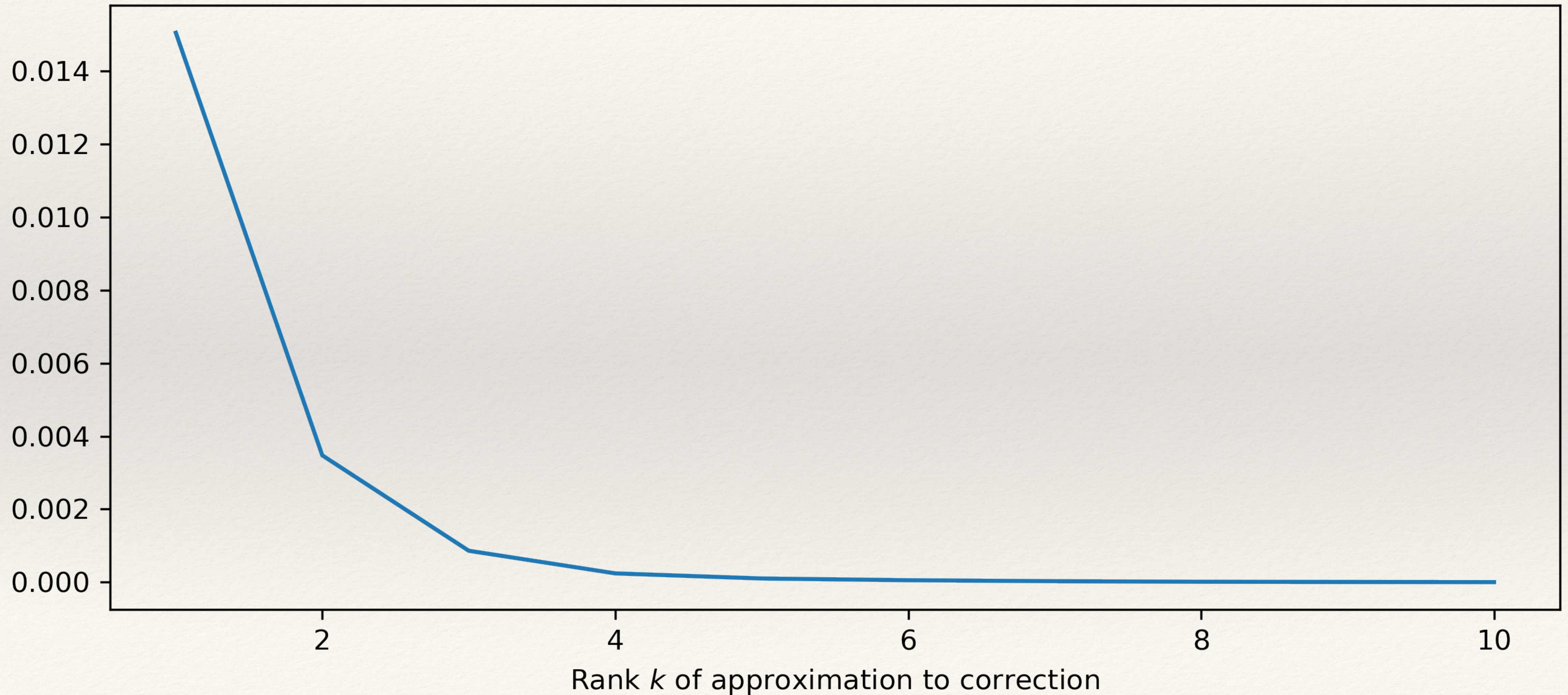
- ❖ We have quasi-Toeplitz representation $\mathbf{J} = T(\mathbf{j}) + \mathbf{E}$ of Jacobians
- ❖ So far, we've used the winding number of \mathbf{j} to assess determinacy
- ❖ Another benefit: use this special form directly to do computations!
 - ❖ key supporting fact: in practice \mathbf{E} well-approximated by low rank
 - ❖ more efficient, no longer fully truncating at T ($T(\mathbf{j})$ in principle infinite)
- ❖ Our inspiration: Bini, Masei, Robol (2019), “Quasi-Toeplitz matrix arithmetic: a MATLAB toolbox”, paper from applied math literature

Why is this math so nice?

- ❖ Suppose we can write $\mathbf{J} = T(\mathbf{j}) + \mathbf{U}\mathbf{V}'$, where \mathbf{U}, \mathbf{V} are $n \times k$ matrices
 - ❖ if k low, we have a *low-rank approximation* of $\mathbf{E} \approx \mathbf{U}\mathbf{V}'$
 - ❖ represents \mathbf{J} in a more concise way
- ❖ Can define algebra of operations on $\mathbf{j}, \mathbf{U}, \mathbf{V}$
 - ❖ e.g. multiplying \mathbf{J}_1 and \mathbf{J}_2 involves multiplying $T(\mathbf{j}_1)$ and $T(\mathbf{j}_2)$, which produces quasi-Toeplitz of form $T(\mathbf{j}_1\mathbf{j}_2) + \mathbf{U}\mathbf{V}'$; and also $T(\mathbf{j}_1)$ times \mathbf{U}_2 , etc.
 - ❖ similar, though a bit more complicated, for inversion

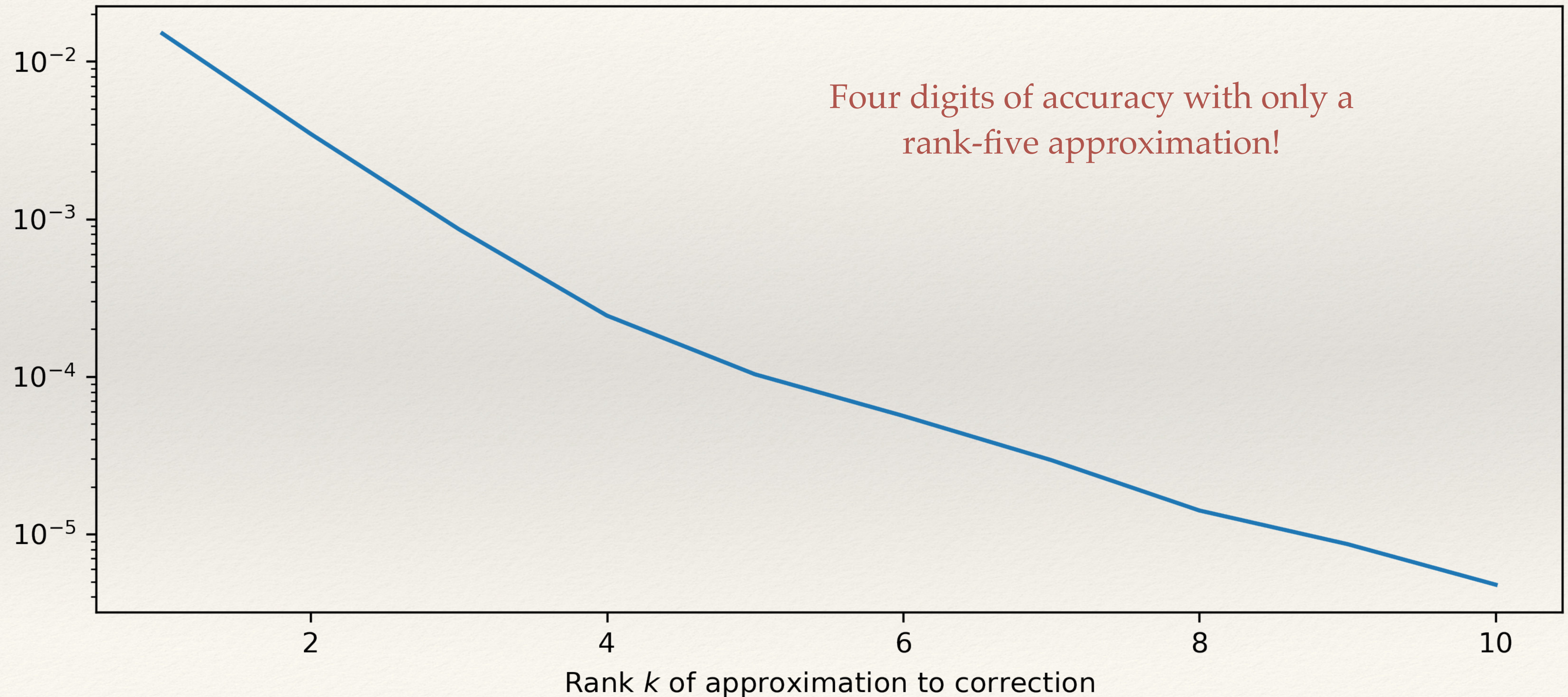
How well can we approximate \mathbf{A} ?

Maximum error in \mathbf{A} by approximation rank



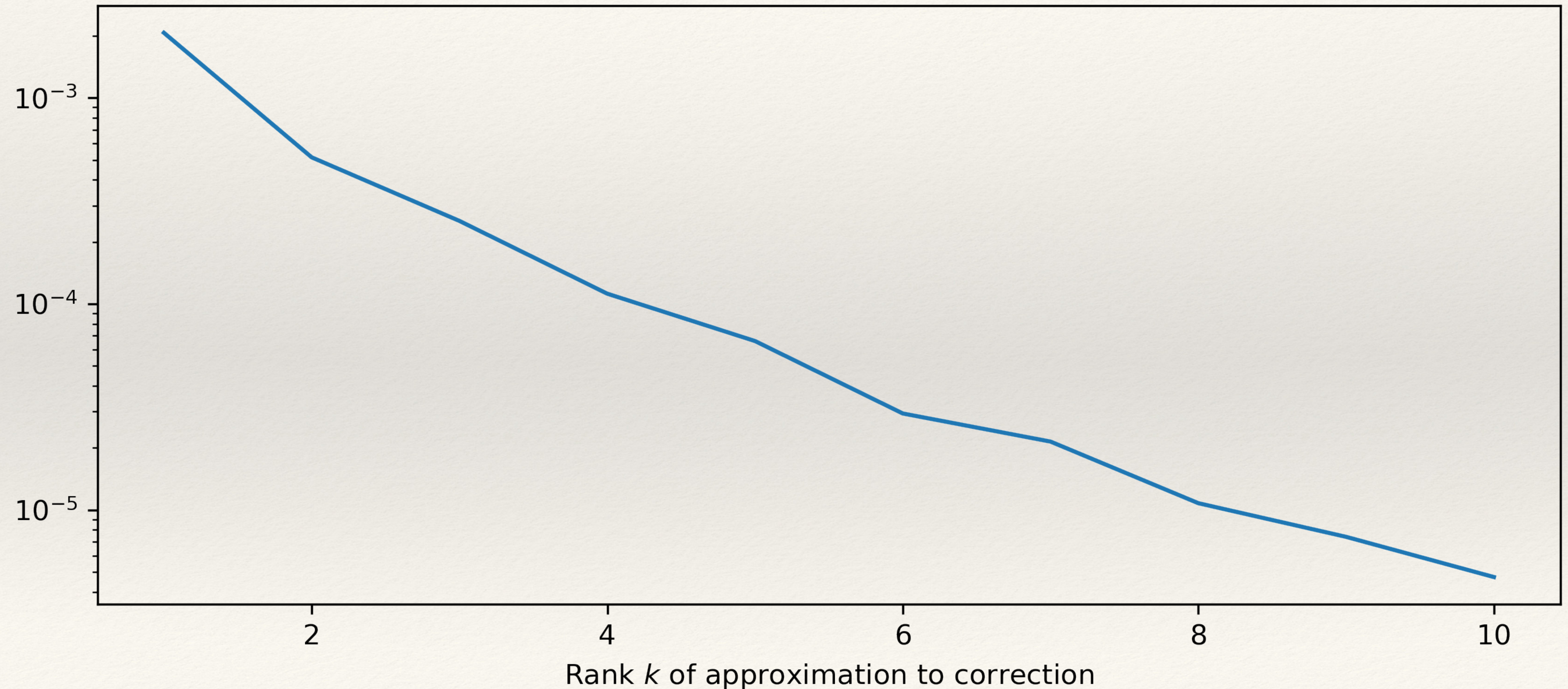
Big easier to visualize with a log scale...

Maximum error in **A** by approximation rank



Low-rank approximation to \mathbf{A}^{-1} also close!

Maximum error in \mathbf{A}^{-1} by approximation rank



How “compressed” is this?

- ❖ Suppose that $T = 1000$, so that \mathbf{A} is 1000×1000 , with 1 million entries!
- ❖ Can store corresponding \mathbf{a} of length only ~ 2000
- ❖ Then \mathbf{U}, \mathbf{V} in a rank-five approximation to \mathbf{E} each size 5000
- ❖ So, can store near-exact approximation with only 12,000 numbers!
 - ❖ Compression of more than 80x
 - ❖ Actually can do **far better**, since many entries in $\mathbf{a}, \mathbf{U}, \mathbf{V}$ near zero
 - ❖ Doing math with this will be more accurate than with truncated \mathbf{A} , since no error from artificial T [implementation details too much for today, though!]

Alternative: use structure for iterative solutions
(and solve giant models in the process!)

First point: easy to get Toeplitz part of inverse

- ❖ Suppose we want to solve $\mathbf{A}d\mathbf{Z} = d\mathbf{B}$
- ❖ \mathbf{A}^{-1} is quasi-Toeplitz of form $T(\mathbf{a}^{-1}) + \mathbf{E}$, with \mathbf{E} low-rank like we saw
- ❖ Key point: \mathbf{a}^{-1} is **really** easy to calculate!
 - ❖ Get $a(z)$ at many z using FFT, then go from $a(z)^{-1}$ to \mathbf{a}^{-1} with inverse FFT
 - ❖ Cost is only $O(T \log T)$, way cheaper than $O(T^3)$ matrix inversion
 - ❖ What can we do with just $T(\mathbf{a}^{-1})$?
 - ❖ [conceptually, \mathbf{a}^{-1} is inverse for infinitely-well-anticipated shocks]

What can we do with \mathbf{a}^{-1} ?

- ❖ Start with $(T(\mathbf{a}) + \mathbf{E})d\mathbf{Z} = d\mathbf{B}$, multiply both sides by $T(\mathbf{a}^{-1})$:

$$T(\mathbf{a}^{-1})(T(\mathbf{a}) + \mathbf{E})d\mathbf{Z} = T(\mathbf{a}^{-1})d\mathbf{B}$$

- ❖ Both $T(\mathbf{a}^{-1})T(\mathbf{a}) - \mathbf{I}$ and $T(\mathbf{a}^{-1})\mathbf{E}$ compact, well-approximated by low rank, so can be written in form

$$(\mathbf{I} + \mathbf{C})d\mathbf{Z} = d\mathbf{y}$$

- ❖ Iterative method (GMRES) **very** good at solving $(\mathbf{I} + \mathbf{C})^{-1}d\mathbf{y}$ if \mathbf{C} low-rank

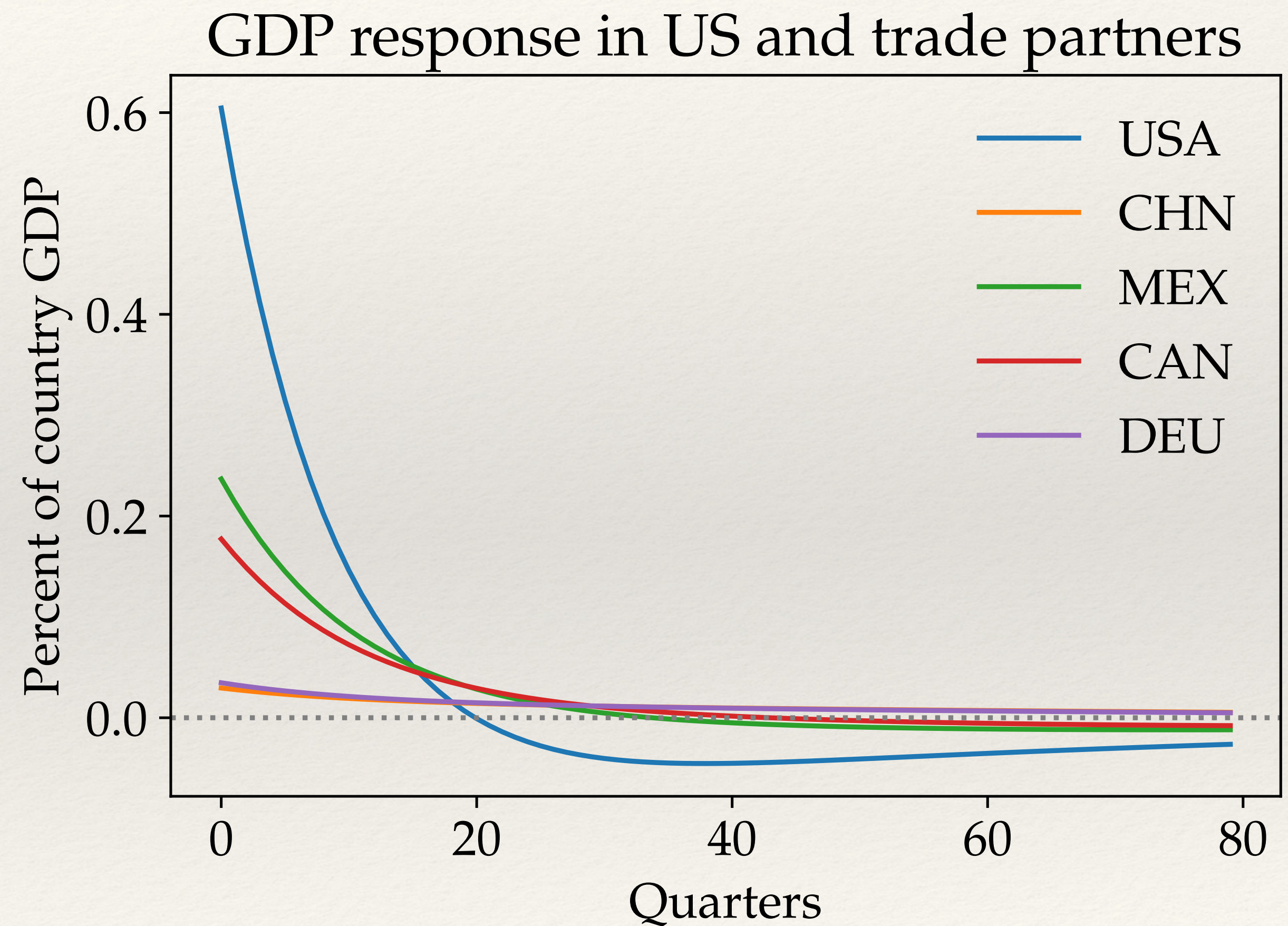
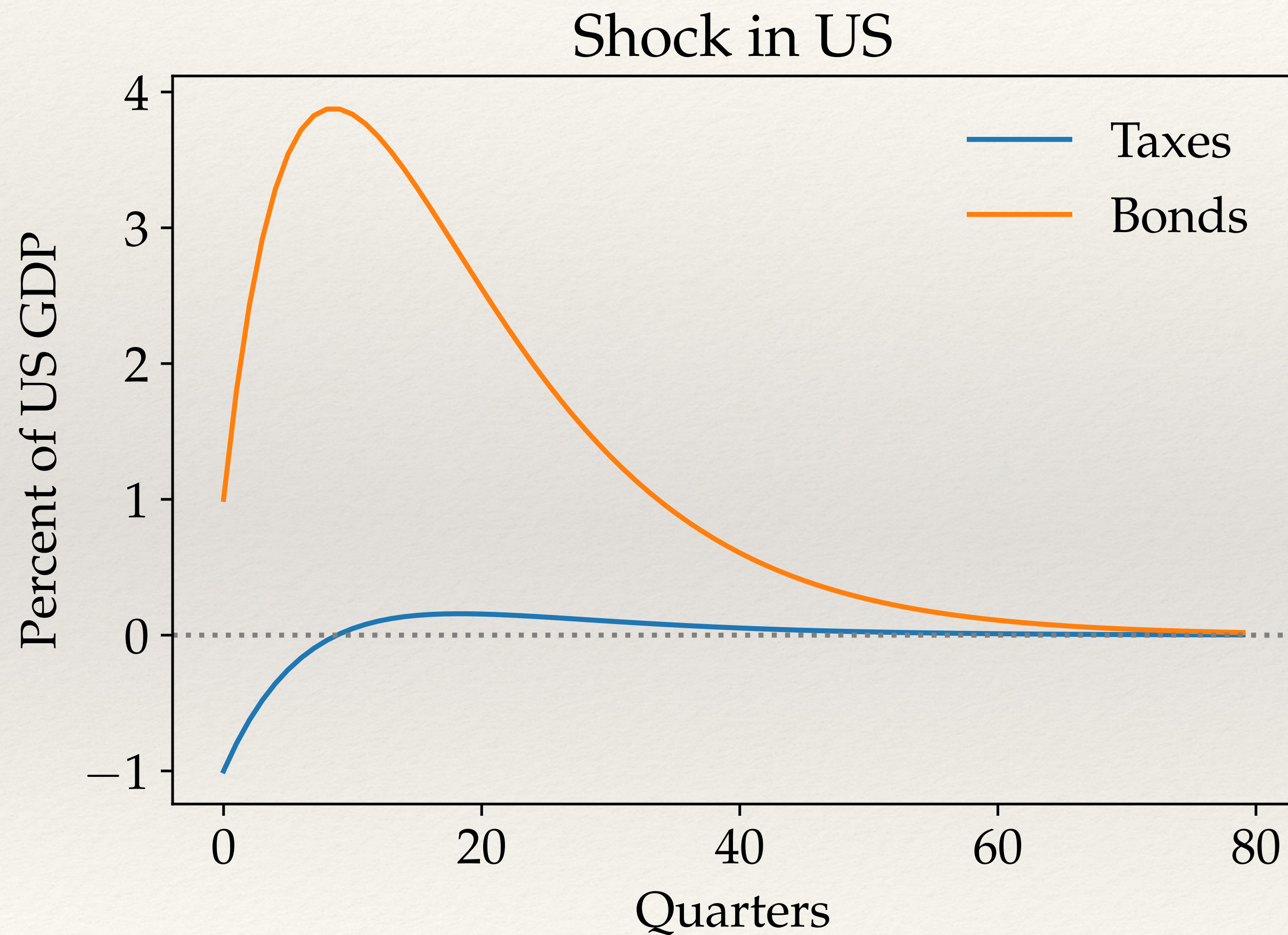
[Multiplying by $T(\mathbf{a}^{-1})$ is called “preconditioning”.]

- ❖ Cheap, doesn't require explicitly forming new matrices like \mathbf{C}

We expand this to HUGE model

- ❖ N -country extension of IKC model, constant r in each country n
- ❖ Fiscal policy in n chooses $\{B_t^n, T_t^n\}$ consistent with budget constraint
- ❖ n spends share $\Pi_{n,n'}$ on output from others n' , take from data for 177 countries
- ❖ Assume same HA model in each n , for simplicity assume all share \mathbf{A}, \mathbf{M}
- ❖ Solve for GDP $\{Y_{nt}\}$ in all N countries, in response to US deficit-financed tax cut, need long horizon $T = 1000$
- ❖ Usual sequence-space approach: Jacobian size $(177,000)^2$: can't even store!
- ❖ With iterative approach, solves in a few seconds on laptop!

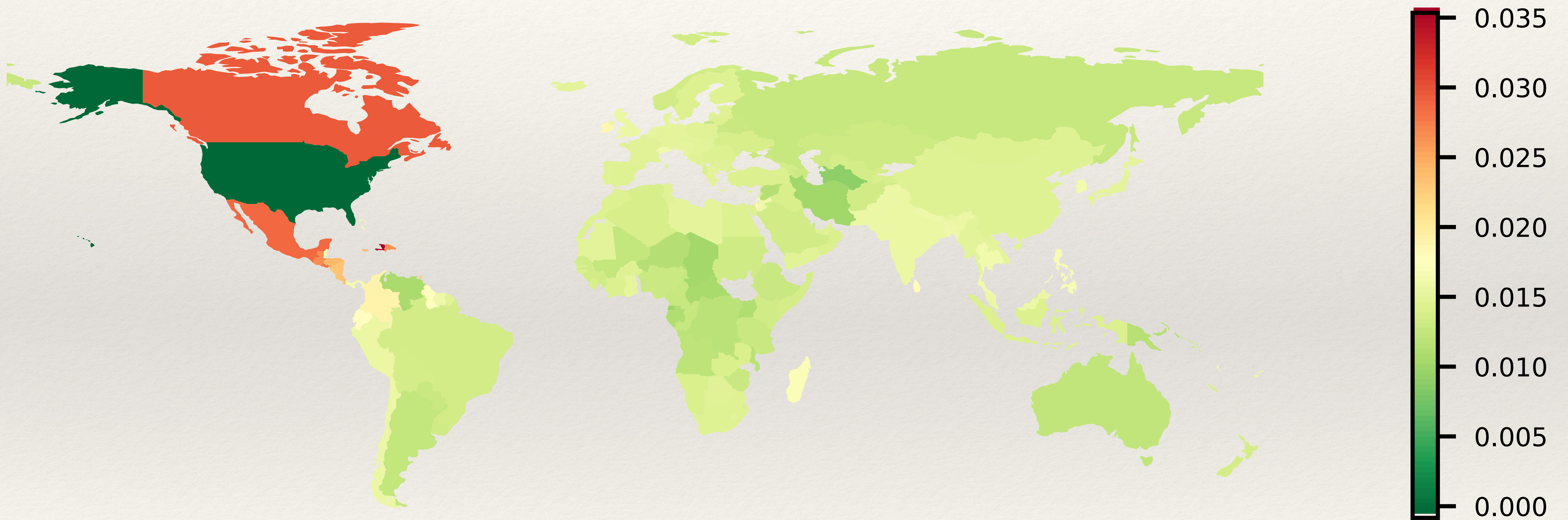
Peek at solution: selected countries over time



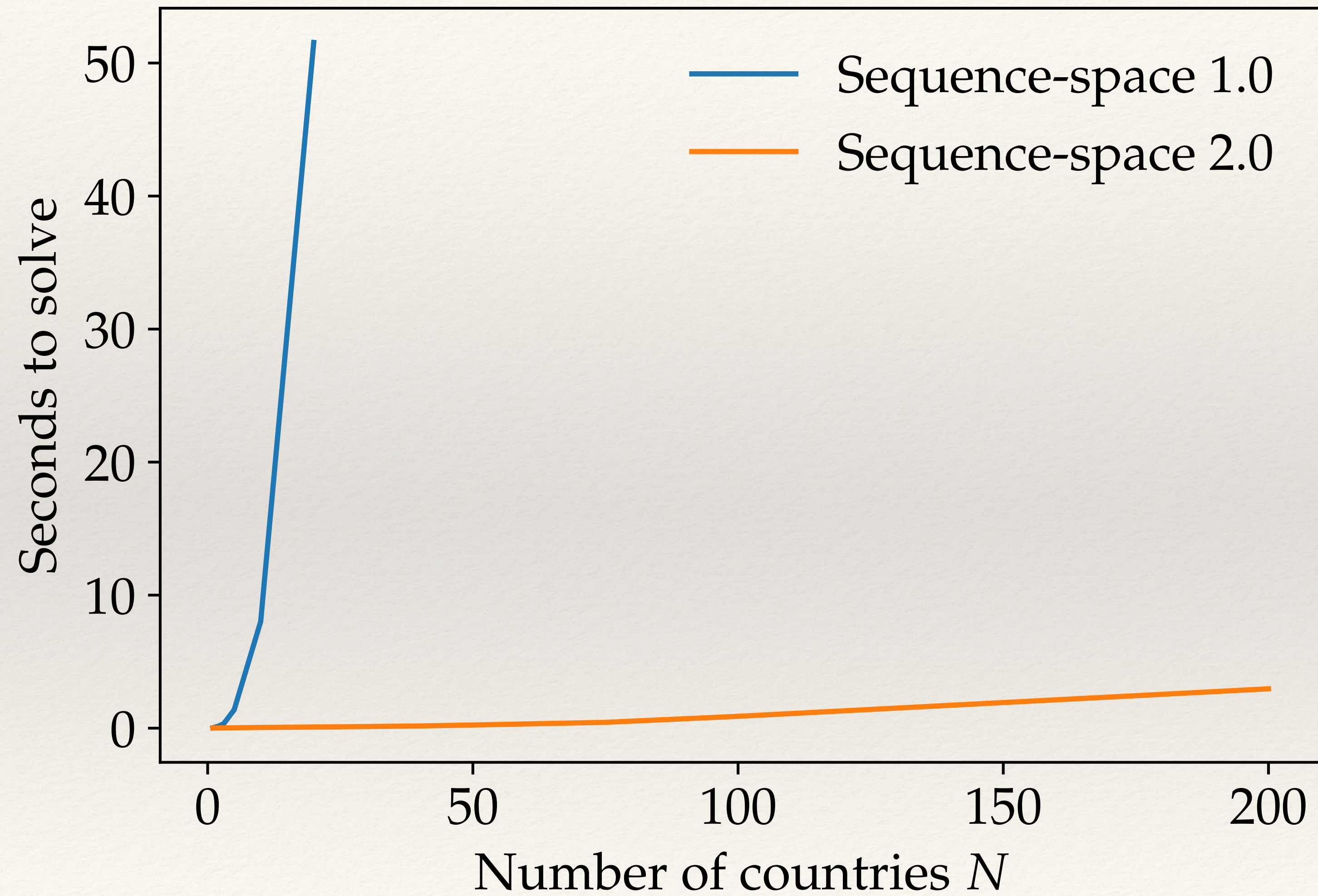
Peek at solution: on impact across countries



Peek at solution: after 20 quarters across countries

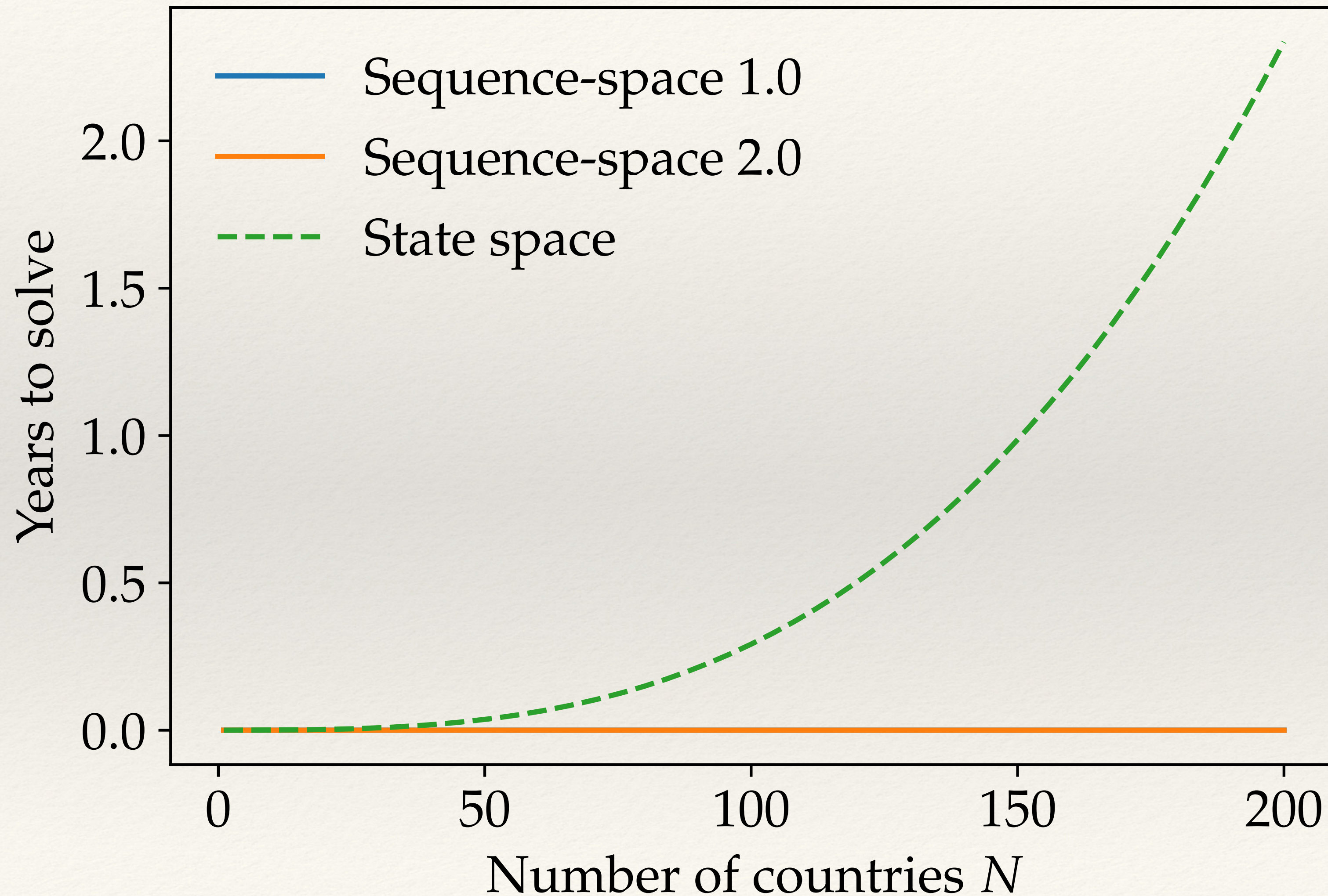


How fast is this?



How fast is this? Compare to state space

(Extrapolation
using cubic
cost of solving
state-space
system!)



Conclusion

- ❖ Quasi-Toeplitz structure of Jacobians delivers:
 - ❖ winding number test for determinacy
 - ❖ **already ready to use!**
 - ❖ faster, truncation-free computations:
 - ❖ **still in development**, bypasses major issue with sequence space
- ❖ extremely fast iterative computations, even in huge models
 - ❖ **also still in development**, but solves 177-country HANK in 3 seconds!!