

# Linear Dynamical Systems

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# Outline

Linear dynamical systems

Population dynamics

Epidemic dynamics

## State sequence

- ▶ sequence of  $n$ -vectors  $x_1, x_2, \dots$ ,
- ▶  $t$  denotes time or period
- ▶  $x_t$  is called *state* at time  $t$ ; sequence is called *state trajectory*
- ▶ (assuming  $t$  is current time)
  - $x_t$  is current state
  - $x_{t-1}$  is previous state
  - $x_{t+1}$  is next state
- ▶ examples:  $x_t$  represents
  - age distribution in a population
  - economic output in  $n$  sectors
  - mechanical variables

# Linear dynamics

- ▶ linear dynamical system:  $x_{t+1} = A_t x_t$
- ▶  $A_t$  are  $n \times n$  dynamics matrices
- ▶  $(A_t)_{ij}(x_t)_j$  is contribution to  $(x_{t+1})_i$  from  $(x_t)_j$
- ▶ called *time-invariant* if  $A_t = A$  doesn't depend on time
- ▶ can simulate evolution of  $x_t$  using recursion  $x_{t+1} = A_t x_t$

## Variations

- ▶  $x_{t+1} = A_t x_t + B_t u_t + c_t$ 
  - $u_t$  is an *input*  $m$ -vector
  - $B_t$  is  $n \times m$  *input matrix*
  - $c_t$  is *offset*
- ▶ used in many application areas

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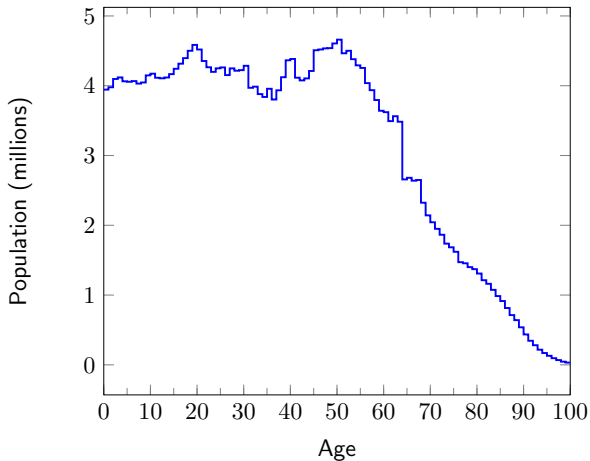
Epidemic dynamics

## Population distribution

- ▶  $x_t \in \mathbf{R}^{100}$  gives population distribution in year  $t = 1, \dots, T$
- ▶  $(x_t)_i$  is the number of people with age  $i - 1$  in year  $t$   
(say, on January 1)
- ▶ total population in year  $t$ :  $\mathbf{1}^T x_t$
- ▶ number of people age 70 or older in year  $t$ :  $(0_{70}, \mathbf{1}_{30})^T x_t$

## Population distribution of the U.S.

(from 2010 census)

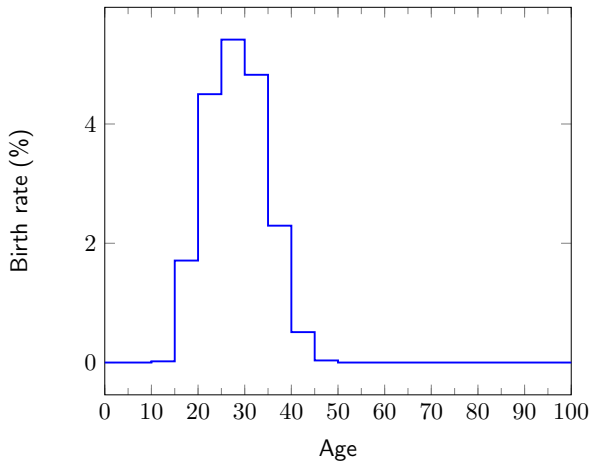




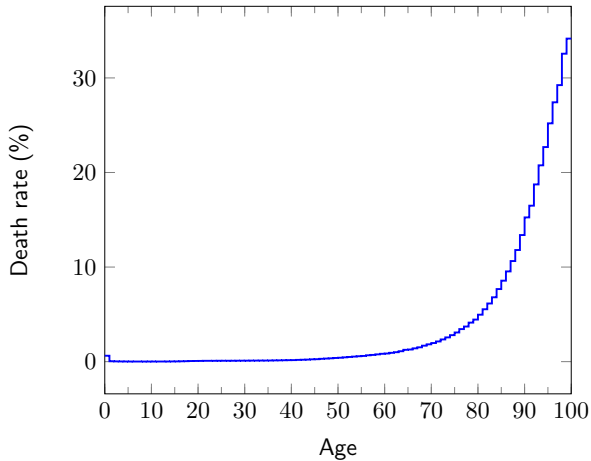
## Birth and death rates

- ▶ birth rate  $b \in \mathbf{R}^{100}$ , death (or mortality) rate  $d \in \mathbf{R}^{100}$
- ▶  $b_i$  is the number of births per person with age  $i - 1$
- ▶  $d_i$  is the portion of those aged  $i - 1$  who will die this year (we'll take  $d_{100} = 1$ )
- ▶  $b$  and  $d$  can vary with time, but we'll assume they are constant

## Birth rate in the U.S.



## Death rate in the U.S.



## Dynamics

- ▶ let's find next year's population distribution  $x_{t+1}$  (ignoring immigration; we'll add that later)
- ▶ number of 0-year-olds next year is total births this year:

$$(x_{t+1})_1 = b^T x_t$$

- ▶ number of  $i$ -year-olds next year is number of  $(i - 1)$ -year-olds this year, minus those who die:

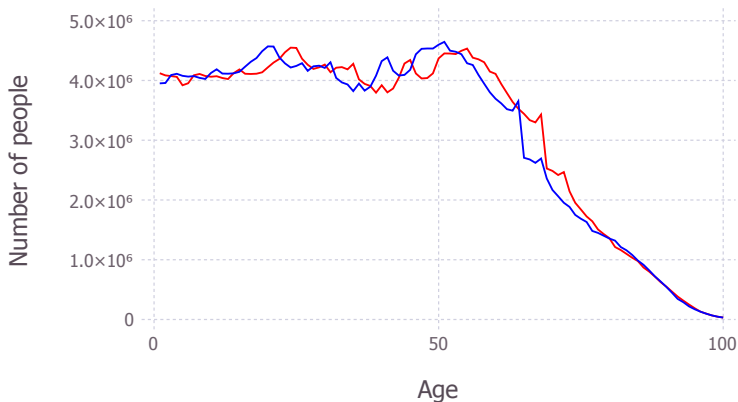
$$(x_{t+1})_{i+1} = (1 - d_i)(x_t)_i, \quad i = 1, \dots, 99$$

- ▶  $x_{t+1} = Ax_t$ , where

$$A = \begin{bmatrix} b_1 & b_2 & \cdots & b_{99} & b_{100} \\ 1 - d_1 & & & \cdots & 0 \\ 0 & 1 - d_2 & & \cdots & 0 \\ \vdots & & \ddots & & \vdots \\ 0 & \cdots & & 1 - d_{99} & 0 \end{bmatrix}$$

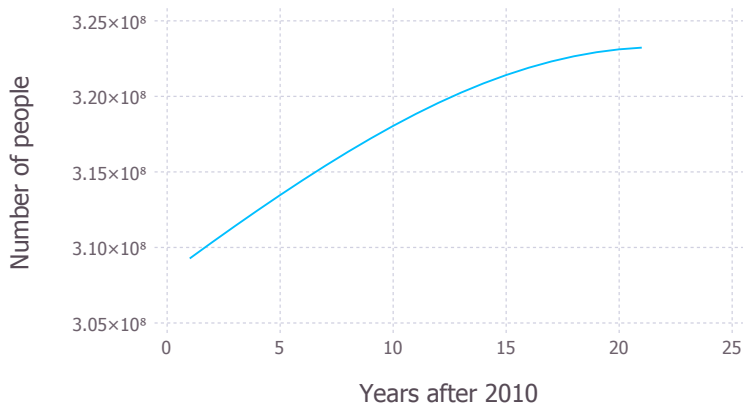
## Predicting future population distributions

predicting U.S. 2015 distribution from 2010 (ignoring immigration)



## Predicting population growth

predicted population growth (ignoring immigration)



## Initial population distributions

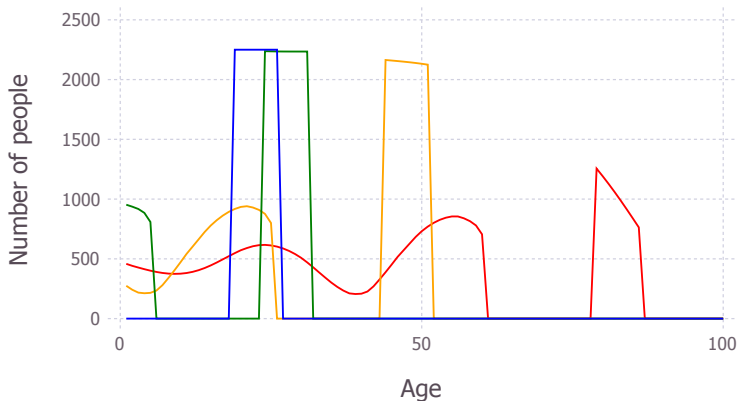
- ▶ what if we changed  $x_0$ ?
- ▶ instead of U.S. Census data, let's use a “college nation”

$$(x_0)_i = \begin{cases} 2200 & i = 19, 20, \dots, 27 \\ 0 & \text{otherwise} \end{cases}$$

(approximate population distribution of Stanford students)

## Predicting future population distributions

predict  $s$  years into the future





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# SIR model

- ▶ 4-vector  $x_t$  gives proportion of 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(sometimes called *SIR model*)
- ▶ e.g.,  $x_t = (0.75, 0.10, 0.10, 0.05)$

## Epidemic dynamics

over each day,

- ▶ 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)

## Epidemic dynamics as linear dynamical system

$$x_{t+1} = \begin{bmatrix} 0.95 & 0.04 & 0 & 0 \\ 0.05 & 0.85 & 0 & 0 \\ 0 & 0.10 & 1 & 0 \\ 0 & 0.01 & 0 & 1 \end{bmatrix} x_t$$

## Simulation from $x_1 = (1, 0, 0, 0)$

