

# Nonlinear Models: Logistic Equation

Department of Mathematics

Salt Lake Community College

(Slides by Adam Wilson)

## Nonlinear Differential Equations

Consider the following nonlinear differential equations.

$$y' = y(1 - y)$$

$$y' = \cos(y - t)$$

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**Analytical:** Sadly, analytical methods cannot always provide formulas for a solutions. Since none of these are linear, the methods we have discussed this chapter won't help us. While the first equation is separable, the other two are not.

**Numerical:** We could apply a numerical method, though this only gives a single approximate solution. Moreover, the further you move from the initial conditions, the less accurate your numerical solution is likely to be.

## Qualitative Analysis

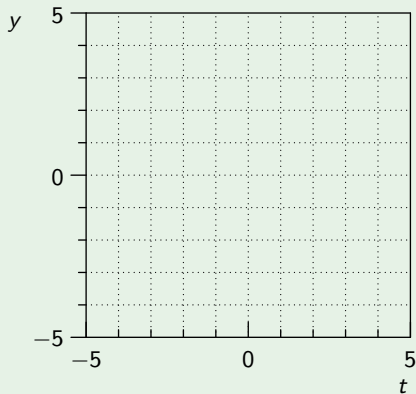
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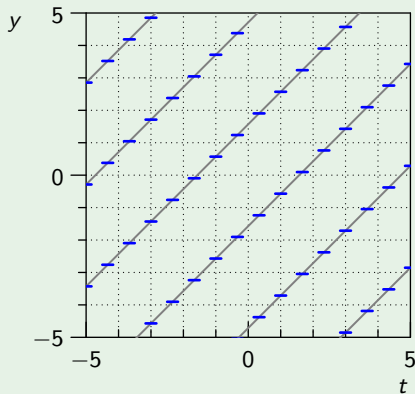
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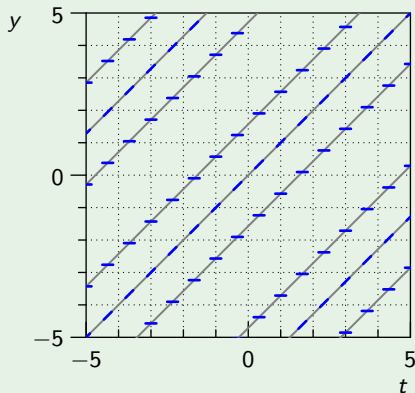
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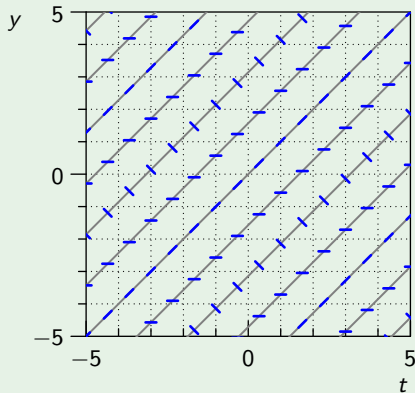
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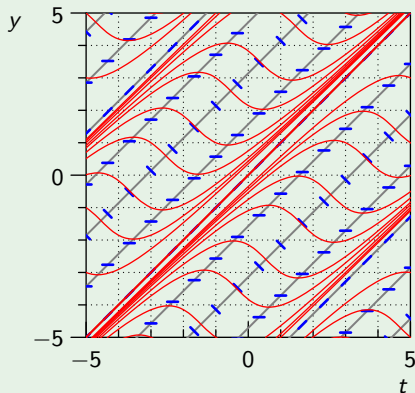


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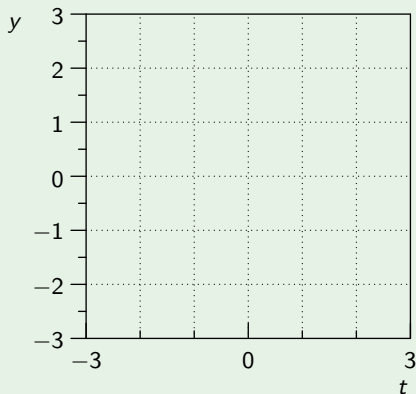


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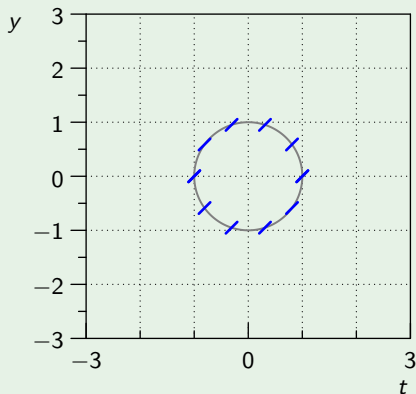
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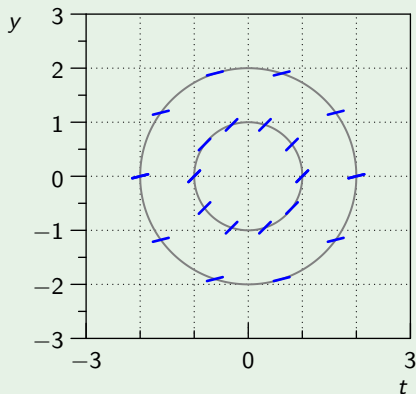


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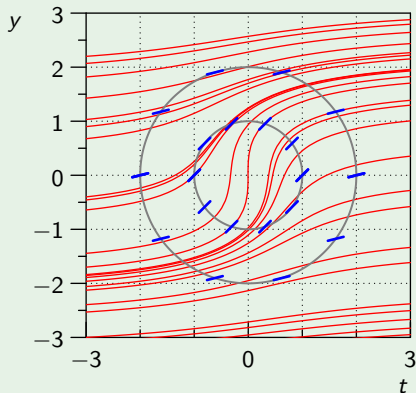


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- When  $y' = 2$ :  $t^2 + y^2 = 2^2$ .
- When  $t^2 + y^2 \rightarrow \infty$ , Slope  $\rightarrow 0$ .
- When  $t^2 + y^2 \rightarrow 0$ , Slope  $\rightarrow$  vertical.

## Autonomous Differential Equation

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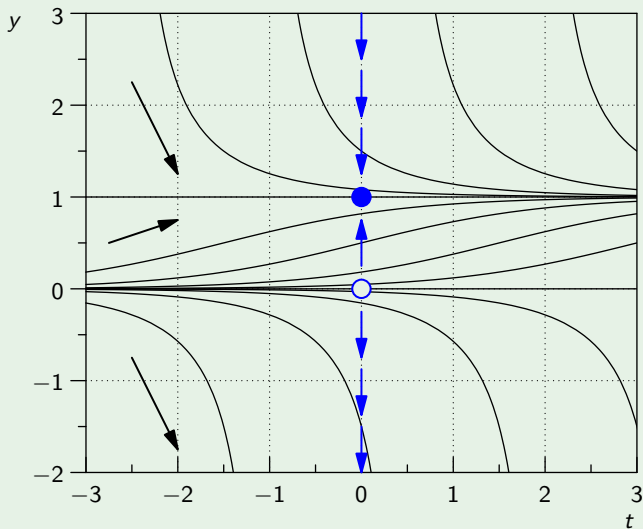
### Phase Line

Thus, for a given  $y$  value, all solutions are horizontal translations. Which means we can encapsulate information about all solutions with a vertical line, called a **phase line**.



### Example 3

$$y' = y(1 - y)$$



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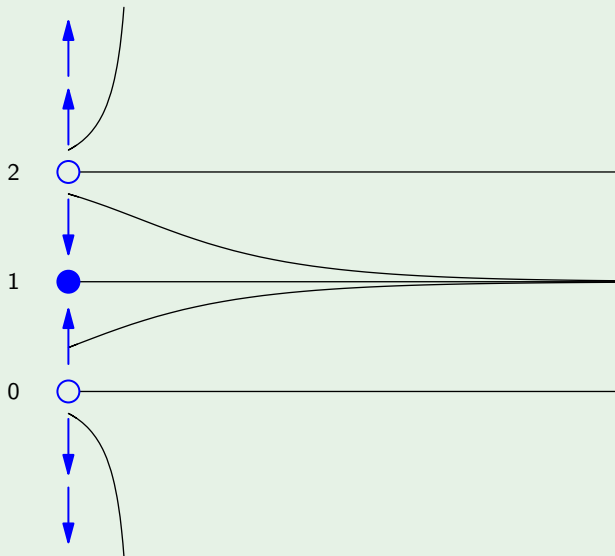
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**Semistable** If the phase-line one of the arrows above or below the equilibrium point towards the equilibrium and the other points away. (Also called a **node**.)

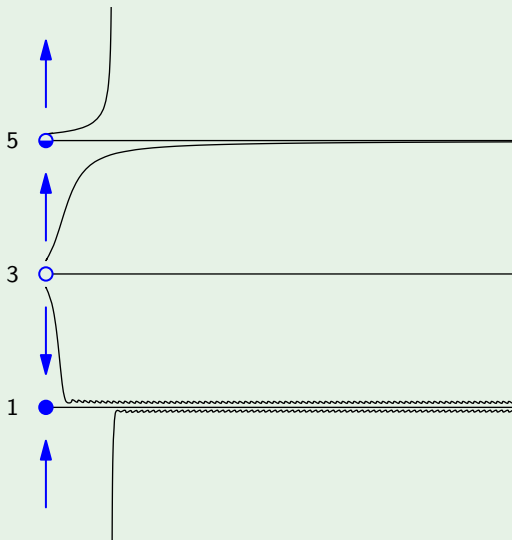
## Example 4

$$y' = y(1 - y)(2 - y)$$



## Example 5

$$y' = (y - 1)(y - 3)(y - 5)^2;$$



## Population Models

Consider the unrestricted growth equation:

$$\frac{dy}{dt} = ky, \quad k > 0$$

which assumes that the rate of growth of a population is always proportional to its size. This equation predicts exponential growth that cannot continue indefinitely.

For long-range predictions we need to consider how the population interacts with its environment. That is, as a population will level off as it reaches a limited food supply, increased disease, crowding, etc.

To build a model that includes these factors we need to replace the constant growth rate  $k$  with a variable growth rate  $k(y)$  that depends on the population size:

$$\frac{dy}{dt} = k(y) \cdot y, \quad k > 0$$

## Logistic Equation

A population may be modeled using

$$\frac{dy}{dt} = r \left( 1 - \frac{y}{L} \right) y$$

where positive parameter  $r$  is called the **initial growth rate** and  $L$  is the **carrying capacity**.



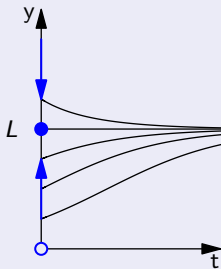
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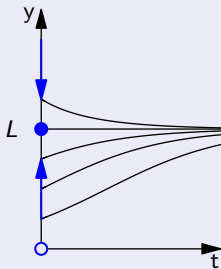
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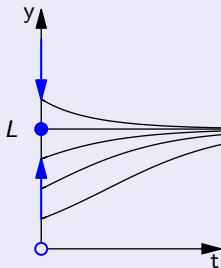
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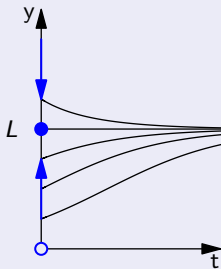
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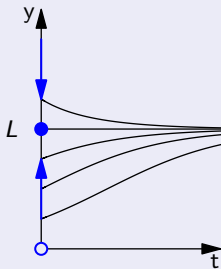
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So, we are really solving

$$\left( \frac{1}{y} + \frac{\frac{1}{L}}{1 - \frac{1}{L}} \right) dy = r \, dt$$



## Analytic Solution of the Logistic Equation

Integrating both sides gives

$$\ln |y| - \ln \left| 1 - \frac{y}{L} \right| = rt + C$$

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Note: If  $y_0 > L$ , we will arrive at the same solution.

## Initial-Value Problem for the Logistic Equation

The solution for  $t \geq 0$  of the logistic IVP

$$\frac{dy}{dt} = r \left(1 - \frac{y}{L}\right) y, \quad y(0) = y_0$$

is given by

$$y(t) = \frac{L}{1 + \left(\frac{L}{y_0} - 1\right) e^{-rt}}$$

where  $r > 0$  is the **initial growth rate** and  $L > 0$  is the **carrying capacity**.

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To model the U.S. population using the logistic equation, we will let  $t = 0$  represent the year 1990 and  $t = 1$  the year 2000.

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Given that we need to find both  $r$ ,  $L$ , and  $y_0$  we will need three data points:

$$y(0) = y_0 = 76.1, \quad y(0.5) = 151.1, \quad y(1) = 271.3$$

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Which means we have the two equations:

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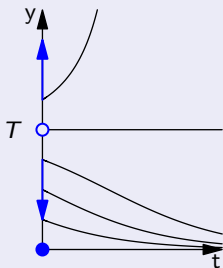
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- The theoretical maximum U.S. population is roughly 774 million.
- The projected population in 2030 is  $y(1.3) \approx 360.7$  million.
- The backward projected population in 1790 is  $y(-1.1) \approx 14.3$  million. (The actual population was 4 million. Why the discrepancy?)

## Threshold Equation

For some species there is a critical population size, such that if the population ever falls below this the species will go extinct. This level  $T$ , called the **threshold level** behaves like a carrying capacity, except solutions need to tend away from  $T$ .



The **threshold equation** is the logistic equation with a negative sign:

$$\frac{dy}{dt} = -r \left( 1 - \frac{y}{L} \right) y$$

## Initial-Value Problem for the Threshold Equation

the solution for  $t \geq 0$  of the threshold IVP

$$\frac{dy}{dt} = -r \left(1 - \frac{y}{L}\right) y, \quad y(0) = y_0$$

is given by

$$y(t) = \frac{T}{1 + \left(\frac{T}{y_0} - 1\right) e^{rt}}$$

where  $r > 0$  is the **initial growth rate** and  $T > 0$  the **threshold level**.