

# Lecture 36: Boundary value problems

We want to adapt the approach we introduced last lecture to solving *boundary value problems*. Consider the following simple example:

$$\begin{aligned}u(0) &= a \\ u(1) &= b \\ u''(x) - a(x)u(x) &= f(x)\end{aligned}$$

Here is an example solution:

In [18]:

```
using ApproxFun

B=dirichlet()
x=Fun([0.,1.])
u=[B;Derivative([0.,1.])^2+1000x^2]\[1.,2.]
ApproxFun.plot(u)
```

Out[18]:

```
Fun([-1.56831,1.42181,0.396509,-2.73726,2.49329,0.521205,-1.24882,
1.76698,2.93991,0.328768 ... 5.95226e-16,-8.27141e-15,-2.50338e-
15,-2.2721e-16,8.4236e-17,3.75925e-17,5.91565e-18,-4.23325e-19,-4
.3985e-19,-9.91228e-20],Chebyshev( [0.0,1.0] ))
```

Unlike initial value problems, where the conditions  $u(0) = a$  and  $u'(0)$  are specified at a single point, in boundary value problems the conditions are specified at two *different* points. This means we can't view their solution as "time-stepping": we have to solve the problem globally. We will do so by using the approach advocated last lecture of constructing discrete derivatives.

## Discrete Second Derivative

Recall the midpoint discrete derivative

$$D_n : \begin{matrix} \text{Values at} \\ x_0, \dots, x_n \end{matrix} \rightarrow \begin{matrix} \text{Values at} \\ x_{1/2}, \dots, x_{n-1/2} \end{matrix}$$

which is an  $n \times n + 1$  matrix with entries

$$D_n \triangleq \frac{1}{h} \begin{pmatrix} -1 & 1 & & & \\ & -1 & 1 & & \\ & & \ddots & \ddots & \\ & & & -1 & 1 \end{pmatrix}$$

where  $h = 1/n$  and  $x_k = kh$ . We will construct an approximate second derivative by now creating another midpoint discrete derivative

$$D_{n-1} : \begin{array}{c} \text{Values at} \\ x_{1/2}, \dots, x_{n-1/2} \end{array} \rightarrow \begin{array}{c} \text{Values at} \\ x_1, \dots, x_{n-1} \end{array},$$

which is  $(n-1) \times n$ .

Because the spacing between the nodes is still  $h = 1/n$ , when we approximate data at  $x_{1/2}, \dots, x_{n-1/2}$  by trapezoids, differentiate, and evaluate at the grid  $x_1, \dots, x_{n-1}$  we get the entries:

$$D_{n-1} \triangleq \frac{1}{h} \begin{pmatrix} -1 & 1 & & & \\ & -1 & 1 & & \\ & & \ddots & \ddots & \\ & & & -1 & 1 \end{pmatrix}$$

where  $h$  is still  $1/n$ .

The  $(n-1) \times (n+1)$  discrete second derivative is then specified by

$$D_n^2 \triangleq D_{n-1} D_n.$$

This satisfies

$$D_n^2 : \begin{array}{c} \text{Values at} \\ x_0, \dots, x_n \end{array} \rightarrow \begin{array}{c} \text{Values at} \\ x_1, \dots, x_{n-1} \end{array},$$

that is, we map from all the nodes to the interior nodes. The entries are given by matrix multiplication as

$$D_n^2 \triangleq \frac{1}{h^2} \begin{pmatrix} 1 & -2 & 1 & & & \\ & 1 & -2 & 1 & & \\ & & \ddots & \ddots & \ddots & \\ & & & 1 & -2 & -1 \end{pmatrix}$$

**Remark** Matrices with constant diagonals are called *Toeplitz matrices*. They have been studied extensively, with a special emphasis on studying the eigenvalues and their behaviour as the dimension tends to infinity.

We can construct this matrix as D2 here:

In [1]:

```
# discrete first derivative
function D(h,n)
    ret=zeros(n,n+1)
    for k=1:n
        ret[k,k]=-1/h
        ret[k,k+1]=1/h
    end
    ret
end

# discrete second derivative
D2(h,n) = D(h,n-1)*D(h,n)
```

Out[1]:

D2 (generic function with 1 method)

We verify that  $D2 \cdot f(x)$  gives an approximation of the second derivative at the interior nodes:

In [2]:

```
n=10
h=1/n

f=x->cos(x)
fpp=x->-cos(x)    # second derivative of f

x=linspace(0.,1.,n+1) # domain nodes
r=x[2:end-1]          # range nodes

norm(D2(h,n)*f(x) - fpp(r),Inf)
```

Out[2]:

0.0008288937970136745

**Exercise** Estimate the rate of convergence by finding  $\alpha$  so that the error decays like  $Cn^\alpha$ .

# Multiplication operator

We now set up the multiplication operator correspo representing multiplication by  $a(x)$ . This is the  $n - 1 \times n + 1$  matrix

$$A_n : \begin{matrix} \text{Values at} \\ x_0, \dots, x_n \end{matrix} \rightarrow \begin{matrix} \text{Values at} \\ x_1, \dots, x_{n-1} \end{matrix}$$

with entries given by

$$A_n \triangleq \begin{pmatrix} 0 & a(x_1) & & & \\ & & a(x_2) & & \\ & & & \ddots & \\ & & & & a(x_{n-1}) & 0 \end{pmatrix}.$$

We can set this up as follows:

In [3]:

```
function A(a::Function,h,n)
    ret=zeros(n-1,n+1)
    for k=1:n-1
        ret[k,k+1]=a(k*h)
    end
    ret
end
```

Out[3]:

A (generic function with 1 method)

In this case, the operator is in fact, exact:

In [4]:

```
a=x->sin(x)

A(a,h,n)
norm(A(a,h,n)*f(x) - a(r).*f(r),Inf)
```

Out[4]:

1.1102230246251565e-16

So the operator  $L = D^2 - a(x)$  is discretized as

$$L_n = D_n^2 - A_n$$

which is a map

$$L_n : \begin{matrix} \text{Values at} \\ x_0, \dots, x_n \end{matrix} \rightarrow \begin{matrix} \text{Values at} \\ x_1, \dots, x_{n-1} \end{matrix}$$

In [6]:

```
L=D2(h,n) - A(a,h,n)
```

Out[6]:

```
9x11 Array{Float64,2}:
 100.0  -200.1   100.0      0.0  ...    0.0      0.0      0.0
    0.0
  0.0   100.0  -200.199   100.0      0.0      0.0      0.0
    0.0
  0.0      0.0   100.0   -200.296      0.0      0.0      0.0
    0.0
  0.0      0.0      0.0   100.0      0.0      0.0      0.0
    0.0
  0.0      0.0      0.0      0.0      0.0      0.0      0.0
    0.0
  0.0      0.0      0.0      0.0  ...   100.0      0.0      0.0
    0.0
  0.0      0.0      0.0      0.0   -200.644   100.0      0.0
    0.0
  0.0      0.0      0.0      0.0   100.0   -200.717   100.0
    0.0
  0.0      0.0      0.0      0.0      0.0   100.0   -200.7
83  100.0
```

## Boundary conditions

We need to represent  $u(0)$  and  $u(1)$  where  $u$  is given at the grid  $x_0, \dots, x_n$ . We see that this is accomplished via the  $1 \times n + 1$  row vectors

$$B_n^0 \triangleq [1, 0, \dots, 0]$$

In [7]:

```
B0 = [1 zeros(1,n)]

B0*f(x) - f(0)
```

Out[7]:

```
1-element Array{Float64,1}:
 0.0
```

and

$$B_n^1 \triangleq [0, 0, \dots, 1]$$

In [8]:

```
B1 = [zeros(1,n) 1]
B1*f(x) - f(1)
```

Out[8]:

```
1-element Array{Float64,1}:
 0.0
```

## Constructing the discrete boundary value problem

We now discretize the *operator*

$$Mu = \begin{pmatrix} u(0) \\ u'' - a(x)u \\ u(1) \end{pmatrix}.$$

by

$$M_n = \begin{pmatrix} B_n^0 \\ D_n^2 - A_n \\ B_n^1 \end{pmatrix}$$

We put the boundary conditions at the top and bottom so that  $M_n$  is tridiagonal (that is, only has three non-zero bands).

In [9]:

```
L=D2(h,n) - A(a,h,n)
M=[B0;
    L;
    B1]
```

Out[9]:

```
11x11 Array{Float64,2}:
 1.0  0.0  0.0  0.0  ...  0.0  0.0  0.0
 0.0
100.0 -200.1 100.0  0.0  0.0  0.0  0.0
 0.0
 0.0 100.0 -200.199 100.0  0.0  0.0  0.0
 0.0
 0.0  0.0 100.0 -200.296  0.0  0.0  0.0
 0.0
 0.0  0.0  0.0 100.0  0.0  0.0  0.0
 0.0
 0.0  0.0  0.0  0.0  ...  0.0  0.0  0.0
 0.0
 0.0  0.0  0.0  0.0 100.0  0.0  0.0
 0.0
 0.0  0.0  0.0  0.0 -200.644 100.0  0.0
 0.0
 0.0  0.0  0.0  0.0 100.0 -200.717 100.0
 0.0
 0.0  0.0  0.0  0.0  0.0 100.0 -200.7
83 100.0
 0.0  0.0  0.0  0.0  ...  0.0  0.0  0.0
 1.0
```

We test that it approximates  $M$ :

In [10]:

```
norm(M*f(x) - [f(0.); fpp(r)-a(r).*f(r); f(1.)],Inf)
```

Out[10]:

```
0.0008288937970233334
```

**Exercise** Estimate the rate of convergence.

We now approximate the boundary value problem

$$Mu = \begin{pmatrix} a \\ f(x) \\ b \end{pmatrix}$$

by solving the discretized problem

$$M_n \mathbf{w} = \begin{pmatrix} a \\ f(\mathbf{x}[2 : end - 1]) \\ b \end{pmatrix}$$

to find  $w_k = \mathbf{e}_k^\top \mathbf{w}$  that approximates  $u(x_k)$ .

Here we solve the same equation as at the top:



In [13]:

```
using PyPlot
```

```
n=100
```

```
h=1/n
```

```
a=x->-1000x^2
```

```
x=linspace(0.,1.,n+1)
```

```
B0 = [1 zeros(1,n)]
```

```
B1 = [zeros(1,n) 1]
```

```
L=D2(h,n) - A(a,h,n)
```

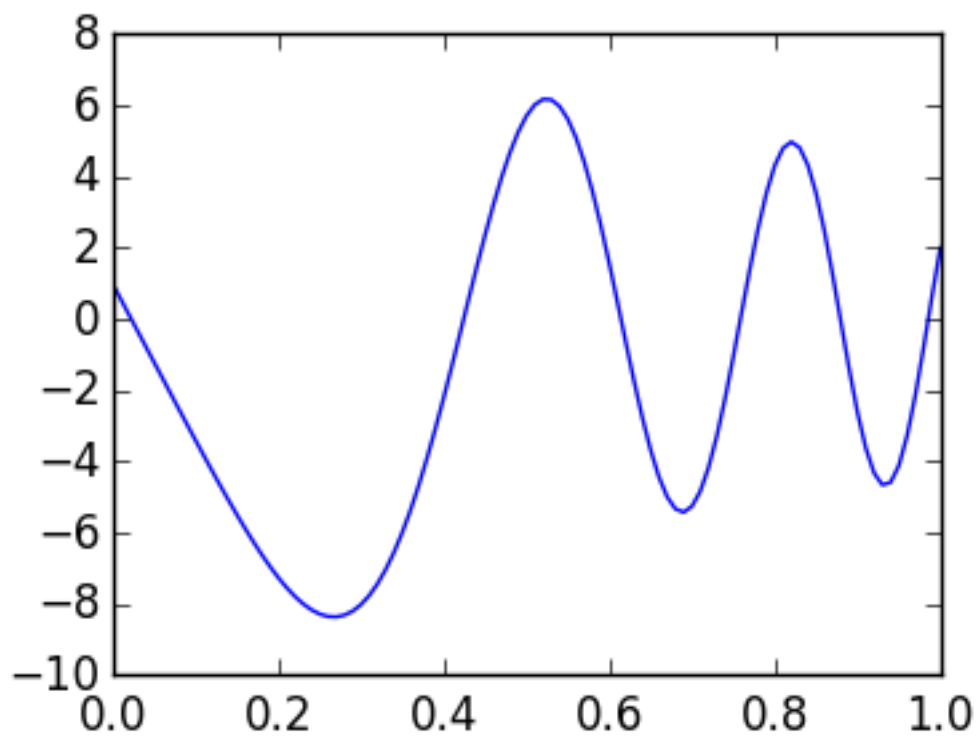
```
M=[B0;
```

```
    L;
```

```
    B1]
```

```
w=M\[1.;zeros(n-1);2.]
```

```
plot(x,w);
```



We observe empirically that the method converges:

In [19]:

```
n=4000
h=1/n

a=x->-1000x^2

x=linspace(0.,1.,n+1)

B0 = [1 zeros(1,n)]
B1 = [zeros(1,n) 1]

L=D2(h,n) - A(a,h,n)
M=[B0;
    L;
    B1]

w=M\[1.;zeros(n-1);2.]

norm(w-u(x),Inf)  # the exact solution u was calulated above
```

Out[19]:

0.0003873070855124894

**Exercise** What property of  $M_n$  guarantees that we converge to the solution  $u$  as fast as

$$\|M_n f(\mathbf{x}) - (Mf)(\mathbf{x}[2 : end - 1])\|_\infty$$

converges?