

# Math 337 - Elementary Differential Equations

## Lecture Notes – Introduction to Differential Equations

Joseph M. Mahaffy,  
([jmahaffy@sdsu.edu](mailto:jmahaffy@sdsu.edu))

Department of Mathematics and Statistics  
Dynamical Systems Group  
Computational Sciences Research Center  
San Diego State University  
San Diego, CA 92182-7720

<http://jmahaffy.sdsu.edu>

Spring 2022

# Outline

- 1 The Class — Overview
  - Contact Information, Office Hours
  - Text & Topics
  - Grading and Expectations
- 2 The Class...
  - MatLab
  - Formal Prerequisites
- 3 Introduction
  - Malthusian Growth
  - Examples
  - Definitions - What is a Differential Equation?
  - Classification
- 4 Applications of Differential Equations
  - Checking Solutions and IVP
  - Evaporation Example
  - Nonautonomous Example
  - Introduction to Maple

# Contact Information



## Professor Joseph Mahaffy

Office	GMCS-593
Email	<a href="mailto:jmahaffy@sdsu.edu">jmahaffy@sdsu.edu</a>
Web	<a href="http://jmahaffy.sdsu.edu">http://jmahaffy.sdsu.edu</a>
Phone	(619)594-3743
Office Hours	MW: 11:30-12:45 at GMCS 593 and by Appointment

## Basic Information: Text/Topics

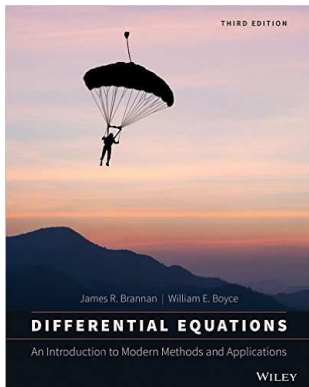
**Text:** The text is optional and old editions are fine.

**Brannan and Boyce:**  
**Differential Equations:**  
An Introduction to Modern  
Methods and Applications.

Wiley 2015.

ISBN 978-1-118-53177-8

**Lecture Notes** available at Bookstore



## Basic Information: Text/Topics

- 1 Introductory Definitions
- 2 Qualitative Methods and Direction Fields
- 3 Linear Equations
- 4 Separable Equations
- 5 Exact and Bernoulli Equation
- 6 Existence and Uniqueness
- 7 Numerical Methods
- 8 2D Linear Systems
- 9 Second Order Differential Equations
- 10 Laplace Transforms
- 11 Power Series

## Other Differential Equation Courses

**Differential Equations and Dynamical Systems:** Several courses extend the material from this class. Courses from the Nonlinear Dynamical Systems Group.

- **Math 531:** *Partial Differential Equations*
- **Math 537:** *Ordinary Differential Equations*
- **Math 538:** *Discrete Dynamical Systems and Chaos*
- **Math 542:** *Introduction to Computational Ordinary Differential Equations*

# Basic Information: Grading

## Approximate Grading

Homework, including WeBWork	30%
Lecture Activities/Computer Labs	25%
3 Exams	27%
Final	18%

- Homework is done in WeBWork and written problems (most inside WW problems) are submitted to Gradescope. Critical to **keep up** on HW after each lecture.
- Lecture Activities are written problems after lectures, which are submitted to Gradescope.
- Exams are based heavily on HW problems and examples from lectures.
- Final: Friday, May 6, 13:00 – 15:00

# Expectations and Procedures, I

- Most class attendance is **OPTIONAL** — Homework and announcements will be posted on the class web page.  
If/when you attend class:
  - You **must wear a mask** according to University rules.
  - Please be on time and pay attention.
  - Please turn off mobile phones.
  - Please be courteous to other students and the instructor.
  - Abide by university statutes, and all applicable local, state, and federal laws.



## Expectations and Procedures, II

- Please, turn in assignments on time. (The instructor reserves the right not to accept late assignments, and there is a maximum of **2** extensions of WeBWorK during the semester.)
- The instructor will make special arrangements for students with documented learning disabilities and will try to make accommodations for other unforeseen circumstances, *e.g.* illness, personal/family crises, etc. in a way that is fair to all students enrolled in the class. ***Please contact the instructor EARLY regarding special circumstances.***
- Students are expected ***and encouraged*** to ask questions in class!
- Students are expected ***and encouraged*** to to make use of office hours! If you cannot make it to the scheduled office hours: contact the instructor to schedule an appointment!

## Expectations and Procedures, III

- Missed midterm exams: Don't miss exams! The instructor reserves the right to schedule make-up exams, modify the type and nature of this make-up, and/or base the grade solely on other work (including the final exam).
- Missed final exam: Don't miss the final! Contact the instructor ASAP or a grade of incomplete or F will be assigned.
- *Academic honesty*: Submit your own work. Any cheating will be reported to University authorities and a **ZERO** will be given for that HW assignment or Exam.

# MatLab

- Students can obtain **MatLab** from EDORAS Academic Computing.
- Google **SDSU MatLab** or access <https://edoras.sdsu.edu/download/matlab.html>.
- **MatLab** and **Maple** can also be accessed in the **Computer Labs GMCS 421, 422, and 425**.
- A discounted student version of **Maple** is available.

# Math 337: Formal Prerequisites

Math 254 or Math 342A or AE 280

- These courses all require **Calculus 151**.
  - Assume good knowledge of *differentiation* and *integration*.
  - Understand series techniques (especially *Taylor's Theorem*)
  - Recall *Partial Fractions Decomposition*.
- These courses all have sections on basic **Linear Algebra**.

# Introduction

## Introduction

- Differential equations frequently arise in modeling situations
- They describe population growth, chemical reactions, heat exchange, motion, and many other applications
- Differential equations are continuous analogs of discrete dynamical systems

# Malthusian Growth

1

## Discrete Malthusian Growth Model:

- Let the initial population,  $P(t_0) = P_0$
- Define  $t_n = t_0 + n\Delta t$  and  $P_n = P(t_n)$
- Let  $r$  be the per capita growth rate per unit time
- The Discrete Malthusian Growth Model satisfies:

$$P_{n+1} = P_n + r\Delta t P_n = (1 + r\Delta t)P_n$$

- New population = Old population + per capita growth rate  $\times$  length of time  $\times$  Old population

# Malthusian Growth

2

**Discrete Malthusian Growth:**  $P_{n+1} = (1 + r\Delta t)P_n$ , so

$$P_1 = (1 + r\Delta t)P_0$$

$$P_2 = (1 + r\Delta t)P_1 = (1 + r\Delta t)^2 P_0$$

$$P_3 = (1 + r\Delta t)P_2 = (1 + r\Delta t)^3 P_0$$

$$\vdots$$

$$P_n = (1 + r\Delta t)P_{n-1} = (1 + r\Delta t)^n P_0$$

The solution of this discrete model is

$$P_n = (1 + r\Delta t)^n P_0,$$

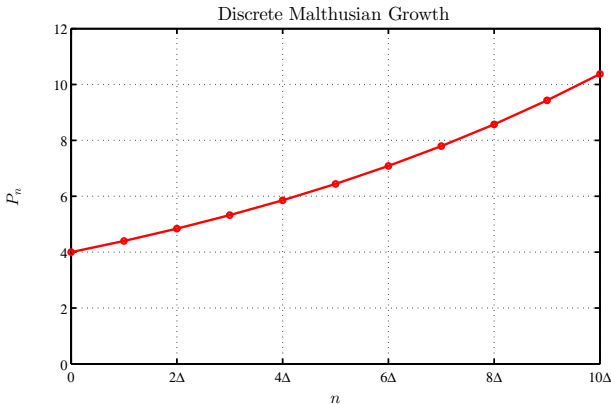
which is an exponential growth

# Malthusian Growth

3

## Discrete Malthusian Growth:

$$P_{n+1} = (1 + 0.1\Delta t)P_n \quad P_0 = 4$$





# Malthusian Growth

4

**Malthusian Growth:** Let  $P(t)$  be the population at time  $t = t_0 + n\Delta t$  and rearrange the model above

$$\begin{aligned}
 P_{n+1} - P_n &= r\Delta t P_n \\
 P(t + \Delta t) - P(t) &= \Delta t \cdot rP(t) \\
 \frac{P(t + \Delta t) - P(t)}{\Delta t} &= rP(t)
 \end{aligned}$$

Let  $\Delta t$  become very small

$$\lim_{\Delta t \rightarrow 0} \frac{P(t + \Delta t) - P(t)}{\Delta t} = \frac{dP(t)}{dt} = rP(t),$$

which is a **Differential Equation**

# Malthusian Growth

5

**Solution of Malthusian Growth Model:** The Malthusian growth model

$$\frac{dP(t)}{dt} = rP(t)$$

- The rate of change of a population is proportional to the population
- Let  $c$  be an arbitrary constant, so try a solution of the form

$$P(t) = ce^{kt}$$

- Differentiating

$$\frac{dP(t)}{dt} = cke^{kt},$$

which if  $k = r$  is  $rP(t)$ , so satisfies the differential equation

# Malthusian Growth

6

**Solution of Malthusian Growth Model** The Malthusian growth model satisfies

$$P(t) = ce^{rt}$$

- With the initial condition,  $P(t_0) = P_0$ , then the unique solution is

$$P(t) = P_0 e^{r(t-t_0)}$$

- Malthusian growth is often called exponential growth

# Example 1: Malthusian Growth

1

**Example 1: Malthusian Growth** Consider the Malthusian growth model

$$\frac{dP(t)}{dt} = 0.02 P(t) \quad \text{with} \quad P(0) = 100$$

Skip Example

- Find the solution
- Determine how long it takes for this population to double

## Example 1: Malthusian Growth

2

**Solution:** The solution is given by

$$P(t) = 100 e^{0.02t}$$

Since  $P(0) = 100$ , satisfying the initial condition, then by computing

$$\frac{dP}{dt} = 0.02(100 e^{0.02t}) = 0.02 P(t),$$

we find that this solution satisfies the differential equation

The population doubles when

$$\begin{aligned} 200 &= 100 e^{0.02t} \\ 0.02t = \ln(2) &\quad \text{or} \quad t = 50 \ln(2) \approx 34.66 \end{aligned}$$

## Example 2: *E. coli* Study

1

**Example 2: *E. coli* Study** In this class we connect the *ordinary differential equations (ODEs)* to real world examples.

This requires fitting our model ODE to actual data.

Consider a culture of *Escherichia coli* growing in rich media at 25°C, which satisfies conditions for Malthusian growth,

$$\frac{dP}{dt} = kP, \quad \text{with } P(0) = P_0.$$

Below is a table of data:

$t$ (min)	OD <sub>420</sub>	$t$ (min)	OD <sub>420</sub>	$t$ (min)	OD <sub>420</sub>
0	0.195	60	0.308	120	0.473
20	0.206	80	0.364	140	0.527
40	0.241	100	0.421	160	0.618

## Example 2: *E. coli* Study

2

**Example 2:** The most common means of fitting a model to data is *minimizing the sum of square errors* between the model and the data.

**Basic statistics** shows how to do fit data to a **straight line** (a common problem in multivariate Calculus).

In general this procedure is significantly more difficult and is usually done **numerically**.

Consider a set of  $n + 1$  **data points**:  $(t_0, P_0), (t_1, P_1), \dots, (t_n, P_n)$ .

Assume the **Malthusian growth model** depends on some parameters,  $p$ :

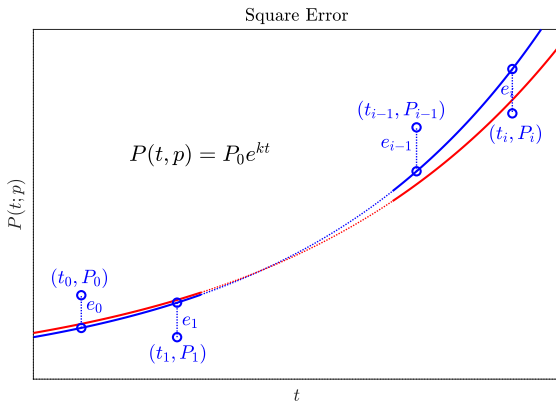
$$P(t; p) = P_0 e^{kt}, \quad \text{with } p = [P_0, k],$$

which depends **nonlinearly** on the parameters,  $p$ .

## Example 2: *E. coli* Study

3

**Example 2:** Below is a figure showing a data set and  $P(t; p)$  with two different values of  $p$ , illustrating the computation of the *sum of square errors*.





## Example 2: *E. coli* Study

4

**Least Squares Best Fit** minimizes the square of the error in the distance between the  $P_i$  data values and the  $P(t_i; p)$  value of the model.

The **error** between the data points and the model satisfies:

$$e_i = P_i - P(t_i; p) = P_i - P_0 e^{kt_i}, \quad i = 0, \dots, n,$$

which depends on  $P_0$  and  $k$ .

The **sums of square errors** function depends on the parameters  $P_0$  and  $k$  of the population model:

$$J(P_0, k) = \sum_{i=0}^n e_i^2 = \sum_{i=0}^n (P_i - P_0 e^{kt_i})^2$$

The **Least Squares Best Fit Model** is the **minimum** of the function  $J(P_0, k)$ .

## Example 2: *E. coli* Study

**Least Squares Best Fit** is found by setting the partials with respect to the parameters equal to zero:

$$\frac{\partial J(P_0, k)}{\partial P_0} = 0 \quad \text{and} \quad \frac{\partial J(P_0, k)}{\partial k} = 0.$$

These equations are highly nonlinear and difficult to solve in general.

**Computer Software Packages** often have numerical methods to approximate the solutions.

We examine two **Computer Software Packages** for finding the least squares best fit Model.

- **Excel's Solver**
- **MatLab's *fminsearch***

## Example 2: *E. coli* Study

6

**Example 2: *E. coli* Study:** Return to the **data** at the beginning of this study and the *Malthusian growth model*,  $P(t; p) = P_0 e^{kt}$ .

There is a hyperlinked **Excel file** showing how this model is fit with *Excel's Solver*

Below is the **MatLab code** for fitting the *Malthusian ODE model*, finding the best initial condition,  $P_0$ , and growth rate,  $k$ .

The *sum of square error program* is given by:

```

1 function J = sumsq_ecoli(p, tdata, pdata)
2 % Function computing sum of square errors for ...
   Malthusian model
3 model = p(1)*exp(p(2)*tdata);
4 error = model - pdata;
5 J = error*error';
6 end

```

## Example 2: *E. coli* Study

7

**Example 2: *E. coli* Study:** The *sum of square error program* is used inside the primary plotting program (line 11):

```

1 clear % Clear previous definitions
2 figure(1) % Assign figure number
3 clf % Clear previous figures
4 hold off % Start with fresh graph
5 mytitle = '\it Escherichia coli';
6 xlabel = '$t$ (min)';
7 ylabel = '$P(t)$ (OD$_{420}$)';
8 td = [0 20 40 60 80 100 120 140 160];
9 pd = [0.195 0.206 0.241 0.308 0.364 0.421 0.473 ...
        0.527 0.618];
10 tt = linspace(0,180,200);
11 [p1,J,flag] = ...
        fminsearch(@sumsq_ecoli,[0.2,0.01],[],td,pd)
12 Pt = p1(1)*exp(p1(2)*tt);

```

## Example 2: *E. coli* Study

**Example 2: *E. coli* Study:** The primary plotting program continues with:

```

13 plot(tt,Pt,'b-','LineWidth',1.5);
14 hold on
15 plot(td,pd,'bo','LineWidth',1.5);
16 grid
17 myeqn=['$P(t)=' ,num2str(p1(1)), 'e^{ ' ...
        ,num2str(p1(2)), 't}$'];
18 text(42,0.53,myeqn,'FontSize',14,'interpreter','latex');
19 legend('Model','Data','Location','southeast');
20 xlim([0 180]);
21 ylim([0 0.8]);
22 fontlabs = 'Times New Roman';
23 xlabel(xlab,'FontSize',14,'FontName',fontlabs, ...
        'interpreter','latex');

```

## Example 2: *E. coli* Study

9

**Example 2: *E. coli* Study:** The primary plotting program finishes with:

```

24 ylabel (ylab, 'FontSize', 14, 'FontName', fontlabs, ...
    'interpreter', 'latex');
25 title (mytitle, 'FontSize', 18, 'FontName', fontlabs, ...
    'interpreter', 'latex');
26 set (gca, 'FontSize', 12);
27 print -depsc ecoli.eps
28 print -djpeg ecoli.jpg
  
```

Significantly, inside the above program we see the line:

```
[p1, J, flag] = ...
```

```
fminsearch (@sumsq_ecoli, [0.2, 0.01], [], td, pd)
```

which invokes *MatLab's nonlinear solver* to **minimize** the *sum of square errors*.

Example 2: *E. coli* Study

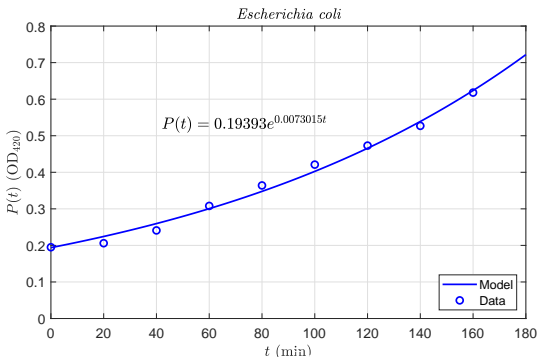
10

**Example 2: *E. coli* Study:** The result gives the *best fitting model*:

$$P(t) = 0.19393e^{0.0073015t},$$

with a *least sum of square errors*,  $J = 0.0015809$ .

The *best fitting model* with the **data** are shown in the graph below:



## Example 2: *E. coli* Study

11

**Example 2: *E. coli* Discrete Model:** Fitting the solution to an **ODE** is a basic *curve fitting* exercise.

Finding the parameters for a *discrete dynamical model* is less straightforward.

The *discrete Malthusian growth model* is the rare discrete model with an explicit answer, so could be solved using a curve fitting routine.

Below we provide the **MatLab program** for fitting this discrete model through a *least squares best fit* using an iterated simulation of the model as parameters vary.

The *discrete Malthusian growth model* satisfies the equation:

$$P_{n+1} = (1 + r)P_n, \quad \text{with initial parameter } P_0,$$

where  $n$  is the discrete time interval of 20 min.

Again we are minimizing the *sum of square errors* between the **model** and the **data** in the table above as the parameters,  $r$  and  $P_0$ , vary.



## Example 2: *E. coli* Study

12

**Example 2: *E. coli* Discrete Model:** The model is simulated,

$$P_{n+1} = (1 + r)P_n, \quad \text{with initial parameter } P_0,$$

for some parameters  $r$  and  $P_0$ , and its square error is computed with the following MatLab program:

```

1 function J = ec_disc_lst(p0,tdata,pdata)
2 % Least Squares fit to Malthusian Growth
3 N = length(tdata);
4 p = p0(1);
5 pop = [p];
6 err = [pdata(1) - p];
7 for i = 2:N % Malthusian iteration
8     p = p*(1+p0(2));
9     pop = [pop,p];
10    err = [err, pdata(i) - p];
11 end
12 J = err*err'; % Sum of square errors
13 end

```

## Example 2: *E. coli* Study

13

**Example 2: *E. coli* Discrete Model:** The **main program** used to compute the best fitting parameters and create a graph is similar to the previous plotting program but varies in a few lines:

```

8  td = [0 20 40 60 80 100 120 140 160];
9  pd = [0.195 0.206 0.241 0.308 0.364 0.421 0.473 ...
        0.527 0.618];
10 [p1,J,flag] = ...
    fminsearch(@ec_disc_lst,[0.2,0.15],[],td,pd)
11 N = length(td);
12 p = p1(1);
13 pop = [p];
14 for i = 2:N+1 % Malthusian iteration
15     p = p*(1+p1(2));
16     pop = [pop,p];
17 end

```

## Example 2: *E. coli* Study

14

**Example 2: *E. coli* Discrete Model:** The primary plotting program continues with:

```

18 plot([td,180],pop,'ro-','LineWidth',1.5);
19 hold on
20 plot(td,pd,'bo','LineWidth',1.5);
21 grid
22 myeqn=['$P_{n+1} = (1+',num2str(p1(2)), '...
        ')P_n,\quad P_0 =$', num2str(p1(1))];
23 text(37,0.63,myeqn,'FontSize',14,'interpreter','latex');
24 legend('Model', 'Data','Location','southeast');

```

As before, inside the above program the line:

```
[p1,J,flag] = ...
```

```
fminsearch(@ec_disc_lst,[0.2,0.15],[],td,pd)
```

invokes MatLab's nonlinear solver, which minimizes the sum of square errors.

## Example 2: *E. coli* Study

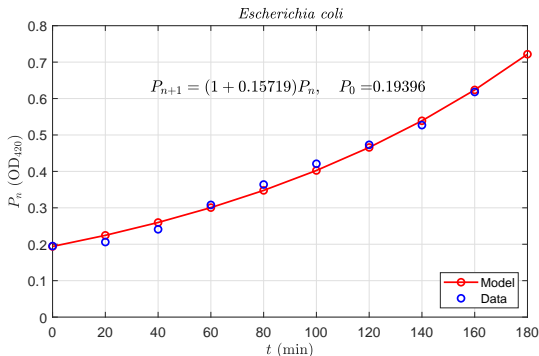
15

**Example 2: *E. coli* Discrete Model:** The result gives the *best fitting model*:

$$P_{n+1} = 1.15719P_n, \quad P_0 = 0.19396,$$

with a *least sum of square errors*,  $J = 0.0015809$ .

The *best fitting model* with the **data** are shown in the graph below:



# What is a Differential Equation?

## What is a Differential Equation?

### Definition (Differential Equation)

An equation that contains derivatives of one or more unknown functions with respect to one or more independent variables is said to be a **differential equation**.

- The classical example is Newton's Law of motion
  - The mass of an object times its acceleration is equal to the sum of the forces acting on that object
  - Acceleration is the first derivative of velocity or the second derivative of position
- In biology, a differential equation describes a growth rate, a reaction rate, or the change in some physiological state

# Types of Differential Equations

- This course considers **Ordinary Differential Equations**, where the **unknown function and its derivatives** depend on a single **independent variable**
- Mathematical physics often needs **Partial Differential Equations**, where the **unknown function and its derivatives** depend on two or more **independent variables**
  - **Example: Heat Equation**

$$\frac{\partial u(x, t)}{\partial t} = D \frac{\partial^2 u(x, t)}{\partial x^2}$$

- This course also examines some **Systems of Ordinary Differential Equations**, where there are several interacting **unknown functions and their derivatives** each depending on a single **independent variable**

# Classification

## Definition (Order)

The **order** of a **differential equation** matches the order of the highest derivative that appears in the equation.

## Definition (Linear Differential Equation)

An  $n^{\text{th}}$  order ordinary differential equation  $F(t, y, y', \dots, y^{(n)}) = 0$  is said to be **linear** if it can be written in the form

$$a_0(t)y^{(n)} + a_1(t)y^{(n-1)} + \dots + a_n(t)y = g(t).$$

The functions  $a_0, a_1, \dots, a_n$ , called the **coefficients** of the equation, can depend at most on the independent variable  $t$ . This equation is said to be **homogeneous** if the function  $g(t)$  is zero for all  $t$ .

Otherwise, the equation is **nonhomogeneous**.

# Applications of Differential Equations

1

**Radioactive Decay:** Let  $R(t)$  be the amount of a radioactive substance

- Radioactive elements transition through decay into another state at a rate proportional to the amount of radioactive material present
- The differential equation is

$$\frac{dR(t)}{dt} = -k R(t) \quad \text{with} \quad R(0) = R_0$$

- This is a **first order, linear, homogeneous differential equation**
- Like the Malthusian growth model, this has an exponential solution

$$R(t) = R_0 e^{-kt}$$



# Applications of Differential Equations

2

**Harmonic Oscillator:** A Hooke's law spring exerts a force that is proportional to the displacement of the spring

- Newton's law of motion: Mass times the acceleration equals the force acting on the mass
- The simplest spring-mass problem is

$$my'' = -cy \quad \text{or} \quad y'' + k^2y = 0$$

- This is a **second order, linear, homogeneous differential equation**
- The general solution is

$$y(t) = c_1 \cos(kt) + c_2 \sin(kt),$$

where  $c_1$  and  $c_2$  are arbitrary constants

# Applications of Differential Equations

3

**Swinging Pendulum:** A pendulum is a mass attached at one point so that it swings freely under the influence of gravity

- Newton's law of motion (ignoring resistance) gives the differential equation

$$my'' + g \sin(y) = 0$$

- The variable  $y$  is the angle of the pendulum,  $m$  is the mass of the bob of the pendulum, and  $g$  is the gravitational constant
- This is a **second order, nonlinear, homogeneous differential equation**
- This problem does not have an easily expressible solution

# Applications of Differential Equations

4

**Logistic Growth:** Most populations are limited by food, space, or waste build-up, thus, cannot continue to grow according to Malthusian growth

- The Logistic growth model has a Malthusian growth term and a term limiting growth due to crowding
- The differential equation is

$$\frac{dP}{dt} = rP \left( 1 - \frac{P}{M} \right)$$

- $P$  is the population,  $r$  is the Malthusian rate of growth, and  $M$  is the carrying capacity of the population
- This is a **first order, nonlinear, homogeneous differential equation**
- We solve this problem later in the semester

# Applications of Differential Equations

5

**The van der Pol Oscillator:** In electrical circuits, diodes show a rapid rise in current, leveling of the current, then a steep decline

- Biological applications include a similar approximation for nerve impulses
- The van der Pol Oscillator satisfies the differential equation

$$v'' + a(v^2 - 1)v' + v = b$$

- $v$  is the voltage of the system, and  $a$  and  $b$  are constants
- This is a **second order, nonlinear, nonhomogeneous differential equation**
- This problem does not have an easily expressible solution, but shows interesting oscillations

# Applications of Differential Equations

6

**Lotka-Volterra – Predator and Prey Model:** Model for studying the dynamics of predator and prey interacting populations

- Model for the population dynamics when one predator species and one prey species are tightly interconnected in an ecosystem
- System of differential equations

$$x' = ax - bxy$$

$$y' = -cy + dxy$$

- $x$  is the prey species, and  $y$  is the predator species
- This is a **system of first order, nonlinear, homogeneous differential equations**
- No explicit solution, but we'll study its behavior

# Applications of Differential Equations

7

**Forced Spring-Mass Problem with Damping:** An extension of the spring-mass problem that includes viscous-damping caused by resistance to the motion and an external forcing function that is applied to the mass

- The model is given by

$$my'' + cy' + ky = F(t)$$

- $y$  is the position of the mass,  $m$  is the mass of the object,  $c$  is the damping coefficient,  $k$  is the spring constant,  $F(t)$  is an externally applied force
- This is a **second order, linear, nonhomogeneous differential equation**
- We'll learn techniques for solving this

# Damped Spring-Mass Problem

1

**Damped Spring-Mass Problem:** Assume a mass attached to a spring with resistance satisfies the second order linear differential equation

$$y''(t) + 2y'(t) + 5y(t) = 0$$

Skip Example

Show that one solution to this differential equation is

$$y_1(t) = 2e^{-t} \sin(2t)$$

# Damped Spring-Mass Problem

2

**Solution:** Damped spring-mass problem

- The 1<sup>st</sup> derivative of  $y_1(t) = 2e^{-t} \sin(2t)$

$$y_1'(t) = 2e^{-t}(2 \cos(2t)) - 2e^{-t} \sin(2t) = 2e^{-t}(2 \cos(2t) - \sin(2t))$$

- The 2<sup>nd</sup> derivative of  $y_1(t) = 2e^{-t} \sin(2t)$

$$\begin{aligned} y_1''(t) &= 2e^{-t}(-4 \sin(2t) - 2 \cos(2t)) - 2e^{-t}(2 \cos(2t) - \sin(2t)) \\ &= -2e^{-t}(4 \cos(2t) + 3 \sin(2t)) \end{aligned}$$

- Substitute into the spring-mass problem

$$\begin{aligned} y_1'' + 2y_1' + 5y_1 &= -2e^{-t}(4 \cos(2t) + 3 \sin(2t)) \\ &\quad + 2(2e^{-t}(2 \cos(2t) - \sin(2t))) + 5(2e^{-t} \sin(2t)) \\ &= 0 \end{aligned}$$

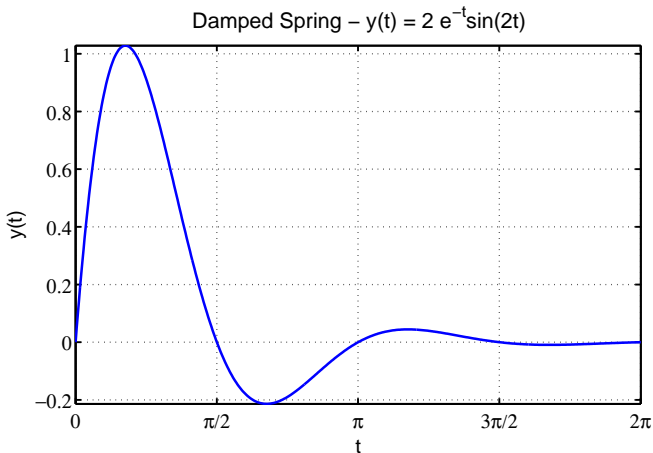
It is often **easy** to check that a solution satisfies a differential equation.



# Damped Spring-Mass Problem

3

## Graph of Damped Oscillator



# Initial Value Problem

## Definition (Initial Value Problem)

An initial value problem for an  $n^{\text{th}}$  order differential equation

$$y^{(n)} = f(t, y, y', y'', \dots, y^{(n-1)})$$

on an interval  $I$  consists of this differential equation together with  $n$  initial conditions

$$y(t_0) = y_0, \quad y'(t_0) = y_1, \quad \dots, \quad y^{(n-1)}(t_0) = y_{n-1}$$

prescribed at a point  $t_0 \in I$ , where  $y_0, y_1, \dots, y_{n-1}$  are given constants.

Under reasonable conditions the solution of an **Initial Value Problem** has a unique solution.

# Evaporation Example

1

**Evaporation Example:** Animals lose moisture proportional to their surface area

Skip Example

- If  $V(t)$  is the volume of water in the animal, then the moisture loss satisfies the differential equation

$$\frac{dV}{dt} = -0.03 V^{2/3}, \quad V(0) = 8 \text{ cm}^3$$

- The initial amount of water is  $8 \text{ cm}^3$  with  $t$  in days
- Verify the solution is

$$V(t) = (2 - 0.01t)^3$$

- Determine when the animal becomes totally desiccated according to this model
- Graph the solution

# Evaporation Example

2

**Solution:** Show  $V(t) = (2 - 0.01t)^3$  satisfies

$$\frac{dV}{dt} = -0.03 V^{2/3}, \quad V(0) = 8 \text{ cm}^3$$

- $V(0) = (2 - 0.01(0))^3 = 8$ , so satisfies the initial condition
- Differentiate  $V(t)$ ,

$$\frac{dV}{dt} = 3(2 - 0.01t)^2(-0.01) = -0.03(2 - 0.01t)^2$$

- But  $V^{2/3}(t) = (2 - 0.01t)^2$ , so

$$\frac{dV}{dt} = -0.03 V^{2/3}$$

# Evaporation Example

3

**Solution (cont):** Find the time of total desiccation

- Must solve

$$V(t) = (2 - 0.01t)^3 = 0$$

- Thus,

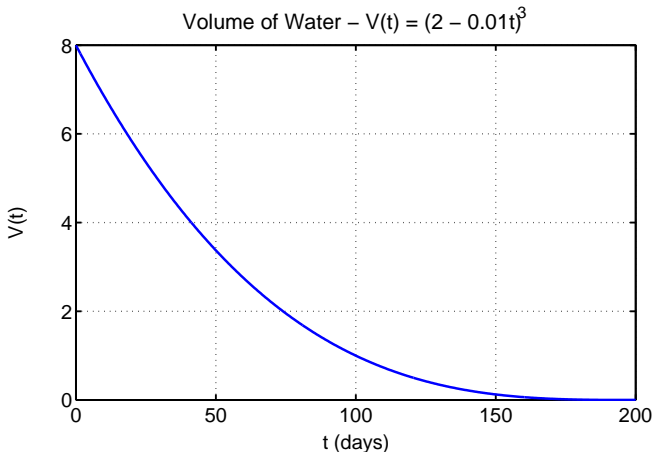
$$2 - 0.01t = 0 \quad \text{or} \quad t = 200$$

- It takes 200 days for complete desiccation

# Evaporation Example

4

## Graph of Desiccation



# Nonautonomous Example

1

**Nonautonomous Example:** Consider the nonautonomous differential equation with initial condition (**Initial Value Problem**):

$$\frac{dy}{dt} = -ty^2, \quad y(0) = 2$$

- Show that the solution to this differential equation, including the initial condition, is

$$y(t) = \frac{2}{t^2 + 1}$$

- Graph of the solution

# Nonautonomous Example

2

**Solution:** Consider the solution

$$y(t) = \frac{2}{t^2 + 1} = 2(t^2 + 1)^{-1}$$

- The initial condition is

$$y(0) = \frac{2}{0^2 + 1} = 2$$

- Differentiate  $y(t)$ ,

$$\frac{dy}{dt} = -2(t^2 + 1)^{-2}(2t) = -4t(t^2 + 1)^{-2}$$

- However,

$$-ty^2 = -t(2(t^2 + 1)^{-1})^2 = -4t(t^2 + 1)^{-2}$$

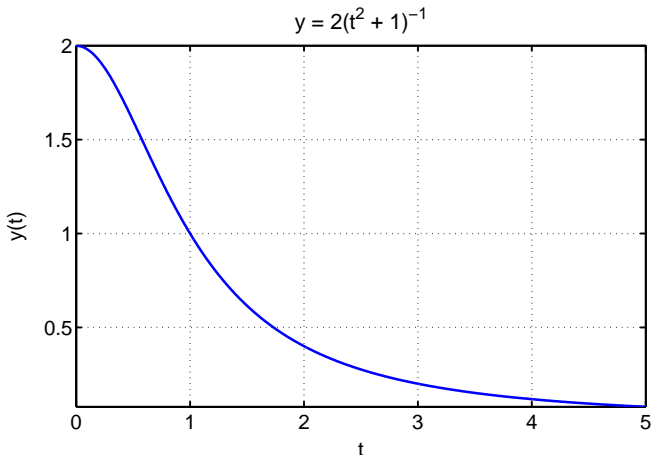
- Thus, the differential equation is satisfied



# Nonautonomous Example

3

## Solution of Nonautonomous Differentiation Equation



# Introduction to Maple

1

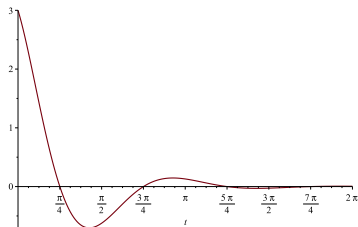
## Introduction to Maple: A Symbolic Math Program

We enter a function  $y(t) = 3e^{-t} \cos(2t)$ ,

```
y := t → 3 · exp(-t) · cos(2 · t);
```

The arrow is `→` and `>` and multiplication is `*`. To plot this function

```
plot(y(t), t = 0..2 · Pi);
```



# Introduction to Maple

2

We have the function:  $y(t) = 3e^{-t} \cos(2t)$ ,

This can be differentiated (and stored in variable  $dy$ ) by typing

```
 $dy := diff(y(t), t);$ 
```

Maple gives:

$$dy := -3e^{-t} \cos(2t) - 6e^{-t} \sin(2t)$$

The absolute minimum and a relative maximum are found with Maple:

```
 $tmin := fsolve(dy = 0, t = 1..2); \quad y(tmin);$ 
```

```
 $tmax := fsolve(dy = 0, t = 2.5..3.5); \quad y(tmax);$ 
```

The result was an **absolute minimum** at  $(1.33897, -0.703328)$ .

The result was a **relative maximum** at  $(2.90977, 0.1462075)$ .

# Introduction to Maple

With  $y(t) = 3e^{-t} \cos(2t)$ , we can solve

$$\int 3e^{-t} \cos(2t) dt \quad \text{and} \quad \int_0^5 3e^{-t} \cos(2t) dt$$

These can be integrated by typing

`int(y(t), t);`   `int(y(t), t = 0..5);`   `evalf(%);`

For the indefinite integral, Maple gives:

$$-\frac{3}{5}e^{-t} \cos(2t) + \frac{6}{5}e^{-t} \sin(2t)$$

For the definite integral, Maple gives:

$$\frac{3}{5} - \frac{3}{5}e^{-5} \cos(10) + \frac{6}{5}e^{-5} \sin(10) = 0.59899347$$

# Introduction to Maple

Show  $y(t) = 3e^{-t} \cos(2t)$  is a solution of the differential equation

$$y'' + 2y' + 5y = 0.$$

The function and derivatives are entered by

```
y := t → 3 · exp(-t) · cos(2 · t);
```

```
dy := diff(y(t), t);
```

```
sdy := diff(y(t), t$2);
```

If we type

```
sdy + 2 · dy + 5 · y(t);
```

Maple gives **0**, which verifies this is a solution.

# Introduction to Maple

Maple finds the general solution to the differential equation

```
de := diff(Y(t), t$2) + 2 * diff(Y(t), t) + 5 * Y(t) = 0;
dsolve(de, Y(t));
```

Maple produces

$$Y(t) = C_1 e^{-t} \sin(2t) + C_2 e^{-t} \cos(2t)$$

To solve an initial value problem, say  $Y(0) = 2$  and  $Y'(0) = -1$ , enter

```
dsolve({de, Y(0) = 2, D(Y)(0) = -1}, Y(t));
```

Maple produces

$$Y(t) = \frac{1}{2} e^{-t} \sin(2t) + 2 e^{-t} \cos(2t),$$

which is made into a useable function by typing

```
Y := unapply(rhs(%), t);
```