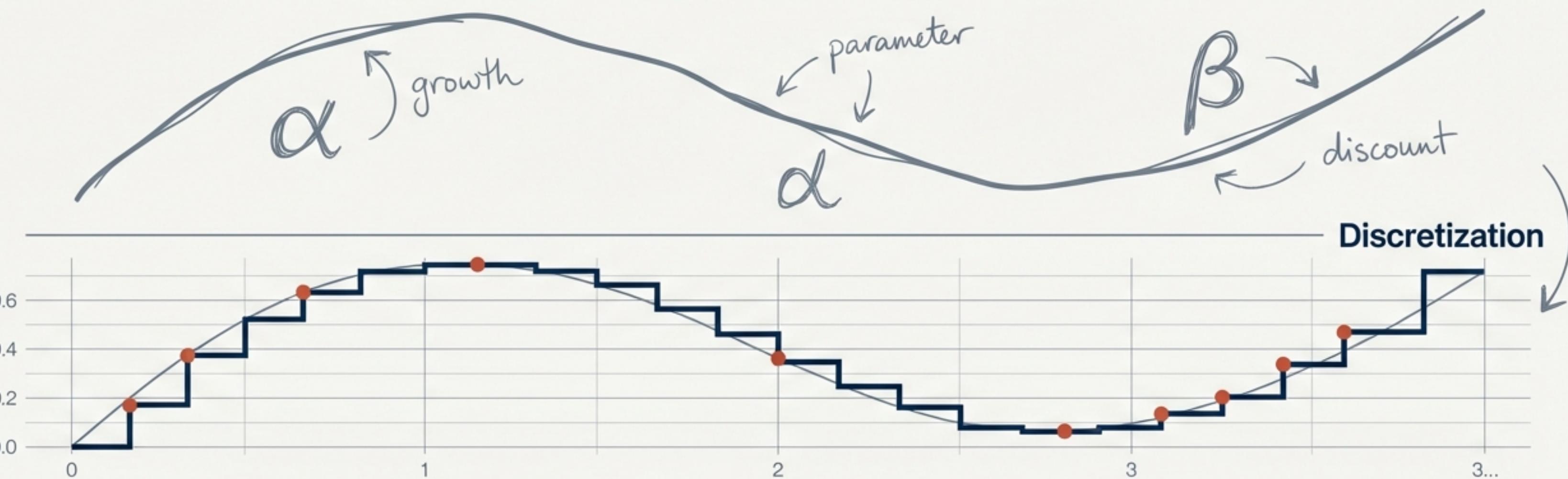


# Computational Macroeconomics: The Toolkit

*From Function Approximation to Dynamic Programming*



## The Imperative

Modern dynamic models rarely have analytical paper-and-pencil solutions. Complexities such as non-linearities, uncertainty, and high dimensionality require robust numerical techniques for analysis.

## The Method

Translate economic theory into executable code using numerical algorithms. This involves techniques for function approximation, numerical integration, and root-finding to solve equilibrium conditions.

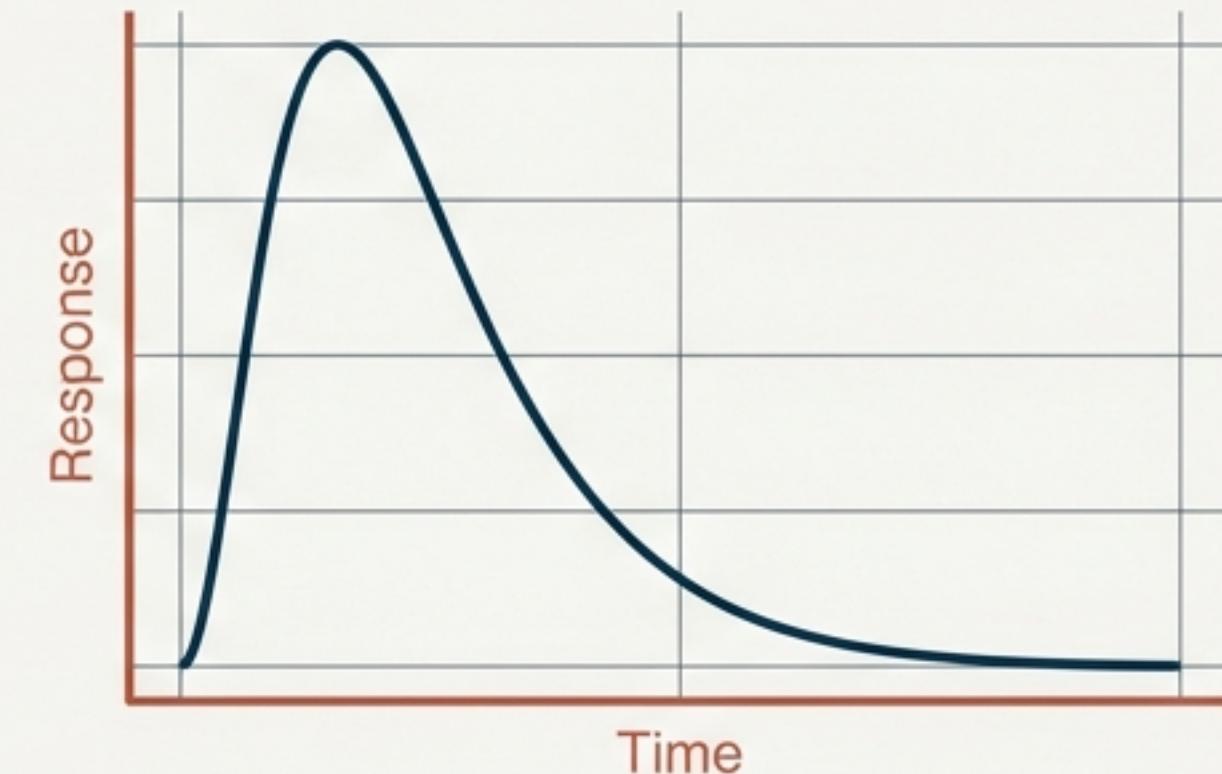
## The Goal

Deconstruct the atomic 'building blocks' required to simulate complex economies. This includes solving bellman equations, policy function iteration, and simulating time series for model validation.

# The Computational Imperative

$$\int_{-\infty}^{+\infty} \max \left( f(x, y), \sum_{i=0}^{\infty} \beta(x, y) V(x') \right) dx + m \delta \sum_{i=0}^{\infty} \beta(x, y) \varepsilon(x) dy$$

Intractable Analytical Solutions



Numerical Approximations

*“Our task as I see it...is to write a FORTRAN program that will accept specific economic policy rules as ‘input’ and will generate as ‘output’ statistics describing the operating characteristics of time series we care about.”*

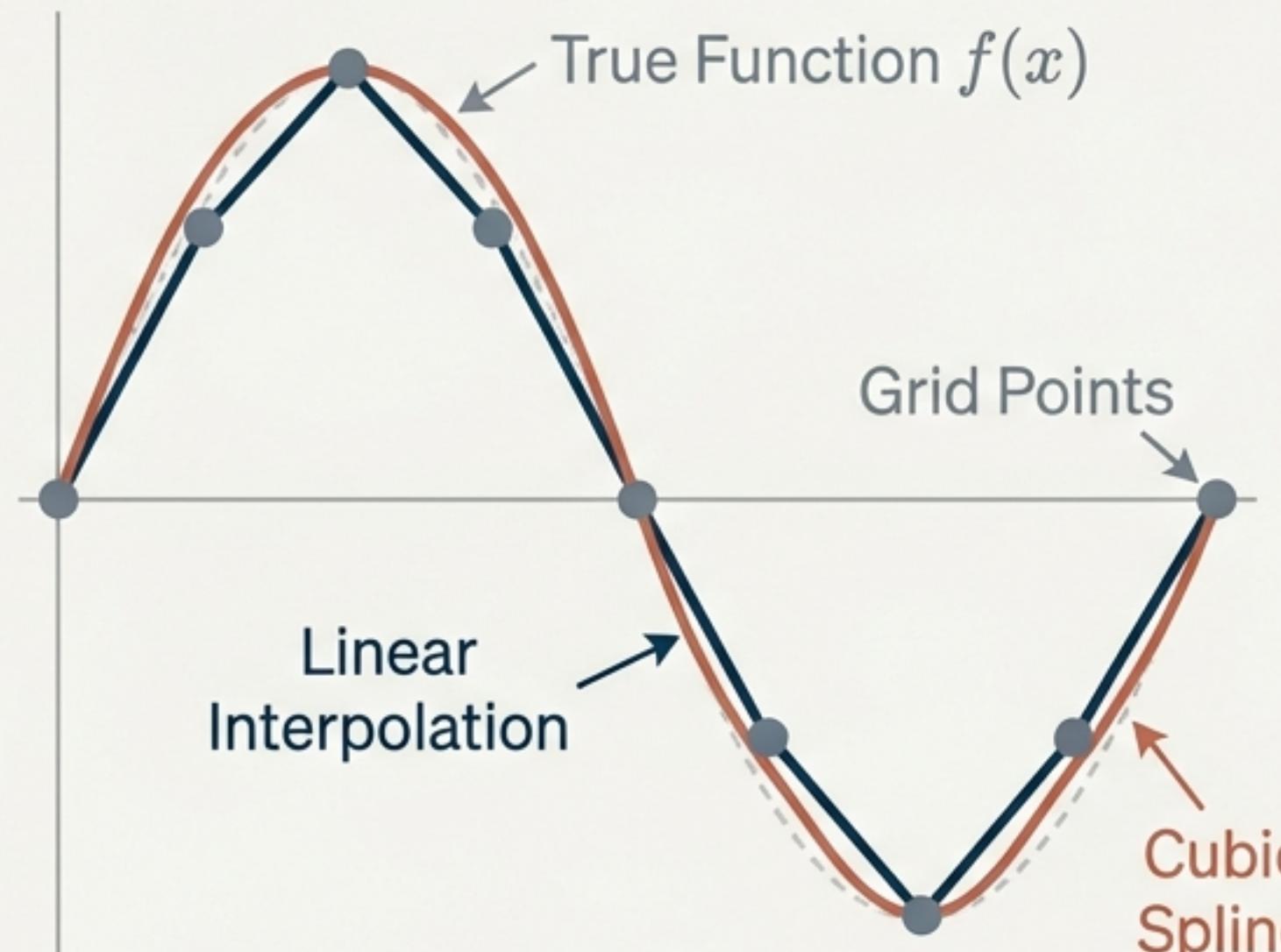
— Robert E. Lucas Jr. (1980)

# Tool 1: Approximating Functions

How do we store infinite points in finite memory?

## Interpolation (Grid Points)

- **Linear:** Connects dots. Robust, local information only. Jagged (non-differentiable).
- **Cubic Spline:** Ensures continuous 1st and 2nd derivatives. Smoother, but errors propagate globally.



$$\hat{f}(x) = \sum a_i \phi_i(x)$$

## Known Functional Forms

- Weighted sum of polynomials (e.g., Chebyshev).
- Uses regression logic to minimize distance between true function and approximation.

# Tool 2: Root Finding (The Robust Approach)

## Bisection: The Tortoise

### The Problem:

Find scalar  $x$  where  $f(x) = 0$ .

### The Algorithm:

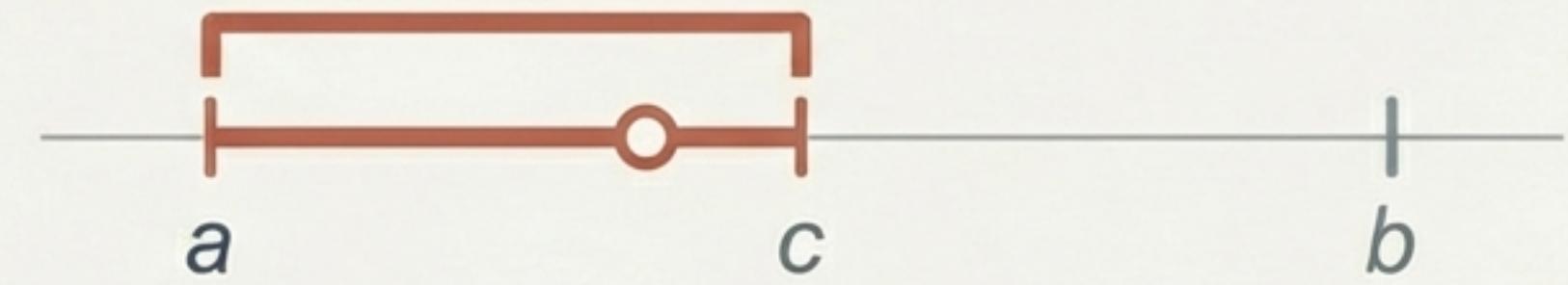
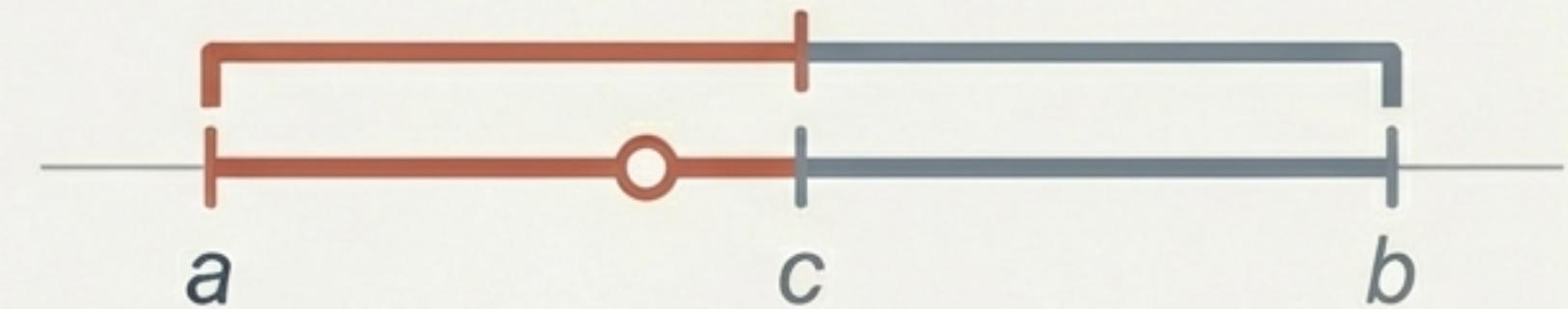
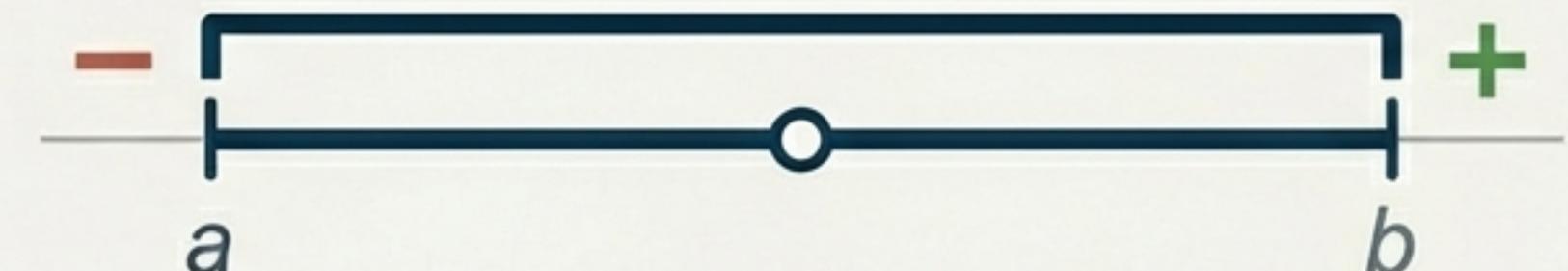
1. Bracket the root between  $a$  and  $b$  (where signs differ).
2. Evaluate midpoint  $c$ .
3. Replace bound to keep root bracketed.
4. Repeat until  $|b - a| < \text{Tolerance}$ .

### Strategic Profile:

**Pros:** Guaranteed convergence.

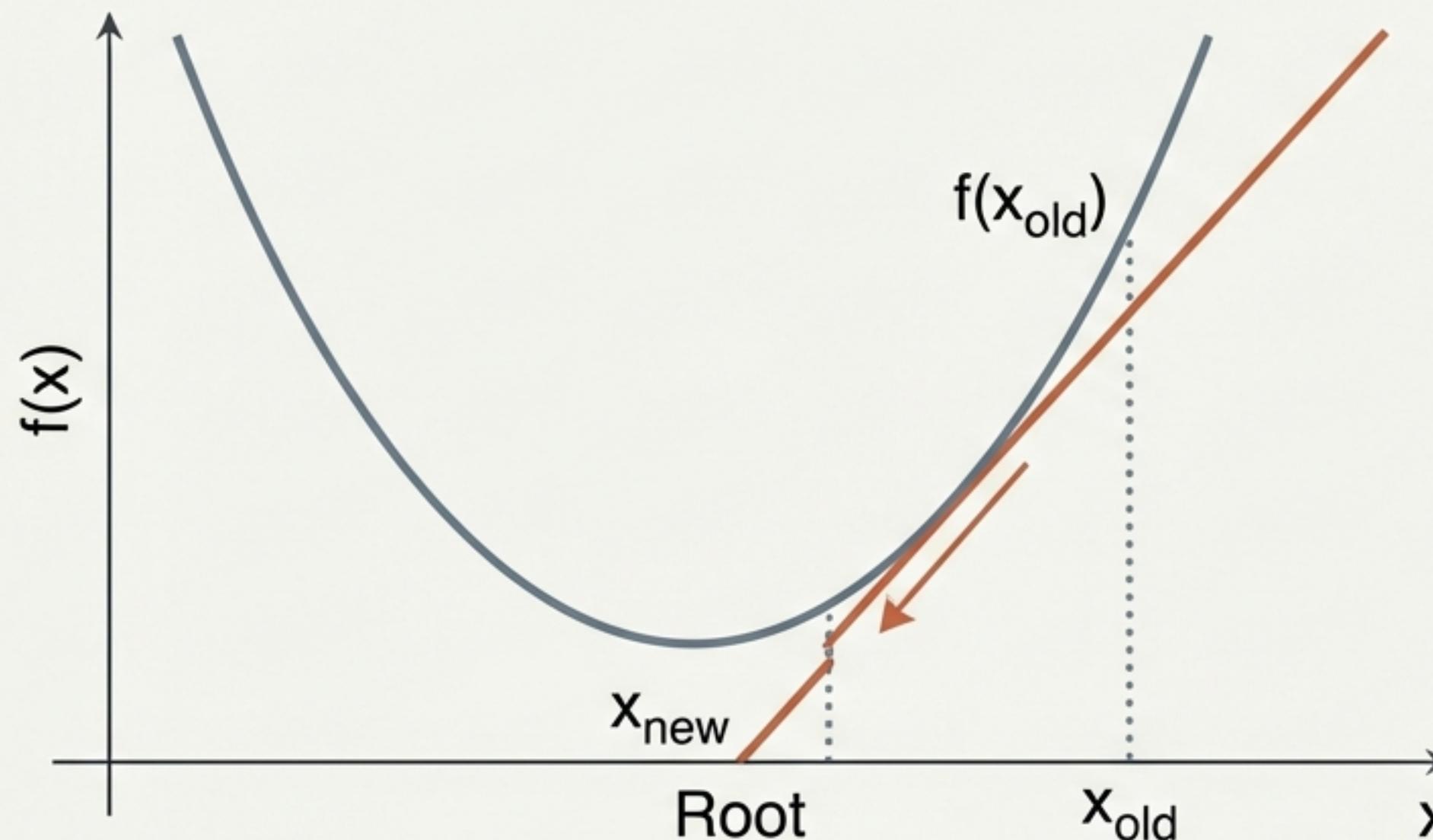
No derivatives needed.

**Cons:** Slow (Linear convergence rate).



# Tool 2: Root Finding (The Fast Approach)

## Newton-Raphson: The Hare



$$x_{new} = x_{old} - \frac{f(x_{old})}{f'(x_{old})}$$

### The Method:

- Approximates function locally using a tangent line (derivative).
- Jumps to where the tangent crosses zero.

### Strategic Profile:

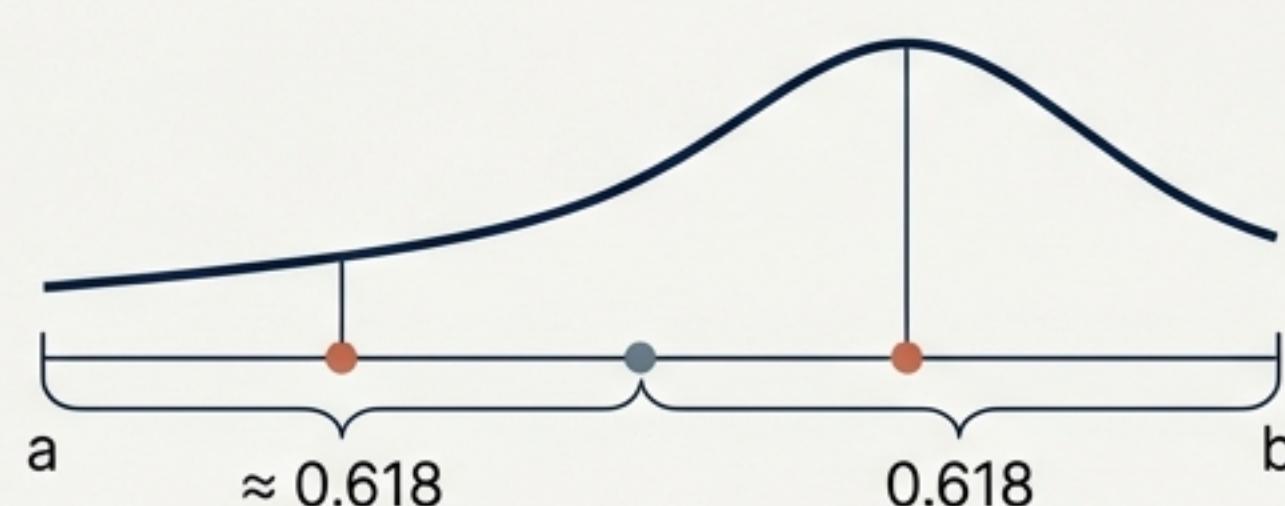
- Pros: Extremely fast (Quadratic convergence).
- Cons: Requires differentiability. Can diverge if function is ill-behaved.

Note: Derivatives often calculated numerically (Finite Difference).

# Tool 3: Optimization ( $\max F(x)$ )

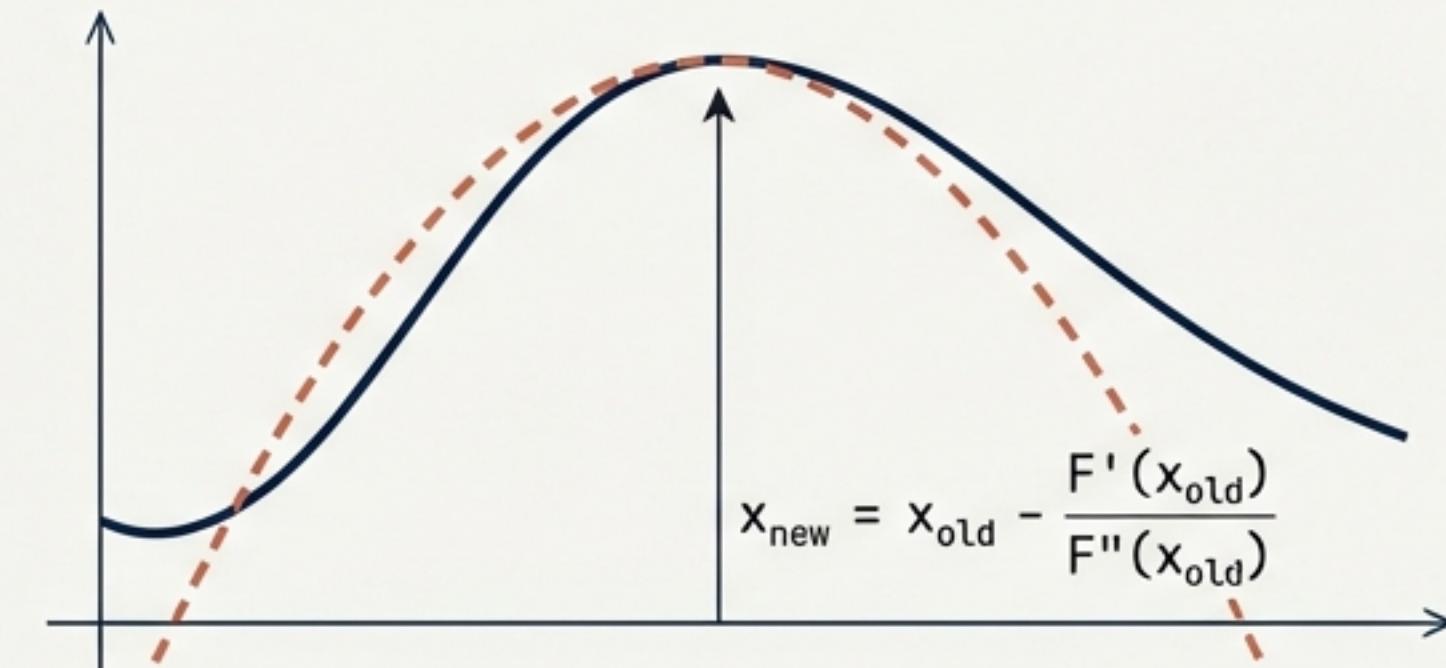
Maximizing  $F(x)$  is computationally equivalent to finding the root of the First Order Condition:  $F'(x) = 0$ .

## Golden-Section Search



Robust, no derivatives. Shrinks interval by golden ratio. Use when function is non-differentiable.

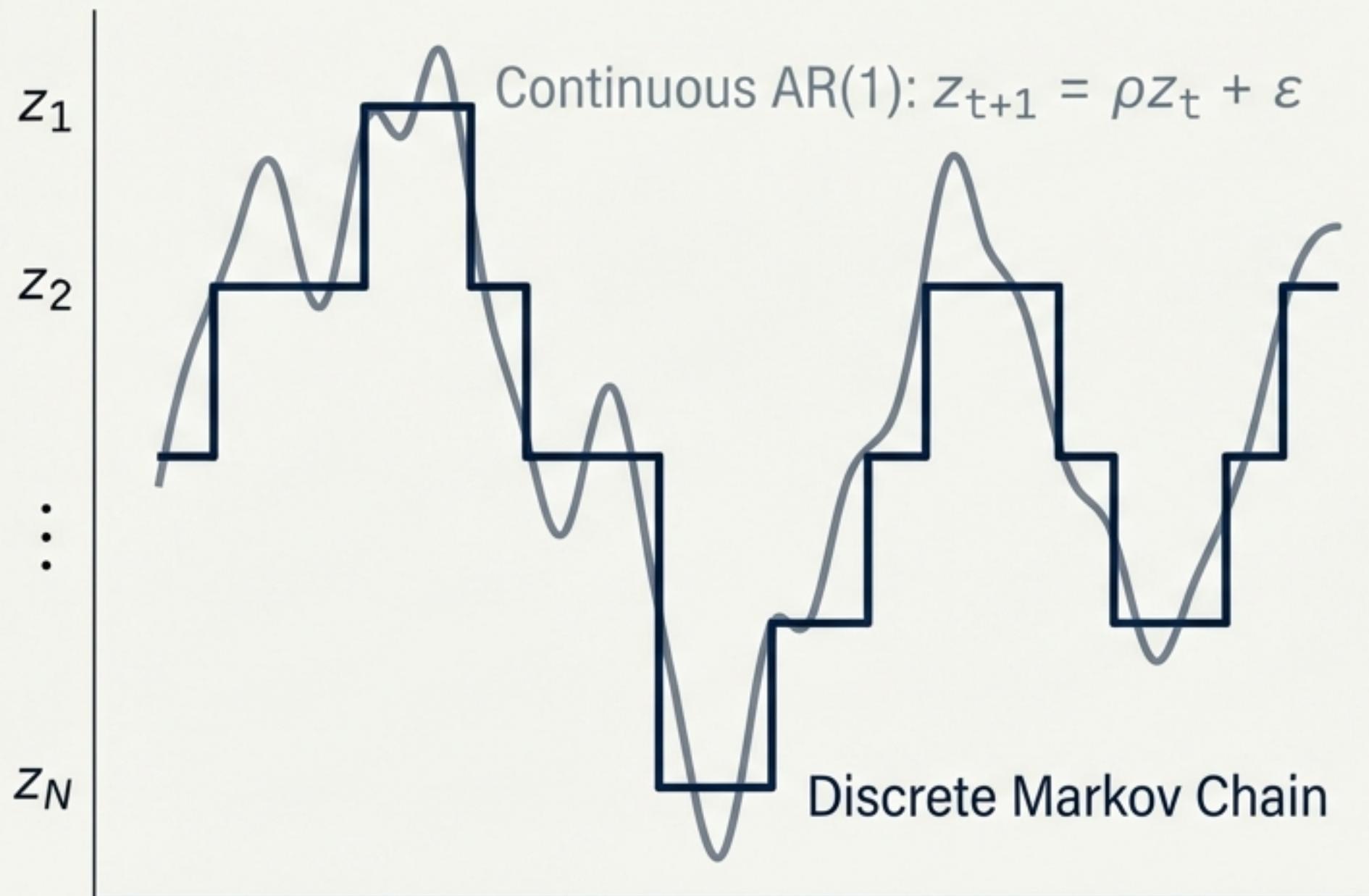
## Newton's Method for Optimization



Uses quadratic approximation. Fast but requires 1st and 2nd derivatives. Use when speed is critical.

# Tool 4: Discretizing Stochastic Processes

Turning continuous shocks into Markov matrices



**The Problem:** Computers solve Dynamic Programming via discrete matrices, but economic shocks are continuous.

## Method A: Tauchen

- Uses integrals over intervals to calculate transition probabilities.
- Best for general purpose approximation.

## Method B: Rouwenhorst

- Matches unconditional mean and variance exactly.
- Best for highly persistent processes ( $\rho$  close to 1), common in macro.

# Solving The Model: Global Methods (VFI)

$$V(k) = \max \{ F(k, k') + \beta V(k') \}$$

Value Function  
(Recursive) 

Optimization Tool 

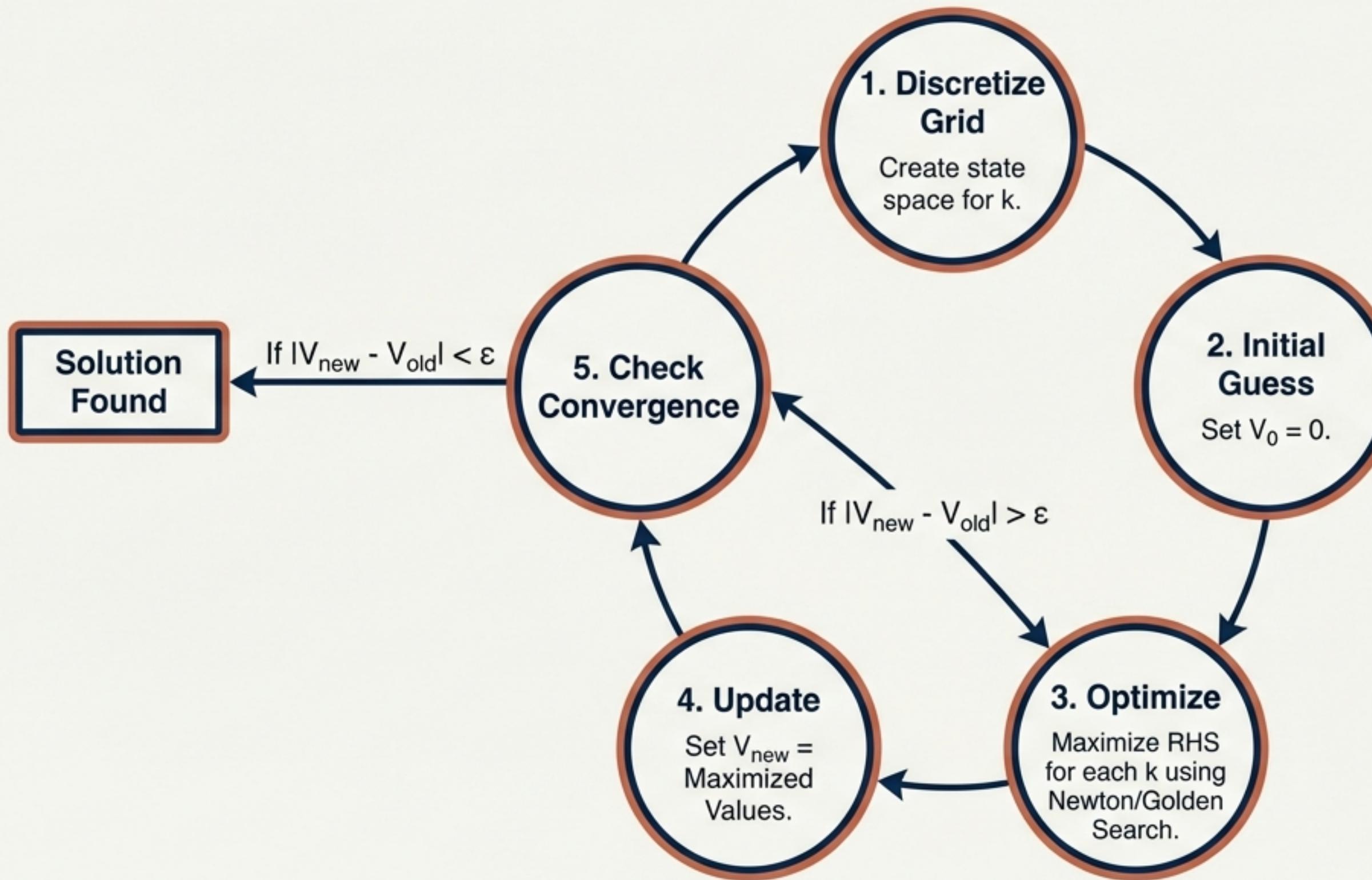
Interpolation Tool 

The Goal: Solve for the Value Function  $V(k)$  and Policy Function.

The Mechanism: Contraction Mapping Theorem

- The operator  $T$  is a contraction mapping.
- Guarantee: Iterating on the Bellman equation converges to the true solution ( $V_i \rightarrow V$ ) from *any* initial guess.

# The Algorithm: Value Function Iteration (VFI)



## Side Note

This process relies on the “Black Box” tools defined earlier:

Approximation for off-grid values and Root Finding/Optimization for the max operator.

# VFI in Practice: The Curse of Dimensionality

## Case Study: The Stochastic Cake Eating Problem

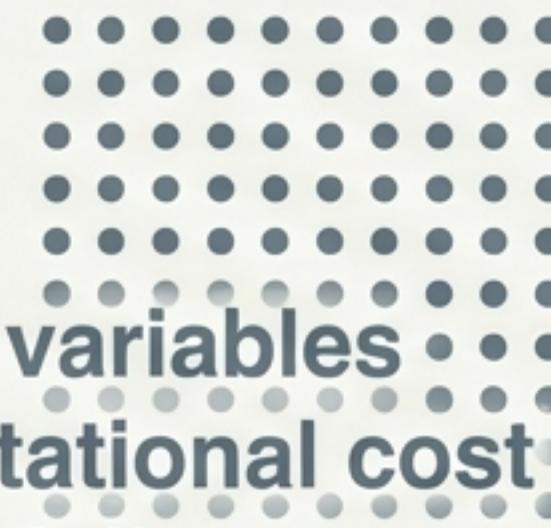
### The Problem:

- Agent maximizes utility of consumption over infinite time.
- Stochastic extension: New cake arrives randomly ( $z$ ).
- Value Function becomes  $V(a, z)$ .

10 Grid Points ( $10^{1}$ )

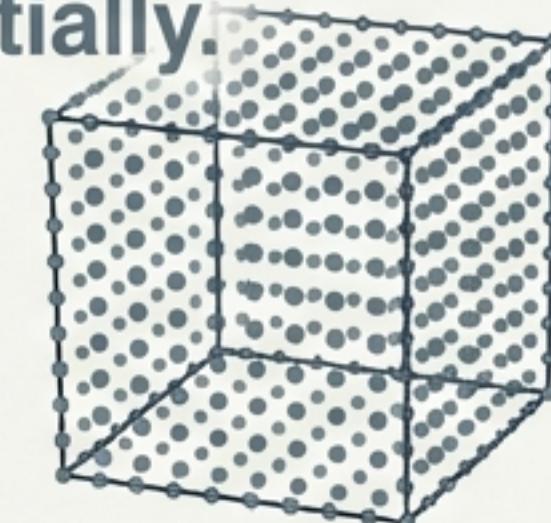


100 Grid Points ( $10^{2}$ )



Adding state variables  
expands computational cost  
exponentially.

1000 Grid Points ( $10^{3}$ )



**Strategic Choice:** VFI is robust (handles risk/non-linearities) but computationally expensive.

# Solving The Model: Local Methods (Linearization)

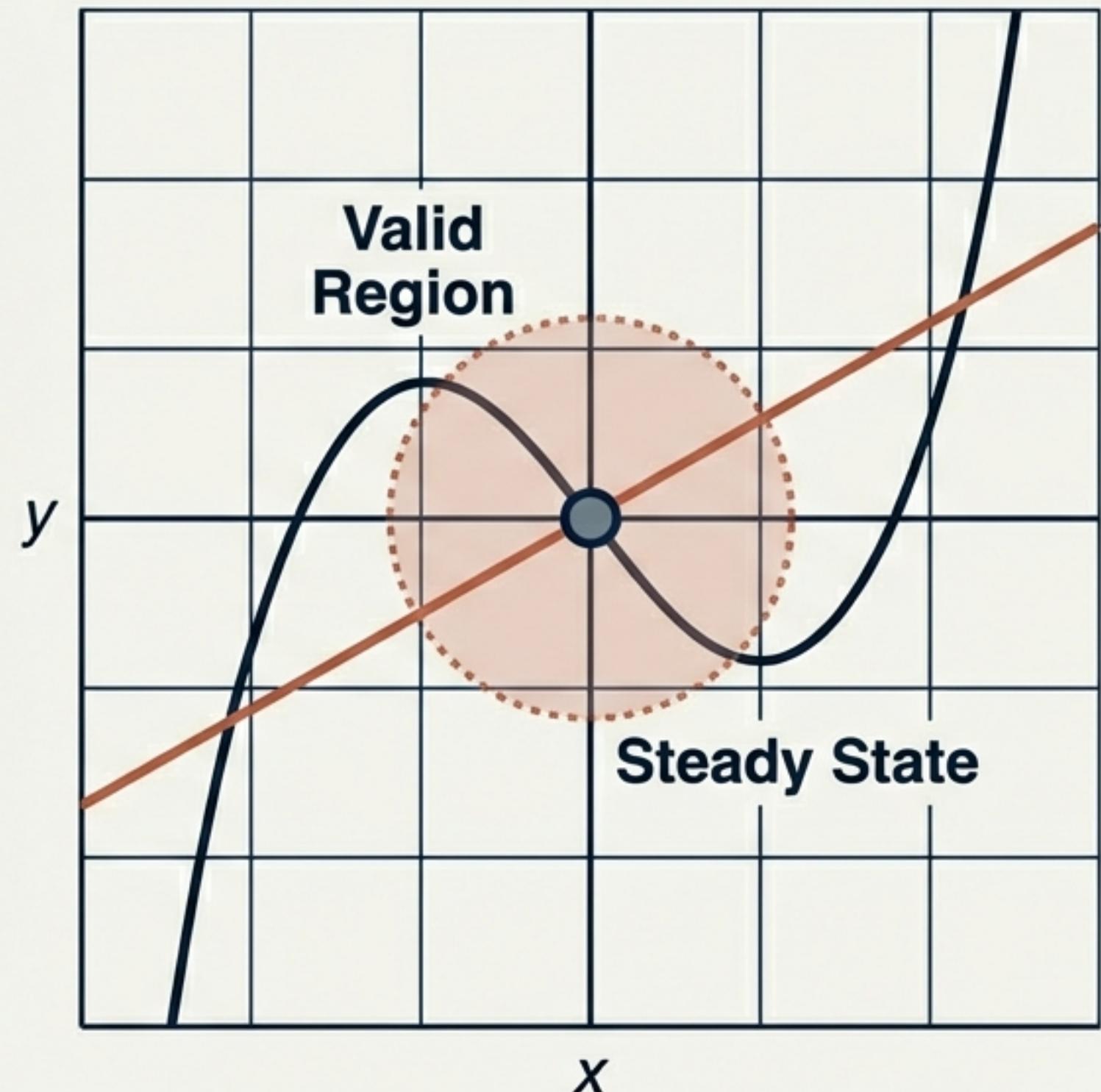
## The Perturbation Approach

### The Concept:

- Instead of solving the whole grid (Global), solve locally around the deterministic Steady State.
- Approximate system using Taylor expansion.

### Key Assumptions:

1. Shocks are small.
2. Economy stays near Steady State.
3. Certainty Equivalence: In linear systems, risk premiums disappear.



# The Linearization Workflow

## 1. Classification

Predetermined Variables ( $x_t$ ):

- Fixed from previous period
- Capital  $K_t$
- Shocks  $z_t$

Jump Variables ( $y_t$ ):

- Determined in current period
- Consumption  $C_t$
- Labor  $H_t$

## 2. Log-Linearization

$$\hat{x}_t = \log(X_t/\bar{X})$$

Convert Euler equations and constraints to log-deviations.

## 3. Matrix System

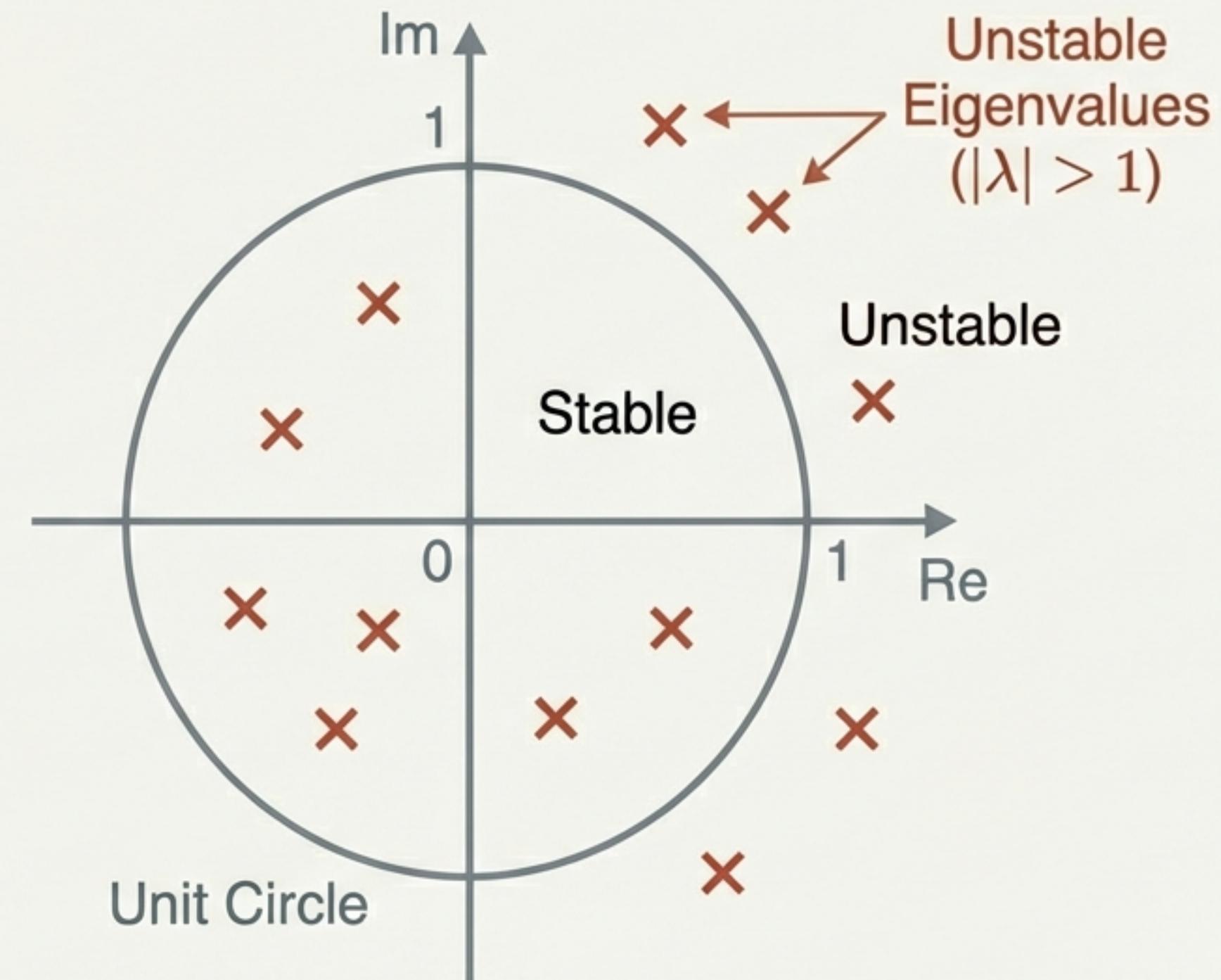
$$A \begin{bmatrix} x_{t+1} \\ E_t[y_{t+1}] \end{bmatrix} = B \begin{bmatrix} x_t \\ y_t \end{bmatrix} + \text{Shocks}$$

# Ensuring Stability: The Blanchard-Kahn Condition

**The Problem:** Linear systems can explode. We need a unique path back to equilibrium.

**The Condition ( $h = m$ ):**

- The number of **unstable eigenvalues** ( $h$ ) must equal the number of **jump variables** ( $m$ ).
- Ensures a unique initial jump for **Consumption** that places the economy on the stable **Saddle Path**.

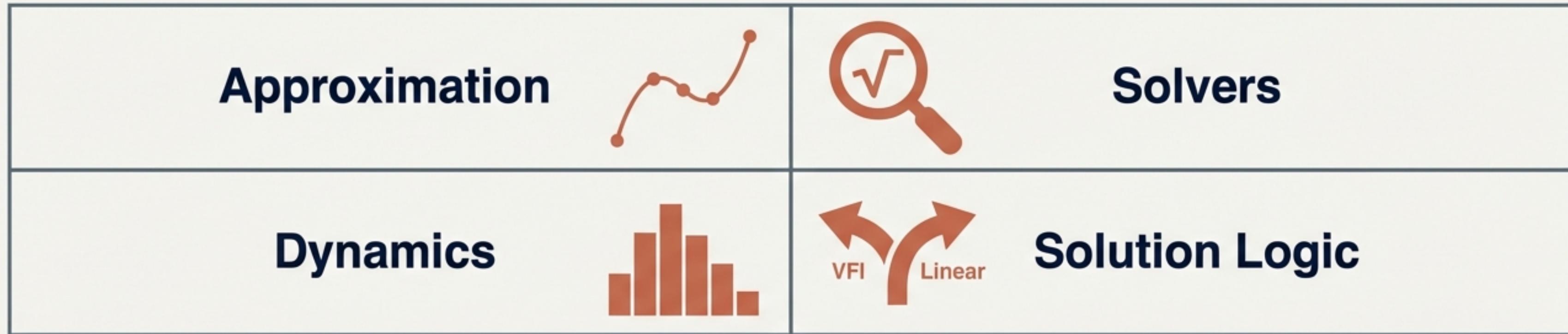


# The Strategic Choice: Global vs. Local

	<b>Global Methods (VFI)</b>	<b>Local Methods (Linearization)</b>
<b>Scope</b>	Valid everywhere in state space.	Valid only near Steady State.
<b>Risk Handling</b>	Accurate. Captures risk aversion & non-linearities.	Certainty Equivalence. Ignores risk premiums.
<b>Cost &amp; Speed</b>	High Cost. Slow. Limited by Curse of Dimensionality.	Low Cost. Instant. Scalable to large systems.

Use VFI for asset pricing/crises. Use Linearization for standard business cycles.

# The Modern Macroeconomist's Toolkit



**Summary:** We have deconstructed the “Black Box”.

- We store functions via Chebyshev/Splines.
- We solve equations via Newton/Bisection.
- We handle shocks via Tauchen/Rouwenhorst.
- We choose solution strategies based on the trade-off between robustness and speed.

**Implementation:** While software (Dynare) automates this, understanding the blocks is essential for diagnosis.

**References:** Chapter 10, Lucas (1980), Judd (1998), Numerical Recipes, QuantEcon.