## Boundary Value Problems

A second order boundary value problem on a closed interval $a \leq x \leq b$ is a differential equation that takes the form

$$
\begin{gathered}
y^{\prime \prime}=f\left(x, y, y^{\prime}\right) \\
y(a)=\alpha \\
y(b)=\beta
\end{gathered}
$$

Given the similarity between this problem and the second order initial value problem one would think that there is not much new here. To a certain extent this is true: as we will see below we can use techniques for solving initial value problems to attack this problem. However, these techniques are not simple.

## The linear case

An important special case of the problem we are studying here is the second order linear boundary value problem. In this version the function $f$ takes a special form, which makes it easier to deal with the problem of matching the boundary conditions.

$$
y^{\prime \prime}=p(x) y^{\prime}+q(x) y+r(x)
$$

What makes this form easy to work with is the fact that the right hand side is linear in the function $y(x)$. Suppose that $y_{1}(x)$ is a solution of

$$
y^{\prime \prime}=p(x) y^{\prime}+q(x) y+r(x)
$$

and $y_{2}(x)$ is a solution of

$$
y^{\prime \prime}=p(x) y^{\prime}+q(x) y
$$

Any linear combination

$$
y(x)=c_{1} y_{1}(x)+c_{2} y_{2}(x)
$$

is also a solution of the original equation:

$$
\begin{gathered}
y^{\prime \prime}=c_{1} y_{1}{ }^{\prime \prime}+c_{2} y_{2}{ }^{\prime \prime} \\
=c_{1}\left(p(x) y_{1}{ }^{\prime}+q(x) y_{1}+r(x)\right)+c_{2}\left(p(x) y_{2}{ }^{\prime}+q(x) y_{2}\right) \\
=p(x)\left(c_{1} y_{1}(x)+c_{2} y_{2}(x)\right)^{\prime}+q(x)\left(c_{1} y_{1}(x)+c_{2} y_{2}(x)\right)+r(x) \\
=p(x) y^{\prime}+q(x) y+r(x)
\end{gathered}
$$

We can take advantage of this property of the linear problem to solve the boundary value problem. Our approach
is to find a function $y_{1}(x)$ that solves the initial value problem

$$
\begin{gathered}
y^{\prime \prime}=p(x) y^{\prime}+q(x) y+r(x) \\
y(a)=\alpha \\
y^{\prime}(a)=0
\end{gathered}
$$

This is a conventional initial value problem, and can be solved by any of the methods from chapter 5 .
We then solve a second initial value problem. $y_{2}(x)$ is the solution to

$$
\begin{gathered}
y^{\prime \prime}=p(x) y^{\prime}+q(x) y \\
y(a)=0 \\
y^{\prime}(a)=1
\end{gathered}
$$

From what we saw above,

$$
y(x)=c_{1} y_{1}(x)+c_{2} y_{2}(x)
$$

Solves the differential equation. Note also that

$$
y(a)=c_{1} \alpha+c_{2} 0=c_{1} \alpha
$$

This tells us that we must pick $c_{1}=1$. To match the boundary condition at the point $x=b$ we require that

$$
y(b)=y_{1}(b)+c_{2} y_{2}(b)=\beta
$$

This is equivalent to requiring that

$$
c_{2}=\frac{\beta-y_{1}(b)}{y_{2}(b)}
$$

Thus,

$$
y(x)=y_{1}(x)+\frac{\beta-y_{1}(b)}{y_{2}(b)} y_{2}(x)
$$

solves the differential equation and matches the boundary conditions.

## The nonlinear case

The technique we saw above works only in the case of a linear equation, because it makes essential use of the fact that a linear combination of solutions is also a solution. For the more general, nonlinear $f\left(x, y, y^{\prime}\right)$ this property no longer holds.

Even so, we will want to develop a method that uses the solution of an initial value problem to solve the boundary
value problem.
Here is a simple idea that will do this. Let $y(x)$ be the solution to the initial value problem

$$
\begin{gathered}
y^{\prime \prime}=f\left(x, y, y^{\prime}\right) \\
y(a)=\alpha \\
y^{\prime}(a)=t
\end{gathered}
$$

If we can find a value of $t$ that causes $y(b)=\beta$ we will have solved our problem.
To emphasize the dependence of $y(x)$ on our choice of $t$, we will write $y(x, t)$.
The problem we have to solve now is finding a value of $t$ such that

$$
y(b, t)=\beta
$$

This idea is called the shooting method, because we are shooting a function with a particular slope at $x=a$ and trying to hit a target value at $x=b$. The shooting method boils down to a root-finding problem. We seek a value of $t$ such that

$$
v(t)=y(b, t)-\beta
$$

has a root. The only complication here is that $v(t)$ is a very unwieldy function to work with. Consider what is required to evaluate $v(t)$ for some choice of $t$ :

1. Specify your choice for $t$.
2. Use a numerical solution technique to solve the initial value problem

$$
\begin{gathered}
y^{\prime \prime}=f\left(x, y, y^{\prime}\right) \\
y(a)=\alpha \\
y^{\prime}(a)=t
\end{gathered}
$$

3. Evaluate that numerical solution at $b$ to get $v(t)=y(b, t)-\beta$.

This makes for a very expensive and cumbersome function evaluation!
Given the practical difficulty of evaluating $v(t)$ we will have to use a root finding technique that is easy to use. One possibility is to use the secant method. In this method we pick two values of $t, t_{1}$ and $t_{2}$, and construct the secant line passing through the points $\left(t_{1}, v\left(t_{1}\right)\right)$ and $\left(t_{2}, v\left(t_{2}\right)\right)$ and then determine the $t_{3}$ where that secant line crosses the axis. We then repeat the process with $t_{2}$ and $t_{3}$ to get a $t_{4}$, and so on. This will produce a sequence of approximate values for $t$ that should converge to the root.

## Using Newton's method

Can we use Newton's method to find a root for $v(t)$ ? At first glance this seems impossible, because the function $v(t$ ) is a ridiculously complicated function. How will we ever compute $v^{\prime}(t)$ to apply the Newton iteration formula?

$$
t_{k+1}=t_{k}-\frac{v\left(t_{k}\right)}{v^{\prime}\left(t_{k}\right)}
$$

Amazingly, it is actually possible to compute $v^{\prime}\left(t_{k}\right)$. The trick is to notice that $v(t)$ comes from the solution of a differential equation, and that we can differentiate that equation with respect to the parameter $t$ :

$$
\frac{\partial y^{\prime \prime}(x, t)}{\partial t}=\frac{\partial f\left(x, y(x, t), y^{\prime}(x, t)\right)}{\partial t}=\frac{\partial f}{\partial x} \frac{\partial x}{\partial t}+\frac{\partial f}{\partial y} \frac{\partial y(x, t)}{\partial t}+\frac{\partial f}{\partial y^{\prime}} \frac{\partial y^{\prime}(x, t)}{\partial t}
$$

Two facts help us to rewrite this. The first is that the variables $x$ and $t$ are independent of each other, so that

$$
\frac{\partial x}{\partial t}=0
$$

The second fact is that

$$
\frac{\partial y^{\prime}(x, t)}{\partial t}=\frac{\partial}{\partial t} \frac{\partial}{\partial x} y(x, t)=\frac{\partial}{\partial x} \frac{\partial y(x, t)}{\partial t}
$$

and also

$$
\frac{\partial y^{\prime \prime}(x, t)}{\partial t}=\frac{\partial^{2}}{\partial x^{2}} \frac{\partial y(x, t)}{\partial t}
$$

These facts transform the equation above into

$$
\frac{\partial^{2}}{\partial x^{2}} \frac{\partial y(x, t)}{\partial t}=\frac{\partial f}{\partial y} \frac{\partial y(x, t)}{\partial t}+\frac{\partial f}{\partial y^{\prime}}\left(\frac{\partial}{\partial x} \frac{\partial y(x, t)}{\partial t}\right)
$$

If we introduce

$$
z(x, t)=\frac{\partial y(x, t)}{\partial t}
$$

This becomes an equation

$$
z^{\prime \prime}(x, t)=\frac{\partial f}{\partial y} z(x, t)+\frac{\partial f}{\partial y^{\prime}} z^{\prime}(x, t)
$$

The original initial conditions

$$
\begin{aligned}
& y(a)=\alpha \\
& y^{\prime}(a)=t
\end{aligned}
$$

become

$$
z(a, t)=0
$$

$$
z^{\prime}(a, t)=1
$$

What we have now is a coupled system of two equations in two unknowns, $y(x, t)$ and $z(x, t)$ :

$$
\begin{gathered}
y^{\prime \prime}(x, t)=f\left(t, y(x, t), y^{\prime}(x, t)\right) \\
y(a, t)=\alpha \\
y^{\prime}(a, t)=t \\
z^{\prime \prime}(x, t)=\frac{\partial f}{\partial y} z(x, t)+\frac{\partial f}{\partial y^{\prime}} z^{\prime}(x, t) \\
z(a, t)=0 \\
z^{\prime}(a, t)=1
\end{gathered}
$$

The equations are coupled, because $y$ and $y^{\prime}$ terms will still appear in the equation for $z$. This mess can be handled by techniques from chapter 5 . Specifically, we can convert this into a first order system of four equations in four unknowns. Solving this system will allow us to compute value for $y(b, t)$ and $z(b, t)$.

Finally, note that

$$
v^{\prime}(t)=\frac{\partial}{\partial t}(y(b, t)-\beta)=z(b, t)
$$

We now have everything we need to apply Newton's method to the original problem of finding a root of $v(t)$.

1. Given a $t_{k}$, we start by solving the system

$$
\begin{gathered}
y^{\prime \prime}\left(x, t_{k}\right)=f\left(x, y\left(x, t_{k}\right), y^{\prime}\left(x, t_{k}\right)\right) \\
z^{\prime \prime}\left(x, t_{k}\right)=\frac{\partial f}{\partial y} z\left(x, t_{k}\right)+\frac{\partial f}{\partial y^{\prime}} z^{\prime}\left(x, t_{k}\right) \\
y\left(a, t_{k}\right)=\alpha \\
y^{\prime}\left(a, t_{k}\right)=t_{k} \\
z\left(a, t_{k}\right)=0 \\
z^{\prime}\left(a, t_{k}\right)=1
\end{gathered}
$$

2. We use our numerical solution technique to estimate $y\left(b, t_{k}\right)$ and $z\left(b, t_{k}\right)$.
3. If $\left|y\left(b, t_{k}\right)-\beta\right|$ is small enough, we stop.
4. Otherwise, we compute the next $t$ and repeat:

$$
t_{k+1}=t_{k}-\frac{v\left(t_{k}\right)}{v^{\prime}\left(t_{k}\right)}=t_{k}-\frac{y\left(b, t_{k}\right)-\beta}{z\left(b, t_{k}\right)}
$$

