

Advanced Macroeconomics I

「上級所得理論 I」 for graduate program
「Advanced Income Theory I」 for G30 and NUPACE

Lecture 1

Introduction to advanced macroeconomics

4/13, 2023

About Me

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- Macroeconomist
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Goals

- Today, we have two goals.
- In the Course Description part, we understand the key differences between the undergraduate macroeconomics and the graduate-level macroeconomics.
- In the Linear Scalar Difference Equations part, we study a new mathematical tool for macroeconomic analysis.

Course Description

Where We Stand

- Since the 1980s, macroeconomics has made a remarkable progress especially in terms of its methodology.
- **The gap between undergraduate macro and graduate macro has become enormous.**
- As a result, it is **entirely hopeless** to understand any of the research papers written in the last 40 years without a strong methodological background.
- What do latest academic papers look like?

What do academic papers look like?

- Take a brief look at my papers published in 2019 and in 2023:
 - <https://tact.ac.thers.ac.jp/x/x15Oeu>
 - <https://tact.ac.thers.ac.jp/x/DDmji7>
- There are two major departures from undergrad macro:
 1. Dynamics: Variables change their values over time.
 2. Optimization: All decisions are derived, not assumed.

Undergraduate Macro Model

- The model is static (= concept of time is absent).

- National income identity (closed economy):

$$Y = C + I + G$$

- Consumption function:

$$C = aY + b$$

- The equilibrium (= solution) of the model is obtained by solving a system of equation.

Difference Equations

- National income identity: $Y_t = C_t + I + G$
- Consider: $C_t = aY_{t-1} + b$
- The equilibrium GDP satisfies
$$Y_t = aY_{t-1} + b + I + G$$
- There are Y_t and Y_{t-1} in one equation.
- This is an example of **difference equation**.

Undergraduate Macro Model

- National income identity: $Y = C + I + G$
- Consumption function: $C = aY + b$
- Consumption function model is ad-hoc (= the household's action is arbitrarily chosen by us, not by individuals in the model).
- What does **microeconomics** tell us?
 - Households make decisions to maximize utility.

Dynamic Optimization

- Instead of we choosing consumption, we introduce an individual who chooses his/her consumption.
- A typical dynamic optimization looks like:

$$\begin{aligned} & \max_{\{c_t\}_{t=0}^{\infty}, \{k_{t+1}\}_{t=0}^{\infty}} \sum_{t=0}^{\infty} \beta^t u(c_t) \\ & \text{subject to} \\ & k_{t+1} = f(k_t) + (1 + \delta)k_t - c_t \\ & \text{for } t = 0 \dots \infty. \end{aligned}$$

Goal for this Semester

- We focus on particularly important methodological skills that are often used in modern macroeconomic research:
 - 1. difference equations** for describing variables that evolve over time, and
 - 2. dynamic optimization** for describing the optimal allocation of resources over time
- After this semester, you will be able to read professional articles and be ready to catch up with the research frontier. This is our goal.

Long Run Goals

- Requirement for MA thesis
 - Fully understand and replicate some results published in a leading professional journal.
 - (Optional) Modify the analysis (by a new data set or by a new parameter set) to add some originality.
- Requirement for Ph.D. dissertation
 - Introduce innovative assumptions or methods to the existing analysis to find something **new**.
 - Complete 3 papers, one of which must be published.

Target

- This course is designed for first-year graduate students.
 - But, all highly-motivated students are welcome.
 - However, please fully understand that this course is demanding (= many time-consuming assignments).
- To help you acquire skills, this course is intentionally made painful for you.

Prerequisites

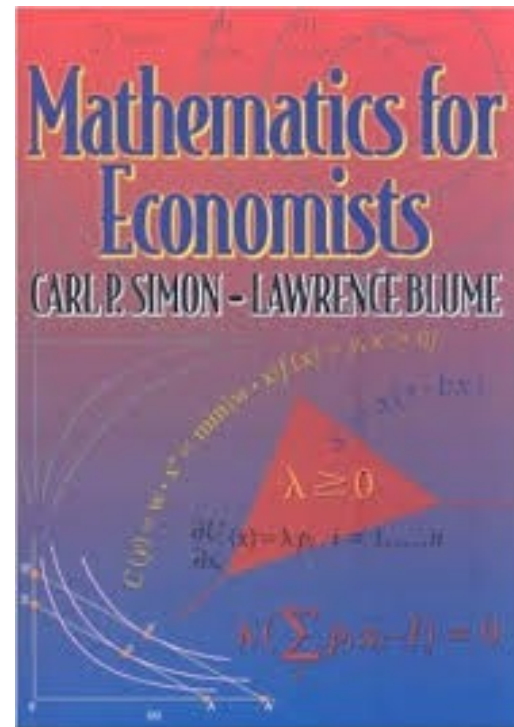
- I will assume that you are familiar with:
 - Constrained maximization problems using the Lagrangian and Karush-Kuhn-Tucker (formally known as Kuhn-Tucker) methods.
 - Partial differentiation and total differentiation
 - Excel and IT related tools.
- If you have no idea about these topics, study them immediately, or consider dropping the course.

Textbooks

- There are many good textbooks.
- But, it will take too much time to go over one book.
 - Many textbooks are about 1,000 pages long.
 - It is much more efficient to give lectures without textbooks.
 - I will focus on solving problems.
 - Learning through examples.
- I encourage you to read some famous textbooks on your own (during or after this semester).

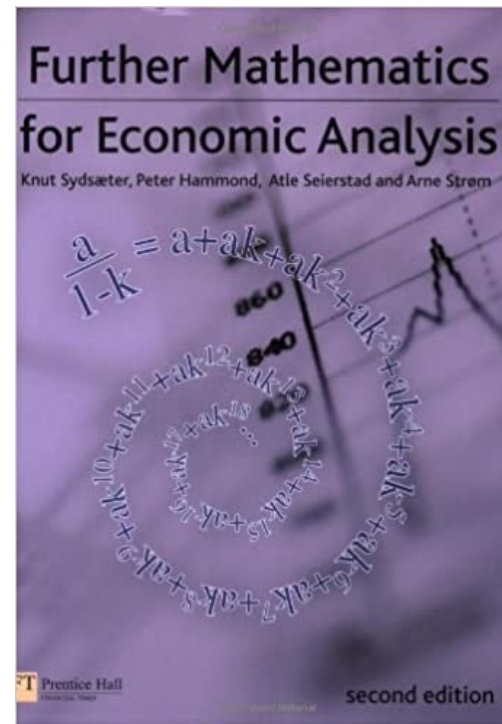
Mathematics for Economists

- Carl Simon & Lawrence Blume, *Mathematics for Economists*, Norton, 1994.
- For advanced undergraduate students and entry-level graduate students.
- Many examples and exercises.
- See if you can solve all the problems in the book.
- My favorite and I highly recommend this book.



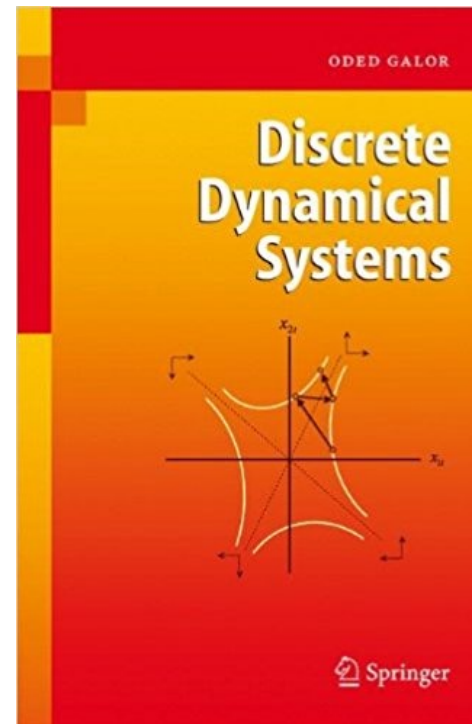
Mathematics for Economists

- Sydsaeter, Hammond, Seierstad, & Strom, *Further Mathematics for Economic Analysis*, 2nd edition, Prentice Hall, 2008.
- Excellent for first-year graduate students.



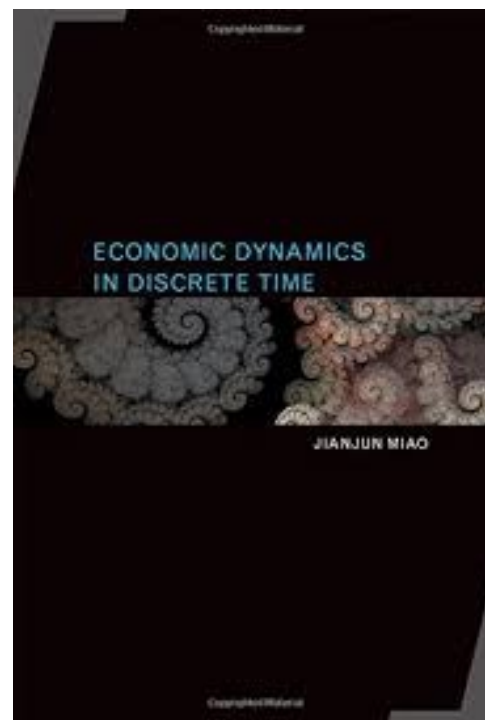
Difference Equations for Macroeconomics

- Oded Galor, *Discrete Dynamical Systems*, Springer, 2010.
- Closely related to the first part of my lecture plan.
- About 150 pages.



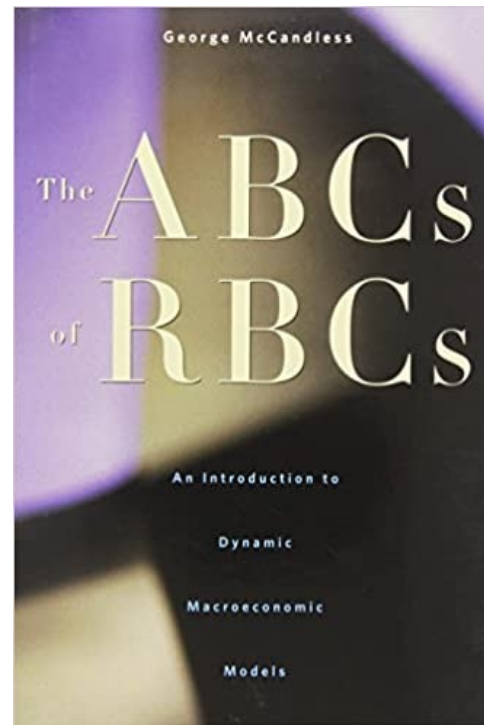
Macroeconomics

- Jianjun Miao, *Economic Dynamics in Discrete Time*, MIT Press, 2014.
- Somewhat related to my lecture plan.
- About 700 pages.



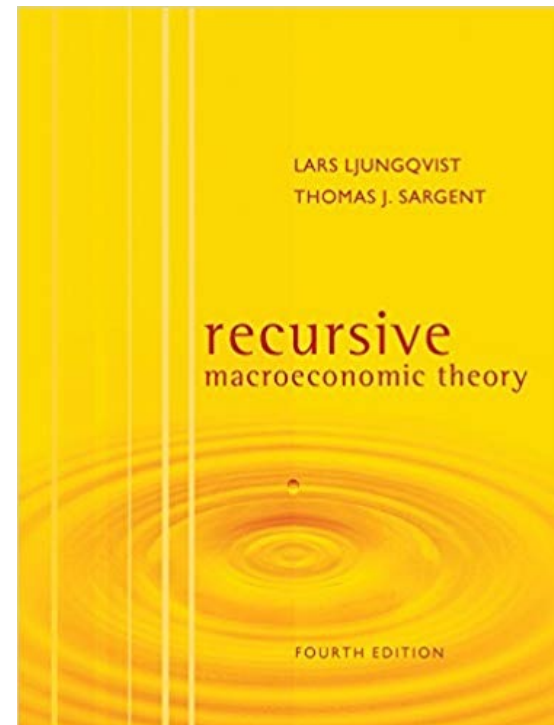
Solving Macro Models

- McCandless, *The ABCs of RBCs: An Introduction to Dynamic Macroeconomic Models*, Harvard University Press, 2008.
- A practical textbook on how to formulate and solve typical macro models.
- Somewhat related to my lecture plan.



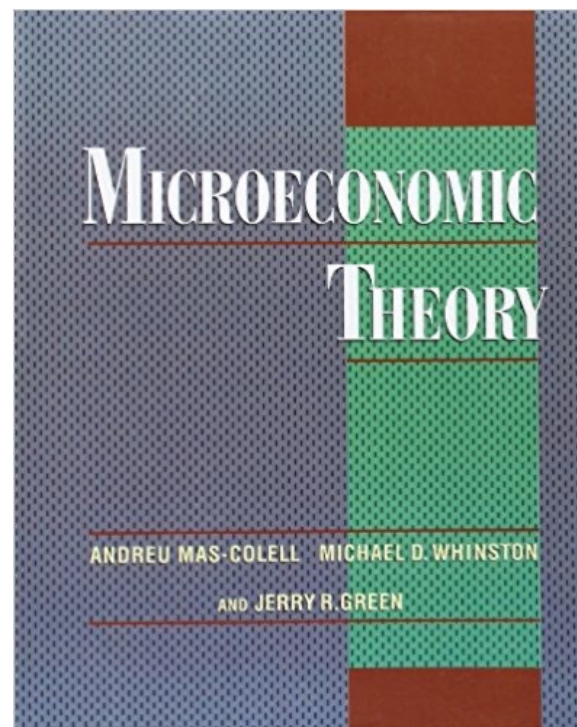
Macroeconomics

- Lars Ljungqvist & Thomas Sargent, *Recursive Macroeconomic Theory*, 4th edition, MIT Press, 2018.
- Extremely famous.
- Starts with dynamic programming.
 - You should wait until the end of this semester.
- Over 1,400 pages!!



Microeconomics

- Andreu Mas-Colell, Michael D. Whinston, & Jerry R. Green, Microeconomic Theory, Oxford University Press, 1995.
- This is the one that all first-year grad students in all grad schools in all countries read. Don't get behind.
- Micro is the foundation of all fields of economics.
- Try to solve all problems in this textbook.



Computer Languages

- 40 years ago, it was OK to write a paper without any computer simulation. Not any more. The old age has gone.
- It is increasingly more important to develop a complicated (realistic) model and solve it using computer.
- It is essential (especially for young students like you) to learn (many) computer languages.

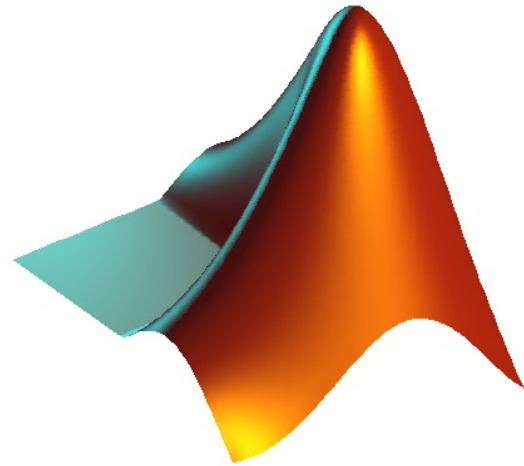
Excel

- Very intuitive and useful.
- We start with Excel to compute dynamic macro models.



Matlab

- Extremely popular among engineering people.
- Especially useful for studying matrix systems.
- The dominant package for simulating macro models.
- Somewhat expensive, but student version is available.



Octave

- **Free** and is a **clone** of Matlab.
 - Octave can execute any Matlab file.
 - Meaning, you do not need to buy Matlab.
 - Slower than Matlab, but don't complain.



Dynare

- Dynare is a package that assists you simulate your dynamic macro models.
 - It will write Matlab (Octave) codes for you.
- **Dynare runs on Matlab or Octave.**

Mathematica

- Very powerful math package.
- Especially useful for solving nonlinear equations.
- Very popular but very expensive.
- You can execute short mathematica codes at <https://www.wolframalpha.com/>



Maxima

- A close cousin of Mathematica and this one is **free**.
- This is **not a clone** of Mathematica, but the languages are similar.



Python

- The language of the next generation.
- Available for free.
- For more about this language, please visit <https://quantecon.org/>



About Assignments

Basic Idea

- My goal is to help you acquire the key analytical methods for macroeconomic research so that you will be able to read professional articles after this semester.
- To acquire new skills, we must focus on solving (many) (time-consuming) questions.

How to Submit Your Work

- All homework assignments must be submitted online at TACT.
 - I will accept PDF files only.
- For many assignments, your answers must be hand-written.
 - Many scanner applications for iPhone/android are available for free.
 - You may also consider using a pen tablet.
 - I strongly suggest that you test each method to see if you can create a PDF of your hand-written letter.

If you have absolutely no idea...

- Try one of the following:
 1. My favorite is Microsoft OneNote, which is free and available under many environments.
 - I find it particularly useful on iPad with Apple Pencil.
 2. My favorite scanner app is Scannable (by Evernote), but there are many apps.
 - Do not use your photo apps as they create a gigantic PDF file. File size matters.
 3. Bring your hand-written notes to a convenience store to manually create a PDF file using a multi-purpose printer there. Costly, not recommended.

Assignment (not to be graded)

- Create a PDF file with your hand-written note.
- Try to find a way to minimize your file size. A PDF from your iPhone camera has a large file size.

Linear Scalar Difference Equations

Varieties of Difference Equations

- Given a function f , suppose x_t evolves according to
$$x_{t+1} = f(x_t)$$
- This is called a **first-order** difference equation.
 - Once you know the current value of x , you immediately know the value of x in the next period.
- This equation is said to be **autonomous**.
 - **Linear** autonomous equation: $x_{t+1} = ax_t + b$ (a, b are parameters)
 - An example of **non-autonomous** equation:
$$x_{t+1} = 2x_t + t$$
 (it depends explicitly on t)

Varieties of Difference Equations

- Consider $x_{t+1} = ax_t^2 + b$. This is an example of (first-order) **nonlinear** difference equation.
- Consider $x_{t+1} = ax_t + bx_{t-1}$. This is an example of **second-order** (linear) difference equation.
- Consider $x_{t+1} = ax_t$. This is an example of **homogeneous** difference equation.
 - This is the simplest difference equation.

Solving $x_{t+1} = ax_t$

- Let us start with a homogeneous equation, which is the simplest case.
- Example) $a = 0.5$.
- Thus, $x_{t+1} = 0.5x_t$.
- Suppose that you know the initial value, $x_0 = 1$.
- Then, what is x_1 ?

Solving $x_{t+1} = ax_t$

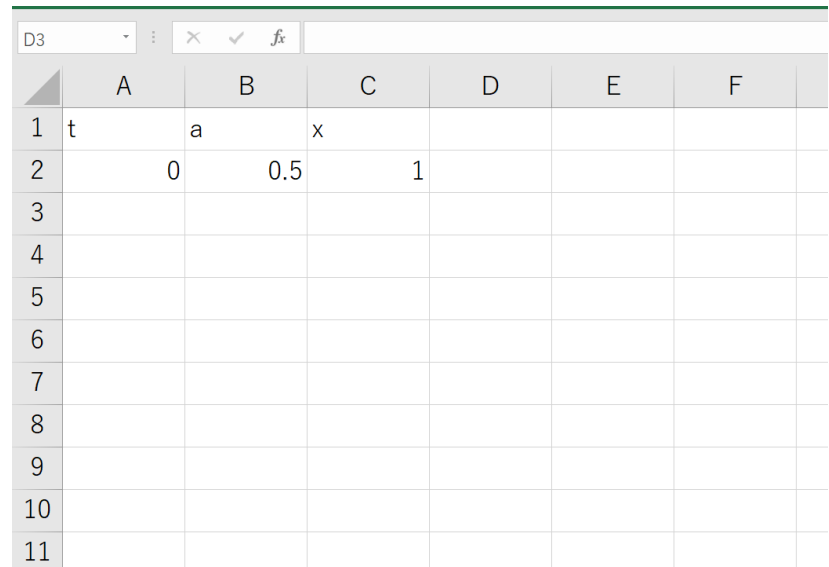
- Substitute $t = 0$ into $x_{t+1} = 0.5x_t$ to obtain
$$x_1 = 0.5 \times x_0 = 0.5 \times 1 = 0.5.$$
- What is x_2 then?
- What is x_t ?

Solving $x_{t+1} = ax_t$

- It is now clear that given the initial condition (i.e., the value of x_0), the difference equation $x_{t+1} = 0.5x_t$ generates a sequence of numbers.
- Because the length this sequence is infinite, it is hopeless to write down the sequence by paper and pencil.
- Let us use Excel (or the like) to continue.

Solving $x_{t+1} = ax_t$

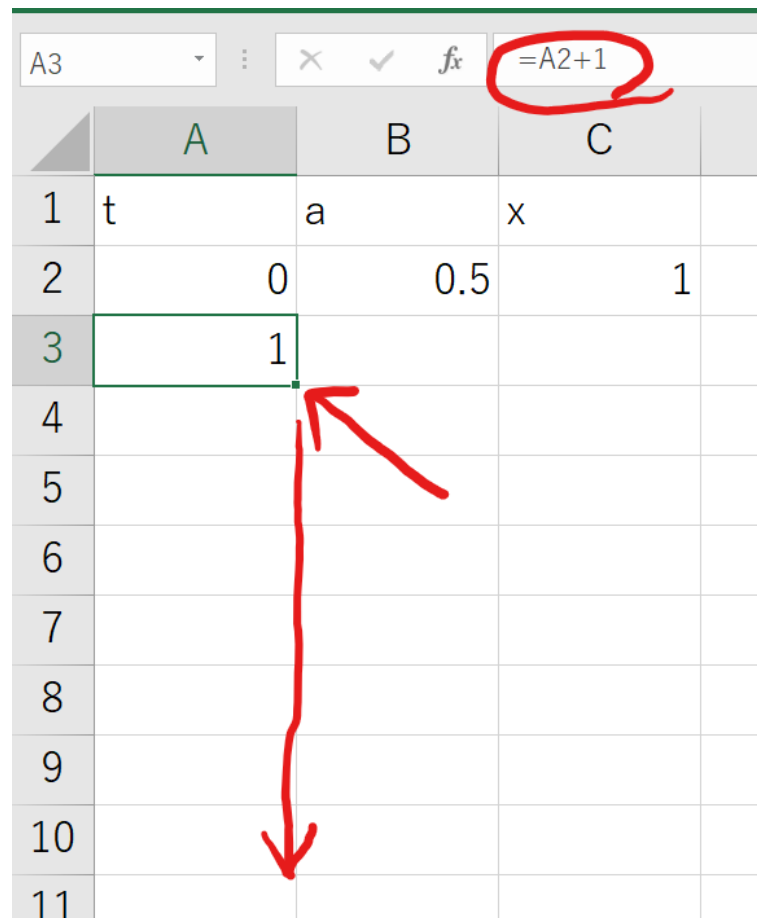
- Column A = period (t)
- Column B = value of a
- Column C = value of x_t
- Cell A2 = 0
- Cell B2 = 0.5 = a
- Cell C2 = 1 = x_0
- You do not have to fill in all the cells manually.



	A	B	C	D	E	F
1	t	a	x			
2	0	0.5	1			
3						
4						
5						
6						
7						
8						
9						
10						
11						

Solving $x_{t+1} = ax_t$

- Cell A3 is “=A2+1”
- Drag the corner of A3 down to A10, say, to see what happens.



	A	B	C
1	t	a	x
2	0	0.5	1
3	1		
4			
5			
6			
7			
8			
9			
10			
11			

Solving $x_{t+1} = ax_t$

- Cells from A3 to A10 are filled automatically.
- Do the same for column B?
- No, there a better way.
- At B2 cell, double-click on the corner button.

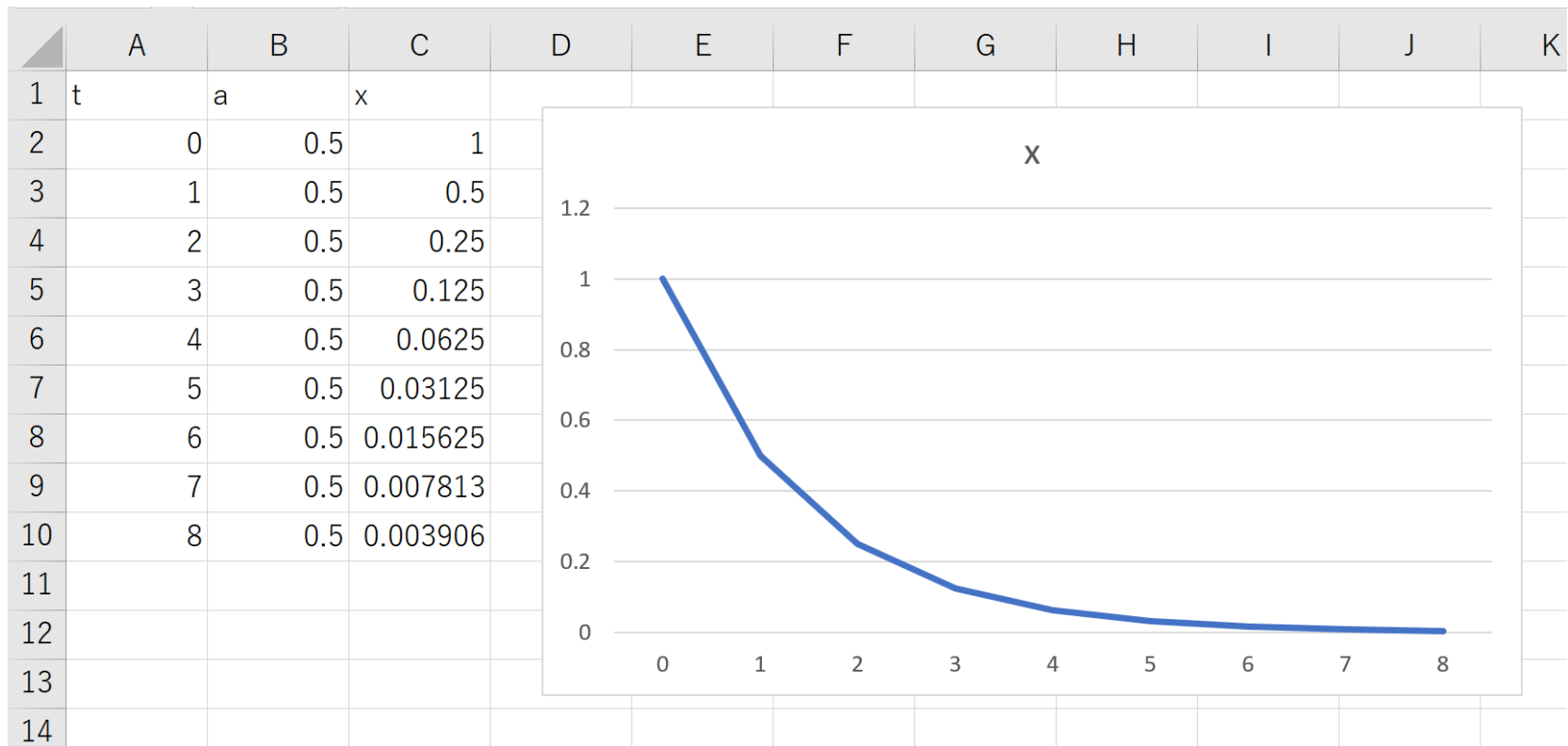
	A	B	C
1	t	a	x
2	0	0.5	1
3	1		
4	2		
5	3		
6	4		
7	5		
8	6		
9	7		
10	8		
11			
12			

Solving $x_{t+1} = ax_t$

- Cell C3 is “=B2*C2”
- Finally, double click on the button at the corner of C3 cell.

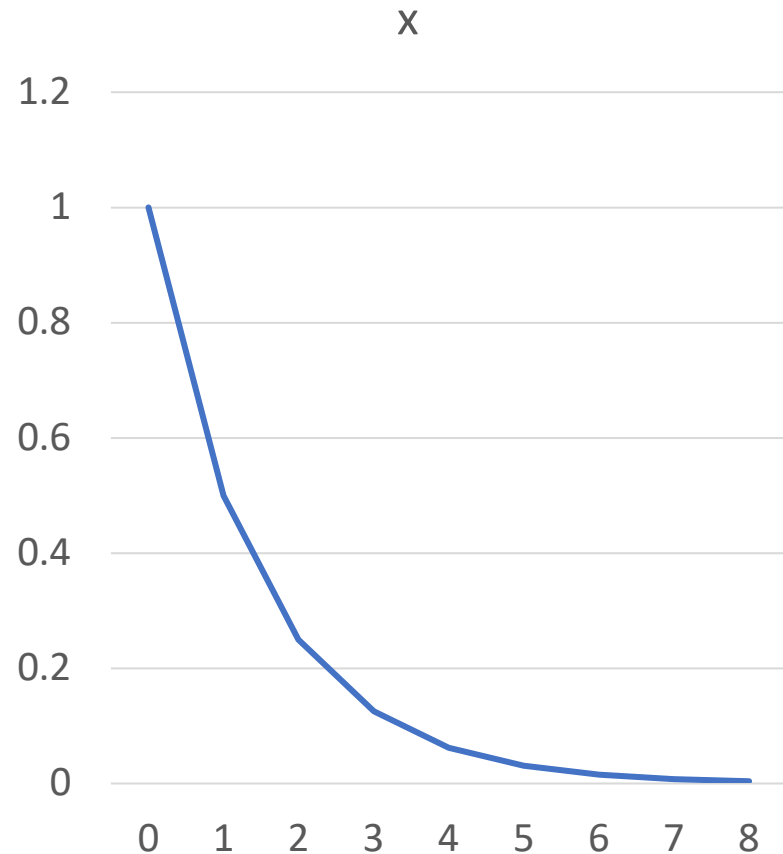
SUM				✕	✓	f_x	=B2*C2
	A	B	C				
1	t	a	x				
2	0	0.5	1				
3	1	0.5	=B2*C2				
4	2	0.5					
5	3	0.5					
6	4	0.5					
7	5	0.5					
8	6	0.5					
9	7	0.5					
10	8	0.5					
11							

Solving $x_{t+1} = ax_t$



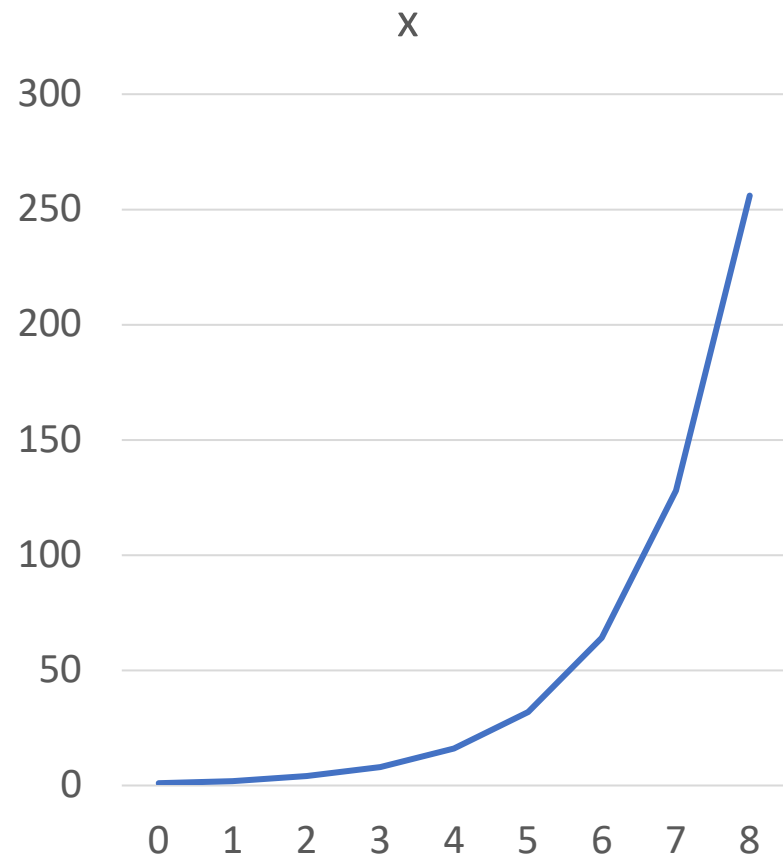
Solving $x_{t+1} = ax_t$

- When $0 < a < 1$, the trajectory exhibits **monotone convergence** to $x = 0$.



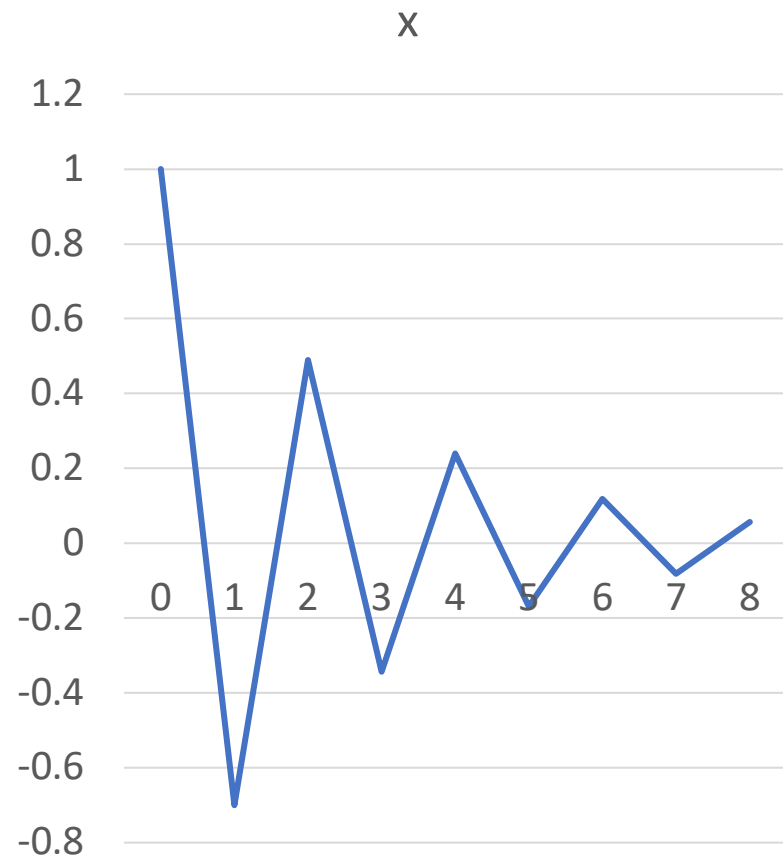
Solving $x_{t+1} = ax_t$

- When $a > 0$, the trajectory exhibits **monotone divergence**.
- The path is said to be **explosive**.



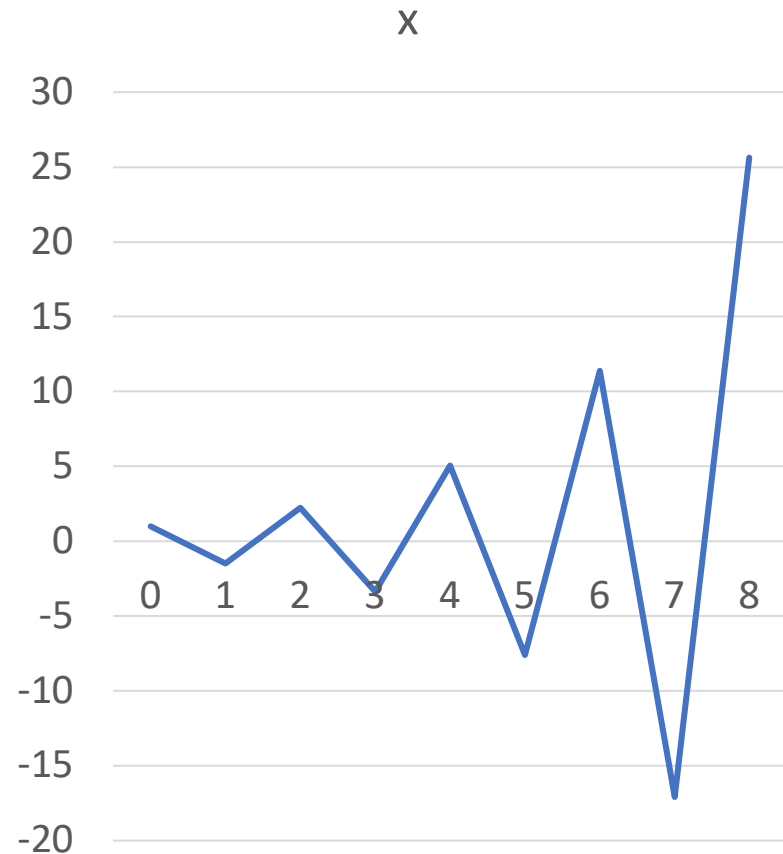
Solving $x_{t+1} = ax_t$

- When $-1 < a < 0$, the trajectory exhibits **damped oscillations**.
- This is convergent.



Solving $x_{t+1} = ax_t$

- When $a < -1$, the trajectory exhibits **explosive oscillations**.



Observations

- Consider $x_{t+1} = f(x_t)$.
- Given a value of x_0 (called the initial condition), we obtain an infinite sequence of numbers satisfying the difference equation.
- The sequence $\{x_t\}_{t=1}^{\infty}$ is the **solution** to the difference equation.
- Two types of solution:
 - **Numerical solution** = computer-generated numbers.
 - **Analytical solution** = a function that reduces the equation to an identity.

Solving $x_{t+1} = ax_t$ Analytically

- Consider $x_{t+1} = ax_t$ with initial condition x_0 .
 - This means that a and x_0 are parameters.
- $x_1 = ax_0$. This a number.
- $x_2 = ax_1 = a(ax_0) = a^2x_0$. Again, a number.
- $x_3 = ax_2 = a(a^2x_0) = a^3x_0$.
- We can now guess that
$$x_t = a^t x_0$$
- This is a function that reduces the equation to an identity, so this must be the solution!

Solving $x_{t+1} = ax_t$ Analytically

- How can we verify that $x_t = a^t x_0$ is the solution?
- To see if $x_t = a^t x_0$ solves $x_{t+1} = ax_t$, let us check whether $x_{t+1} - ax_t$ is zero or not:

$$x_{t+1} - ax_t = a^{t+1}x_0 - a \times a^t x_0 = 0.$$

- Thus, **the solution to $x_{t+1} = ax_t$ given x_0 is**

$$x_t = a^t x_0$$

- It is extremely important that you remember (and never forget) this result, as we will come back to this many times.

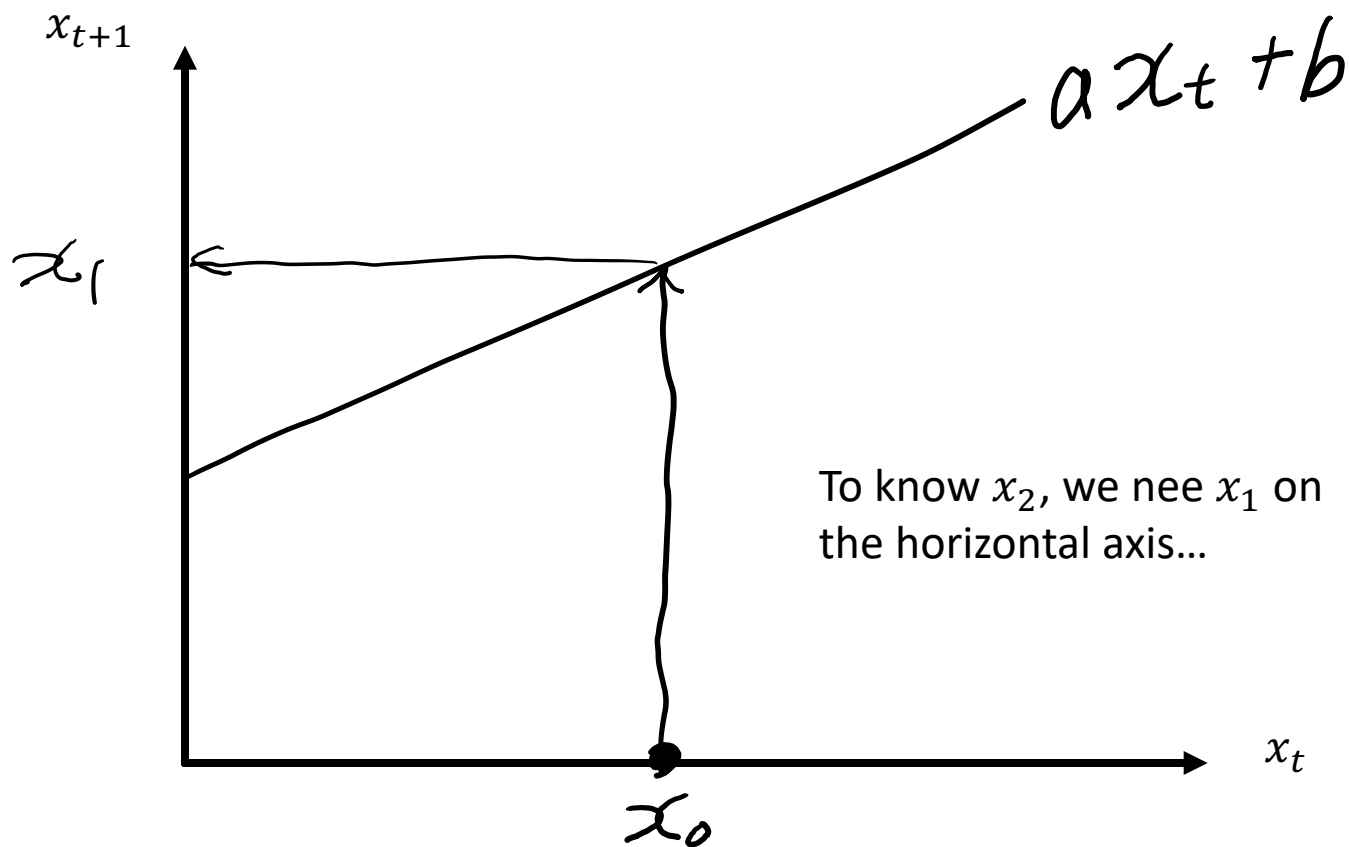
Solving $x_{t+1} = ax_t + b$

- Now consider the general first-order linear scalar difference equation:

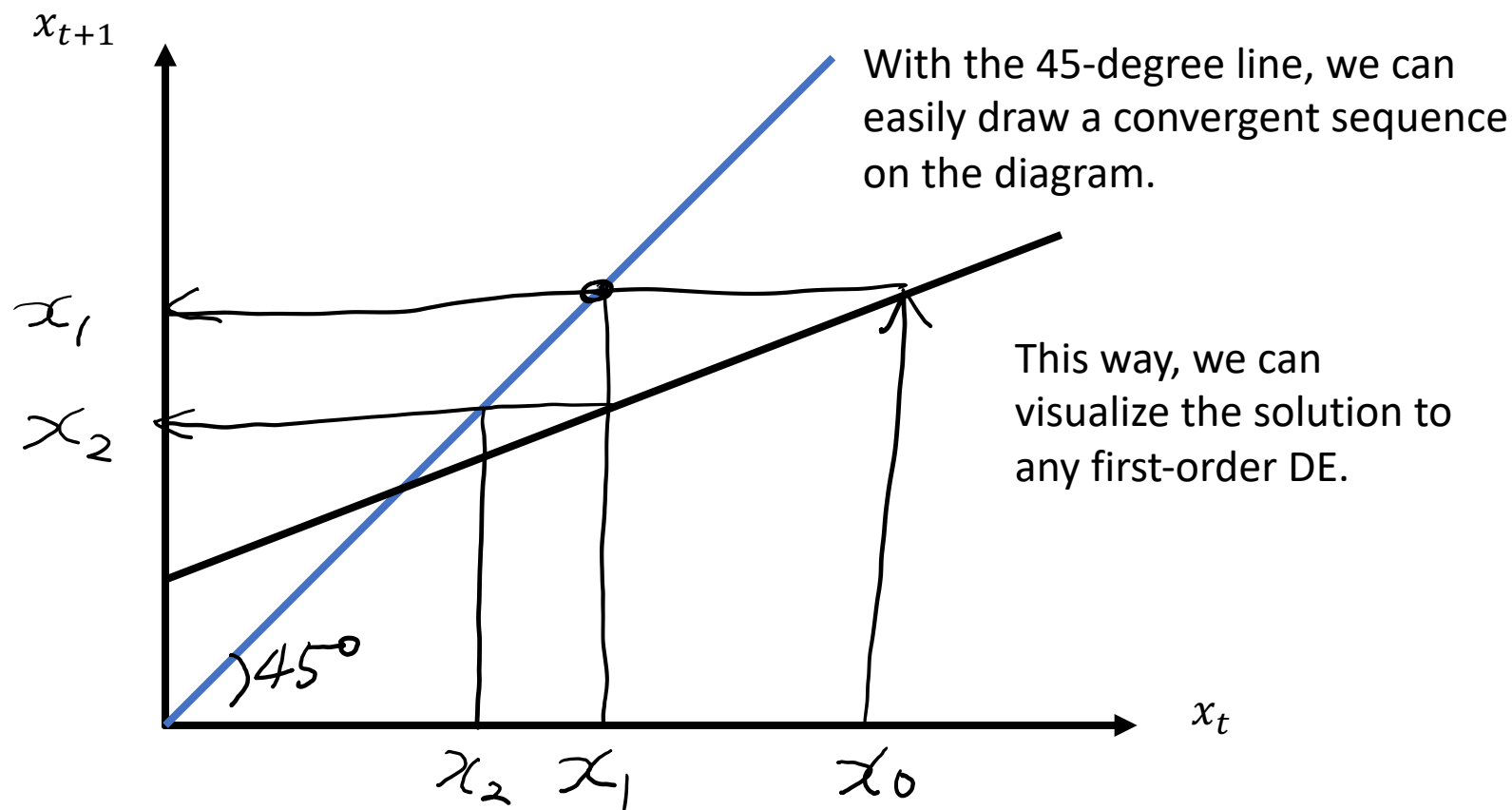
$$x_{t+1} = ax_t + b$$

- Suppose $0 < a < 1$ and $b > 0$.
- Let us now draw a diagram.
- x_{t+1} is on the vertical axis.
- x_t is on the horizontal axis.

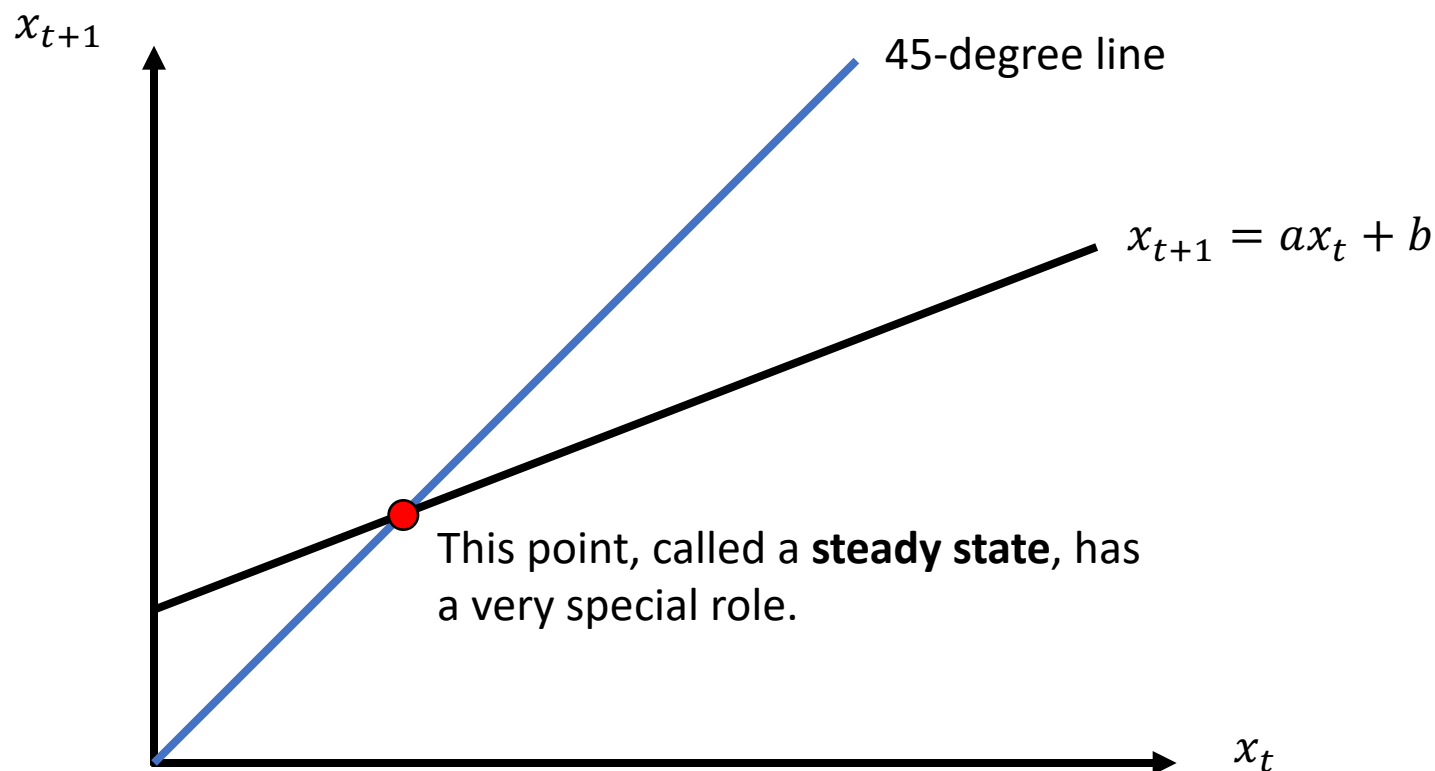
Solving $x_{t+1} = ax_t + b$



Solving $x_{t+1} = ax_t + b$



Solving $x_{t+1} = ax_t + b$



Solving $x_{t+1} = ax_t + b$

- Definition: A **steady state** (or, stationary state) of a dynamical system is a solution to the difference equation such that $x_t = x_{ss}$ for all t .
 - x_{ss} is a number.
 - Notation can be anything, but people prefer \bar{x} , x^* , x_{ss} , or simply x .
- Finding a steady state is easy.
- Apply the definition to our DE to obtain
$$x_{ss} = ax_{ss} + b$$
- Note that this is a simple equation.

Solving $x_{t+1} = ax_t + b$

- The solution to $x_{ss} = ax_{ss} + b$ (when $a \neq 1$) is

$$x_{ss} = \frac{b}{1-a}$$

- Note that this is a solution to $x_{t+1} = ax_t + b$ as $x_t = x_{ss}$ for all t satisfies the equation.

Solving $x_{t+1} = ax_t + b$

- Consider $x_{t+1} = ax_t + b$.
- How can we solve it analytically?
- **Basic idea: Transform the equation to the one we know the solution.**
- Let us subtract x_{ss} from both sides of the equation (Remember that $x_{ss} = ax_{ss} + b$):

$$\begin{aligned}x_{t+1} - x_{ss} &= ax_t + b - x_{ss} \\&= ax_t + b - (ax_{ss} - b) \\&= a(x_t - x_{ss})\end{aligned}$$

Solving $x_{t+1} = ax_t + b$

- Now we have:

$$x_{t+1} - x_{ss} = a(x_t - x_{ss})$$

- Define

$$\hat{x}_t = x_t - x_{ss}$$

- Then, our non-homogeneous DE reduces to a homogeneous DE:

$$\hat{x}_{t+1} = a\hat{x}_t$$

- What is the solution to this new DE?
- You should be able to solve it within a second.

Solving $x_{t+1} = ax_t + b$

- The solution to $\hat{x}_{t+1} = a\hat{x}_t$ is $\hat{x}_t = a^t \hat{x}_0$.
- Unfortunately, we are not done yet because this is only the solution to the transformed equation, not the original one.
- Our next (and final) step is to recover the solution to the original DE.
- Substitute $\hat{x}_t = x_t - x_{ss}$ back into $\hat{x}_t = a^t \hat{x}_0$ to obtain

$$x_t - x_{ss} = a^t(x_0 - x_{ss})$$

Solving $x_{t+1} = ax_t + b$

- From our previous expression $x_t - x_{ss} = a^t(x_0 - x_{ss})$, we obtain
$$x_t = x_{ss} + a^t(x_0 - x_{ss})$$
- This is the solution to the original DE.
- When $b = 0$, we have $x_{ss} = 0$ and the solution reduces to

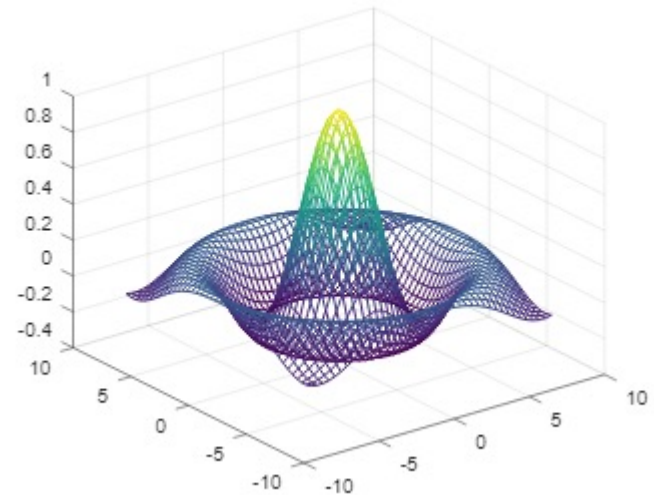
$$x_t = a^t x_0$$

Stability

- Definition: x_{ss} is **stable** if
$$\lim_{t \rightarrow \infty} x_t = x_{ss} \text{ for any } x_0$$
- For any value of b ,
 - $|a| < 1 \Leftrightarrow x_{ss}$ is stable.
 - $|a| > 1 \Leftrightarrow x_{ss}$ is unstable.
 - $a > 0 \Rightarrow \{x_t\}_{t=0}^{\infty}$ is monotonic.
 - $a < 0 \Rightarrow \{x_t\}_{t=0}^{\infty}$ displays oscillations.
- For stability, a plays the central role, while b only determines the level of the steady state.

Get Ready for Octave

- Install Octave on your PC from this site (not the octave site):
- <https://www.dynare.org/download/>
 - Win: You should choose **Octave 8.1.0**
 - Mac: visit https://www.dynare.org/resources/quick_start/



Get Ready for Python (Optional)

- Install Anaconda (= Python) on your PC.
 - https://python-programming.quantecon.org/getting_started.html
- Or, get a google account.
 - Many of you already have one, I suppose.
- Python is optional, but I recommend that you be multilingual.

