

Reconstructing Sparse Trigonometric Functions

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Recently, the growing field of compressed sensing has attracted a great deal of attention to sparsity. In many signal processing applications, a compressible signal can be represented as a sparse trigonometric polynomial

$$p(x_1, \dots, x_n) = \sum_{(\theta_{j_1}, \dots, \theta_{j_n}) \in J_1} a_{j_1 \dots j_n} \cos(\theta_{j_1} x_1 + \dots + \theta_{j_n} x_n) \\ + \sum_{(\phi_{j_1}, \dots, \phi_{j_n}) \in J_2} b_{j_1 \dots j_n} \sin(\phi_{j_1} x_1 + \dots + \phi_{j_n} x_n), \quad J_1, J_2 \subset \mathbb{Z}_{\geq 0}^n,$$

and one seeks to reconstruct a sparse signal from a small number of samples.

The Chebyshev polynomial basis gives a recursive relation among trigonometric functions. In exact arithmetic, Lakshman and Saunders gave a sparse Chebyshev interpolation algorithm [7] that can be regarded as a generalization of the Ben-Or/Tiwari algorithm [1]. In a finite precision environment, a connection between the Ben-Or/Tiwari sparse interpolation and Prony's classical method [9] was observed by Giesbrecht, Labahn, and Lee [4]. A reformulation of Prony's method as a generalized eigenvalue problem [5] leads to a robust numerical sparse interpolation algorithm [4]. A generalized eigenvalue reformulation of sparse Chebyshev interpolation is presented in [3].

Based on Rutishauser's qd-algorithm [10], an iterative algorithm for multivariate sparse polynomial interpolation, in the standard power basis, is developed [2]. This is due to a connection [8] between the generating polynomial computed via the Berlekamp/Massey algorithm in the Ben-Or/Tiwari sparse interpolation and the convergence of Hadamard polynomials, defined in terms of the determinants of the associated Hankel matrices and polynomials [6, p. 625], in the qd-algorithm.

We extend such connection to trigonometric polynomials and the corresponding structured matrices and show how it is used for the reconstruction of a sparse trigonometric polynomial in the form $p(x_1, \dots, x_n)$. Our trigonometric reconstruction leads to an iterative algorithm for sparse trigonometric interpolation. The Hankel-plus-Toeplitz matrix appearing in the sparse Chebyshev interpolation problem [7] and its generalized eigenvalue reformulation [3], can be factored as a structured matrix product. From the solution of the generalized eigenvalue problem the frequencies $(\theta_{j_1}, \dots, \theta_{j_n}) \in J_1$ and $(\phi_{j_1}, \dots, \phi_{j_n}) \in J_2$ are obtained. The numerical rank of the structured matrices is connected to the number of terms in $p(x_1, \dots, x_n)$, in other words to the sparsity.

Example Consider the interpolation of

$$p(x) = a_1 \cos \theta_1 x + a_2 \cos \theta_2 x + b_1 \sin \phi_1 x + b_2 \sin \phi_2 x$$

from the evaluation of $p(x)$. The values of $a_1, a_2, \theta_1, \theta_2$ can be obtained from

$$\begin{aligned} p(0) &= a_1 + a_2, & p(1) + p(-1) &= a_1(2 \cos \theta_1) + a_2(2 \cos \theta_2), \\ p(2) + p(-2) + 2p(0) &= a_1(2 \cos \theta_1)^2 + a_2(2 \cos \theta_2)^2, \\ p(3) + p(-3) + 3(p(1) + p(-1)) &= a_1(2 \cos \theta_1)^3 + a_2(2 \cos \theta_2)^3; \end{aligned}$$

and b_1, b_2, ϕ_1, ϕ_2 from

$$\begin{aligned} p(1) - p(-1) &= 2b_1 \sin \phi_1 + 2b_2 \sin \phi_2, \\ p(2) - p(-2) &= 2b_1 \sin \phi_1(2 \cos \phi_1) + 2b_2 \sin \phi_2(2 \cos \phi_2), \\ p(3) + p(-3) + f(1) - f(-1) &= 2b_1 \sin \phi_1(2 \cos \phi_1)^2 + b_2 \sin \phi_2(2 \cos \phi_2)^2, \\ p(4) + p(-4) + 2(f(2) - f(-2)) &= 2b_1 \sin \phi_1(2 \cos \phi_1)^3 + 2b_2 \sin \phi_2(2 \cos \phi_2)^3. \end{aligned}$$

We interpolate $p(x)$ from 9 evaluations: $p(-4), \dots, p(0), p(1), \dots, p(4)$. \square

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