# Math 6630: Numerical Solutions of Partial Differential Equations Polynomial spectral methods, II <br> See Shen, Tang, and Wang 2011, Chapter 4, <br> Hesthaven, S. Gottlieb, and D. Gottlieb 2007, Chapters 7, 8 

Akil Narayan ${ }^{1}$<br>${ }^{1}$ Department of Mathematics, and Scientific Computing and Imaging (SCI) Institute University of Utah

April 12, 2023

## Non-periodic problems

We have considered PDE's with periodic boundary conditions using Fourier Series.
Our main goal now is to modify our procedures appropriately for nonperiodic problems, e.g., ones of the form,

$$
-u^{\prime \prime}(x)=f(x), \quad u(-1)=u(1)=0, \quad \chi \in[-|,|]
$$

Our main technique to accomplish will be to approximate the solution as a polynomial.
Orthogonal polynomials are the main computational tool we'll employ.

- For general weight functions $\omega$ :
- three-term recurrence
- Gaussian quadrature through recurrence coefficients
- For "classical" weight functions $\omega$ :
- Sturm-Liouville operator $\rightarrow$ Best $L_{\omega}^{2}$ convergence rates
- Rodrigues' formula $\rightarrow$ explicit formulas for recurrence coefficients
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- Classical weight functions yield Legendre, Hermite, Chebyshev, Gegenbauer, Jacobi, and Laguerre polynomial families
Now we can address solving differential equations.


## The high-level ideas

The particular tools for solving differential equations will not change.
In particular, we'll still use the weighted residual methods strategy:

- For (linear) stationary problems:
- We introduce an appropriate bilinear form that yields a well-posed weak formulation for the PDE
- We discretize the bilinear form via Galerkin or collocation methods (or Petrov-Galerkin methods)


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Of particular note: all the theory we've covered applies quite directly, including Ceá's
Lemma, Lax-Milgram, and stability for time-dependent problems.

## Overall types of methods

We will use either Galerkin or collocation methods. Both of these approaches require defining a weight function $\omega(x)$.

Bilinear forms will be defined using the $L_{\omega}^{2}$ inner product, and so our trial/test spaces will be spanned by $L_{\omega}^{2}$-orthonormal polynomials $p_{0}, p_{1}, \ldots$.

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We will describe methods with the Galerkin/collocation label along with the name of the basis corresponding to $\omega$. E.g., :

- A Legendre-Galerkin method is a Galerkin approach where $\omega(x)=1$, and we will use Legendre polynomials as a basis for the trial/test space.
- A Chebyshev-collocation method is a collocation approach where $\omega(x)=\left(1-x^{2}\right)^{-1 / 2}$, and we will use Chebysheypolynomials as a basis.


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The main new (and nontrivial) issue we must contend with are boundary conditions.


## Three approaches for boundary conditions

For concreteness, we'll consider the following ODE as an example:

$$
-u^{\prime \prime}(x)=f(x)
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$$
u(-1)=u(1)=0
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However, the procedures we describe next for handling boundary conditions apply to essentially any type of PDE.

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- "Natural" handling of homogeneous boundary conditions: the weak form is constructed to weakly enforce boundary conditions.


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- "Natural" handling of homogeneous boundary conditions: the weak form is constructed to weakly enforce boundary conditions.
- "Tau" handling of boundary conditions: trailing (high-frequency) cofficients are altered to satisfy the boundary conditions
We'll mainly describe nonhomogeneous boundary conditions and the essential boundary conditions strategy.


## "Lifting" for non-homogeneous boundary conditions

Linearity is really nice.
Consider solving,

$$
-u^{\prime \prime}(x)=f(x), \quad u(-1)=3, \quad u(1)=-2,
$$

One strategy to address such boundary conditions is to realize that

$$
w(x)=-2 \frac{1+x}{2}+3 \frac{1-x}{2}=-\frac{5}{2} x+\frac{1}{2},
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With this information, we can define

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U(x):=u(x)-w(x)
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One we solve this equation with homogeneous boundary conditions for $U$, then we simply define

$$
u(x)=U(x)+w(x) .
$$

## "Essential" treatment of boundary conditions

Because of the previous result we can, for many cases, specialize to homogeneous boundary conditions,

$$
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A standard Galerkin approach would utilize the trial/test space,

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P_{N}=\operatorname{span}\left\{x^{j} \mid j=0, \ldots, N\right\},
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but $u \in P_{N}$ need not satisfy the boundary conditions.
One way to address boundary conditions is to define a space so that functions must be from this space:

$$
P_{N, 0}=\left\{p \in P_{N} \mid p(-1)=0, \quad p(1)=0\right\} .
$$

The dimension of this space is $\operatorname{dim} P_{N, 0}=\operatorname{dim} P_{N}-2=N-1$.
Building the boundary conditions explicitly into the approximation space is called an essential treatment of boundary conditions.

## Essential boundary conditions

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There are two popular computational ways to realize $P_{N, 0}$. The first defines,

$$
q_{n}(x):=p_{n}(x)-p_{n}(-1) \frac{1-x}{2}-p_{n}(1) \frac{1+x}{2}, \quad n \geqslant 2
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where $p_{n}$ are Legendre polynomials. Thus, $q_{n}( \pm 1)=0$, and we take,

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The second approach is similar, using instead high-degree Legendre polynomials,

$$
r_{n}(x):=p_{n}(x)-\gamma_{n} p_{n-1}(x)-\delta_{n} p_{n-2}(x), \quad n \geqslant 2,
$$

where $\gamma_{n}$ and $\delta_{n}$ are chosen to satisfy,

$$
r_{n}( \pm 1)=0 \Longrightarrow\left(\begin{array}{cc}
p_{n-1}(-1) & p_{n-2}(-1) \\
p_{n-1}(1) & \left.p_{n-2}(2)\right)
\end{array}\right)\binom{\gamma_{n}}{\delta_{n}}=\binom{p_{n}(-1)}{p_{n}(1)} .
$$

Finally,

$$
P_{N, 0}=\operatorname{span}\left\{r_{n} \mid n=2, \ldots, N\right\}
$$

There ore other options" e.g.

$$
\begin{aligned}
& S_{n}(x)=\left[1-x^{2}\right] p_{n}(x), \quad n=0,1,2 \ldots N-2 \\
& \text { span }\left\{s_{n}\right\}_{n=0}^{N-2} C P_{N}
\end{aligned}
$$

## A simple example

Let's construct a Legendre-Galerkin method for,

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-u^{\prime \prime}(x)+u(x)=f(x), \quad u(-1)=u(1)=0
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To identify an appropriate bilinear form, we consider

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H_{0}^{1}([-1,1]):=\left\{u \in H^{1}([-1,1]) \mid u( \pm 1)=0\right\} .
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With $\langle\cdot, \cdot\rangle$ the standard $L^{2}$ inner product on $[-1,1]$ and taking $u, v \in H_{0}^{1}([-1,1])$, we have

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\left\langle-u^{\prime \prime}, v\right\rangle+\langle u, v\rangle=\langle f, v\rangle \quad \Longrightarrow \quad\left\langle u^{\prime}, v^{\prime}\right\rangle+\langle u, v\rangle=\langle f, v\rangle .
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Then the (infinite-dimensional) weak form is,
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To discretize things, we'll use the trial/test space $P_{N, 0}$ with basis $q_{n}$. Then our scheme is Find $u \in P_{N, 0}$ such that $\left\langle u^{\prime}, v^{\prime}\right\rangle+\langle u, v\rangle=\langle f, v\rangle$ for all $v \in P_{N, 0}$

## Implementation

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Find $u \in P_{N, 0}$ such that $\left\langle u^{\prime}, v^{\prime}\right\rangle=\langle f, v\rangle$ for all $v \in P_{N, 0}$
This scheme can be implemented as follows:

$$
u(x)=\sum_{j=2}^{N} \widehat{u}_{j} q_{j}(x)
$$

and setting $v=q_{n}$ for each $n=2, \ldots, N$, we obtain a linear system for the coefficients $\boldsymbol{u}=\left(\widehat{u}_{j}\right)_{j=2}^{N}$ :

$$
(S+M) u=f
$$

$$
f=\left(f_{n}\right)_{n=2}^{N}, \quad f_{n}=\left\langle f, q_{n}\right\rangle
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where the stiffness and mass matrices have entries, respectively,

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(S)_{n, j}=\left\langle q_{n}^{\prime}, q_{j}^{\prime}\right\rangle, \quad(M)_{n, j}=\left\langle q_{n}, q_{j}\right\rangle, \quad n, j=2, \ldots, N
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There are two main options to compute these matrices:

- Analytical methods - use the three-term recurrence and some related properties to compute exact expressions.
- Computational methods - use Legendre-Gauss quadrature.
- The integrand is a polynomial of degree at most $2 N$, so an $N+1$-point rule is sufficient.
- To evaluate $q_{n}^{\prime}$, we need to evaluate $p_{n}^{\prime} \longrightarrow$ differentiate the three-term recurrence to obtain a modified recurrence relation that can evaluate $p_{n}^{\prime}(x)$.


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In implementation, choosing $q_{n}$ versus $r_{n}$ as a basis does make a difference. E.g., choosing $r_{n}$ in this case makes $S$ diagonal.

[^0]
## Beyond essential boundary conditions

The previous approach works similarly for time-dependent problems.
We'll briefly describe alternative approaches. For things like Robin/mixed boundary conditions, e.g.,

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\alpha u(1)+\beta u^{\prime}(1)=0, \quad \alpha, \beta \neq 0,
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then it's more appropriate to "naturally" enforce boundary conditions. In particular if $u, v$ satisfy this condition, then via integration by parts:

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\begin{aligned}
\left\langle-u^{\prime \prime}, v\right\rangle & =\left\langle u^{\prime}, v^{\prime}\right\rangle-\left.u^{\prime}(x) v(x)\right|_{-1} ^{1} \\
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Note that this bilinear form has constant (boundary) terms, but is still linear in both $u$ and $v$.

An appropriate trial/test space would not enforce any conditions essentially at $x=+1$.
Instead, the scheme would enforce the condition through the bilinear form.
Such a treatment of boundary terms is a natural treatment of boundary conditions.

## "Tau" methods

Another approach to handling boundary conditions is through a "tau" approach.
The main idea: strictly enforce a boundary condition in the scheme, but remove one weak Galerkin condition.
E.g., for our favorite problem,

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-u^{\prime \prime}(x)=f(x), \quad u( \pm 1)=0,
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then $a(n$ unrecommended) tau scheme would be,

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\text { Find } u \in P_{N} \text { such that }\left\langle u^{\prime}, v^{\prime}\right\rangle+\langle u, v\rangle=\langle f, v\rangle \text { for all } v \in P_{N-2},
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One generally should not use tau methods if there are better approaches (e.g., essential handling of boundary conditions).

However, tau methods are very useful for time-dependent problems with time-depedent boundary conditions, e.g., $u(1, t)=h(t)$.

## Collocation methods

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Unfortunately, there are never any Gauss quadrature points lying on $\pm 1$.
One can instead sacrifice quadrature accuracy for the freedom to place nodes at $\pm 1$. This results in the Gauss-Lobatto rule. In particular, this quadrature rule satisfies,

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\int_{-1}^{1} p(x) \omega(x) \mathrm{d} x=\sum_{j=1}^{N} w_{j} p\left(x_{j}\right), \quad p \in P_{2 N-3}, x_{1} \quad=-1, \not c_{N} \quad=1
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\int_{-1}^{1} p(x) \omega(x) \mathrm{d} x=\sum_{j=1}^{N} w_{j} p\left(x_{j}\right), \quad p \in P_{2 N-3, x_{1}} \quad=-1, x_{N} \quad=1 .
$$

If we use the Legendre-Gauss-Lobatto points $\left\{x_{j}\right\}_{j=1}^{N}$ for collocation, along with the Lagrange polynomial representation,

$$
u(x)=\sum_{j=1}^{N} u_{j} \ell_{j}(x), \quad \quad \ell_{j}\left(x_{k}\right)=\delta_{j, k}
$$

then the scheme:

- Enforces zero PDE residual at $x_{j}$ for $j=2, \ldots, N-1$
- Sets $u_{1}=0$, and $u_{N}=0$.

The inverse transform (back to orthogonal polynomial coefficient space) on the Gauss-Lobatto points is not unitary, but it's well-conditioned.

## References I



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[^0]:    A. Narayan (U. Utah - Math/SCI)

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