Math 221
Topics for first exam

Chapter 1: Introduction

Background
A differential equation is an equation involving an (unknown) function \( y \) and some of its derivatives. The basic goal is to solve the equation, i.e., to determine which function or functions satisfy the equation. Differential equations come in several types, and our techniques for solving them will differ depending on the type.

Ordinary vs. partial: If \( y \) is a function of only one variable \( t \), then our differential equation will involve only derivatives w.r.t. \( t \), and we will call the equation an ordinary differential equation. If \( y \) is a function of more than one variable, then our differential equation will involve partial derivatives, and we will call it a partial differential equation. We will deal almost exclusively with ordinary differential equations in this class.

Systems: Sometimes the rates of change of several functions are inter-related, as with the populations of a predator \( y(t) \) and its prey \( x(t) \), where \( x' = ax - ax y \) and \( y' = \gamma xy - cy \). We call this a system of differential equations, and its solution would involve finding both \( x(t) \) and \( y(t) \).

Order: Techniques for solving differential equations differ depending upon how many derivatives of our unknown function are involved. The order of a differential equation is the order of the highest derivative appearing in the equation. The Implicit Function Theorem tells us that we can rewrite our equation so that it equates the highest order derivative with an expression involving lower order terms:

\[
y^{(n)} = F(t, y, y', \ldots, y^{(n-1)})
\]

Linear vs. non-linear: A differential equation is linear if it can be written as

\[
a_0(t)y^{(n)} + \cdots + a_{n-1}(t)y' + a_n(t)y = g(t)
\]

(i.e., the function \( F \) is linear in the variables \( y, y', \ldots, y^{(n-1)} \), although it need not be linear in \( t \)). A differential equation is non-linear if it isn’t linear! E.g.,

\[
y' = y^2
\]

is non-linear, while

\[
y' = (\sin t)y/(1 + t^2) - \cos(\cos t)
\]

is linear.

Solutions and Initial Value Problems

Solving a differential equation means to determine which function or functions satisfy the equation. Our solutions come in two flavors: explicit solutions \( y = y(t) \) which provide a function of \( t \) which satisfies the equation, and implicit solutions which provide an equation \( g(y, t) = 0 \) which any explicit solution would have to satisfy. The idea is that we can treat \( g(y, t) = 0 \) as implicitly defining \( y \) as a function of \( t \); given a specific value \( t = c \) for \( t \), we solve (numerically?) \( g(y, c) = 0 \) for \( y \) to determine the value of the solution to the differential equation at \( c \).

In general, a differential equation \( y' = f(t, y) \) will have many solutions; but typically one particular solution can be specified by requiring one additional condition be met; that \( y \)
take a specific value \( y_0 \) at a specific point \( t_0 \). If we think of the time \( t_0 \) as the time at which we “start” our solution, then we call the pair of equations

\[
y' = f(t, y) \quad y(t_0) = y_0
\]

an initial value problem (or IVP). There is a general result which gives conditions guaranteeing that an IVP has a solution:

If \( y' = f(t, y) \) is a differential equation with both \( f \) and \( \frac{\partial f}{\partial y} \) continuous for \( a < t < b \) and \( \alpha < y < \beta \), and \( t_0 \in (a, b) \) and \( y_0 \in (\alpha, \beta) \), then for some \( h > 0 \), the initial value problem

\[
y' = f(t, y), \quad y(t_0) = y_0
\]

has a unique solution for \( t \in (t_0 - h, t_0 + h) \).

In general, however, the size of the interval where we can guarantee existence (and uniqueness) can be very small, and often depends on the choice of initial value! For example, for the equation

\[
y' = y^2
\]

the righthand side is continuous everywhere (as is the partial derivative), but the interval we can choose for the solutions \( y = -1/(t + c) \) depends on \( c \), which will depend on the initial condition! And it can never be chosen to be the entire real line.

Failure to satisfy the hypotheses of the result can easily kill both existence and uniqueness. For example, the equation

\[
y' = y^{1/3}
\]

has many solutions with the initial condition \( y(0) = 0 \), such as \( y = 0 \) and \( y = (2t/3)^{3/2} \).

**Direction Fields**

In many cases, especially for first order differential equations, we can ‘see’ what a solution should look like without actually finding the solution. For first order equations, \( y' = f(t, y) \), a solution \( y(t) \) will satisfy \( y'(t) = f(t, y(t)) \), and so we can think of \( f(t, y) \) as giving the slope of the tangent line to the graph of \( y(t) \) at the point \( (t, y(t)) \). But since the function \( f \) is already known, we can draw small line segments at ‘every’ point of the \( t-y \) plane with slope \( f(t, y) \) at the point \( (t, y) \); this is called the direction field for our differential equation. A solution to our differential equation is simply a function whose graph is tangent to each of these line segments at every point along the graph. Thinking of the direction field as a velocity vector field (always pointing to the right), our solution is then the path of a particle being pushed along by the velocity vector field. From this point of view it is not hard to believe that every (first order ordinary) differential equation has a solution, in fact many solutions; you just drop a particle in and watch where it goes. Where you drop it is important (it changes where it goes), which really is what gives rise to the notion of an initial value problem; we seek to find the specific solution with the additional initial value \( y(t_0) = y_0 \).

**The Approximation Method of Euler**

Most first order equations cannot be solved by the methods we will present here; the function \( f(y, t) \) is too complicated. For such equations, the best we can often do is to approximate the solutions, using numerical techniques. One method is the tangent line
method, also known as Euler’s method. The idea is that our differential equation \( y' = f(t, y) \) tells us the slope of the tangent line at every point of our solution, and the tangent line can be used to approximate the graph of a function, at least close to the point of tangency. In other words, for a solution to our differential equation,

\[
y(t) \approx y(t_0) + y'(t_0)(t - t_0) = y_0 + f(t_0, y_0)(t - t_0)
\]

for \( t - t_0 \) small. If we wish to approximate \( y(t) \) for a value of \( t \) far away from our initial value \( t_0 \), we use the above idea in several steps. We cut up the interval into \( n \) pieces of length \( h \) (called the stepsize), and then set

\[
\begin{align*}
y_1 &= y_0 + f(t_0, y_0)h, \quad t_1 = t_0 + h \\
y_2 &= y_1 + f(t_1, y_1)h, \quad t_2 = t_1 + h \\
y_3 &= y_2 + f(t_2, y_2)h, \quad t_3 = t_2 + h
\end{align*}
\]

and continue until we reach \( y_n \), which will be our approximation to \( y(t) = y(t_n) \). Each step can be thought of as a mid-course correction, using information about the direction field at each stage to determine which way the solution is tending.

Calculus teaches us that at each stage the error introduced is approximately proportional to the square of \( h \). So with a stepsize half as large, we will require twice as many steps, but each introduces an error only about one-fourth as large, so overall we get an error only half as large. This leads us to conclude that as the stepsize goes to 0, the error between our approximate solution \( y_n \) and \( y(t_n) \) goes to 0.

**Chapter 2: First Order Differential Equations**

**Separable Equations**

There is a class of first order equations for which we can readily find solutions by integration; there are the separable equations. A differential equation is separable if it can be written as

\[
y' = A(t)B(y)
\]

This allows us to ‘separate the variables’ and integrate with respect to \( dy \) and \( dt \) to get a solution:

\[
\frac{1}{B(y)}dy = A(t) \, dt \quad \text{; integrate both sides}
\]

In the end, our solution looks like \( F(y) = G(t) + c \), so it defines \( y \) implicitly as a function of \( t \), rather than explicitly. In some cases we can invert \( F \) to get an explicit solution, but often we cannot.

For example, the separable equation \( y' = ty^2 \), \( y(1) = 2 \) has solution

\[
\int \frac{dy}{y^2} = \int t \, dt + c
\]

so solving the integrals we get \((-1/y) = (t^2/2) + c\), or \( y = -2/(t^2 + 2c) \); setting \( y = 2 \) when \( t = 1 \) gives \( c = -1 \).

**Linear Equations**

Perhaps the most straightforward sort of differential equation to solve is the first order linear ordinary differential equation

\[
y' = a(t)y + b(t)
\]
We will typically (following tradition) write such equations in standard form as
\[ y' + p(t)y = g(t) \] (**)

For example, near the earth and in the presence of air resistance, the velocity \( v \) of a falling object obeys the differential equation \( v' = g - kv \), where \( g \) and \( k \) are (positive) constants.

There is a general technique for solving such equations, by trying to think of the left-hand side of the equation as the derivative of a single function. In form it looks like the derivative of a product, and by introducing an integrating factor \( \mu(t) \), we can actually arrange this. Writing
\[(\mu(t)y)' = \mu(t)(y' + p(t)y)) = \mu(t)g(t)\]
we find that (where \( \exp(\text{blah}) \) means \( e \) raised to the power ‘blah’)
\[ \mu(t) = \exp(\int p(t) \, dt) \]
and so
\[ \mu(t)y = \int \mu(t)g(t) \, dt = \int (\exp(\int p(t) \, dt)) g(t) \, dt + c \]
which we can then solve for \( y \).

Putting this all together, we find that the solutions to (** are given by
\[ y = e^{-\int p(t) \, dt} (\int e^{\int p(t) \, dt} g(t) \, dt + c) \]
For example, the differential equation
\[ ty' - y = t^2 + 1 \]
, after being rewritten in standard form as
\[ y' - (1/t)y = t + (1/t) \]
has homogeneous solution
\[ y_h = \exp(\int 1/t \, dt) = \exp(\ln t) = t \]
so we have
\[ y = t(\int 1 + 1/t^2 \, dt) = t(t - (1/t) + c) \]
and so our solutions are \( y = t^2 - 1 + ct \), where \( c \) is a constant.

But what is \( c \)? Or solution is actually a family of solutions; a particular solution (i.e., a particular value for \( c \)) can be found from an initial value \( y(t_0) = y_0 \). For example, if we wished to solve the initial value problem
\[ ty' - y = t^2 + 1 \ , \ y(2) = 5 \]
we can plug \( t = 2 \) and \( y = 5 \) into our general solution to obtain \( c = 1 \).

**Chapter 3: Mathematical Models**

**Compartmental Analysis**

In many instances, the rate of change of a quantity can be best analysed by treating the factors that make the quantity go up separately from those that make it go down; each can often be easily understood in isolation. We can then build a differential equation modeling the behavior of the quantity \( y = y(t) \) as
\[ y' = \text{(things that make } y \text{ go up)} - \text{(things that make } y \text{ go down)} \]
As a basic example, we have mixing problems. The basic setup has a solution of a known concentration mixing at a known rate with a solution in a vat, while the mixed solution is poured off at a known rate. The problem is to find the function which gives concentration in the vat at time \( t \). It turns out that it is much easier to find a differential equation which describes the amount of solute (e.g., salt) in the solution (e.g., water), rather than the concentration.
If the concentration pouring in is $A$, at a rate of $N$, while the solution is pouring out at rate $M$ with concentration $A(t) = \frac{x(t)}{V(t)}$, then if the initial volume is $V_0$, we can compute $V(t) = V_0 + (N - M)t$. The change in the amount $x(t)$ of solute can be computed as (rate falling in)−(rate falling out), which is

$$x' = AN - A(t)M = AN - \frac{x}{V_0 + (N - M)t}M$$

This is a linear equation, and so we can solve it using our techniques above.

We can also deal with a succession of mixing problems, the output of one becoming the input of the next, by treating them one at a time; the only change in the setup above is that the incoming concentration for the next vat (to solve for $x_{i+1}(t)$) would be the concentration $x_i(t)/V_i(t)$ found by solving the equation for the previous vat.

Another situation where this kind of analysis proves successful is in modeling population growth. The idea is that if $y$ is the population at time $t$, then

$$y' = \text{(birth rate)} - \text{(death rate)}$$

Typically, the birth rate is proportional to the population, i.e. is $ry$, while the death rate is either modeled as being proportional to the population (Malthusian model) or is a sum (logistic model); one part is proportional to the population (death by “natural causes”), the other is proportional to the square of the population (this typically represents contact between individuals, arising from competition for food, overcrowding, etc.), i.e., is $ky^2$. Put together, and combining the two terms proportional to population, we obtain

$$y' = ry$$

for the Malthusian model, and

$$y' = ry - ky^2$$

for the logistic model.

Both equations are separable, and so we can use phase lines to understand their long-term behavior, as well as finding explicit solutions (using partial fractions, for the logistic equation).

**Heating and Cooling**

**Newton’s Law of Cooling:** This states that the rate of change of the temperature $T(t)$ of an object is proportional to the difference between its temperature and the ambient temperature of the air around it. The constant of proportionality depends upon the particular object (and the medium, e.g., air or water) it is in. In other words,

$$T' = k(A - T)$$

Since a cold object will warm up, and a warm object will cool down, this means that the constant $k$ should be positive. Writing the equation as

$$T' + kT = kA$$

we find the solution (after solving the IVP)

$$T(t) = A + (T(0) - A)e^{-kt}$$

Typically, $k$ is not given, but can be determined by knowing the temperature at some other time $t_1$, by plugging into the equation above and solving for $k$.

**Newtonian Mechanics**

If we wish to model the motion of an object, whose position at time $t$ is given by $x(t)$, then (setting $v(t) = x'(t)$) Newton’s Second Law of Motion tells us that
\[ mv' = \text{the sum of the individual forces acting on the object} \]

When we can understand these forces, in terms of \( t \) and \( v \), we can build a first order differential equation, which we can then bring our techniques to bear to solve. Typical forces include:

- gravity: \( F_g = mg \) or \( F_g = -mg \), depending upon whether we think of the positive direction as down (giving +) or up (giving -). \( g = 9.8 \text{ m/sec}^2 = 32 \text{ ft/sec}^2 \) (approximately)
- air resistance: this is typically modeled either as \( F_a = -kv \) (for smallish velocities) or \( F_a = -kv^2 \) (for large velocities). It always acts to push our velocity towards 0, hence the - sign.
- external force: \( F_e = g(t) \); this represents a force that “follows along” the object and tries to push it in a direction that is “pre-programmed” in time.

With these sorts of forces, we get a general equation

\[ mv' = \pm mg - kv + g(t) \]

which we can solve by the methods we have developed. For example, ignoring external forces and assume the positive direction is “down”, we have the initial value problem

\[ mv' = mg - kv \quad v(0) = v_0 \]

with solution

\[ v(t) = \frac{mg}{k} + (v_0 - \frac{mg}{k})e^{-\frac{kt}{m}} \]

As \( t \to \infty \), \( v(t) \to \frac{mg}{k} = \text{the terminal velocity} \).

**Chapter 4: Linear Second Order Equations**

**Linear Differential Operators**

Basic object of study: second order linear differential equations. Standard form:

\[ y'' + p(t)y' + q(t)y = g(t) \quad (*) \]

Initial value problem: we need two initial conditions

\[ y(t_0) = y_0 \quad \text{and} \quad y'(t_0) = y'_0 \]

Basic existence and uniqueness: if \( p(t), q(t), \) and \( g(t) \) are continuous on an interval around \( t_0 \), then any initial value problem has a unique solution on that interval. Our Basic goal: find the solution!

\( (*) \) is called homogeneous if \( g(t) = 0 \); otherwise it is inhomogeneous. \( (*) \) is an equation with constant coefficients if \( p(t) \) and \( q(t) \) are constants.

Our main new technique for exploring these equations will be operator notation. We write \( L[y] = y'' + p(t)y' + q(t)y \) (this is called a linear operator), then a solution to \( (*) \) is a function \( y \) with \( L[y] = g(t) \). Some familiar linear operators: \( D^n[y] = y^{(n)} \) (the \( n \)-th derivative operator). The operator is called linear because

\[ L[cy] = cL[y] \quad \text{and} \quad L[y_1 + y_2] = L[y_1] + L[y_2] \]

For a linear differential equation, \( L[c_1y_1 + c_2y_2] = c_1L[y_1] + c_2L[y_2] \), and so if \( y_1 \) and \( y_2 \) are both solutions to \( L[y] = 0 \) then so is \( c_1y_1 + c_2y_2 \). \( c_1y_1 + c_2y_2 \) is called a linear combination of \( y_1 \) and \( y_2 \). This fact is called the Principle of Superposition: more generally, for a linear operator, if \( L[y_1] = g_1(t) \) and \( L[y_2] = g_2(t) \), then \( L[y_1 + y_2] = g_1(t) + g_2(t) \).
Fundamental Solutions of Homogeneous Equations

Basic idea: with (the right) two solutions \(y_1, y_2\) to a homogeneous linear equation

\[ y'' + p(t)y' + q(t)y = 0 \quad (***) \]

we can solve any initial value problem, by choosing the right linear combination: we need to solve

\[ c_1y_1(t_0) + c_2y_2(t_0) = y_0 \quad \text{and} \quad c_1y_1'(t_0) + c_2y_2'(t_0) = y_0' \]

for the constants \(c_1\) and \(c_2\); then \(y = c_1y_1 + c_2y_2\) is our solution. This we can do directly, as a pair of linear equations, by solving one equation for one of the constants, and plugging into the other equation, or we can use the formulas

\[
 c_1 = \begin{vmatrix} y_0 & y_2(t_0) \\ y_1'(t_0) & y_2'(t_0) \end{vmatrix} \quad \text{and} \quad c_2 = \begin{vmatrix} y_1(t_0) & y_0 \\ y_1'(t_0) & y_2(t_0) \end{vmatrix}
\]

where \(\begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc\). This makes it clear that a solution exists (i.e., we have the ‘right’ pair of functions), provided that the quantity

\[ W = W(y_1, y_2)(t_0) = \begin{vmatrix} y_1(t_0) & y_2(t_0) \\ y_1'(t_0) & y_2'(t_0) \end{vmatrix} \neq 0\]

\(W\) is called the Wronskian (determinant) of \(y_1\) and \(y_2\) at \(t_0\). The Wronskian is closely related to the concept of linear independence of a collection \(y_1, \ldots, y_n\) of functions; such a collection is linearly independent if the only linear combination \(c_1y_1 + \cdots + c_ny_n\) which is equal to the 0 function is the one with \(c_1 = \cdots = c_n = 0\).

Two functions \(y_1\) and \(y_2\) are linearly independent if their Wronskian is non-zero at some point; for a pair of solutions to (***), it turns out that the Wronskian is always equal to a constant multiple of

\[ e^\int p(t) \, dt \]

and so is either always 0 or never 0. We call a pair of linearly independent solutions to (***), a pair of fundamental solutions. By our above discussion, we can solve any initial value problem for (***), as a linear combination of fundamental solutions \(y_1\) and \(y_2\). By our existence and uniqueness result, this gives us:

If \(y_1\) and \(y_2\) are a fundamental set of solutions to the differential equation (***), then any solution to (***), can be expressed as a linear combination \(c_1y_1 + c_2y_2\) of \(y_1\) and \(y_2\).