

Embeded silicon crystals: Calculating properties from Raman Spectra

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Abstract

In the present document we explore the characterization by means of Raman spectroscopy of a quartz matrix with embeded silicon nano crystals (Si-NC) A total of 20 Raman spectrums where taken along the x axis. The spectrums show a change in the shape of the LO and TO transitions due to a change in the crystal size. The analysis of the Raman Spectra where made using a python script available online to further discussion. The algorithm first takes an interval defined by the user k1 and k2 to work with it, denoise the signal, performs a base line calculation, and fits a multi peak signal which is the sum of three gaussian curves, but the algorithm can be extended to n gaussians. This information is collected in a plot of the x position and the position of the peaks related to the transitions and also the sigma (FWHM) related to the gaussian profiles.

1 Introduction

Silicon is an indirect band gap semiconductor (1.1 eV). Silicon has a high refractive index and low light absorption in the visible spectrum. It is transparent to infrared radiation, which makes it useful in optics and solar cell technologies. Crystalline silicon has a diamond-like crystal structure, known as a diamond cubic structure, the lattice constant is approximately 0.357 nm.

Silicon nanocrystals, also known as silicon quantum dots, possess unique properties at the nanoscale that make them useful in a variety of applications like: optoelectronics, where silicon nanocrystals can emit light in the visible and near-infrared range due to their quantum confinement effect; bioimaging and bioengineering, their small size, chemical stability, and tunable emission wave-

lengths make them ideal for labeling and tracking biological entities.

It is possible to identify the silicon crystal structure analyzing its Raman spectrum, the conservation of phonon momentum q in crystalline silicon leaves as Raman active mode only the zone center ($q = 0$) optical phonon at, which produces a single peak of high intensity. In the other hand, amorphous silicon exhibits a q -selection rule relaxation due to the loss in long range order, which produces broadening on the silicon main peak and shifting toward lower wave numbers.

To create silicon nanocrystals using sputtering, a silicon target is typically bombarded with inert gas ions, such as argon (Ar). The high-energy ion bombardment causes silicon atoms to be dislodged from the target surface. These silicon atoms

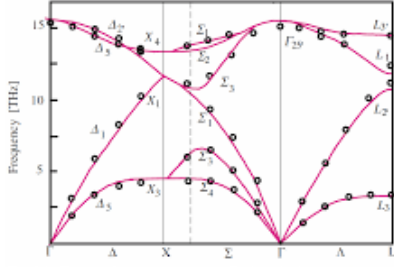


Figure 1: Silicon phonon dispersion diagram.

then condense and nucleate on the substrate, forming nanocrystals. Various parameters can be controlled during the sputtering process to influence the growth of silicon nanocrystals. These parameters include sputtering power, sputtering gas pressure, substrate temperature, and deposition time. By adjusting these parameters, researchers can tailor the size, density, and distribution of silicon nanocrystals in the thin film.

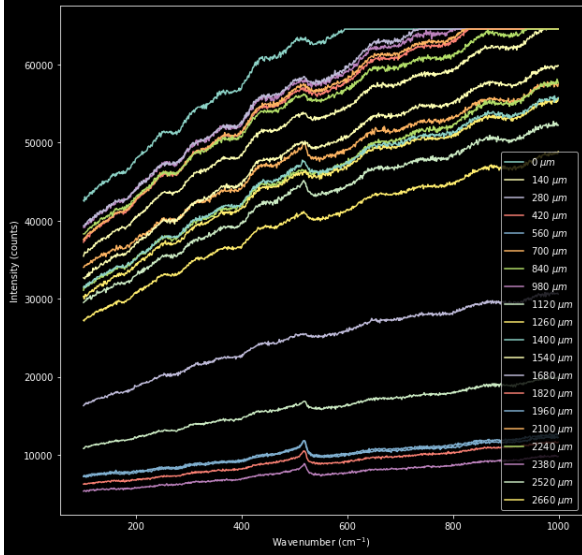


Figure 2: Raman spectra for the sample taken at different x positions.

2 Experimental Details

The sample in which the analysis was carried on is a quartz matrix with embedded silicon nanocrystals processed by Sputtering RF, with the density of the crystals varying along the x axis. A High Resolution Raman Spectrometer was used in this experiment (LabRam HR600) with an excitation wavelength of 633nm, Acquisition time of 10 seconds, 2 accumulations, wavenumber between (100,1000) and the Hole at 75%. The spectrums were taken with a separation in the x axis of $140\mu m$



Figure 3: Manipulation of the Raman Spectrometer to focus the excitation source on the surface of the sample

2.1 Analysis of the signal

As 'Figure 2' shows, the signals present a baseline that is a constant emission present in all the wavenumbers, to start the analysis we pick a range of wavenumber between 350 cm^{-1} and 600 cm^{-1}

After this we can start to fit the signal to a sum of three gaussian curves, this are defined by the following equation:

3 Results and discussion

The training process of GPT involves pre-training on large corpora and subsequent fine-tuning on specific downstream tasks. We describe the pre-training phase, which involves predicting masked tokens in a language modeling objective. We dis-

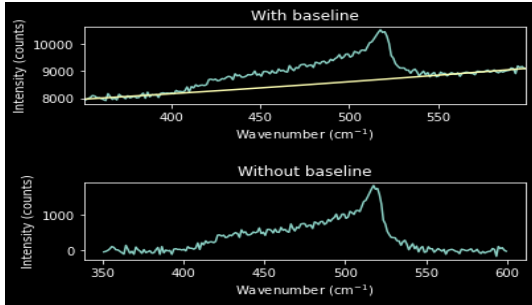


Figure 4: Raman Spectrum before and after doing the base line correction, in the software we fit a polinomial of degree 3 but it can be changed.

cuss the challenges and considerations in training GPT on massive datasets, as well as techniques to enhance its performance and efficiency. Furthermore, we delve into the fine-tuning process, where GPT is adapted to specific tasks through supervised learning.

$$\int_a^b \quad (1)$$

4 Applications of GPT

GPT has demonstrated remarkable versatility across a wide array of NLP applications. We provide an overview of the various tasks where GPT has excelled, such as language translation, text summarization, question answering, and sentiment analysis. For each task, we discuss the methodologies employed and the corresponding performance of GPT. We also examine the ethical implications and potential biases that arise when using GPT in real-world applications.

5 Strengths and Limitations

While GPT has achieved impressive results, it also exhibits certain limitations. We analyze the strengths and weaknesses of GPT in different contexts, including long-range dependency modeling,

handling rare or out-of-distribution words, and dealing with biased or harmful content. We highlight the challenges that researchers and practitioners face when working with GPT and propose potential avenues for improvement.

6 Conclusion

In this paper, we have presented a comprehensive analysis of GPT, examining its architecture, training process, and applications. GPT has significantly advanced the field of NLP, demonstrating its capabilities in generating coherent and contextually appropriate text. However, challenges remain in mitigating biases, improving long-range dependency modeling, and addressing ethical concerns. As the field progresses, further research and innovation will pave the way for the continued advancement and responsible use of GPT in diverse real-world applications.