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### Numerical Methods for CSE

### Problem Sheet 6

## Problem 1. Evaluating the derivatives of interpolating polynomials (core problem)

In [1, Section 3.2.3.2] we learned about an efficient and "update-friendly" scheme for evaluating Lagrange interpolants at a single or a few points. This so-called Aitken-Neville algorithm, see [1, Code 3.2.31], can be extended to return the derivative value of the polynomial interpolant as well. This will be explored in this problem.

$$dp = dipoleval(t, y, x)$$

that returns the row vector  $(p'(x_1), \dots, p'(x_m))$ , when the argument x passes  $(x_1, \dots, x_m)$ ,  $m \in \mathbb{N}$  small. Here, p' denotes the *derivative* of the polynomial  $p \in \mathcal{P}_n$  interpolating the data points  $(t_i, y_i)$ ,  $i = 0, \dots, n$ , for pairwise different  $t_i \in \mathbb{R}$  and data values  $y_i \in \mathbb{R}$ .

HINT: Differentiate the recursion formula [1, Eq. (3.2.30)] and devise an algorithm in the spirit of the Aitken-Neville algorithm implemented in [1, Code 3.2.31].

**Solution:** Differentiating the recursion formula [1, (3.2.30)] we obtain

$$p_{i}(t) \equiv y_{i}, \qquad i = 0, \dots, n,$$

$$p'_{i}(t) \equiv 0, \qquad i = 0, \dots, n,$$

$$p_{i_{0},\dots,i_{m}}(t) = \frac{(t - t_{i_{0}})p_{i_{1},\dots,i_{m}}(t) - (t - t_{i_{m}})p_{i_{0},\dots,i_{m-1}}(t)}{t_{i_{m}} - t_{i_{0}}},$$

$$p'_{i_{0},\dots,i_{m}}(t) = \frac{p_{i_{1},\dots,i_{m}}(t) + (t - t_{i_{0}})p'_{i_{1},\dots,i_{m}}(t) - p_{i_{0},\dots,i_{m-1}}(t) - (t - t_{i_{m}})p'_{i_{0},\dots,i_{m-1}}(t)}{t_{i_{m}} - t_{i_{0}}}.$$

The implementation of the above algorithm is given in file dipoleval\_test.m.

- (1c) For validation purposes devise an alternative, less efficient, implementation of dipoleval (call it dipoleval\_alt) based on the following steps:
  - 1. Use MATLAB's polyfit function to compute the monomial coefficients of the Lagrange interpolant.
  - 2. Compute the monomial coefficients of the derivative.
  - 3. Use polyval to evaluate the derivative at a number of points.

```
Use dipoleval_alt to verify the correctness of your implementation of dipoleval with t = linspace(0, 1, 10), y = rand(1, n) and x = linspace(0, 1, 100). Solution: See file dipoleval test.m.
```

### Problem 2. Piecewise linear interpolation

- [1, Ex. 3.1.8] introduced piecewise linear interpolation as a simple linear interpolation scheme. It finds an interpolant in the space spanned by the so-called tent functions, which are *cardinal basis functions*. Formulas are given in [1, Eq. (3.1.9)].
- (2a) Write a C++ class LinearInterpolant representing the piecewise linear interpolant. Make sure your class has an efficient internal representation of a basis. Provide a constructor and an evaluation operator () as described in the following template:

HINT: Recall that C++ provides containers such as std::vector and std::pair.

**Solution:** See linearinterpolant.cpp.

(2b) • Test the correctness of your code.

# Problem 3. Evaluating the derivatives of interpolating polynomials (core problem)

This problem is about the Horner scheme, that is a way to efficiently evaluate a polynomial in a given point, see [1, Rem. 3.2.5].

(3a) Using the Horner scheme, write an efficient C++ implementation of a function

```
1 template <typename CoeffVec>
2 std::pair<double, double> evaldp ( const CoeffVec & c, double x )
```

which returns the pair (p(x), p'(x)), where p is the polynomial with coefficients in c. The vector c contains the coefficient of the polynomial in the monomial basis, using Matlab convention (leading coefficient in c [0]).

Solution: See file horner.cpp.

(3b) • For the sake of testing, write a naive C++ implementation of the above function

```
template <typename CoeffVec>
std::pair<double,double> evaldp_naive ( const CoeffVec & c, double x )
```

which returns the same pair (p(x), p'(x)). This time, p(x) and p'(x) should be calculated with the simple sums of the monomials constituting the polynomial.

**Solution:** See file horner.cpp.

(3c) • What are the asymptotic complexities of the two implementations?

**Solution:** In both cases, the algorithm requires  $\approx n$  multiplications and additions, and so the asymptotic complexity is O(n). The naive implementation also calls the pow() function, which may be costly.

(3d)  $\odot$  Check the validity of the two functions and compare the runtimes for polynomials of degree up to  $2^{20} - 1$ .

Solution: See file horner.cpp.

### Problem 4. Lagrange interpolant

Given data points  $(t_i, y_i)_{i=1}^n$ , show that the Lagrange interpolant

$$p(x) = \sum_{i=0}^{n} y_i L_i(x), \quad L_i(x) := \prod_{\substack{j=0 \ j \neq i}}^{n} \frac{x - t_j}{t_i - t_j}$$

is given by:

$$p(x) = \omega(x) \sum_{j=0}^{n} \frac{y_j}{(x - t_j)\omega'(t_j)}$$

with  $\omega(x) = \prod_{j=0}^{n} (x - t_j)$ .

**Solution:** Simply exploiting the chain rule of many terms:

$$\omega'(x) = \sum_{i=0}^{n} \prod_{\substack{j=0\\j\neq i}}^{n} (x - t_j)$$

Since  $(t_i - t_j) = \delta_{i,j}$ , it follows  $\omega'(t_i) = \prod_{\substack{j=0 \ j \neq i}}^n (t_i - t_j)$ . Therefore:

$$p(x) = \omega(x) \sum_{j=0}^{n} \frac{y_j}{(x - t_j)\omega'(t_j)} = \sum_{j=0}^{n} \frac{y_j}{(x - t_j) \prod_{\substack{j=0 \ j \neq i}}^{n} (t_i - t_j)} \prod_{\substack{j=0 \ j \neq i}}^{n} (x - t_j)$$
$$= \sum_{j=0}^{n} \frac{y_j}{\prod_{\substack{j=0 \ j \neq i}}^{n} (t_i - t_j)} \prod_{\substack{j=0 \ j \neq i}}^{n} (x - t_j) = \sum_{j=0}^{n} y_j \prod_{\substack{j=0 \ j \neq i}}^{n} \frac{x - t_j}{t_i - t_j}.$$

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