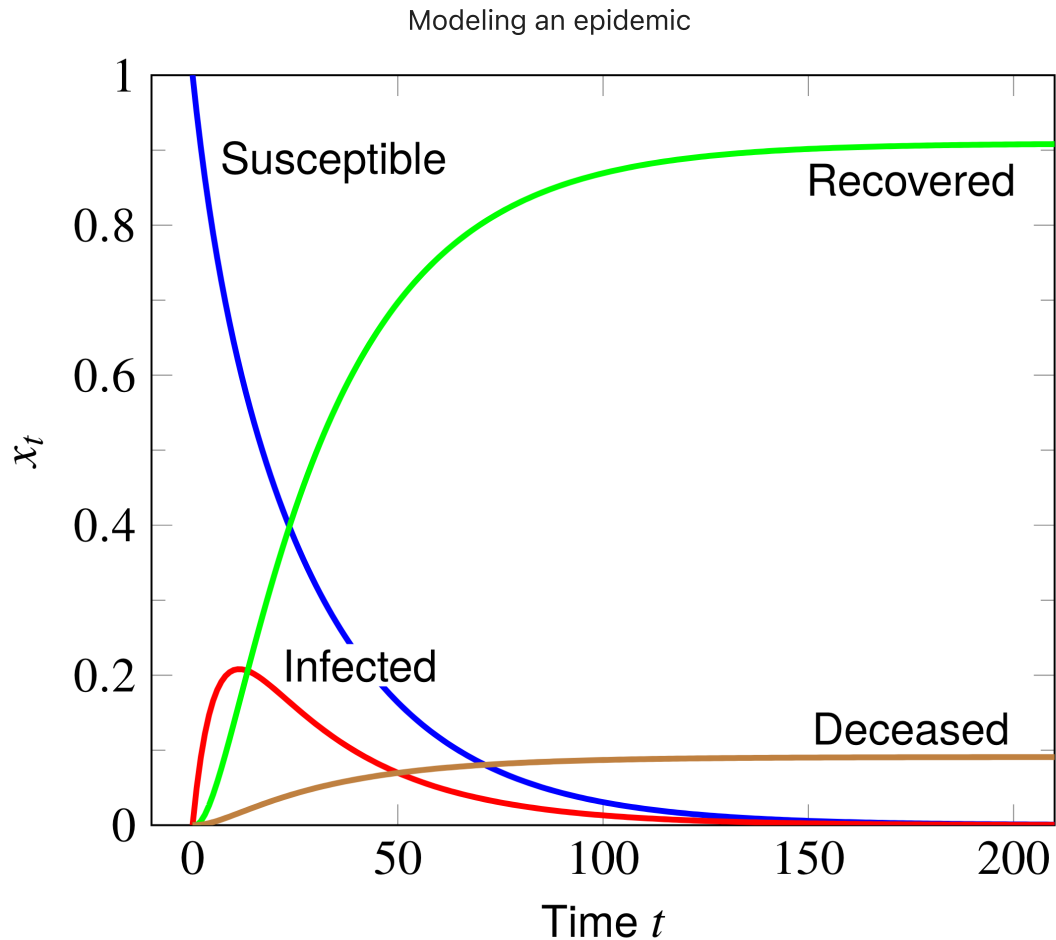


# 07 Linear Dynamical Systems



Unit 1: Vectors, Book ILA Ch. 1-5

Unit 2: Matrices, Book ILA Ch. 6-11 + Book IMC Ch. 2

- **06 Matrices**
- **07 Linear Equations**
- 08 Linear Dynamical Systems
- 09 Matrix Multiplication
- 10 Matrix Inverse

Unit 3: Least Squares, Book ILA Ch. 12-14 + Book IMC Ch. 8

Unit 4: Eigen-decomposition, Book IMC Ch. 10, 12, 19

## Exercises

**Exercise:** Let  $0_n$  be the  $n \times n$  zeroes matrix and  $x$  an  $n$ -vector. Compute  $0_n x$ .

**Exercise:** Consider a function  $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$ , that is affine. What is the matrix  $A$  and the vector  $b$  such that:  $f(x) = Ax + b$ ?

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## Outline: 08 Linear Dynamical Systems

- [Linear dynamical systems](#)
- [Epidemic dynamics](#)

## State sequence

**Definition:** Consider a sequence of  $n$ -vectors  $x_1, x_2, \dots, x_t, \dots$  where  $t$  denotes the time or period. Then:

- $x_t$  is called a state at time  $t$ , or a time-point
- the sequence  $x_1, x_2, \dots, x_t, \dots$  is called a state trajectory or a time-series.

Assuming  $t$  is the current time:

- $x_t$  is the current state,
- $x_{t-1}$  is the previous state,
- $x_{t+1}$  is the next state.

## Covid-19 in Santa Barbara

In [69]:

```
import pandas as pd

filename = 'https://raw.githubusercontent.com/CSSEGISandData/COVID-19/master/css
df_covid = pd.read_csv(filename)
df_sb = df_covid[df_covid["Admin2"] == "Santa Barbara"]
df_sb.head()
```

Out [69]:

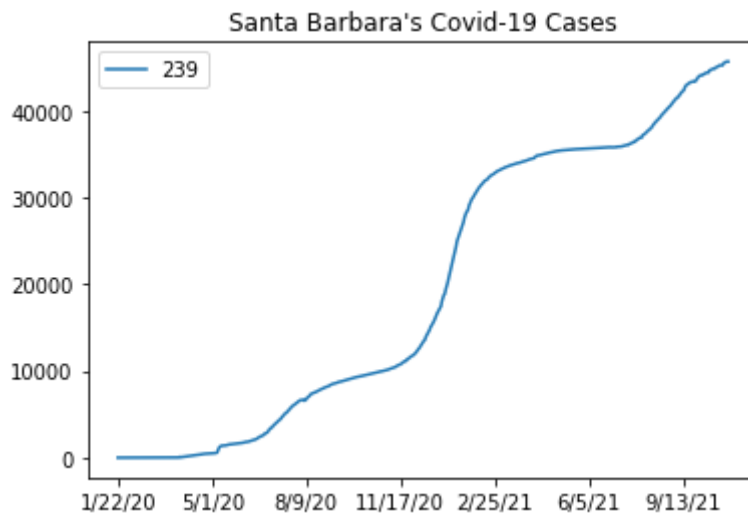
UID	iso2	iso3	code3	FIPS	Admin2	Province_State	Country_Region	Lat
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	UID	iso2	iso3	code3	FIPS	Admin2	Province_State	Country_Region	Lat
239	84006083	US	USA	840	6083.0	Santa Barbara	California	US	34.653295

1 rows x 660 columns

In [71]:

```
df_timeseries = df_sb.drop(
    columns=[
        "Combined_Key", "UID", "iso2", "iso3", "code3",
        "FIPS", "Admin2", "Province_State", "Country_Region",
        "Lat", "Long_"
    ])
df_timeseries.transpose().plot(
    title="Santa Barbara's Covid-19 Cases");
```



## Other Examples

- epidemiology: susceptible, infected, recovered and deaths at  $t$
- mechanics: position and velocity of plane through time  $t$
- finance: return of  $n$  assets at time  $t$

**Exercise:** In these examples, what is the state  $x_t$ ?

## Linear Dynamical System

**Definition:** A linear dynamical system is a linear equation describing the evolution of a state over time, and written as:

$$x_{t+1} = A_t x_t \quad t = 1, 2, \dots$$

where:

- $A_t$  are  $n \times n$  dynamics matrices,
- $(A_t)_{ij}(x_t)_j$  is contribution to  $(x_{t+1})_i$  from  $(x_t)_j$ .

The system is called time-invariant if  $A_t = A$  does not depend on time.

**In Python:** We can simulate evolution of  $x_t$  using for loops to compute:

$$x_{t+1} = A_t x_t$$

recursively (see later slide on the epidemics model).

## Variations: Linear Dynamical System with Input

**Definition:** The following system is called a linear dynamical system with input:

$$x_{t+1} = A_t x_t + B_t u_t + c_t \quad t = 1, 2, \dots$$

- $u_t$  is an input  $m$ -vector
- $B_t$  is  $n \times m$  input matrix
- $c_t$  is offset.

## Variations: $K$ - Markov Model

**Definition:** The following system is called a  $K$ -Markov model or an auto-regressive model :

$$x_{t+1} = A_1 x_t + \dots + A_K x_{t-K+1} \quad t = K, K+1, \dots$$

- next state depends on current state and  $K - 1$  previous states
- for  $K = 1$ :
  - time-invariant linear dynamical system  $x_{t+1} = A x_t$ .

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## SIR Model

**Definition:** The Susceptible-Infected-Recovered (SIR) model of an epidemic is a linear dynamical system defined on a 4-vector state  $x_t$ , that gives the proportion of the population in 4 infection states:

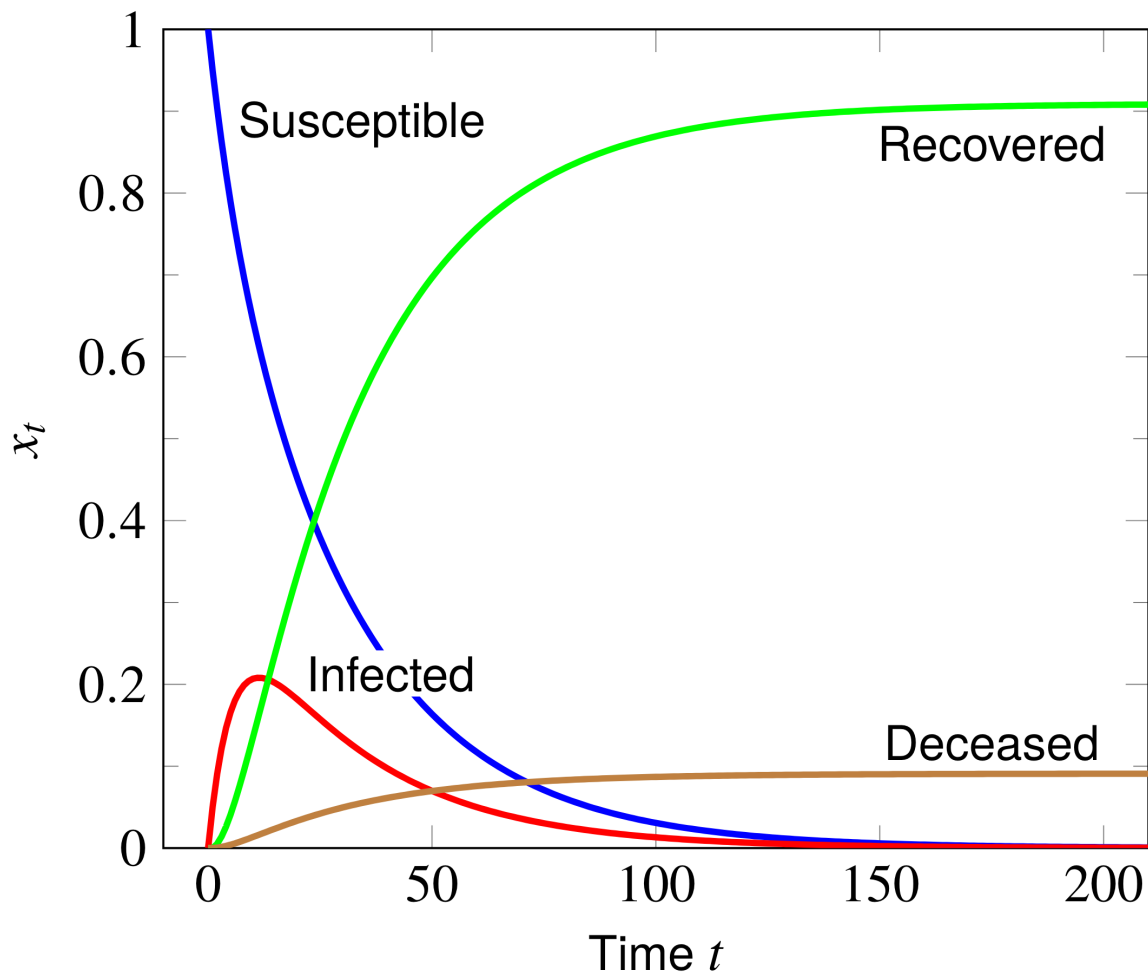
- Susceptible: can acquire the disease the next day
- Infected: have the disease
- Recovered: had the disease, recovered, now immune
- Deceased: had the disease, and unfortunately died

**Example of state:**  $x_t = (0.75, 0.10, 0.10, 0.05)$

**Exercise** Write the linear dynamical system associated to this epidemic.

- among susceptible population
  - 5% acquires the disease
  - 95% remain susceptible
- among infected population
  - 1% dies
  - 10% recovers with immunity
  - 4% recover without immunity (i.e., become susceptible)
  - 85% remain infected
- 100% of immune and dead people remain in their state

Simulation from  $x_0 = (1, 0, 0, 0)$



**Exercise:** Recreate this plot.

In [72]:

```
import matplotlib.pyplot as plt
import numpy as np

x0 = np.array([1, 0, 0, 0])
A = np.array([
    [0.95, 0.04, 0, 0],
    [0.05, 0.85, 0, 0],
    [0, 0, 1, 0],
    [0, 0, 0, 1]
```

```
[0, 0.1, 1, 0],  
[0, 0.01, 0, 1]  
])  
  
...
```

Out[72]: Ellipsis

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Resources: Book ILA Ch. 09