AMCS 394E: Contemp. Topics in Computational Science. Computing with the finite element method

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For example:

The solution u(x) of

$$-\Delta u = f(x), \quad \forall x \in (0,1)$$

describes some distribution (e.g., of heat) at every point between (0, 1).

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* Represent the solution in finitely many points.

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Alternatives:

- * Represent the solution in finitely many points.
- * Use finite information to approximate the solution.

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- * how do I obtain the solution?

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Global approximations:

For example via Taylor series:

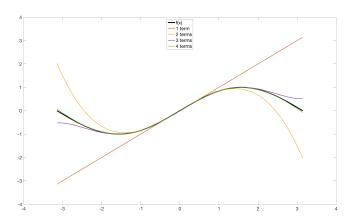
$$f(x) = \sum_{n=0}^{\infty} \frac{f^{(n)}(x_0)}{n!} (x - x_0)^n$$
$$\approx \sum_{n=0}^{N} \alpha_n (x - x_0)^n$$

Exercise:

Consider f(x) = sin(x) and approximate it via Taylor series.

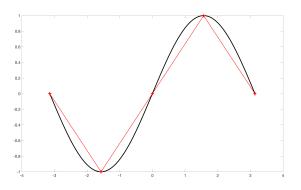
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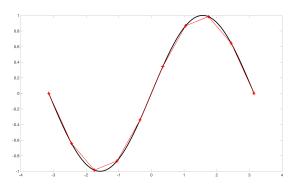
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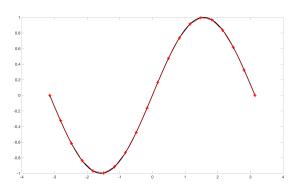
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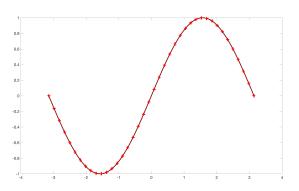
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Questions:

- * Why can we consider polynomials?
- * Should we use global or local approximations?

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Is a set of objects that can be added together and multiplied by scalars (of a given field) and yield a new object also in the set.

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Important concepts of a vector space:

Basis and dimension.

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Requirement:

The elements of the basis *B* must be linearly independent.

Dimension of a vector space

The dimension of the vector space is the number of elements in the basis. Note that any basis of V has the same number of elements.

Examples of vector spaces and basis

Example 1:

Three dimensional euclidean space.

- * Elements in the space are all vectors in \mathbb{R}^3 .
- * Example of a basis is the set of orthonormal vectors $(\hat{i}, \hat{j}, \hat{k})$.
- * Dimension of the space is 3.

Examples of vector spaces and basis

Example 2:

Polynomial space of degree *p*.

- * Elements in the space are all polynomials with degree at most *p*.
- * Example of a basis is the set of monomials $\{1, x, x^2, ..., x^p\}$.
- * Dimension of the space is p + 1.

Examples of vector spaces and basis

Example 3:

Space of infinitely smooth functions in 1D: $C^{\infty}(\mathbb{R})$.

- * Elements in the space are all 1D infinitely smooth functions.
- * Examples of basis functions are:
 - set of monomials $\{1, x, x^2, ...\}$.
 - trigonometric functions $\{e^{2\pi inx}\}_{n=-\infty}^{\infty}$.
- Dimension of the space is infinity.

Examples of vector spaces and basis

Example 4:

square integrable functions in 1D: $L^2(\mathbb{R})$.

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Some definitions

Dense set

A set S is dense in V if every element in V is either in S or is a limit point of S.

Limit point of a set S

The limit point of S is any point that can be approx. by points in S.

Example:

The rationals are dense in the real numbers.

Weierstrass approximation theorem:

Let f(x) be a real valued and continuous function in an interval [a,b]. Then for any $\epsilon > 0$, there is a polynomial p(x) in [a,b] s.t.

$$|f(x)-p(x)|<\epsilon,\quad \forall x\in[a,b]$$

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- * Smoothness properties of the function
- * Computational cost.

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Relevance within finite elements:

Since $L^2(\Omega)$ is more general than $C^0(\Omega)$, solutions with discontinuities can be represented by the L^2 space.

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In simple words, ... we can approximate for

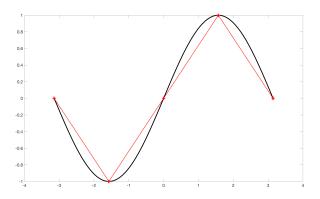
we can approximate functions in $L^2(\Omega)$ via (sequences of) continuous functions.

Summary so far:

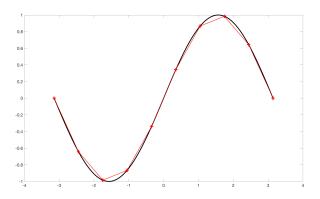
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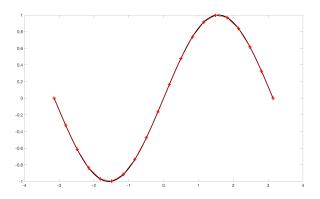
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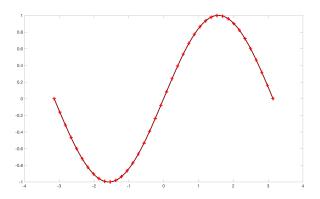
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- * Consider the interval [a, b].
- * Let f(x) be the function to be approximated.
- * Let $f_h(x)$ be a piecewise polynomial approximation over [a, b]. Then,

$$||f(x)-f_h(x)||_{L^2([a,b])} \leq C(b-a)^{p+1},$$

where p is the degree of the polynomial and C is a constant independent of u and |b-a|.

The most common (finite dimensional) space in finite elements is

$$\underbrace{V_h := \{ v \in C^0(\Omega) \mid v|_K \in \mathbb{P}^p(K) + \mathsf{BCs} \}}_{\text{(finite dim.) space of the finite element soln.}} \subset \underbrace{V := \{ v \in L^2(\Omega) + \mathsf{BCs} \}}_{\text{(infinite dim.) space of the exact soln.}}$$

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- * We need dim(V_h) linearly independent functions $\varphi \in V_h$.
- * Consider the 'hat' functions.
- * Interpolatory versus non-interpolatory basis functions.

Finite element solution

Given the basis $\{\varphi_i, j = 1, ..., \dim(V_h)\}$, the finite element solution is

$$u_h(x) = \sum_{n=1}^{\dim(V_h)} a_i \varphi_i(x),$$

where a_i are called degrees of freedom (DoFs) or control points.

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Note: if the basis is interpolatory; i.e., if $\varphi_i(x_i) = \delta_{ii}$, then

$$a_i = u_h(x_i)$$