

Universidade do Minho Escola de Ciências

Fourier Analysis: Phonons in hBN

Alexandre Silva

João Henriques

Tiago Rodrigues

Laboratórios Avançados de Física

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Objectives

- Introduce the basics of Fourier analysis and the case of study (probing phonons in hBN using EELS);
- Apply Fourier analysis to the experimental and theoretical data;
- Understand what processes might have caused the differences between the experimental and simulated data;
- **Ultimate goal:** Understand how Fourier analysis is used to validate theoretical

models with experimental data;

Introduction...

Fourier Analysis

The Fourier transforms are defined as:



Fourier Analysis

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Case of Study: What & Why?

Case of study: Probing phonons in hexagonal boron nitride (hBN) using electron energy loss spectroscopy (EELS) - R. J. Nicholls *et. al*. PRB **99** 2019.

Phonons: Crystal lattice excitations. Why? To understand lattice Dynamics.

hBN: Most stable form of BN polymorphs, widely used in its 2D configuration.

EELS: Beam of electrons impinges on a sample and the transmitted electrons' energy is measured. **Why?** Allows us to study momentum depedence and nanometer spatial resolution.

DFT: Density functional theory is a powerful, widely spread method at the heart of today's computational physics research... and it is like a black box to us.

Case of Study: Details

Theory:

- 1. Scattering formalism is used to compute cross sections;
- 2. To simplify the problem some approximations are introduced;
- Density functional perturbation theory was used to compute the phonon eigenvalues.

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Experiment:

- A transmission electron microscope (STEM) was used to perform EELS;
- 2. The apparatus was callibrated no minimize electron damage;
- 3. No denoising or deconvolution routines were used.

Validation...

Learning The Tools



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Learning The Tools



It Works!

Uncertainty Principle



Smaller $\Delta t \Longrightarrow$ Bigger $\Delta \omega$

Uncertainty Principle



Cleaning the Signal...

Raw Data





General features of the data (R. J. Nicholls *et. al.* PRB **99** 2019):

- Two independent peaks;
- Different broadening;
- Initial noise.





Looking at the noise:





It is noise...

Data Fits



Experimental



Simulated





Data Fits



Smearing the Data...

Convolution

We start by convoluting the first peak with a Lorentzian. Why?

To broaden the peak. Produce a better fit to experimental data.



Overview

- We started by validating the Fourier analysis method with controllable examples;
- A frequency filter was applied to the experimental data, removing the noise and

smoothing the results;

• Experimental and simulated data were fitted with different distributions

indicating different line width regimes;

• The simulated data was broadened by means of a convolution with a Lorentzian presenting good agreement with experimental data.

Backup Slides...

Let us define the discrete Fourier transform as:

$$f(t) = \sum_{n=-\infty}^{\infty} \frac{1}{2\pi} C_n e^{i\omega nt}$$

The spectral coefficients C_n are easily obtained using:

$$\int_{-\infty}^{\infty} f(t)e^{-i\omega mt}dt = \sum_{n=-\infty}^{\infty} \frac{1}{2\pi} \int_{-\infty}^{\infty} C_n e^{i\omega(n-m)t}dt$$
$$\int_{-\infty}^{\infty} f(t)e^{-i\omega mt}dt = \sum_{n=-\infty}^{\infty} C_n \delta_{m,n}$$
$$\int_{-\infty}^{\infty} f(t)e^{-i\omega mt}dt = C_m$$

Let us once again use the definition:

$$2\pi f(t) = \sum_{n=-\infty}^{\infty} C_n e^{i\omega nt}$$

which may be re-written as:

$$2\pi f(t) = \sum_{n=1}^{\infty} C_n e^{i\omega nt} + \sum_{n=1}^{\infty} C_{-n} e^{-i\omega nt} + C_0$$

we now note that $e^{i\omega t} = cos(\omega t) + i sin(\omega t)$. Thus:

$$2\pi f(t) = \sum_{n=1}^{\infty} C_n [\cos(\omega t) + i \sin(\omega t)] + \sum_{n=1}^{\infty} C_{-n} [\cos(n\omega t) - i \sin(n\omega t)] + C_0$$

From here it is simple to show that f(t) can be put in the form:

$$f(t) = \frac{1}{2\pi} \left[C_0 + \sum_{n=1}^{\infty} A_n \cos(n\omega t) + B_n \sin(n\omega t) \right]$$

where $A_n = C_n + C_{-n}$ and $B_n = i(C_n - C_{-n})$. Some remarks:

- If f(t) is an even function of t, then all B_n must vanish and thus C_n = C_{−n};
- ▶ If f(t) is an odd function of t, then all A_n (and C_0) must vanish and thus $C_n = -C_{-n}$;
- The parity is conserved between the dual spaces;
- \triangleright C_0 gives the baseline.

Yet another way we could handle this expression is the following:

$$f(t) = \frac{1}{2\pi} \left[C_0 + \sum_{n=1}^{\infty} A_n \cos(n\omega t) + B_n \sin(n\omega t) \right]$$
$$= \frac{1}{2\pi} \left[C_0 + \sum_{n=1}^{\infty} D_n \cos(\phi_n) \cos(n\omega t) + D_n \sin(\phi_n) \sin(n\omega t) \right]$$
$$= \frac{1}{2\pi} \left[C_0 + \sum_{n=1}^{\infty} D_n \cos(n\omega t - \phi_n) \right]$$

An analogous procedure could be applied for a sin instead of a cos.